

FOOD AND DRUG ADMINISTRATION
CENTER FOR DRUG EVALUATION AND RESEARCH

MEETING OF THE
DRUG SAFETY AND RISK MANAGEMENT ADVISORY COMMITTEE

8:05 a.m.

Thursday, December 4, 2003

Holiday Inn
Two Montgomery Village Avenue
Gaithersburg, Maryland

ATTENDEES

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ATTENDEES (Continued)

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SUZANNE COFFMAN, PHARM.D.
CLEMENT GALLUCCIO
BRUCE LAMBERT, PH.D.
PATRICIA STAUB, J.D., R.PH.
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C O N T E N T S

Issue: Current Screening Methods to Assess
 Sound-alike and Look-alike Proprietary
 Drug Names in order to Reduce the Incidence
 of Medication Errors Resulting from
 Look-alike and Sound-alike Names

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P R O C E E D I N G S

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DR. GROSS: Good morning, everybody. I'd like to start the meeting. If you plan on going home today, we should start the meeting now.

I am the chair of the Drug Safety and Risk Management Advisory Committee. My name is Peter Gross. I'm the Chair of the Department of Medicine, Hackensack University Medical Center.

We have a very interesting agenda today.

I'd like to go around and introduce the members of our advisory committee or have them introduce themselves. We will start with Brian Strom at my left.

DR. STROM: I'm Brian Strom from the University of Pennsylvania School of Medicine.

DR. CRAWFORD: Good morning. Stephanie Crawford, University of Illinois, Chicago, College of Pharmacy.

DR. HOLMBOE: Eric Holmboe from Yale University.

DR. LEVIN: Arthur Levin, Center for Medical Consumers.

DR. MORRIS: Lou Morris, Louis A. Morris and Associates.

MR. BLOOM: I'm Jeff Bloom from Washington,

1 D.C. I'm an AIDS patient advocate in Washington, D.C.

2 DR. DAY: Ruth Day, Duke University.

3 DR. COHEN: Mike Cohen, Institute for Safe
4 Medication Practices.

5 DR. GARDNER: Jacqueline Gardner, University of
6 Washington, School of Pharmacy.

7 DR. FURBERG: Curt Furberg, Wake Forest
8 University.

9 MS. SHAPIRO: Robyn Shapiro, Center for the
10 Study of Bioethics, Medical College of Wisconsin.

11 MS. JAIN: Shalini Jain, Executive Secretary
12 for the advisory committee, representing the FDA.

13 DR. GROSS: The two people from the FDA that
14 are at our table are Dr. Paul Seligman, who is Director of
15 the Office of Pharmacoepidemiology and Statistical Science,
16 and Acting Director of the Office of Drug Safety, and to
17 his left is Jerry Phillips, Associate Director of
18 Medication Error Prevention at the FDA.

19 Shalini Jain now will go over the conflict of
20 interest statement.

21 MS. JAIN: Good morning, everyone, and thanks
22 for attending our meeting today.

23 The following announcement addresses the issue
24 of conflict of interest with respect to this meeting and is
25 made a part of the record to preclude even the appearance

1 of such at this meeting.

2 The topic of today's meeting is an issue of
3 broad applicability. Unlike issues before a committee in
4 which a particular product is discussed, issues of broader
5 applicability involve many industrial sponsors and academic
6 institutions.

7 All special government employees have been
8 screened for their financial interests as they may apply to
9 the general topic at hand. Because they have reported
10 interests in pharmaceutical companies, the Food and Drug
11 Administration has granted general matters waivers of broad
12 applicability to the following SGEs, or special government
13 employees, which permits them to participate in today's
14 discussion: Dr. Michael R. Cohen, Dr. Ruth S. Day, Dr.
15 Curt D. Furberg, Dr. Peter A. Gross, Dr. Louis A. Morris,
16 Dr. Brian L. Strom.

17 A copy of the waiver statements may be obtained
18 by submitting a written request to the agency's Freedom of
19 Information Office, room 12A-30 of the Parklawn Building.

20 Because general topics could involve so many
21 firms and institutions, it is not prudent to recite all
22 potential conflicts of interest, but because of the general
23 nature of today's discussions, these potential conflicts
24 are mitigated.

25 In the event that the discussions involve any

1 other products or firms not already on the agenda for which
2 FDA participants have a financial interest, the
3 participants' involvement and their exclusion will be noted
4 for the record.

5 With respect to all other participants, we ask
6 in the interest of fairness that they address any current
7 or previous financial involvement with any firm whose
8 product they may wish to comment upon.

9 Thank you.

10 DR. GROSS: For the record, I'll read the main
11 issue being discussed today. Current screening methods to
12 assess sound-alike and look-alike proprietary drug names in
13 order to reduce the incidence of medication errors
14 resulting from look-alike and sound-alike names.

15 Now I'd like to reintroduce you to Dr. Paul
16 Seligman, Director of the Office of Pharmacoepidemiology
17 and Statistical Science and Acting Director of the Office
18 of Drug Safety.

19 DR. SELIGMAN: Good morning. It's a pleasure
20 this morning to welcome back our Drug Safety and Risk
21 Management Advisory Committee, those of you who are going
22 to be making presentations this morning, as well as all of
23 you who will be participating in today's discussion. Today
24 we have a full committee assembled, and I thank you all for
25 your time and effort and consideration in being here today.

1 Peter has introduced the topic up for
2 discussion for today which is to look at current screening
3 methods to assess similarities amongst proprietary drug
4 names. As some of you may realize, this topic was
5 scheduled for discussion on September 19th of this year.
6 This discussion seems to bring along the weather.
7 Unfortunately, the meeting was canceled because Hurricane
8 Isabel came roaring through and forced the last-minute
9 cancellation, and I apologize to those of you who either en
10 route or actually had arrived here in Washington just prior
11 to that last-minute cancellation.

12 At today's session we're going to be hearing
13 from several speakers who will elaborate on a number of
14 different drug screening methods. I'm looking forward to
15 exploring this issue with the help of Dr. Gross and the
16 other advisory committee members, as well as our guest
17 speakers. There are a number of questions that we will
18 formally pose to the committee for consideration which will
19 be presented at the end of these presentations and prior to
20 this afternoon's discussion. So once again, I'd like to
21 take this opportunity to welcome everyone again and thank
22 our committee.

23 With that, I think I will start the program
24 this morning by teeing up the first topic, which is
25 advancing the science of screening proprietary drug name

1 review.

2 The underlying basis for our discussion today
3 is that there are a substantial number of medication errors
4 that result from confusions caused by look-alike and sound-
5 alike names and confusing packaging and drug labeling.

6 In the 1999 report from the Institute of
7 Medicine, *To Err is Human*, the IOM report proposed that the
8 FDA require drug companies to test proposed drug names for
9 confusion.

10 In November of 2002, the Department of Health
11 and Human Services Committee on Regulatory Reform called
12 for the FDA to shift the responsibility for conducting this
13 kind of review and testing to the industry.

14 In June of this year, in cooperation with PhRMA
15 and the Institute for Safe Medication Practices, we held a
16 well-attended and interesting public discussion here in
17 Washington, which was really the first attempt to explore
18 the current methods to screen proprietary drug names for
19 similarities. It was an outstanding, interesting,
20 engaging, and robust discussion, and basically what we
21 heard was that the current approach, which is largely
22 qualitative, isn't consistent, nor can most approaches at
23 present be validated or reproduced. I'm going to talk in
24 greater detail about more of the comments that we heard in
25 that meeting, but that was sort of the overall message that

1 we got out of that discussion.

2 There is a variety of approaches that can and
3 have been used to screen drugs for proprietary names.
4 You're going to be hearing experts this morning sort of
5 delve into those particular topics. I'm going to take a
6 moment this morning to sort of talk about some of those and
7 some of the concerns and issues raised by these particular
8 methods.

9 The first method is the use of basically expert
10 committees, people knowledgeable in pharmacy, people
11 knowledgeable in issues related to behavioral sciences, et
12 cetera. Basically in the area of expert committees, which
13 is essentially assembling groups of 8 to 12 participants to
14 look at names, I think one of concerns that we have is that
15 there's not much research in these areas. If experts
16 panels are to be successful, they need to be run
17 consistently to be useful. There has to be an
18 establishment or clear understanding of what the baseline
19 level of expertise that is needed for these expert
20 committees. And as always, whenever you assemble groups of
21 people together to review things, there is a tendency for
22 group thinking, if you will.

23 There is a whole host of challenges related to
24 surveys and questionnaire designs, including how to design
25 surveys in anticipation of marketing a product prior to

1 that product actually being available, limits on experts'
2 ability to predict errors, the need to consider how one
3 might develop simulated circumstances that accurately
4 reflect a pharmacy or prescribing environment, and what
5 ways one might consider the use of focus groups in
6 generating ideas, although clearly these are approaches
7 that are, by their nature, weak in evaluating individual
8 reactions to stimuli.

9 The engineering world uses a variety of
10 approaches in failure mode and effect analysis that range
11 from picking expert committees and teams to detailing of
12 flow charting processes to determine root cause analyses of
13 errors, to using using tools that systematically go through
14 each step to determine essentially what's not working and
15 why it's not working, and to assign a level of severity, as
16 well as visibility, for a particular problem. The degree
17 to which these kinds of techniques can be applied to
18 evaluating and assessing proprietary names has yet to be
19 tested, but I think there are many lessons to be learned
20 from the world of failure mode and effect analyses.

21 There is a variety of handwriting recognition
22 techniques that combine certain basic elements of
23 handwriting that are similar to all handwriting techniques
24 that involve pattern recognition of writing a proposed name
25 and developing databases of graphic patterns for all

1 existing drug names to make a comparison. So we'll hear
2 about some of these today as well.

3 There are also computational linguistic
4 techniques that can be applied. This is an area that we at
5 the FDA have been particularly interested and have worked
6 closely with a contractor to develop a system which allows
7 us to systematically screen the names using a software
8 algorithm that allows us to look at phonetic strings and
9 groups of letters and to do essentially orthographic and
10 phonological matching and screening of names.

11 It's also possible to consider standard study
12 design and sampling techniques. You'll hear a little bit
13 this morning of the approach that we use at the FDA to
14 essentially conduct our own internal sampling of names.
15 Although this is the approach that we use and I think we've
16 used it with some degree of success, there clearly needs to
17 be some standardization of this approach, tests for
18 reliability and reproducibility and validity since the work
19 that we do at the FDA, while valuable, does not have a gold
20 standard against which we can measure the results of our
21 work.

22 As I indicated, there is also a variety of
23 computer-assisted decision-based analyses that can be a
24 powerful driver in terms of looking at prescribing
25 frequency, looking at potential harm that certain name

1 confusions can cause, as well as developing objective
2 measures to demonstrate reliability and predict the
3 probability of human error.

4 Another key issue for us in this era of risk
5 management is what role risk management programs play. Are
6 there situations where certain name confusions, because of
7 the potential risks of the drugs, may be more acceptable
8 than in other situations where a potential name confusion
9 can be devastating or life-threatening?

10 Clearly in an era where we are looking at all
11 elements of managing risk and how to validate and
12 understand how these elements and tools function and how
13 well these plans work, we're clearly interested in knowing
14 as well whether risks associated with names and naming can
15 also be managed in the post-marketing environment and
16 whether one could design risk management plans around
17 limiting errors associated with potential confusions of
18 names. Many of the elements in our upcoming risk
19 management guidance talk about the need to demonstrate
20 baselines of error, demonstrating goals for programs and
21 measuring the success of these programs. Can these
22 techniques and principles be applied as well to errors and
23 problems caused by name confusion?

24 So basically at the public hearing last June,
25 we heard I think the following major themes.

1 First, the need to adopt a more systematic
2 process with standardized tools for evaluating proprietary
3 names.

4 Second, we heard that all products made
5 available to patients, whether they are prescription or
6 over-the-counter drugs, should be held to the same standard
7 of testing.

8 There is a need to try to simulate these kinds
9 of situations that reflect real-life drug order situations
10 to really evaluate in a realistic fashion the potential for
11 problems in naming confusion.

12 Indeed, the study designs, to the degree they
13 can, should replicate medication order situations where
14 there are known error vulnerabilities.

15 And how medication orders, for example, are
16 communicated can either be improved to reduce the potential
17 for errors and how current medication order communication
18 scenarios contribute to the propagation or continuation of
19 those errors.

20 Particularly in the area of pediatrics, if one
21 is looking at pediatric patients, it's important to not
22 only look at confusions associated with the name, but also
23 issues related to how well communication is managed in
24 terms of the strength, the quantity, and the directions of
25 use, as well as critical prescribing information, such as

1 patient age and weight.

2 There must be study methods that can be
3 scientifically validated, reproduced, and that are
4 objective and transparent to all.

5 One of the issues that was also raised at the
6 public hearing, which we are not going to address today, is
7 the issues of suffixes and prefixes associated with drug
8 names which also have the potential and, indeed, to
9 contribute to the problem of medication errors, nor will we
10 be dealing today issues associated with over-the-counter
11 family names and drug names that are marketed based on
12 consumer recognition that lead also to consumer confusion.

13 So basically the major theme is that we feel
14 that there is inconsistency in how name testing is
15 currently conducted, that there is the need to produce
16 valid and reproducible findings. You'll hear today that
17 while all methods offer some value, we need to think about
18 how to use these methods probably in a complementary
19 fashion to come up with ways to prevent unneeded confusion
20 once a product is marketed.

21 Following this open public meeting today, we
22 will take both the results of the input we receive from the
23 public as well as from our advisory committee, summarize
24 these, as well as what we learned from June, and then look
25 at the degree to which we can come up with a guidance to

1 industry that will provide them direction on how best to
2 conduct pre-marketing testing and to communicate those
3 results and data to the FDA.

4 Today following my presentation, we're going to
5 hear from Jerry Phillips about the way we approach name
6 testing at the Food and Drug Administration. We'll be
7 hearing from a representative from PhRMA to talk about
8 industry's approach, and then hear from five experts who
9 are listed on the agenda talking about a variety of
10 techniques that are currently being used to evaluate names.

11 We've asked each one of our expert panelists to
12 provide an overview of each method, to discuss how that
13 method should be validated, to determine how a study design
14 can be used to evaluate how drug names can be studied to
15 reduce medication errors, and the strengths and weaknesses
16 of each of those methods.

17 Today we will consider the pros and cons of
18 also taking a risk-based approach to testing proprietary
19 names, to identifying the critical elements of each method
20 to be included in good naming practices as part of a
21 guidance document, to describe circumstances when field
22 testing would be important and should be required to
23 indicate whether one method should stand alone, and to
24 describe circumstances when it would be appropriate to
25 approve a proprietary drug name contingent on a risk

1 management program.

2 Thank you all very much and I will now turn the
3 proceedings over to Dr. Gross.

4 DR. GROSS: Thank you, Dr. Seligman.

5 The next speaker is Robert E. Lee, Jr.,
6 Assistant General Patent Counsel at Eli Lilly and Company.

7 He is going to talk on views on trademark evaluation.

8 He's representing PhRMA.

9 MR. LEE: Thank you for this opportunity to
10 share PhRMA views on pharmaceutical trademarks.

11 I would like to start with an echo from the
12 June 26th, 2003 public meeting that PhRMA was honored to
13 co-sponsor with FDA and ISMP. Among the points in my
14 closing comments at that session was the observation that
15 the role of trademarks in medication errors remains
16 unknown. We do know that trademarks are part of most
17 medication error reports, not necessarily as the cause, but
18 as a convenient identifier for the products involved.
19 PhRMA companies are interested as anybody in seeing
20 medication errors eliminated. We believe that methods used
21 by most PhRMA sponsors are an effective method for
22 developing trademarks that help prevent medication errors.
23 We are willing to work with the FDA and others on
24 validated, improved methods, if it is possible that such
25 can be developed.

1 Pharmaceutical trademarks are very visible and
2 because they are so visible, they make an easy target for
3 blame and criticism. The expression, "trademarks cause
4 medication errors," has become an unchallenged part of
5 regulatory language. Since PhRMA has not been able to find
6 scientific support for the assumption, we think that this
7 characterization is an overstatement and this is the time
8 and place for it to be respectfully challenged.

9 Individuals inside and outside the FDA may
10 unknowingly criticize trademarks when they use and overuse
11 the expression "problem name pairs." For example, during
12 the June 26th public meeting, Cozaar and Capoten were
13 described as a problem name pair because they were involved
14 in medication error. Cozaar and Capoten may have been
15 involved in a medication error, but we do not agree that
16 they are confusingly similar.

17 I have five points I'd like to cover this
18 morning.

19 Point number one. Pharmaceutical trademarks
20 support medication safety. The very essence of a trademark
21 is to distinguish one manufacturer's goods from those of
22 another. To do this effectively, trademarks must be
23 distinctive and unique. It is this distinctiveness that
24 serves to avoid confusion among current users and future
25 users. This benefits both the manufacturer of the product

1 and the consumer of the product. Later on I will discuss
2 in more detail the hard work that many manufacturers expend
3 to develop pharmaceutical trademarks.

4 Distinctive and unique pharmaceutical
5 trademarks support medication safety because there are no
6 better product identifiers than trademarks. Nonproprietary
7 names such as USANs and INNnS use a stem system that is
8 designed to group products together that have therapeutic
9 class similarity. This creates a built-in similarity for
10 generic names using the same stems.

11 Numbers would be a poor choice for product
12 identifiers, and combinations of numbers and letters would
13 probably be worse. Note that public internet addresses
14 changed from the internet protocol addresses that used
15 strings of numbers and letters to mainly alphabetical
16 domain names that are easy to pronounce and remember.

17 As noted earlier, we are not able to find solid
18 scientific data to show the role that trademarks play in
19 medication errors, but it is easy to find public
20 statements, news reports, and trade publications that echo
21 the assumption that 12.5 percent of medication errors
22 reported to FDA are a result of confusion between drug
23 names. Yes, trademarks are involved in medication errors,
24 but the involvement is most often in the convenient
25 reporting of the errors, not the causes.

1 For example, the name pair Clinoril and Oruvail
2 is among the several hundred problem name pairs listed in
3 the USP Quality Review publication. Another pair among
4 those listed is Cozaar and Zocor. We can all assume that
5 well-meaning practitioners reported errors or near misses
6 involving these trademarks, but we should not assume that
7 these trademarks are so confusingly similar that they
8 caused the problem.

9 FDA states that there are more than 700 problem
10 name pairs, but only some of them contain two trademarks.
11 Some contain a trademark and a generic name, and still
12 others contain two generic names.

13 Rather than having the profession and public
14 believe that trademarks cause medication errors, shouldn't
15 we pause to perform a differential analysis to better
16 understand the relative roles of the many factors involved
17 in medication errors? PhRMA agrees that more work must be
18 done to prevent or minimize medication errors. However,
19 putting an inappropriate focus on trademarks, while
20 ignoring other factors, gives a false sense of security
21 that something significant is being done to reduce
22 medication errors, while the underlying causes continue to
23 put patients at risk.

24 Improvements at the prescription level are
25 needed. One such initiative is legislation enacted in July

1 2003 in Florida that requires physicians to print
2 prescriptions legibly. Another is similar legislation
3 enacted by Washington State.

4 A number of promising improvements at the
5 dispensing level were described by the late Dr. Tony Grasha
6 at the University of Cincinnati. His research demonstrated
7 that dispensing errors can be reduced by changes in the
8 pharmacy work environment such as the use of prescription
9 copyholders at eye level, limiting pharmacist workload,
10 adequate lighting, improved equipment, et cetera.

11 These and other initiatives at the prescribing
12 and dispensing areas hold promise to reduce medication
13 errors.

14 Point number two. There is a highly effective
15 method for developing pharmaceutical trademarks. The
16 current method used by sponsors for developing new
17 trademarks has been refined over the course of two
18 centuries under the common law and trademark statutes. It
19 is the most reliable method we know for determining whether
20 two trademarks are likely to be confused by prescribers,
21 dispensers, or consumers of the product.

22 During the early years, the central issue of
23 likelihood of confusion was generally decided by comparing
24 the various characteristics, similarities, and
25 dissimilarities of the marks and the goods. But over time,

1 analysis of likelihood of confusion became more
2 sophisticated and continues to evolve.

3 For example, in recent years most PhRMA
4 companies seek input from health practitioners on the front
5 lines so as to take into account various factors such as
6 the frequency of prescribing, the consequences if products
7 are mixed up, the dosage form, dosage strength, dosing
8 regimen, delivery system, dispensing environment, the end
9 user, et cetera.

10 Fact-based expert opinions made by trademark
11 attorneys are also enhanced by continuous feedback from the
12 judicial system. This judicial experience on issues of
13 confusing similarity teaches us that the likelihood of
14 confusion is a fact-driven expert determination.
15 Similarity is a factor, but only one factor. Ultimately,
16 trademark attorneys and judges apply many factors to all of
17 the facts to reach a decision, and the decisions rest on
18 the reliability and the relevance of the facts.

19 Through the research and writings of Dr. Bruce
20 Lambert, we have some evidence that the industry is doing a
21 reasonably good job of safely adding new trademarks to
22 those already in use. Using various research tools to
23 measure orthographic similarity, like trigram analysis, Dr.
24 Lambert concluded that contrary to some impressions that
25 the drug lexicon is getting too crowded, the evidence

1 presented suggests that most pairs of drug names are not
2 similar to one another. This was in Dr. Lambert's paper,
3 An Analysis of the Drug Lexicon.

4 Point number three, creative development and
5 related activities. Creating distinctive and unique
6 trademarks is a carefully constructed process that begins
7 as long as four to six years before product launch and
8 involves a great deal of sponsor resources.

9 There are some differences among sponsors, but
10 the overall approach begins with creating long lists of
11 candidates. These can come from internal resources or from
12 outside vendors with extensive experience in trademark
13 creation. It is not unusual for the initial list to
14 contain several hundred candidates. These long lists are
15 narrowed through an internal process where the emphasis is
16 on eliminating candidates because they have potential
17 safety risks or other problems. As the list is narrowed to
18 a workable number of about 30 candidates that the sponsor
19 believes are appropriate for the product profile, they are
20 put through a more intensive screening process with
21 increasing emphasis on similarity to other trademarks,
22 generic names, medical terms, et cetera. Trademark
23 candidates must survive the safety screens along with
24 evaluations from legal, regulatory, linguistic, and
25 commercial perspectives.

1 Trademark clearance is a detailed process that
2 involves four stages, each of which weeds out candidates
3 that have an unacceptable similarity to other trademarks
4 based on an experienced analysis of the data. We not only
5 compare candidates with trademarks that are on the market,
6 but also those in the official trademark registration files
7 in the U.S. and other countries around the world.

8 Stage one deals primarily with look-alike and
9 sound-alike similarity and relies on search engines that
10 are powered by sophisticated algorithms. For example, a
11 typical approach is to sort trademarks by prefix, infix,
12 and suffix using Boolean logic to combine letter strings
13 into various configurations. This is an interactive
14 process whereby the expert searcher changes the searching
15 strategy depending on the results from the previous search
16 run. This process continues until the searcher is
17 convinced that the most relevant preexisting marks have
18 been found in the database.

19 Another approach relies more on sophisticated
20 phoneme analysis to measure phonetic similarity. Pat
21 Penyak was going to be here from Thompson & Thompson to
22 speak a little bit at the public session on what Thompson &
23 Thompson does researching. Unfortunately, Pat was in an
24 automobile accident, so she's not going to be here. I
25 understand she's fine. I think there will be someone else

1 from T&T here today.

2 Comprehensive search reports are the raw data
3 that is analyzed by trademark attorneys who perform an
4 expert evaluation of similarity issues from both the visual
5 and phonetic perspectives.

6 Stage two of the clearance process involves
7 input from front-line practitioners who supply insights
8 into how the trademarks will be used in a clinical setting.
9 In addition to name similarity, the input from the clinical
10 environment covers such elements as: frequency of
11 prescribing, that is, popularity of the product; route of
12 administration, dosage form, dosage strength, the usual
13 regimen, clinical indications which hold important
14 information about patient issues, storage, special
15 preparation requirements, dispensing environment, generic
16 name.

17 Stage three deals with forming the expert
18 opinion. Once the searching and fact-gathering are
19 complete, the sponsor team, comprising various disciplines
20 such as legal, regulatory, clinical, and marketing, applies
21 these various factors to all the facts available.

22 Pharmacists provide relevant input about the
23 clinical and dispensing environment.

24 The legal searching provides insights into the
25 look-alike and sound-alike similarity of other trademarks

1 with earlier priority rights.

2 Marketing and linguistic input identifies marks
3 that are suitable for the relevant universe of prescribers,
4 dispensers, and patients.

5 All of these inputs provide the resources for a
6 fact-driven expert judgment about the suitability of the
7 trademark for use on the product under consideration. It
8 is only after all of this work is completed and all the
9 results reviewed that a decision is made on which
10 trademark, among the few survivors, will be adopted and
11 moved to the next stage.

12 Stage number four, the final stage in the
13 process, involves the filing of an application for
14 registration in the U.S. Patent and Trademark Office. Even
15 with all the searching and fact-gathering that formed the
16 basis for the selection decision, there are more reviews
17 and hurdles ahead. Typically all pharmaceutical trademarks
18 are filed in class 5 at the Patent and Trademark Office.
19 This class contains more than 150,000 applications or
20 registrations in the U.S. alone, more than a million
21 worldwide.

22 PTO examiners who are experienced in reviewing
23 pharmaceutical trademarks conduct an independent search of
24 the candidate trademark for confusing similarity. These
25 examiners, working in class 5, apply a higher standard for

1 pharmaceutical trademarks due to public health concerns.

2 If the examiner finds the trademark acceptable
3 under the PTO review standards, the trademark is published
4 in the Official Gazette, a weekly publication that contains
5 all trademarks recently filed. Competitors and others
6 routinely review the Official Gazette to see if any of the
7 trademarks published might be unacceptably similar to their
8 own marks.

9 If a published trademark is determined to be
10 unacceptably similar to the owner of the trademark with a
11 priority right at the PTO, the owner can file a notice of
12 opposition which stops the PTO approval process until the
13 opposition is resolved by adjudication or settlement.

14 In a situation where an issue of confusing
15 similarity arises between two trademark applications, it is
16 necessary to determine who has the right to register the
17 mark. In the U.S. and all other countries, trademark laws
18 provide that the first to file an application has priority
19 over the later-filed trademark application.

20 The national trademark systems are tied
21 together by treaty so that priority is assigned to the
22 first filed application in any one of the treaty countries.
23 This is an important matter and has legal implications if
24 overridden by a priority scheme not endorsed by Congress.

25 Point number five, promise and pitfalls of

1 computer technology. We learned that FDA is working with
2 the Project Performance Corporation to develop a web-based
3 drug comparison system called POCA, an acronym for Phonetic
4 and Orthographic Computer Analysis. New and improved
5 software tools and databases can support the process of
6 trademark selection. PhRMA looks forward to being part of
7 the development of the new software so that it can be
8 integrated into work being done by commercial vendors with
9 similar interests.

10 We do see some serious pitfalls with the POCA
11 project. The first is the fear that FDA would not openly
12 share the system with sponsors. We think it is important
13 for sponsors to have the option of integrating any new FDA-
14 sponsored software into existing trademark evaluation
15 processes. The second is the fear that FDA would use
16 output from POCA to second guess the decisions about
17 trademark acceptability made by sponsors who follow the
18 processes that I described earlier.

19 Recommendations. In closing, I would like to
20 make four recommendations.

21 One, FDA should recognize the intrinsic value
22 of trademarks that make it possible for billions of
23 prescriptions to move through the dispensing and
24 administration process error-free. In addressing the small
25 percentage of prescriptions that result in medication

1 error, FDA and others should focus resources on the major
2 unaddressed causes of these errors.

3 This is number two. For all the reasons I've
4 given today, FDA should recognize the value of the current
5 methods employed by sponsors to develop clear and adopt new
6 trademarks for pharmaceutical products as an effective
7 working model of good naming practices. The current
8 process includes review and judgment by front-line
9 practitioners, the sponsor trademark attorney, the PTO
10 examiner, and competitors before a trademark is adopted.
11 Careful consideration should be given to the extent of
12 further trademark review by FDA so as to avoid moving
13 beyond the point of diminishing returns.

14 Number three, FDA has an interest in making
15 sure that pharmaceutical product names are chosen with care
16 and should exercise its regulatory leverage in seeing to it
17 that sponsors select trademarks carefully. FDA should
18 establish guidelines, based on the sponsor process
19 described earlier and insure that the guidelines are
20 followed.

21 FDA should encourage the development of
22 improved computer software tools, more comprehensive
23 databases, and additional research so long as FDA
24 recognizes that the process for determining the suitability
25 of a new trademark is largely a fact-based expert judgment

1 that should be made by those who have the professional
2 expertise.

3 Thank you for your kind attention, and I'll be
4 here all day for any questions.

5 DR. GROSS: Thank you very much, Mr. Lee.

6 Next we will hear from Jerry Phillips who is
7 Associate Director of Medication Error Prevention at the
8 Office of Drug Safety. He will present the FDA's approach
9 to proprietary name evaluation.

10 MR. PHILLIPS: Thank you. I'm going to talk a
11 little bit about a couple of things. I'm going to give
12 some definitions. I'm going to tell you a little bit about
13 our perspective as far as the seriousness of the issue and
14 then our process for evaluation at FDA.

15 First, let's start off with the definition of a
16 medication error. This definition comes from the National
17 Coordinating Council for Medication Error Reporting and
18 Prevention and it has also been proposed in the SADR rule
19 by FDA. Basically the key word here is that it's a
20 preventable event that may cause or lead to inappropriate
21 medication use or patient harm while the medication is in
22 the control of a health care professional, a patient, or a
23 consumer.

24 FDA focuses on medication errors that relate to
25 the safe use of a drug product. In its perspective, that

1 includes the naming, the labeling, and/or packaging of a
2 drug product that might contribute to an error.

3 A proprietary name by definition is a name
4 that's owned by a company or an individual and is used for
5 describing its brand of a particular product. It's also
6 known as a brand name or a trademark.

7 We just heard some of the statistics on the 700
8 name pairs. I acknowledge that both proprietary and
9 generic names are part of that list. Some of those are
10 actual errors and some of them are potential errors that
11 are on this USP list of 700 drug names.

12 To date about 25,000 medication error reports
13 have been received by FDA. When we look at the database,
14 we do a root cause analysis of those events and determine
15 the causes of those. From the aggregate data,
16 approximately 12.5 percent of the errors are related to the
17 names. This is from the reporter's perspective of the
18 cause of the event.

19 FDA, myself and others on the staff, publish
20 mortality data that was collected from 1993 to 1998 and was
21 published in the American Journal of Health System
22 Pharmacists on October 1, 2001. Of this data, we had 469
23 fatalities due to medication errors. A breakdown of this
24 is 16 percent of the deaths were due to receiving the wrong
25 drug product. Now, receiving the wrong drug product

1 doesn't mean it's necessarily related to the wrong name. A
2 physician could write for the wrong drug and that product
3 could be administered. But if we look at proprietary name
4 confusion and generic name confusion, 5 percent of the
5 deaths were caused by proprietary names and 4 percent by
6 generic names.

7 There are many, many causes of medication
8 errors such as lack of communication, use of abbreviations,
9 handwriting, lack of knowledge. There are many, many
10 reasons.

11 Some of the other reasons include similar
12 labels and labeling. In this particular picture, what you
13 see is a blue background. You see red lettering. You see
14 a standardized format on these particular bottles, and this
15 can lead to selection errors.

16 In this particular case, these are ophthalmic
17 drug products manufactured by one particular company, and
18 you can see the similarity across the different products
19 that increases the chance for selection errors.

20 This is an example of an over-the-counter drug
21 product. This is that OTC family trade name issue that
22 we're not going to talk about today. But basically it's a
23 similar labeling and packaging. These two drug products
24 have different active ingredients. One is oxymetazalone.
25 The other one is phenylephrine. They both have different

1 durations of action, and it has led to confusion.

2 Names that don't seem to be similar, Avandia
3 and Coumadin, when written sometimes do look very, very
4 similar and have resulted in errors. This is an example of
5 a prescription written for Avandia 4 milligrams every day
6 and Coumadin 4 milligrams every day. The similarity,
7 having both identical strengths, both being written for
8 every morning increases the risk of a medication error when
9 these names are written together and have resulted in
10 errors.

11 So what is FDA looking for when we look at
12 trade names? There are basically two things. We look for
13 sound-alike/look-alike properties of that name and we also
14 look for promotional and misleading claims associated with
15 that proprietary name.

16 For sound-alike/look-alike properties we're
17 looking at currently marketed and unapproved drug products
18 that we have in the pipeline. We're also looking to other
19 medicinal products and to commonly used medical
20 abbreviations, medical procedures, and lab tests.

21 So what's the information that we need in order
22 to do our risk assessment? Of course, we need to know the
23 proprietary or trademark and its established name. We also
24 need to know how it's going to be dosed, its strength, its
25 dosing schedule, its use and its indication, its labels and

1 labeling. If there's a device involved, we ask for the
2 working device model, and we also look at the formulation
3 and the packaging proposed, along with the trademark.

4 This is a busy schematic flow of the process at
5 FDA. There's a request for a proprietary name consult that
6 comes from the product sponsor, and that is at any time
7 from phase II of an IND to the filing of the NDA, the
8 sponsor requests the name through that IND or NDA, and it
9 is then filed in the reviewing division. A project manager
10 will consult the Office of Drug Safety or the Division of
11 Medication Errors and technical support in that office.

12 The review, which I'll go into a little bit
13 more detail, is a multi-faceted review that starts off with
14 an expert panel. We use computer analysis, POCA, which was
15 mentioned earlier, and prescription drug studies. Then a
16 risk assessment by a safety evaluator on DMETS's staff is
17 done that takes into account all this data. The review
18 goes to a team leader, a deputy director, and the associate
19 office director. Recommendation is then given back to the
20 reviewing division who reviews our consult. They either
21 agree or disagree with it and then provide that information
22 back to the sponsor.

23 As I just mentioned, the analysis consists of
24 an expert panel, a computer analysis which looks at the
25 orthographic/phonetic similarities of a name. We search

1 other external computer databases. We perform prescription
2 drug studies. These are simulated prescription studies
3 that try to simulate the real world as far as prescribing
4 practices, which include a verbal order, an outpatient
5 written prescription, and an inpatient written
6 prescription. And then we provide an overall risk/benefit
7 assessment based upon the information that we've collected.

8 The expert panel consists of approximately 12
9 of the DMETS safety evaluators. This includes a physician,
10 pharmacists, nurses, and one DDMAC representative. That's
11 for advertising that renders an opinion for misleading or
12 promotional claims.

13 There is a facilitator in this expert panel
14 that is randomly selected and rotated.

15 Each expert panel member reviews reference
16 texts, computers, and provides a relative risk rating for
17 each name prior to the meeting.

18 Then there is a group discussion at the expert
19 panel and there's a consensus that's built on each
20 particular name.

21 From this, we design prescription drug studies.

22 From the expert panel, there may be several names that
23 have been identified by those experts of marketed drug
24 products that might be confusingly similar. And from that,
25 we design these studies where we will write an outpatient

1 prescription with the proposed name and an inpatient
2 prescription written and also a verbal order.

3 The prescription study designs are developed
4 specifically for failure mode. In other words, we stress
5 the tester by randomly selecting different types of
6 handwritings, using actual practice standards. Instead of
7 putting an indication on a prescription, we would leave
8 that indication off because putting the indication on
9 necessarily doesn't reflect normal current practice and it
10 would also lead the analysis in a different direction so
11 that you wouldn't get an error necessarily.

12 We have various staff members that are asked to
13 write sample prescriptions for each name. There is a
14 marketed drug or control prescription that's also included
15 in the prescriptions so that the tester knows that they're
16 evaluating unapproved drug products, but also we'll put in
17 some marketed drugs. Sometimes we'll include marketed drug
18 products that are known error pairs to validate the
19 prescription studies.

20 The prescription is scanned and then they're e-
21 mailed to a subset of FDA health care workers. Their
22 interpretations are e-mailed back to us in writing.

23 There are about 130 FDA physicians, nurses, and
24 pharmacists across the centers that respond by this e-mail
25 system with their interpretations and comments. To

1 eliminate any one reviewer from reviewing a name more than
2 once, we divide the entire group into thirds where the n is
3 approximately 43 to review each verbal order, written, and
4 outpatient prescription order. The response rate is
5 usually around 70 percent.

6 This is an example of a product that we had on
7 a scientific round. This was not a proposed name by a drug
8 company. It was called Novicar. The top prescription is
9 an example of the prescriptions that we normally scan for
10 our participants. In this case, we had written out the
11 patient's name and the date, Novicar 40 milligrams, 1 PO
12 every day, #30, and Dr. Opdra at that time.

13 The bottom is example of an inpatient order
14 that we wrote for this study that gives the diet of the
15 patient, blood work, a DC order, and the Novicar is put in
16 there also. The lined orders on an inpatient order present
17 different types of errors because of the lined orders, and
18 that's why we duplicate both.

19 Just to back up, on this particular study we
20 actually discovered that there were lots of errors with
21 Novicar with -- oh, shoot. I just forgot. I'll come back
22 to it.

23 VOICE: Narcan.

24 MR. PHILLIPS: What was it?

25 VOICE: Narcan.

1 MR. PHILLIPS: Narcan.

2 On verbal orders, randomly selected DMET staff
3 are asked to record a verbal prescription via telephone
4 recorder. An example. This is Dr. Dee Mets and I'm
5 calling in a prescription for Jane Doe for Novicar 40
6 milligrams. I want to give 30 with two refills. And
7 that's recorded and then sent to the group of physicians
8 and nurses and pharmacists on the prescription drug
9 studies. Then after they hear that, they e-mail us back
10 their interpretations.

11 We also use a phonetic and orthographic
12 computer analysis. This is a recent software that we have
13 contracted. We abbreviate it as POCA. It's a set of
14 phonetic and orthographic algorithms that are used for an
15 automated and computerized method for evaluating trade
16 names for their similar sound-alike and their look-alike
17 properties. The prototype has been completed and is in
18 operation currently and is being used routinely in DMETS's
19 reviews. We are also working on validating this prototype
20 and hope to have that completed soon.

21 POCA provides a percentage ranking of
22 orthographic and phonetic similarity between the proposed
23 name and the database of existing trade names that it
24 compares itself to. It also considers the similar
25 strengths and dosage forms when looking at a name.

1 Now, the safety evaluator also does a risk
2 analysis and they examine the data from the expert panel
3 that was originally done, the prescription studies, any
4 computerized searches, POCA to establish any risks for
5 confusion. They also evaluate the potential safety risk
6 associated with two identified drug products being confused
7 with each other due to that similarity and examine their
8 post-marketing data -- that's preventable adverse drug
9 event data -- their clinical and regulatory experience and
10 any literature reports. It's important to take the lessons
11 that we've learned from post-marketing into this evaluation
12 also.

13 Some contributing factors for name confusion
14 include similar indications, having the two drug products
15 prescribed in the same patient population, having identical
16 formulations, overlapping strengths or directions, being
17 stored in the same area.

18 We also look at what's the potential for harm
19 when we look at the two trademarks. What are the
20 consequences if a patient misses the pharmacological action
21 of the intended drug? We ask these questions routinely.
22 And then we ask, what are the pharmacological actions and
23 toxicities of the unintended drug product?

24 There is a final review done. There are
25 actually basically two reviews that are done on trade names

1 at FDA: first, the initial one that I just described which
2 was a multi-faceted review, and a final review that's done
3 approximately 90 days before the action on the application.
4 We don't repeat the extensive evaluation that I just
5 mentioned. We're only looking for any confusion with names
6 that have been approved since the initial review was done
7 and to the time in which the application is going to be
8 approved for FDA approved names during that interval.

9 I thank you very much.

10 DR. GROSS: Thank you, Mr. Phillips.

11 The next speaker is Dr. Bonnie Dorr, Assistant
12 Professor, Department of Computer Sciences at the
13 University of Maryland. She will talk about automatic
14 string matching for reduction of drug name confusion.

15 DR. DORR: And make that Associate Professor.

16 DR. GROSS: Congratulations.

17 (Laughter.)

18 DR. DORR: It's seven years ago now. Thanks.

19 So I'm going to talk about automatic string
20 matching, some of the things that you've heard already that
21 are part of the technology behind POCA, and I'll also talk
22 about other analyses that are done that, combined with some
23 of that technology, could potentially get improved results.

24 So these are the questions, just to remind you,
25 that we were asked to address. I will be giving an

1 overview, some of which you've probably seen before -- but
2 it never hurts to review -- of phonological string matching
3 for ranking. Also, I will be looking at orthographic
4 string ranking.

5 And validation of a study method. What we use
6 is precision and recall against a gold standard to
7 determine the effectiveness of the different matching
8 approaches.

9 I'll talk about an optimal design of a study,
10 and interface for assessing appropriateness of the newly
11 proposed drug name.

12 And then finally, strengths and weaknesses.
13 Each algorithm can miss some correct answers and also get
14 too many that may not be appropriate. So we'll learn more
15 about that.

16 So this is the overview. String matching is
17 used to rank similarity between drug names through two
18 different techniques. Some of these were mentioned.
19 Orthographic compares strings in terms of spelling without
20 reference to sound. Phonological compares strings on the
21 basis of a phonetic representation or how they sound.
22 Within those, each of them has two different types of
23 matching that are done. One is by virtue of distance. How
24 far apart are the two strings? And the other is by
25 similarity. How close are the two strings? If two drug

1 names are confusable, of course, we want the distance to be
2 small and the similarity to be big. So that's the basic
3 idea.

4 I'll give some examples briefly of different
5 orthographic and phonological approaches, both with
6 distance and similarity.

7 Under the heading of orthographic, we have a
8 couple of distance metrics that are actually related, the
9 Levenshtein distance and the string-edit distance. There's
10 a function between those, so they come out to be about the
11 same when you do an analysis.

12 I'll talk about LCSR which is the Longest
13 Common Subsequence Ratio, and Dice. The LCSR and Dice are
14 similarity metrics, all under the heading of orthographic.

15 Under the heading of phonological, I'll talk
16 about a distance metric that is based on sounds called
17 Soundex that's been around for a long time versus a
18 similarity metric under the heading of phonological called
19 ALINE. You may see some typos floating around. Sometimes
20 it's spelled A-L-I-G-N, but this is actually the name that
21 was used for the system.

22 When we want to compare distance and
23 similarity, we want to sort of look at, okay, what do you
24 mean how far apart or how close? Can I look at those two
25 and say whether there's a relation between them? Usually

1 what you do is you say the distance between two strings,
2 two drug names, is comparable in some way to 1 minus their
3 similarity. It's the number between 0 and 1, so if you
4 subtract it from 1, you get a number that allows you to
5 compare these.

6 Orthographic distance. Essentially with the
7 Levenshtein and string-edit distances, you're counting up
8 the number of steps it takes to transform one string into
9 the other. Some examples are given here where, as you can
10 see, the bold-faced pieces here indicate the places where
11 the two strings are different, and the remainder is the
12 same. So you're actually counting the number of places
13 that you're different. That's the Levenshtein or string-
14 edit distance.

15 Also, if you look at Zantac and Xanax, you can
16 see that the X's are counted as different. Even though
17 certainly the initial X sound sounds the same as the Z at
18 the beginning here, they're taken to be different. So the
19 number is 3. Then typically what we do to get sort of a
20 global distance is we divide by the length of the longest
21 string. So we actually know that this distance is really
22 .33 because you have to factor in the length of the string
23 as well; whereas, for the latter one, you're talking about
24 a distance of .5. This is actually a counterintuitive
25 result. If you use Levenshtein or string-edit, Zantac and

1 Xanax are more distant than Zantac and Contac, and that's
2 not a result that you want. So we'll talk about that.

3 LCSR. In this approach, you double the length
4 of the longest common subsequence and divide by the total
5 number of characters in the string. What does that mean in
6 terms of these same examples? You're looking at the
7 similarity in this case, because before we were looking at
8 distance, so we were highlighting the Z and the A. Now
9 we're actually going to highlight the rest of the string.
10 We're going to look at where they're the same. We're going
11 to do a doubling operation here. That's 2 times 4. We're
12 going to divide out. We get .67 here, whereas with Zantac
13 and Xanax, highlighting the characters again that are the
14 same, you get .55. Now, in this case this are reversed.
15 You're talking about similarity. So we're actually in this
16 case saying that Zantac and Contac are more similar than
17 Zantac and Xanax, which also is not a result that you want
18 to get.

19 Dice doubles the number of shared bigrams.
20 What are bigrams? That's just two characters that occur
21 together, and you divide by the total number of bigrams in
22 each string. Some examples are shown here. If you take
23 Zantac and you sort of pull out all its bigrams, and then
24 Contac and pull out all its bigrams, and then you do this
25 doubling operation again, you divide by the total number of

1 bigrams in each string, you get .6. Whereas, if you do the
2 same thing with Zantac and Xanax, you're going to get .22.

3 Again, these are similarity metrics which means you really
4 kind of want Zantac and Xanax to be close, and they aren't
5 close. They're .22 compared to Zantac and Contac which are
6 actually .6. So, again, we're getting a result that we
7 don't particularly want. But these are common techniques
8 that have been used in the literature.

9 Another technique, now moving to the
10 phonological approaches, moving away from look-alike and
11 getting into sound-alike. Here what you do is you
12 transform all but the first consonant to numeric codes.
13 You delete 0's and truncate resulting string to four
14 characters. This is a character conversion that's referred
15 to here. You're actually sort of mapping the vowels to
16 nothing. The 0 means they just drop out. These consonants
17 here kind of sound alike, so they get a 1 and so on. So
18 each of these sets of consonants is going to get a
19 particular number.

20 To give you some concrete examples to work
21 with, this allows you to say "king" and this sort of
22 version of "khyngge," sort of an archaic version. They
23 sound alike and they each get the same code: k52, k52. So
24 those, indeed, look the same.

25 Unfortunately, if you really apply this

1 thoroughly, you get "knight" and "night" aren't the same
2 because one of them is k523 and the other is n23.

3 And even worse, things like "pulpit" and
4 "phlebotomy" come out to be the same when they are
5 radically different, and so you get some pretty bad results
6 there.

7 So the same thing with Zantac and Xanax.
8 You're missing out on that commonality between the initial
9 Z or X sound.

10 Also, an alternative approach to sound-alike
11 that has been used that's been reported in the literature
12 is to compare, instead of using phonological distance of
13 this type, the syllable count, the initial and final
14 sounds, and the stress locations. But this has been shown
15 to miss out on some confusable pairs like Sefotan and
16 Seftin because that has a different number of syllables,
17 and Gelpad and hypergel, where you sort of swap things
18 around, and "gel" is at the beginning of one and at the end
19 of the other.

20 So really, what you need is something to
21 provide that -- the pronunciation for sound-alike -- you
22 need to be able to capture what's going on there for those
23 types of similarities. So ALINE is something developed by
24 Greg Kondrak in the year 2000 to use phonological features
25 for comparing words by their sounds. Some characters are

1 missing here but it doesn't matter much. Those two lines
2 right there are telling you that an ending X sound sounds
3 like KS as in Xanax, but an initial X sound sounds like Z.

4 So if you take those and break them down into the features
5 of what those phonological symbols mean, really you can
6 talk about the pronunciation, the position of the tongue in
7 the mouth and where it stands with respect to the teeth and
8 the back of the mouth, and that's what those features mean
9 in here, without going into detail.

10 The point is that you're going to use, instead
11 of a part of a string as in Soundex, the entire string.
12 Instead of dropping vowels as in Soundex, you're actually
13 going to keep them and they are going to be more
14 significant in drug names. And you're going to use
15 decomposable features in determining the sorts of
16 confusions that people get.

17 This was developed originally for identifying
18 cognates and vocabularies of related languages such as
19 "colour" versus "couleur" in French. But the feature
20 weights can be tuned for a specific application, which
21 is what we've done with this system.

22 In this approach, phonological similarity of
23 two words is reduced to an optimal match between their
24 features. So what we do is we take something like Zantac
25 and Xanax and we align the characters by virtue of going

1 through the decomposed features of this form.

2 Just to show you another example. This is
3 Osmitrol and Esmolol. This is a schwa. It's missing. It
4 isn't missing in mine, but they don't always port over to
5 other people's machines.

6 So the approach that's being used here is to
7 sum up the weight of the match on each sound. In fact, you
8 can align the characters of the strings by looking at their
9 underlying phonological sound. The E in the Esmolol is
10 actually a sound. You take an alignment and you balance
11 out across the features of each of those. If you've got a
12 good match, you get a higher score. So the M and the M get
13 a very high score. In fact, that's a maximal score,
14 whereas this vowel sound in here is close. It's certainly
15 higher than a 5, but it's not up to a 10, and so on. And
16 then you add up and you get a 58 here, and then you
17 normalize it by the total maximum score which would be 80
18 in this case. You could get a potential score of 80 if
19 they were identical strings to get a number like .73.

20 So this approach identifies identical
21 pronunciation of different letters like the M that we saw.
22 It also identifies non-identical but similar sounds such
23 as this one at the head of the two words.

24 Of course, I have to show you a picture of a
25 head with a tongue and teeth, just to make sure that you

1 know that I'm a computational linguist. But the idea is
2 that there are positions within the mouth that -- sound is
3 produced through the vocal tract and also involves the
4 position of the lips, the tongue, the teeth, the hard
5 palate, the soft palate. That's all called place of
6 articulation. Everything bundles up under place of
7 articulation. But also the manner in which air passes
8 through the oral cavity which we call manner of
9 articulation. So there are a lot of other features too,
10 but the top two that we really like to focus on are place
11 and manner.

12 These are some examples of places of
13 articulation. So here is where the two lips are together.
14 That's called bilabial. Here's where the tongue is right
15 behind the teeth like a D or a T. That's alveolar, and so
16 on. Here's a K sound where the back of the tongue is
17 raised. This is called place of articulation.

18 And we can assign particular values. Each
19 individual value within that feature is given a particular
20 weight. So bilabial is really important for drug name
21 matching, for example, and the other ones may be less
22 important.

23 I said place of articulation and manner of
24 articulation. There are also some others that I won't go
25 into. These two are the heaviest weighted values. We

1 really focus on those and give them the highest score if we
2 get a match there.

3 Just to give you some examples. So these are
4 showing the Zantac/Contac comparison that I gave you
5 earlier with Edit, Dice, and LCSR. I already had given you
6 those scores and I showed how they were computed. In the
7 case of ALINE, we actually have Zantac and Xanax as the
8 highest scoring pair out of the three different pairs, the
9 three different combinations that you can get, which is
10 much closer to what we would like to see. We'd like to see
11 that we're looking at the initial sound as something that
12 humans consider to be phonologically equivalent even if the
13 characters are different. So that one actually gets a
14 higher score, whereas Zantac and Xanax in the others do not
15 get the highest score, come in sort of second place.

16 Question number two was how do we validate this
17 approach, and the answer for this is to use something
18 called precision which is counting up the number of matches
19 your algorithm found. We could try this with Edit, Dice,
20 ALINE and so on. Take each one of those algorithms, count
21 up how many matches that it got, and take that over the
22 number of correct matches that you could possibly get, and
23 that's precision.

24 Recall is the number of correct matches in your
25 problem space versus how many does your algorithm determine

1 to be a match. So that's the notion of recall.

2 We use the USP Quality Review as our gold
3 standard. This is necessary in order to determine
4 precision and recall. There were 582 unique drug names,
5 399 true confusion pairs, and if you multiply these out,
6 combinatorically you could get 169,000 possible pairs. You
7 can then rank all of those pairs according to -- in this
8 case I'm not using ALINE. I just put Dice up here. You
9 could rank them according to whether they match with that
10 particular algorithm.

11 So Atgam and ratgam was the one that came out
12 the highest. Using Dice, it came out with a score of .889.

13 It has a plus sign in front of it, which means it did
14 occur in the USP Quality Review as a confusable name pair.

15 It also was the top ranking one.

16 Our next ranking one also has a plus sign,
17 which means it did occur in the USP Quality Review as a
18 confusable pair.

19 The next one down did not occur in the USP
20 Quality Review but maybe it should. It looks like it's a
21 typo. But in any case.

22 Quinidine and quinine. I'm not an expert on
23 pronunciation of these particular drugs, but that was the
24 next one down, and it did occur, and so on.

25 So you can figure out on the basis of these,

1 and how often you're getting the correct answer out of your
2 gold standard, what your precision and recall values are.
3 If you map that out, the way to do it is to compare
4 precision at different values of recall. So the precision
5 is along this axis. How precise are you being with your
6 answers? How many correct answers are you getting? Over
7 how many correct answers out of the problem space are you
8 getting. If you take those two together, you get a graph
9 that looks like this. ALINE is the top score over here
10 with the sound-alike version.

11 If you turn ALINE into the look-alike version
12 -- there is a version that you can just take out all the
13 pronunciation -- it still gets a pretty high score. In
14 fact, it even gets higher than the sound-alike version in
15 one place. But they look pretty much the same for several
16 values of recall, whereas LCSR is lower-performing. Edit
17 is the blue line here, and Dice is down here.

18 At least we have a feel for the idea that
19 somewhere in this manner and place, the places of
20 articulation in the mouth, the way air passes through the
21 mouth, is doing something to get us closer to the USP
22 Quality Review, with the caveat that there are a lot of
23 other errors recorded in the USP Quality Review, of course.

24 In fact, we had to do some studies that are not reported
25 here on cases where it wasn't such a large list of many

1 names that people had speculation and other things
2 factoring into it. So we worked with another list as well
3 and got similar results, but I haven't brought that in
4 here.

5 We really do need to make sure of the
6 transcription into the sound form isn't what's getting the
7 full power of our matching. That is, if we gave Dice and
8 LCSR that same ability to look at sound, would they perform
9 as well as ALINE. It turns out they don't. The sound and
10 the non-sound versions of Dice and the sound and the non-
11 sound versions of LCSR perform lower than ALINE with its
12 phonetic transcription. There's something going on with
13 the weighting and the tuning of the parameters based on
14 articulation points that gets us the higher value.

15 So what would an optimal design of a study be?

16 I actually agree with Dr. Lee that a system should be
17 openly shared, that an optimal study would involve the
18 development and use of a web-based interface that allows
19 applicants to enter newly proposed names. That same
20 software should be used by FDA to ensure consistency of
21 scoring so that everybody is looking at the same scoring
22 mechanism. And that design would ensure that updated
23 versions of software would be continuously available to
24 potential applicants.

25 So the interface would display a set of scores

1 produced by each approach individually, as well as combined
2 scores based on the union of all the approaches. That's
3 something I want to get into. Even though ALINE is the
4 highest-scoring one, there are reasons to look at the
5 combinations of the different approaches to figure out the
6 best answer.

7 The applicant could compare the score to a pre-
8 determined threshold to assess appropriateness, or that
9 threshold could be set community-wide.

10 In advance, running experiments with different
11 algorithms and their combinations against the gold standard
12 would help to determine the appropriateness for the
13 threshold and also allow for fine-tuning, calculating the
14 weights for the drug name matching.

15 Just continuing along that last point there,
16 right now the parameters have default settings for cognate
17 matching, but they may not be appropriate for drug name
18 matching. Something that we might want to do as a part of
19 this is to calculate the weights for drug name matching and
20 then use hill climbing to search against a gold standard to
21 get the values that we're giving for the articulation
22 points closer to what we need for drug name matching.

23 For our initial experiments, we did tune the
24 parameters for the drug name task, looking at things like
25 maximum score, which has to be a high threshold for cognate

1 matching, but should be lower for drug name matching
2 because we ended up with things where it was too risky to
3 consider certain pairs to be the same. Like the "puh" and
4 the "kuh" sound should not be considered the same for drug
5 name matching, whereas in cognate matching, they should be.

6 Also there was something called an insertion and deletion
7 penalty, which should be low for the cognate task but
8 higher for drug name matching. Because confusable names
9 are frequently the same length, a vowel penalty which for
10 cognates, the vowel penalty is low. Vowels are less
11 important than consonants, but that's not true of the drug
12 name matching. Again, we're taking this from a field and
13 moving it into a whole different application, so this type
14 of tuning is necessary. Phonological feature values for
15 drug name matching, place distinctions should be ranked as
16 high as manner distinctions.

17 Last question. Strengths and weaknesses. Just
18 sort of repeating something Dr. Seligman said, all methods
19 offer value and should be used complementarily.

20 So here are some ALINE matches. ALINE gets
21 these sort of pairs, but others don't because ALINE doesn't
22 care whether there are shared bigrams or subsequences. It
23 really is looking at the phonetic features associated with
24 these. Again, these are pairs that I took out of the USP
25 Quality Review.

1 On the other hand, Dice matches with these
2 particular pairs, but others don't because Dice is able to
3 match pairs of words that are similar with bigrams. If it
4 can find that the S and the I is here and the S and the I
5 is here, it's looking at that sort of thing. So ALINE
6 would potentially have trouble with that. And it can do
7 that even though the remaining parts are not the same. So
8 gel and gel show up here, but the remaining parts are not
9 the same, but Dice gets those.

10 LCSR gets these, but others don't because the
11 number of shared bigrams is small for these types of pairs,
12 Edecrin and Eulexin. I'm sorry for the pronunciation that
13 I'm giving. Except for the "in" right here, there are no
14 shared bigrams in this particular pair, but LCSR is able to
15 find that as a potential confusable drug name pair.

16 Just to elaborate on each of those really from
17 the previous slide telling you what's going on, ALINE,
18 using interpolated precision, gets the highest score. It's
19 easily tuned to the task and matches similar sounds even if
20 there's a difference in initial characters like Ultram and
21 Voltaren, but it misses words with high bigram count, as I
22 mentioned.

23 And potentially the weight-tuning process may
24 induce overfitting to the data, so if we get it trained up
25 so that it gets this pair here, it may also get a false

1 pair, the Brevital and ReVia pair which is not one of the
2 confusable ones.

3 Dice, on the other hand, matches parts of the
4 words to detect confusable names that would otherwise be
5 dissimilar, like Gelpad and hypergel, but misses similar
6 sounding names like the ones that ALINE can get, the Ultram
7 and Voltaren pair with no shared bigrams.

8 LCSR matches words where the number of bigrams
9 is small like this pair I showed you on the last slide, but
10 misses similar sounding names like Lortab and Luride that
11 have a low subsequence overlap.

12 So the previous slide showed the weaknesses and
13 strengths, but we think that taking a combined approach --
14 and in fact, we have some initial experiments from the last
15 week or two that are not shown here, that the best approach
16 is to use a combination of all of these to get closest to
17 the gold standard. So we want to continue experimentation
18 with different algorithms and their combinations against
19 the gold standard.

20 Fine-tuning based on comparisons with that gold
21 standard. So, of course, we still need to look at
22 reweighting phonological features specifically for the drug
23 naming task.

24 We believe that taking the phonological
25 approach that has been designed in ALINE by itself and also

1 in combination with other algorithms provides a strong
2 foundation for search modules in automating the
3 minimization of medication errors.

4 And again, just reiterating that a combined
5 approach that benefits from the strengths of all the
6 algorithms, increased recall, without severe degradation in
7 precision, that is, the false positives, is the way to go
8 in my opinion.

9 DR. GROSS: Well, thank you, Dr. Dorr, for
10 clarifying that confusing field for people who aren't in
11 it.

12 (Laughter.)

13 DR. GROSS: We have time for some questions.
14 Brian.

15 DR. STROM: I have three questions for Jerry.
16 We heard from Mr. Lee that there wasn't a problem. We're
17 hearing from you that there is. Let me ask each of the
18 three separately. How often do you get a name from
19 industry that FDA ends up rejecting?

20 MR. PHILLIPS: We reject about one-third of the
21 trade names, and we review about 300 names a year.

22 DR. STROM: Second. How do you know which one
23 was correct? In other words, were they correct in
24 originally thinking it was safe, or was FDA's approach
25 correct in rejecting it?

1 MR. PHILLIPS: That's difficult. I have case
2 examples where we suspected problems of a drug name prior
3 to approval, and for reasons, it got approved, and sure
4 enough, we had post-marketing data that confirmed our
5 opinions. I also have evidence that things that we had
6 concerns about got into the marketplace and we never saw
7 that come forth. So it's difficult to know who's right and
8 who's wrong at times.

9 DR. STROM: A third question which is related.
10 Dr. Dorr just gave us an elegant presentation versus a gold
11 standard, the gold standard being the USP list of names.
12 Why is that a gold standard, and what does that list
13 represent? Clearly the idea of testing these methods
14 against a gold standard make enormous sense. What I'm
15 questioning is how gold is the gold standard?

16 MR. PHILLIPS: Well, the gold standard is from
17 the reports that the USP has received of medication errors
18 associated with both generic and trademark confusion. So
19 that list is a representation of all the reports that have
20 come in. Some of those reports are potential errors and
21 some of them are actual. So the gold standard probably
22 should be applied to those errors that occurred with
23 trademark confusion pairs that actually occurred in an
24 error and not a potential error. That's the reason why we
25 chose that as the gold standard because it's actually based

1 upon actual clinical experience of people being injured or
2 being involved in an error with those names.

3 DR. GROSS: Michael Cohen.

4 DR. COHEN: Thank you. I have a few questions
5 too for the different speakers. I'll ask them as quickly
6 as possible.

7 First for Mr. Lee, as you know, ISMP actually
8 contributes to the FDA Medwatch database as well. The USP
9 and ISMP together we actually have received many, many
10 error reports with trademarks. I agree with you. They're
11 always multi-factorial. There are many contributing
12 factors besides the drug name. But would PhRMA acknowledge
13 that at least one of the contributing factors clearly might
14 be a trademark? Otherwise, how could you explain a change
15 in a trademark totally eliminating the problem? For
16 example, Losec and Lasix. It's gone. We never had another
17 problem with that. Levoxine, gone when the name was
18 changed to Levoxyl. So from that standpoint, I need that
19 clarification to make sure that we're on the same page here
20 -- the committee, that is, and PhRMA.

21 MR. LEE: Yes, I think there are certainly
22 examples of name pairs on the marketplace that are more
23 similar than others, but I would think the modern day
24 practice, let's say, by PhRMA companies takes into account
25 the clinical settings. I think with that screening with

1 the clinical settings, we should see less occurrence of the
2 kind of name pairs like Lasix and Losec.

3 DR. COHEN: A second thing. This is for Jerry
4 I guess. I wanted to know if he would acknowledge -- I
5 agree with Bob and you -- I don't agree with you that the
6 percentage of errors related to trademarks in the FDA
7 Medwatch database is actually a true reflection of what's
8 happening out there, and I think that should be pointed out
9 because really what it is I think the reporters
10 characteristically see FDA as a repository or an
11 organization that can effect change with product-related
12 issues. So the types of reports that you would get I think
13 more than practice-related issues would be product-related
14 issues and the kinds of things that you would get reported
15 would be things that practitioners who report to the
16 program think can be addressed by FDA. So I just wanted to
17 point that out. We do see that figure quite frequently and
18 it could be misleading unless you use it correctly, which
19 is what you did, you said reported to FDA. You didn't say
20 that's the actual percentage out there.

21 MR. PHILLIPS: I acknowledge that. That's the
22 data based upon what we've received, and we have a system
23 that collects data on drug products and more serious
24 adverse events. So it is skewed in one direction.

25 I would mention that Medmarx has released its

1 annual report this year. I think there was some 8 percent
2 of their reports of 192,000 reports that had something to
3 do with name confusion. Some 4,000 patients were involved
4 in errors. So I think there is some evidence outside FDA's
5 reporting system that it still is a problem.

6 DR. COHEN: I'm not trying to minimize it. I'm
7 just saying that it may not be 12.5 percent.

8 The other thing, for Dr. Dorr, I had two quick
9 questions. Do you think systems like yours could be used
10 as a sole method for testing?

11 DR. DORR: I don't know if you mean the
12 technique, the methodology.

13 DR. COHEN: Yes.

14 DR. DORR: Right. So what we're experimenting
15 with right now -- we actually have a pretty good result --
16 is bringing in a combined version of Dice, ALINE, LCSR, and
17 so on. By the way, this is only for look-alike and sound-
18 alike. So we have an orthographic version of it and we
19 have a phonetic version of it. So we don't pretend to try
20 to -- I guess that was 16 percent or 12 percent somebody
21 said of the overall problem. So I agree with your comments
22 about the USP Quality Review as taking in too many things
23 that have nothing to do with that type of matching.

24 But I believe that taken alone, the phonetic
25 approach, if you had to choose one, is the best one. We've

1 got some definitive, repeatable results on that. But you
2 can get better than any of the approaches alone, including
3 ALINE, if you take a combination of the different
4 algorithms.

5 DR. COHEN: Then finally for you, what
6 databases do you actually use?

7 DR. DORR: The only one was that USP Quality
8 Review.

9 DR. COHEN: I see.

10 DR. DORR: Yes. Although more recently we have
11 looked at something that was a proprietary database. I'm
12 working with PPC, and so they had given us a smaller
13 version of just names that are not in this sort of broader
14 category of any medication error. And we were getting
15 similar results on that one, but I couldn't put any of that
16 on the slides.

17 DR. COHEN: Thank you.

18 DR. GROSS: Robyn Shapiro has a question.

19 MS. SHAPIRO: Yes. I still am somewhat
20 confused about the underlying assumption, being a newcomer
21 to this whole topic. To me the data about the causation is
22 very weak. For example, Dr. Phillips, in your comments,
23 the 12.5 percent by reporter, is the reporter always the
24 individual who we think is responsible for that error? And
25 if not, then how good is that data in and of itself?

1 And the confusion about the underlying
2 assumption is important not only for us to kind of think
3 about why we're here, but also where we're going. In other
4 words, if a risk management approach really had to do with
5 how we see these prescriptions written out, then the
6 transcription would be the subject of our focus as opposed
7 to the actual name.

8 So I'd like to know from the FDA how confident
9 you feel about the causation of these med errors being
10 attributable to the name itself.

11 MR. PHILLIPS: I feel pretty confident about
12 the data that I have and the causation, that there is a
13 contributing factor with similarity of trademarks, that
14 they can definitely be associated with the event. There
15 may be other contributing factors, but there is a definite
16 association between similarities of names that contribute
17 to errors.

18 MS. SHAPIRO: Based on data? You feel
19 confident because you have data about that?

20 MR. PHILLIPS: That's correct.

21 MS. SHAPIRO: Could we see it?

22 MR. PHILLIPS: Within our Adverse Event
23 Reporting System and the data that I cited, the analysis
24 that was done over the 6-year period?

25 MS. SHAPIRO: Yes. Again, I'm interested in

1 pulling it apart so that we know, if we can, that these
2 errors we feel confident are on account of the name as
3 opposed to all these other factors that go into med errors.
4 That would help me to think about a risk management
5 approach.

6 MR. PHILLIPS: Usually when a reporter reports
7 on a medication error, they're going to give a narrative of
8 the event itself and usually will provide some causes of
9 that event. That doesn't necessarily mean that reporter is
10 correct. The reporter may not actually be involved in the
11 error, as you cited. They may be reporting the event. A
12 risk manager may be reporting the analysis that was done at
13 a facility, and according to that facility, these were the
14 contributing factors associated with that medication error.

15 There are always more than one factor involved in an
16 error. So just to say that it was just trade name was
17 probably not true for the whole event. But if you do look
18 at the narratives in the cases and look at these -- and you
19 can run those similarities through an analysis yourself,
20 and we do that -- you will see the similarities and the
21 contributing factors.

22 DR. GROSS: We have three more questions. I'm
23 taking more time for the discussion because it's beginning
24 to get at the crux of the problem. Ruth Day.

25 DR. DAY: I have a couple of questions for Dr.

1 Dorr. First of all, you're comparing across these
2 different computational linguistic methods. They all have
3 their strengths and weaknesses, and taken together, they do
4 a lot. It's great to see.

5 I'm concerned, however, they all depend on an
6 initial phonetic transcription. So one part of that is who
7 does the transcription. I have seen within companies, as
8 they go forward with a given name, there are alternative
9 pronunciations even within the company. We heard from you
10 this morning quinine. Others say quinine. You could also
11 say quinine and so on. So you might say there are these
12 alternative pronunciations, and so once you decide on a
13 phonetic transcription, you've decided on one. So there
14 could be some consequences for this.

15 So number one, who does the transcription and
16 who decides that's the one to go forward with?

17 DR. DORR: So there are two questions.

18 First, who does the transcription? I should
19 clarify. These were all automatically transcribed, which
20 means a choice was made and probably the wrong choice in
21 many cases. One deterministic choice was made. So there
22 was no human involved in that. On the basis of information
23 on English in general, we know that -- and in fact, it
24 probably would have come out with quinine. Who knows? But
25 based on what it has available in general, we have an

1 automatic transcriber.

2 However, the second question is, what do we do
3 with these different variants? What do we do with
4 different pronunciations within a dialect? And then what
5 do we do when you have different dialects entering into the
6 picture? That's sort of the next phase of what we're
7 trying to look at. We need to be able to train on
8 different dialects in getting the variations of
9 particularly vowel sounds. Those tend to be the ones that
10 people trip up on the most. And even in different
11 languages, which is another area that we want to look at
12 next. Right now, there is just one deterministic answer
13 and it could be the wrong one.

14 DR. DAY: Even within the same dialect -- in
15 our lab, we have people just pronounce drug names and we
16 find great variation even within very narrow sets of
17 people, all highly educated, excellent readers, and so on.
18 There are alternative pronunciations. Since what we're
19 looking at is comparison of phonological similarity across
20 pairs, if we don't have a sense of the alternative
21 pronunciations and their relative probabilities of each one
22 to begin with, then I don't know what we're comparing.

23 DR. DORR: No. That's exactly how you want to
24 do it. You want to have differing probabilities with
25 alternatives that are available to you, and what you rely

1 on is that if some vowel sound was wrong, that the
2 remainder of the word would get you close enough that
3 there's at least some hint that something could be going on
4 here. But you do need to have more than one pronunciation,
5 and as I mentioned, definitely within dialects, you do get
6 these variations and people having the same education level
7 will pronounce them differently. So I agree that that's
8 something we are not doing now that needs to be done.

9 DR. DAY: Okay. And just my second question
10 and last question. You've done a great job with the
11 different features for producing the different sounds.
12 There's often an interaction across features. So, say, for
13 example, place and manner of articulation define stop
14 consonants, and there's a huge psycholinguistic literature
15 that shows that people make systematic errors in perceiving
16 them. So these are sounds like "puh," "tuh," "kuh," "buh,"
17 "duh," "guh." And when people listen to those and make
18 mistakes under noise or under good hearing conditions, you
19 can predict what mistakes they're going to make. So
20 they're more likely to confuse "puh" and "buh" than "puh"
21 and "guh." These are direct calculations based on the
22 number of features that vary.

23 So have you taken into account these well-known
24 interactions of features in these computational linguistic
25 methods?

1 DR. DORR: That's exactly what the decomposable
2 features are supposed to give you, that you're not just
3 taking "puh" as one sound, but you're breaking it down
4 into, say, eight or nine different features. So that's
5 where you can get that multiplicative interaction, that you
6 have so many of them that it describes really a bunch of
7 different dimensions along which you can compare another
8 vector of features so that they differ in two of those
9 features, but if seven out of the nine match, then that's a
10 very highly likely confusable pair. And that's based on
11 the phonetic literature.

12 DR. DAY: So how do you determine those
13 weights? We saw 40 and 50 for place versus manner or vice
14 versa.

15 DR. DORR: Right. That's tuning that was used
16 initially for the cognate matching task for determining
17 across language pairs like French and English whether there
18 are certain similarities like couleur and colour, and those
19 had to be retuned and adjusted so that, for example, manner
20 and place are now given a higher weight than they were in
21 the cognate matching task based on what we found in the
22 data from the drug name pairs. So you can actually fine-
23 tune it for your particular application.

24 As I said, the caveat is we were training on
25 data that had other things playing into it that had nothing

1 to do with either look-alike or sound-alike names. A lot
2 of these were reports and not real errors that actually
3 occurred. So we were training on sort of noisy data, and
4 we'd like to have a better training set to do that.

5 DR. GROSS: We have two more questioners and
6 then we'll have to move on. Jeff Bloom.

7 MR. BLOOM: Yes. Dr. Dorr, can you come back
8 up for just a second please? Thank you.

9 Picking up on what Dr. Day said -- and I would
10 quibble a little bit with the vowel situation. We are
11 living increasingly in a multi-cultural society, including
12 not only just patients, but also doctors, nurses, health
13 care practitioners, where there are particular diphthongs
14 that are not native to their natural language, if English
15 is not their first language. The R's and L's are
16 particularly difficult for people to say. I don't know how
17 that could be formulated in to figure out how to do that in
18 what you're doing, but I think it's an important issue.

19 DR. DORR: And that's exactly what we're going
20 to be doing next. We have a phonetic transcription table
21 for Spanish, and we're looking at one for French. Again,
22 these are superimposed on top of -- well, they're not
23 really English names. They're some sort of brand name. So
24 we're taking kind of what people would think a Spanish
25 speaker would say an English pronunciation, and that is the

1 next phase. It's not a part of the work we've done so far.

2 It's the next phase of the work. It's very important.

3 DR. GROSS: Stephanie Crawford.

4 DR. CRAWFORD: Dr. Dorr, please stay.

5 (Laughter.)

6 DR. CRAWFORD: I have two questions. First,
7 when you were discussing the tests of orthographic distance
8 and similarity, several times when you made the comparisons
9 with Contac versus Zantac and Xanax, you stated it was not
10 the result that you wanted to get. I'm a little confused
11 with that because through objectivity, do you have presumed
12 results you wish to get? That's the first question, and
13 then I'll have a second one for you.

14 DR. DORR: First question. So we were again
15 looking at a gold standard and did not find Contac and
16 Zantac in there. Did anybody find that pair? If you did,
17 let me know. If it shows up -- by the way, it will show up
18 in the list. It will just be ranked lower, and so it
19 depends where your threshold is. But Xanax and Zantac is a
20 confusable pair and Contac and Zantac were not among the
21 confusable pairs. The reported pairs. So that's what I'm
22 saying. It seems that that's the result we wouldn't want.

23 DR. CRAWFORD: And my last question. I
24 appreciate the very fine comparisons you did with the three
25 approaches, ALINE, Dice, and LCSR. I wanted to ask, are

1 these the only approaches? If not, how were they the three
2 that you selected for comparisons, and if they're the only
3 ones you're considering.

4 DR. DORR: So LCSR and Levenshtein are actually
5 related. There are also other versions. Like there are
6 bigram and trigram versions of these. I put the sort of
7 simplest cases up there, but we did take the string
8 matching approaches that were in the computational
9 linguistics literature to be reported the best in our
10 comparison. And then phonological -- the standard -- when
11 we began studying this with Soundex or its sort of relative
12 Phonex which we also looked at. We just started with what
13 was reported to be best in the literature for each of these
14 types.

15 DR. GROSS: Thank you all for those excellent
16 questions.

17 The next speaker is Dr. Richard Shangraw, Jr.
18 who is CEO of Project Performance Corporation. He will
19 discuss the use of expert panels in evaluating drug name
20 confusion.

21 DR. SHANGRAW: You can tell already we're going
22 to change gears a little bit here. My presentation is sort
23 of at the other end of the spectrum. It's really talking
24 about the use of expert panels as a way of identifying
25 potentially confusing drug name pairs. In some respects,

1 it's going to build on Bob Lee's comments about the use of
2 experts in this problem area. And I'm going to talk a
3 little bit more broadly about the problem. In fact, when I
4 got the questions for this presentation, I interpreted the
5 question about how does this method compare to others to be
6 a broader question about how does expert panels, for
7 example, compare to computational linguistic approaches or
8 experimental pharmaceutical approaches. So I'm going to
9 have a sort of broader perspective on the problem.

10 Before I get into the problem set, let me just
11 give a quick background for those who may not know a lot
12 about the field of expert panels or expert committees.
13 It's an area that has emerged primarily in the '40s and
14 '50s. It grew out of a lot of research on the use of
15 experts in a number of different settings: policy settings
16 where there were some concerns that policy makers here in
17 D.C. were not generating the best policy decisions when
18 they got together to solve problems. That led to a number
19 of formal structure techniques for using expert opinion. I
20 don't think they use them now, but at least there were some
21 thoughts of trying get those structured techniques in
22 place.

23 You'll hear through my presentation today the
24 use of the term Delphi. There's a technique called Delphi
25 that's been used as a nominal group technique that's been

1 used formally for many years, 20-30 years.

2 And there's also been a large application of
3 the use of expert panels in the health care field. In
4 fact, there's a longstanding set of research that's been
5 done by UCLA and the RAND Corporation on using these kinds
6 of expert panels and approaches for looking at appropriate
7 care in hospital settings. NIH uses a consensus-based
8 approach for some of their decision-making.

9 I think Dr. Seligman was accurate in saying
10 that there hasn't been a lot of specific research in this
11 problem set area, that is, the use of expert panels in this
12 drug name confusion area, but there's a load of evidence
13 and research in using expert panels in many other settings.
14 What you're going to hear today is my bringing that amount
15 of expertise and that research that's been done into this
16 problem set area and talking about a process for how it
17 might be used for drug name comparison purposes.

18 I'm going to be very procedurally oriented
19 today. I think the biggest criticism of expert panels and
20 expert committees is the ability to replicate or validate
21 their outcomes. The best improvement that can be made in
22 terms of improving the outcome of an expert panel or an
23 expert committee is by introducing repeatable processes
24 related to the way that these panels or committees are
25 conducted. As you'll see here on my slide -- and this is

1 really going to be the driver behind my presentation here
2 -- I'm going to work through a design on how an expert
3 panel could be conducted that could be replicated and
4 perhaps validated -- and I'll talk about them a little bit
5 later in the presentation -- as a way to ensure that you
6 could get consistent and possibly highly appropriate
7 results coming out of a group of human experts as opposed
8 to a computational system on a computer.

9 I'm going to go through each one of these
10 boxes, but in broad terms, there is a panel that's selected
11 and moderated, and before you can really select and
12 moderate that panel, you have to figure out the definition
13 of who's an expert and you have to figure out what sort of
14 guidelines this panel is going to use in terms of the way
15 they vote or rank decisions through the panel.

16 Most of the literature talks about and most of
17 the research that we've done talks about the use of
18 separating these panels into rounds or phases where you
19 would have the problem set introduced. It's often called
20 the exploratory round or the discovery round where you
21 actually try to just put on the table all the possible
22 alternatives where you might have a confusion with a
23 specific drug name. You would then consolidate and collate
24 those results.

25 Then you would have a second round where you

1 would have a ranking or voting process. In fact, some of
2 the techniques I described earlier, the nominal group
3 technique and Delphi technique, will extend these rounds
4 many times. They'll go three rounds, four rounds, five
5 rounds before they come to an actual decision.

6 Then obviously you'd have some solution set or
7 result coming out of this panel.

8 Perhaps the first problem and probably one
9 that's most challenging here is to make sure you have the
10 right experts participating in the panel. Again,
11 guidelines can be established here. It can be based upon
12 experience. It can be based upon not only years of
13 experience but type of experience, clinical experience. It
14 can be based on education, training, pharmacists, nurses,
15 doctors. But clearly there could be some baseline
16 established here for the type of expert that would be asked
17 to participate in the panel.

18 Second, you have to be concerned about
19 conflicts. This is an interesting problem that you've
20 already discussed this morning in terms of this panel being
21 put together in terms of making decisions. This is clearly
22 an expert panel sitting before us here, and you have to be
23 concerned about those in these kinds of panels also.

24 Personalities is a clear factor of concern
25 that's been introduced through many studies. The concern

1 here is on dominating personalities. Obviously, in the
2 front-end stage, you certainly don't want to select a whole
3 set of dominating personalities to be part of your panel.

4 Then finally, there's some good research now to
5 suggest that the larger the diversity of the panel, the
6 more likely you are to get a broader or more robust result.
7 So, in other words, if the set is all pharmacists, it's
8 probably not as good as a set that has some pharmacists,
9 some nurses, some doctors. You even heard Bob Lee talk
10 about the fact that they introduce legal counsel into their
11 panels and other people that have expertise in this area.

12 The second part, again before you even get
13 started, is laying the groundwork on how you vote and how
14 you rank decisions. This is another very important part of
15 the process. This is probably the part of the process that
16 can lead to the most dynamic changes in the outcomes of
17 panels. These are very simple issues. Does the majority
18 vote win? If you pull a pair up and the expert panel looks
19 at it and the majority thinks it's a problem, is that
20 sufficient? If it's not majority, is it two-thirds? If
21 it's not two-thirds, is it 90 percent? Making those
22 decisions on the front end before you get to the process,
23 obviously makes a process more repeatable.

24 And the second part of that is related to how
25 you collate the results. If we have 10 experts in a room

1 and they're trying to vote on or rank a set of problems
2 associated with a confusing drug name pair, how do you rank
3 or collate the different ranks amongst the experts? There
4 are a number of different techniques out there for doing
5 this. The nominal group technique has an extended process
6 that looks at the way that people rank and combines those
7 ranks together, giving higher priority to first and second
8 ranks. We could spend a long time talking about just how
9 you collate ranks, but suffice it to say there's a process
10 for doing that. There are different ways of doing that.
11 None of them are perfect, but at least you need to
12 establish that on the front end.

13 You've seen some numbers already today from
14 Jerry Phillips about numbers of participants in their
15 expert panels. I think you'll hear some from some of the
16 other speakers today. Dr. Kimel, for example, who's up
17 after me, has a very closely related area and that's use of
18 focus groups, and she'll talk about some of those numbers
19 also. But in general, the size of an expert panel is about
20 8 to 12 participants.

21 The issue of moderator, which I'm actually not
22 going to spend a lot of time on because Dr. Kimel is going
23 to spend some time on it, talking about the role of the
24 moderator. It's also very important in these groups as a
25 way of facilitating the discussion.

1 So now let's break it down into how an expert
2 panel would proceed. Round one. Given the electronic age,
3 most of the expert panels that we're seeing being conducted
4 out there are certainly from a cost perspective in terms of
5 making sure they minimize the cost of conducting these
6 panels are conducting round one's electronically. It's
7 predominantly done through e-mail. An e-mail is sent to a
8 participant. They are given some procedures and processes
9 about how they're to look at different drug names. They're
10 asked to provide a ranked list back to the moderator, and
11 then those ranked lists are collated. Clearly the number
12 of names being processed by an individual, the ranking
13 procedure and process can all affect this stage of the
14 process.

15 There are also clearly some concerns here given
16 this topical area of confidentiality. I'll talk about that
17 a little bit later in terms of strengths and weaknesses of
18 this approach.

19 Once you get the results for round one, you
20 consolidate them, using any of a number of different
21 approaches for taking ranked results and putting them
22 together and displaying them. Some of those approaches
23 simply say let's just focus on the number one rankings from
24 across the experts, and there are also ways of taking those
25 rankings and consolidating them in such a way that you can

1 have a broader list exposed to the participants in round
2 two or a narrower list.

3 Again, this is an area that Dr. Dorr hit on
4 just briefly, and that is the issue of if the system,
5 whether it's an expert panel or a computer system,
6 generates potentially confusing names of 100 potential
7 pairs, it's much more difficult to rank in order and
8 organize those types of results than ones where you see 10
9 or 20 potentially confusing names. This process, while it
10 seems much more human based on the computational methods,
11 can yield the same kind of results where you could have
12 potentially very large sets of potential confusing names
13 coming out of the set of experts, and you have to be
14 concerned about the ability of the experts to process
15 through those names.

16 Round two is really probably the round that is
17 the focus of most of the expert committee/expert panel
18 research and that's really the way that you get at the
19 decisions. It's called the decision round, summary round,
20 the ranking round. It's the part in the round that after
21 the discovery round, round one, that you bring the experts
22 back together and have them now, in a face-to-face
23 situation or increasingly in a computer-facilitated
24 situation, discuss the potential issues associated with
25 name pairs or potentially confusing name pairs.

1 As I said before, this is a process that
2 historically has been done face to face. Experts are flown
3 in, for example, this panel you see before you here. And
4 they are asked to communicate amongst themselves with a
5 moderator, to sort through a set of issues. Increasingly
6 there are web-based tools that are doing this where you
7 have a speaker phone, a teleconference, augmented by a
8 computer screen on the internet where they're able to have
9 conversations through the telephone lines, and they use the
10 computer screen as a way of organizing and facilitating the
11 discussion.

12 Again, there have to be some predetermined
13 rules about voting. This process can be a lengthy process.
14 it can take anywhere from 2 hours to 6 hours to 8 hours
15 depending upon the complexity of the name that's involved.
16 It's also an expensive part of this piece of this process
17 given especially the cost, for example, of flying this
18 group of experts in. You can imagine the cost of doing
19 that across the 300 or 400 names, for example, that Jerry
20 Phillips says has to be reviewed on an annual basis.

21 So can we validate these methods? Obviously,
22 the biggest concern here is can you replicate across expert
23 panels the results of the expert panel. Most of us sitting
24 around here today would say that's a tough problem. Right?
25 Experts have different perspectives. They come from

1 different views. They're moderated differently.

2 I would argue that if the procedures and
3 processes are well established ahead of time and if there's
4 understanding of those processes by the participants, if
5 you have diversity of views, and you have a good moderator,
6 that there is a possibility of replicating these
7 procedures. It could be done two ways from a testing
8 perspective.

9 The first is one which I call reliability.
10 That is, do different panels come up with the same results?
11 That's the first question. So if I have one panel here
12 today and a panel tomorrow and I give the same drug name,
13 will they basically come up with the same result?
14 Obviously, that could be tested. It's expensive to pull
15 those panels together, but nevertheless, it could be
16 tested.

17 Second is the issue of validity or in this case
18 predictability, and that is, if the panel is given a name,
19 do they come up with an answer or a potentially confusing
20 pair that can be compared against some standard? We've
21 talked about this gold standard in the first talk by Dr.
22 Dorr. That again could be replicated giving a panel a set
23 of names that we know have known confusions on and see if
24 they actually generate that same list of names whether
25 there are known confusions. Again, that could be tested.

1 It's expensive, but it can be done.

2 There are some problems, of course, in that
3 second test in terms of what's called the history effect,
4 and that is, if panel members know that there have been
5 known confusions with a name, then we have problems in
6 terms of history, with that effect. But nevertheless, you
7 could perhaps control for that in terms of panel
8 participation.

9 So these are probably the two key pieces that
10 you'd like to look at from an expert panel perspective.

11 So what are the strengths of the design? Well,
12 clearly when Dr. Dorr was asked the question by one of the
13 experts here on the panel is this approach sufficient in
14 and of itself, and that was asked on the computational
15 approaches, I think much the same question could be asked
16 about an expert panel. Is an expert panel sufficient in
17 and of itself to solve this problem or to address this
18 problem?

19 And my answer, being a good social scientist,
20 is that I'd always like to have multiple methods. So a
21 combination of a method, for example, of a computational
22 approach perhaps on the front end for the discovery phase,
23 which is to say, give me the list of potential confusions,
24 and then taking that list and providing it to an expert
25 panel, much like the process that Jerry Phillips describes

1 the way that the FDA does it, seems to me to be a more
2 appropriate and possibly more robust approach to solving
3 the problem because in my opinion the ability of the human
4 expert to digest and to analyze some of the questions that
5 have already been presented by this panel, in terms of the
6 computational approach, could have some value, the ability
7 to sort through dialect by different pronunciations, by
8 misinterpretations, by handwriting. These are all things
9 that the computer is getting pretty good at, but I still
10 think the human has an ability to do some more in that
11 area.

12 Second, I think the other part of this, which
13 is the really interesting piece of this puzzle and that is
14 with a set of experts sitting around a panel talking about
15 potentially confusing pairs, you can ask the panel why do
16 you think that's a confusion. It's hard to do that with a
17 computer. In other words, you can say why is that
18 confusing to you, and you can at least get some elicitation
19 from the expert about why they think there might be a
20 confusion. Now, we could probably dive into the mechanics
21 of why the computer thought it was a confusion, but I think
22 as a group of reasoned experts in a room, you like to hear
23 a human interpretation of that potential confusion.

24 And finally, as you can see, the design is easy
25 to understand. It's pretty straightforward. It has some

1 process pieces to it, but it's relatively easy to
2 understand.

3 Weaknesses. Many weaknesses with this
4 approach.

5 I talked, first of all, about the fact that the
6 panels are susceptible to domineering personalities. We've
7 already talked about that.

8 It's difficult to validate the designs. I
9 proposed some methods, but they are difficult and require a
10 lot of controls.

11 The ability of the group to achieve consensus
12 is a particularly perplexing problem with expert panels, in
13 that even if you establish voting methods, there may be
14 some issues in terms of the ability of the panel to come to
15 some sort of consensus-based conclusion.

16 We've already talked and heard some issues
17 about dialect and concern. If the panel is not diverse
18 enough, there may be some issues there.

19 You can also have wide variability in the
20 results across panels given the expertise of the panels.

21 And finally and probably as important is as we
22 move to these electronic panels, there's always going to be
23 concern of confidentiality, certainly on the part of the
24 pharmaceutical industry in terms of taking these names and
25 putting them across the ether to other people to comment on

1 them.

2 So that's a quick overview of the expert panel
3 and expert committee approach to this problem.

4 DR. GROSS: Thank you very much, Dr. Shangraw.

5 Any questions from the advisory committee?

6 Yes, Eric Holmboe.

7 DR. HOLMBOE: I'd just be curious to know, with
8 regard to expert panels, what data do we have with regard
9 to this issue in the past? You mentioned, Jerry, that
10 about a third of names get rejected. What role have expert
11 panels, if any, played in that particular process?

12 MR. PHILLIPS: The expert panel plays an
13 important role in our process, but it's just one component
14 of a multi-faceted review. So I think if we went back and
15 looked at the recommendations of the expert panel on the
16 final conclusion, that they're going to be pretty
17 consistent.

18 DR. GROSS: Stephanie Crawford, do you have a
19 question?

20 DR. CRAWFORD: Thank you. A very quick
21 question. How do you determine consensus? You said it's
22 not always achievable? By what definition would you have
23 consensus?

24 DR. SHANGRAW: Well, the first problem with
25 consensus is and the failing of many of these panels is

1 they don't decide on the voting method before they have the
2 panel. So if you don't decide on the voting method before
3 you conduct the panel, you will never get consensus.
4 Certainly it's harder to achieve. So the first solution to
5 that is to have an agreed-upon voting method before you go
6 into the panel process.

7 Voting methods can be as simple or as complex
8 as you want them to be. Some use simple one vote
9 mechanisms. Some use majority mechanisms, plurality
10 mechanisms. Some use rolling voting mechanisms. There are
11 a number different techniques. But the most important
12 point here is establishing that ahead of time and having
13 the panel participants agree on that. If you do that, then
14 consensus is easier to accomplish, obviously, because once
15 you get to that point, you hold the vote, and whatever
16 voting method you've decided to use then helps to finalize
17 your consensus.

18 Unfortunately, most panel members, after a long
19 and heated debate, when they get to the point where they're
20 supposed to vote, decide they don't like the voting
21 methods. And then we have another set of problems. But
22 that's the difference in dealing with humans than with
23 computers.

24 DR. GROSS: Michael Cohen.

25 DR. COHEN: Yes. It hasn't been mentioned yet,

1 but I think a large percentage of the practitioner review
2 that Mr. Lee was talking about before is actually done by
3 companies that are separate from the PhRMA company. I
4 think most of those companies, from what I can gather, use
5 a system where they would actually -- first of all, there's
6 more than just one name that's tested for a particular
7 compound. There might be 10 or even 15 or more. But they
8 would use what is considered, I think, an expert group. In
9 other words, there are physicians, nurses, pharmacists that
10 are out there in the field that are working every day, and
11 it might be done by the internet. They would actually look
12 at actual names and listen to them, how they're pronounced,
13 et cetera, whatever, and then provide feedback. And then
14 that information is collated and presented to an expert
15 group that does what is called the failure mode and effects
16 analysis or failure analysis.

17 Is that considered expert panel on both ends?
18 That is not.

19 DR. SHANGRAW: No, absolutely.

20 DR. COHEN: Oh, it is.

21 DR. SHANGRAW: You're going to hear from the
22 next speaker an even broader discussion on focus groups,
23 and we can have a long debate about is an expert panel the
24 same as a focus group. The answer is they all come from
25 the same genre. They all come from the same category of

1 approaches that says let's convene a group of human
2 experts. Let's tap into their brains and let's find
3 solutions to problems. So the next speaker is going to
4 talk about that from a focus group perspective, which in
5 fact some of the third party research groups use focus
6 groups, and she'll be talking more about that.

7 DR. GROSS: Brian Strom has the next question.

8 DR. STROM: We've heard today, it sounds like,
9 an enormous effort underway at FDA and industry, multiple
10 private companies using expert panels. This has been
11 underway for many years, it sounds like. You described for
12 us a very clear, very nice description of the process and
13 how you would test the reliability and validity. Given the
14 huge effort that has been underway all these years, all
15 these drug names, can you tell me what data are available
16 on the reliability and validity of the approach?

17 DR. SHANGRAW: If the question is what's
18 available on the reliability of an approach testing drug
19 names specifically, I do not have any data in that area.
20 That's not to say there's none out there. I'm not aware of
21 any at this point.

22 DR. STROM: Does anybody know? Jerry?

23 MR. PHILLIPS: I'm not aware of any either.

24 DR. COHEN: I don't think there is any.

25 DR. SHANGRAW: It's sad that we don't because

1 you're exactly right. We've been doing this for years and
2 we should have some data, but I haven't seen it yet.

3 DR. GROSS: Jeff Bloom.

4 MR. BLOOM: Thank you.

5 In reading through the meeting materials and
6 also your presentation, one of the things I was wondering
7 about is, has there been any consideration of including
8 patients in any of the expert panels? After all, patients
9 need to understand the drug names and also serve as a check
10 and balance against making sure they're getting the correct
11 drug.

12 DR. SHANGRAW: In many of the health-related
13 expert panels, for example, ones convened by NIH and UCLA,
14 there is a role for the patient in those panels. Obviously
15 that comes into the front part of this discussion where I
16 talked about how you define an expert, and clearly that
17 would be part of that discussion about whether or not a
18 patient would be included. I think there are a number of
19 reasons why you might want to include a patient, but that
20 would have to be determined on the front end.

21 DR. GROSS: There's a question or a comment
22 from Jerry Phillips.

23 MR. PHILLIPS: Rick, the process in which you
24 vote in an open meeting, whether that's privately -- what
25 influence does that have on the decision-making process and

1 how important is that?

2 DR. SHANGRAW: That's a very good question, and
3 I failed to address that. One of the techniques that has
4 been used to deal with the domineering personality problem
5 in expert panels is to use anonymous voting throughout the
6 process. Now, there's been some research on that which
7 says that a completely anonymous voting process, especially
8 in the expert panel, that second phase, which is the
9 decision phase, doesn't lead to the best decision because
10 you have to expose at some point a position and then use
11 that as a basis for discussing the problem. So the general
12 approach has been, in the literature at least at this point
13 and the research, is to have anonymous voting through phase
14 one, which you saw in this process, which is to identify
15 and rank on an anonymous basis through that discovery phase
16 to present the list, but then by phase two, that that
17 voting would become more public as a means of facilitating
18 discussion. There's a longstanding debate about even if
19 you have that open voting process in that second phase, and
20 there are still some that argue to keep it anonymous, but
21 that it is a key piece of the issue of the domineering
22 personality problem.

23 DR. GROSS: We will adjourn and reconvene at
24 10:35.

25 I have a suggestion for FDA and PhRMA. Maybe

1 at lunchtime you could prepare a list of what methods you
2 are currently using to avoid look-alike/sound-alike names
3 so that when it's time for us to make some recommendations,
4 we have that information summarized for us.

5 (Recess.)

6 DR. GROSS: I hope you all had a nice coffee
7 break. We're going to reconvene so we can try to stay on
8 schedule.

9 The next speaker is Miriam Bar-Din Kimel,
10 Senior Project Manager of MEDTAP International, who will
11 talk on the focus group methodology.

12 DR. KIMEL: My presentation will be about focus
13 group methodology and the application to the drug naming
14 process. It will actually build upon similar methods that
15 Dr. Shangraw had discussed in the previous session.

16 First I will review focus group methodology,
17 including strengths and limitations. Then I will describe
18 how focus group methodology may be applied to the drug
19 naming process, and finally discuss conclusions.

20 Focus groups are a form of qualitative research
21 methodology used to address specific research questions
22 that require depth of understanding that cannot be achieved
23 through quantitative methods. Focus groups can be used in
24 various phases of research and in conjunction with various
25 research methods. In the exploratory phase, they can help

1 determine which populations to test and to target. In
2 pretesting, they can help identify and clarify perceptions
3 about specific topics, products, or messages. And in
4 triangulation, also known as convergence of multiple data
5 sources or methodologies, focus groups can be used to
6 support other sources of qualitative data.

7 More specifically, focus groups can be used to
8 gather background information, diagnose problems with
9 programs and processes, stimulate new ideas or identify new
10 relationships, generate hypotheses for future qualitative
11 or quantitative study, evaluate programs, develop
12 qualitative understanding of how individuals view a
13 situation or deal with a phenomenon of interest, or help
14 interpret quantitative results.

15 Focus group methodology can be used as a
16 standalone investigation or as part of a multi-method study
17 in conjunction with other qualitative and quantitative
18 methods. For example, in survey design, focus groups are
19 often used as a first step to identify relevant items in
20 the patient's own words. Once the instrument is developed,
21 quantitative psychometric analysis is then performed to
22 test the instrument properties.

23 Focus group methodology also can be used to
24 supplement the interpretation of quantitative data. For
25 example, a trial may find a large number of asthma patients

1 come to the ER for treatment with minor symptoms, and then
2 a focus group can be conducted afterwards to find out why
3 they come to the ER.

4 There are different types of focus groups that
5 may be used. Traditional focus groups are conducted in
6 person and have a structured format, most often using
7 interview guides to direct the discussion. Brainstorming
8 is also conducted in person but is nondirective and
9 unstructured. Delphi techniques, as previously described,
10 can be done via mail using structured questionnaires to
11 direct participants to identify issues relevant to the
12 topic of interest and then rank the issues in order of
13 importance.

14 Traditional focus groups typically involve 8 to
15 12 individuals who discuss the topic of interest under the
16 direction of a trained moderator. The moderator must be
17 trained in group dynamics and have strong interviewing
18 skills. This is important to avoid domination of
19 aggressive individuals in the group and to include quiet
20 individuals. They are structured and use an interview
21 guide to help direct the discussion. They last from 1 to 2
22 hours depending on the research question and the
23 characteristics of the participants. A recorder is
24 generally used to take field notes during the session.
25 Findings are often transcribed from the recording.

1 For in-person groups, facilities designed for
2 group interviewing are ideal, enabling members of the
3 scientific team to observe the discussion and, if
4 consistent with the study design, provide the moderator
5 with additional questions or queries pursuant to the
6 group's discussion.

7 Focus group participants are chosen based on
8 characteristics that the researcher wants to understand
9 further, also known as break characteristics and control
10 characteristics. The number and nature of the groups and
11 sessions is determined by the purpose of the study, the
12 design complexity. For example, if the characteristic of
13 interest is complex, a researcher may want to conduct
14 several focus groups to make sure all relevant themes are
15 identified. But typically two to three focus groups are
16 conducted in diverse geographic regions, and the nature and
17 number of groups is also based on the resources allocated.

18 Data from focus group include tape recordings,
19 transcriptions, which for a 2-hour session could be up to
20 40 to 50 pages, and field notes which are usually taken by
21 a second researcher during the focus group session.

22 The analysis is driven by the underlying
23 research question and involves a careful review, synthesis,
24 and summary of data from tape recordings, transcription,
25 and field notes. Qualitative data is interpretive and

1 constrained by the context. In addition, the topics are
2 generally linked to the interview guidelines. Data
3 gathered during the focus groups take the form of
4 information, quotations, themes, and issues gathered from
5 the participants during the course of the interview.

6 Steps involved in data analysis are mechanical,
7 such as organizing, and interpretative, such as identifying
8 common themes and patterns within themes and drawing
9 meaningful conclusions. Software such as Ethnograph may be
10 used to help identify themes.

11 Reliability of data may be enhanced by repeated
12 review of the data and by independent analysis by two or
13 more experienced analysts.

14 Results are expressed qualitatively as themes,
15 issues, or concerns and are highlighted with substantiating
16 quotes. Results also may be presented quantitatively such
17 as the number of participants who agreed or disagreed on
18 particular issues and the frequency of themes within the
19 group discussion. The appropriate sample characteristics
20 are also presented so the reader or the reviewer has an
21 understanding of the nature of the participants providing
22 the data.

23 Focus group methodology is only as useful and
24 as strong as its link to the underlying research question
25 and the rigor with which it is applied.

1 Strengths of focus groups are: that they
2 provide concentrated amounts of rich data in the
3 participants' own words on precisely the topic of interest;
4 that the interaction with respondents and interaction among
5 group members add a richness to the data that can be missed
6 in individual interviews; and that the data can provide
7 critical information in the development of hypotheses or
8 the interpretation of quantitative data.

9 The primary limitation of focus group
10 methodology is the relatively small number of participants
11 and the limited generalizability to the larger population.

12 Group dynamics can also be a challenge or a
13 limitation. A group with particularly quiet individuals or
14 aggressive talkers or a group with a tendency toward
15 conformity or polarization can make group dynamics
16 difficult, particularly if the moderator is inexperienced.

17 Careful attention to study design replication using
18 multiple groups within a study and a well-trained
19 experienced moderator can minimize this limitation.

20 In some cases, interpretation can be time-
21 consuming and require several experienced analysts. To
22 enhance the strength of the results, independent analysis
23 by two or more analysts is always preferred.

24 Focus groups may be a useful method for
25 identifying problem areas in testing proprietary drug names

1 to minimize medication errors. For example, this
2 methodology is ideal for understanding potential sources of
3 confusion from the user's perspective, and therefore focus
4 group participants include physicians, pharmacists, and
5 nurses, as well as patients and caregivers.

6 Focus group methodology also can be used to
7 identify situations in which confusion is most likely to
8 occur. For example, in particular patient populations,
9 such as elderly patients taking multiple medications or
10 situations such as pharmacies where drugs are shelved
11 alphabetically by proprietary name.

12 Focus groups can also be used to test
13 conclusions of expert panels about sound-alike medications
14 that pose a threat in the practice or home setting, to
15 develop research methods for testing sound-alike
16 medications quantitatively, and for understanding behaviors
17 underlying prescription practices that can contribute to
18 name-related errors in order to identify high-risk
19 therapeutic areas.

20 Focus groups can also inform quantitative
21 research design; provide qualitative data to aid in the
22 interpretation of quantitative results, for example,
23 explain unexpected areas of confusion; serve as an integral
24 part of a multi-method evaluation program, for example,
25 triangulation with in-depth interviews with physicians,

1 pharmacists, or patients; and provide a useful foundation
2 for designing risk assessment and management studies, for
3 example, identifying potential problems in professional
4 practice and home use patterns.

5 When used appropriately, focus group
6 methodology can provide rich depth of understanding of a
7 problem or phenomenon of interest. Depending on the
8 response question, it can be used in isolation or to
9 complement or supplement quantitative methods. And as is
10 true of all research methodologies, its utility is a
11 function of its link to the research question and the rigor
12 to which it is applied.

13 DR. GROSS: Thank you, Dr. Kimel.

14 Any questions for Dr. Kimel? Yes, Lou Morris.

15 DR. MORRIS: In your conclusion, you say it can
16 be used in isolation, but in all the examples you gave, it
17 seemed to be used in combination. Could you describe a
18 situation where you think it could be used in isolation?

19 DR. KIMEL: In general, I think it could.
20 Probably for the purposes of working with drug naming, I
21 think it would probably be best to be used in combination.

22 DR. GROSS: Any other questions from the panel?

23 (No response.)

24 DR. GROSS: If not, we'll move on to the next
25 speaker. Kraig Schell, Assistant Professor, Department of

1 Psychology at Angelo State University, will discuss use of
2 laboratory and other simulations in assessing drug name
3 confusion.

4 DR. SCHELL: Good morning. Let me start with a
5 couple of preliminary remarks: first, to tell you what a
6 privilege it is to be here with you this morning, and
7 second, to express deep regret that unfortunately Tony
8 Grasha, whom many of you know, who would have been here
9 today, of course passed away about a month ago. So I'm
10 going to do my best to fill his very, very large shoes. A
11 lot of what I'm going to talk about today was research that
12 he and I had worked on for now the past seven years. But,
13 unfortunately, a good part of it is also in his head, and
14 so I'm going to do the best job I can to try and estimate
15 what would have been in his head with respect to some of
16 these topics.

17 The current state of the problem, as we've seen
18 it, he and I, over the past seven years, is clearly that
19 drug name confusion is a component that we need to be
20 concerned about with respect to patient injury and
21 financial loss. Many of the means of assessing drug name
22 confusion are primarily based on rational and
23 reductionistic approaches, such as FMEA and RCA,
24 phonological and orthographical analysis and expert teams
25 and committees, which all three, to some extent, are based

1 on a rational decision-making approach to the problem.
2 Unfortunately, as we know in psychology for quite some
3 time, humans aren't necessarily rational. In fact, we're
4 rather irrational things, and the problem of name
5 confusability is also a broad and less rational problem
6 than might be assumed just by looking at it superficially.

7 Some of the research that we've done over the
8 last seven years has identified many of these factors, as
9 well as several others that I didn't have the room to list,
10 as potential problematic variables that can affect error
11 production and error capture in pharmacy filling and
12 verification tasks done both in our laboratory at the
13 University of Cincinnati and also at Angelo State
14 University where I am and also in field sites that we've
15 worked with over the past few years.

16 Our approach to the problem is based on these
17 following assumptions and observations. Drugs that look
18 and sound similar are not confused with each other or
19 misfilled, at least with the current data we have
20 available, in the same proportions that we would expect
21 based on their similarity indices. For instance, Zantac
22 and Xanax which was talked about before. Obviously very
23 similar phonetically and also has quite a bit of similarity
24 in terms of its bigrams and trigrams, but you would expect
25 that with degree of similarity that we would be misfilling

1 that drug 7-8 times out of 10. Thank God, that's not the
2 case. Actually we're much more accurate than that.

3 That leads me to believe that that variable,
4 although it is important phonologically and
5 orthographically, is not the only problem obviously. And I
6 agree with what Bob Lee said earlier. There are definitely
7 other conditions that need to be included and added into
8 the equation such that perceptual factors are necessary,
9 but not necessary and sufficient explanations for why the
10 problem of human error exists.

11 And the third assumption that we rest on is
12 that human error as a process is not rational. In fact,
13 Dr. Riesen, in his classic work in 1990 on human error,
14 called errors latent pathogens that sit inside systems and
15 processes in every organization and every realm of society
16 that are just waiting for a situation to bring them to the
17 surface and infect it with an error.

18 I'm reminded of the problem that occurred with
19 the USS Vincenz and the Iranian airliner a few years ago in
20 the Persian Gulf, and if you evaluate that particular topic
21 very closely -- and many people have in the psychological
22 literature -- you see that the individual components of
23 that particular event weren't necessarily problematic in
24 and of themselves. It was the combination of those
25 components in that particular given situation that led to

1 the erroneous decision to shoot down that airliner. That's
2 the approach that we're taking, which is much more
3 consistent with a human factors approach to the problem,
4 much broader in its scope.

5 So simulating, as we do in the research we've
6 done for the past seven years, gives us the ability to look
7 at human factors that might interact with the physical
8 characteristics of a drug name. In other words, under what
9 conditions are Zantac and Xanax more or less confusable?

10 One possible thing that we could talk about
11 here -- and I'll mention it again later in the talk -- is
12 the informational context surrounding the drug. For
13 instance, Mr. Phillips talked a little bit about the
14 Avandia and the Coumadin misfill and mentioned in his talk
15 a very important point, that the dosage and the
16 administration of the drug is probably a significant
17 contributing factor to the confusion of Avandia and
18 Coumadin, two words that look, as he said, relatively
19 nothing alike. And it's those kinds of factors and those
20 kinds of issues that we can look at in a simulation
21 paradigm.

22 This is the model that we are proposing that
23 Dr. Grasha and I built and I am proposing it to you today
24 that the simulation structure should take. Along the left-
25 hand side of the slide there, you see what is called the

1 control/realism continuum. Generally speaking, as control
2 increases -- in other words, as experimental control is
3 strengthened -- the realism of the simulation decreases.
4 So the stuff at the top of the pyramid that you see, the
5 lab simulation and the pharmacy school simulations, because
6 of the necessity of experimental control in those
7 paradigms, they're necessarily going to be somewhat
8 artificial and they're going to eliminate sources of
9 variance that could be important.

10 As you progress down the pyramid to the error
11 monitoring stations, there we have a great deal more
12 realism as we're actually working in pharmacies and
13 hospitals around the country, but the control that we have
14 over error production and error capture is lessened. It
15 requires the complete model to get a full and total picture
16 of how medication errors exist and are produced and are
17 captured. Just looking at one of these levels is not going
18 to give us a complete picture.

19 The simulation also allows us to capture what
20 we call a subjective error. Basically what that is is an
21 error that is made and is corrected before it leaves the
22 pharmacy. These are a significant source of error in our
23 research that are not going to be predominantly recorded in
24 self-reporting databases such as USP, et cetera. The
25 objective error would be the error that actually left the

1 pharmacy and then was recorded as one of those that
2 occurred. We call them also process errors because they
3 are errors of the human process that is required in order
4 to fill or verify a script from beginning to end.

5 One very interesting finding that we replicated
6 numerous times in both the laboratory and in retail and
7 outpatient pharmacies is that for every six process errors
8 that we can capture, one of those tends to get by all
9 verification steps and actually leave the pharmacy and be
10 dispensed to consumers. We believe that's a very important
11 ratio because if we can demonstrate that a particular drug
12 is creating an inordinate amount of process errors, that
13 gives us pause and makes us begin to think that if that
14 drug name were allowed to be put into actual pharmacies,
15 running the risk of pharmacists being more vulnerable to
16 moving into an error mode of processing and then, as a
17 result, more of these scripts actually leaving the
18 pharmacy.

19 Another benefit to the simulation is that it's
20 safe. None of these drugs actually go to anyone and they
21 aren't actually taken by anyone during the simulation. So
22 we can make as many errors as we want to and no one is
23 actually harmed by them. In fact, one of the designs that
24 Tony was going to do before his untimely passing that we
25 talked about for several years was to use the simulation to

1 train for errors, as has been done in other fields, where
2 we actually force the participant in the simulation to make
3 the mistake over and over and over again, to build a schema
4 for that mistake so when they do it later, they recognize,
5 wait a minute, it's not right, I shouldn't be doing this,
6 this doesn't feel correct, and they're able to make a
7 correction.

8 It allows us to use a variety of different
9 experimental and quasi-experimental designs. We can do
10 case studies. If we wanted to look at team performance in
11 the pharmacy techs and pharmacists and how they're
12 interacting, we can do that. We can do an actualistic
13 observation design, a variety of different approaches are
14 possible in the simulation.

15 And we can insert drug names that are being
16 evaluated into an existing database of already evaluated
17 and marketed drugs to see if anything currently on the
18 market that maybe we haven't pinpointed up to this point is
19 a source of potential error that we may have overlooked.

20 Three laboratory approaches that I can talk to
21 you about. Two of them we've done already. The third one
22 is in production right now.

23 The full-scale dispensing task is exactly what
24 it sounds like. We use mock materials to allow
25 participants to fill mock orders for these prescriptions.

1 It's actually rather amusing if you were to take a look at
2 it, and I'll show you a picture in a moment. We use things
3 like craft beads and paper clips. We even used cereal at
4 one point in time. We had some Trix cereal on the shelf
5 that we were calling drugs and assigning names to them and
6 having people dispense them as if they were sitting in
7 front of a bench in a pharmacy.

8 The verification task is where the scripts are
9 filled beforehand and an individual takes the scripts in
10 sets, verifies them against a database with the same
11 information that would have been on the label, and then
12 tells us whether this order is correct or this order is not
13 correct. Very similar to what a pharmacist might do going
14 back to through the will-call or the return-to-stock bins
15 to see if anything was erroneous in that sense.

16 And thirdly, the drug name perception task
17 following the methods of Bruce Lambert and also Dr. Dorr,
18 what she's doing. I'm building this currently at Angelo
19 State University to be able to look at drug name confusion
20 from that human factors perspective, being able to add
21 different individual difference factors and see how that
22 influences the confusability of the names.

23 That's a panoramic view of the original
24 pharmacy simulation lab. It didn't reproduce very well in
25 your handout, but essentially it's just portable plastic

1 shelves with a computer work station. The scripts were
2 written on index cards and in various styles of
3 handwriting, and participants simply sat in front of the
4 computer and were able to fill the scripts as if they were
5 working in a pharmacy. They do sit. Pharmacists for the
6 last few years have told us how unfair that is because they
7 always have to stand.

8 (Laughter.)

9 DR. SCHELL: The only explanation I can offer
10 you is we didn't have any tables that were tall enough, so
11 we had to make do with what we had.

12 This is the verification lab I currently run at
13 Angelo State University. On the right-hand side, those are
14 the scripts. We use standard 30-count pill bottles.
15 You'll notice that there is a 3-by-5 index card in each of
16 the bags. We use that to simulate the label that would
17 normally be attached to the bottle, and we chose to do that
18 primarily for convenience. The labels would eventually
19 tear or start to lose their adhesion, and it would become
20 an issue of cost. The index cards are much more durable,
21 so it allows us to keep our costs down.

22 But the individuals simply look at each script,
23 decide whether the correct item is in the bottle, whether
24 the correct amount of that item is in the bottle, and
25 whether the index card information matches a database that

1 they are presented with for that particular script.

2 The drug name confusion task. The interface
3 for this is currently being built, so I'll describe it the
4 best I can. Essentially a drug name would be presented to
5 a participant on the screen and we'll be able to vary the
6 amount of time they'll be able to see that name. Then they
7 have to navigate through a virtual shelf where they have to
8 select first what letter did that name start with. Then
9 that will move them to a new screen where there will be a
10 variety of different drug names starting with that letter,
11 and then they have to select the drug name that they
12 believe they saw.

13 Now, here's the kicker. Once they select one
14 of the letters, they can't go back. So if they select a P,
15 for instance, and then they realize, oh, man, it didn't
16 start with a P, well, they're kind of stuck now. They're
17 going to have to select the one that they think is closest
18 to what they saw, realizing they've already made the error.
19 The reason we make it so that it does that is so that we
20 can separate process errors from committed errors. When
21 each of those occurs, we'll be able to separate them out.

22 We can change the duration of name
23 presentation, the inclusion of informational context. We
24 can add feedback to tell the performer whether they're
25 doing well or whether they're doing poorly at given

1 intervals.

2 The informational context variable I should
3 also mention can be switched to a different domain of
4 knowledge. Since we're looking at basic human performance
5 and we're using primarily naive participants, most of our
6 participants don't know quinine from Celexa. So dosage and
7 administration information is relatively meaningless to
8 them. So we have four different knowledge bases that are
9 more in a college students domain, such as television,
10 movies, sports, and things like that, and then we can
11 provide informational context around those and study
12 basically the same perceptual processes.

13 The pros. Strict control is the biggest
14 advantage to the laboratory simulation. We can tailor that
15 as necessary. We can vary systematically different factors
16 that we believe to be important. What I mean by
17 customizable products is that we can do more than one
18 product name at a time. We can insert 20 different product
19 names into a given experimental design if we wanted to, and
20 provided folks are on task long enough, we could look at a
21 variety of different permutations and combinations of
22 those.

23 The disadvantages. The lack of realism.

24 Shorter versions of the task tend to be overly
25 simplistic, and what I mean by that is the shorter that

1 they're on task -- and believe me, getting a college
2 student to do anything for 2 hours is a chore. We have to
3 eliminate a lot of things that pharmacists do such as take
4 phone calls, be interrupted by customers, have to deal with
5 insurance companies, and those kinds of things. Longer
6 periods of time on task, we can add those things in.

7 It's possible that we might control some causes
8 of name confusion and other sources of error in the
9 experimental design per se. So numerous experimental
10 designs and numerous studies would have to be employed.

11 The movie set simulation, the second tier, is a
12 broader-based pharmacy simulation where the environment is
13 more similar and more exact with respect to an actual
14 pharmacy. The emphasis would be on duplicating the work
15 flow and other conditions under which prescription filling
16 and checking would occur, such as the insurance companies
17 and the multiple scripts at one time, and the irate
18 customers, and those kinds of things. Both objective and
19 subjective data could be collected in this as well.

20 A note of explanation here. By training I am a
21 business psychologist, and one of the things that many
22 corporations do to select managers is something called an
23 assessment center -- maybe some of you are familiar with
24 that -- where management trainees will be placed in an
25 observation tank, basically a large area, and given a set

1 of exercises to do while current managers watch and rate
2 them. In the movie set simulation, we apply the same basic
3 analogous idea to this particular level of the pyramid. We
4 would be able to create exercises that incorporate many of
5 these factors that could impact performance into a series
6 of exercises that then we could do with each of these drug
7 names.

8 So there could be the insurance fiasco
9 exercise, for instance. How does dealing with an insurance
10 company while you're filling a script for that particular
11 drug name impact its confusability?

12 The multiple script exercise.

13 Similar preceding name. Much of what we've
14 done to this point has been on looking at pairs of names
15 simultaneously. Well, what happens when we have a
16 consistent, frequent representation of one name, followed
17 by then a highly confusable name right after that? Is
18 there a perceptual bias toward the name that had been
19 perceived first?

20 Frequent prescription exercise.

21 Stressed out exercise.

22 All these things that you see here could be
23 designed and we could, just like the gauntlet, run a name
24 through a series of these exercises to see how different
25 environmental conditions affect their confusability.

1 The simulations in the colleges of pharmacy are
2 very similar to the movie set simulation, but with one
3 important difference. In the movie set simulation, the
4 emphasis is on researching and pinpointing environmental
5 and individual difference factors that could impact
6 confusability. In the college of pharmacy, we would then
7 take that knowledge into a similar situation in the college
8 of pharmacy and then train new pharmacists on those
9 situations in individual difference factors, being aware of
10 them, understanding that they occur, understanding how they
11 influence confusability, and be able to dedicate a little
12 bit more training toward the confusability factors that
13 enter into doing their job on a daily basis.

14 So in the movie set simulation, really basic
15 research is the emphasis. In the college of pharmacy
16 simulation, training is the emphasis. As a result, it may
17 not be quite as flexible for manipulation and
18 experimentation since training is a little bit different
19 approach than basic research.

20 Finally, the error monitoring station. In
21 automated pharmacies, especially the pharmacist's role is
22 switched from filling to verification largely. As you, I'm
23 sure, are aware, in many States now technicians can do most
24 of the filling tasks by themselves. In Texas I believe a
25 technician can do everything from start to finish. The

1 only thing that's required is that a pharmacist check the
2 script before it leaves. So that's starting to become a
3 trend. So verification is becoming more and more
4 important.

5 This test would insert the new drug into an
6 existing pharmacy that would be, of course, in connection
7 with FDA or the pharmaceutical companies. Controls would
8 be in place to ensure that the drug is not actually
9 dispensed, but we would insert mock orders for this drug
10 into the standard flow of everyday business. Two types of
11 data could be generated here.

12 Of course, objective, end-result data. We're
13 very interested to see if an error with that particular
14 drug makes it out of the verification process.

15 But secondly, we're also interested to see
16 whether the drug creates those process errors that we
17 talked about. The way that we do that is that pharmacists
18 and technicians carry what we call a self-monitoring
19 booklet around with them, and whenever they catch
20 themselves about to make an error with this targeted drug,
21 we simply ask them, when they have a moment, to pull their
22 booklet out and simply note a tally mark, oops, almost
23 messed that one up. We also ask them to monitor those
24 self-corrections for other drugs because we want to look
25 for confusability pairs and see if any of those are there.

1 So both types, subjective and objective data,
2 are recordable.

3 The advantage to the monitoring station is that
4 there's really no conflict of interest in the sense that
5 it's kind of a live test. We're not expecting any kind of
6 result. We know maybe what we should see based on the
7 earlier stages of the model, but there's really no hidden
8 agenda ideally based in that. It's an actual, real-world
9 environment, as realistic as we can make the simulation.
10 That's the goal of the monitoring station.

11 There are marketing ramifications as well.
12 Drug companies could get some information about how these
13 drugs may be marketed in a different way than they
14 currently are or would be. There could be some information
15 that comes out of the simulation with respect to that.

16 The disadvantages. There is a risk of
17 accidental dispensation, the risk being that there's an
18 actual order for drug A, the test drug gets dispensed to
19 that person by mistake. That risk is there. It could be
20 correctable with observers on site from the testing
21 authorities.

22 There is a use of self-report data, and the
23 process errors are completely self-report. We know from
24 just human nature that sometimes we are not very quick to
25 recognize the fact that we almost made a mistake,

1 especially if that mistake is one that could have caused
2 potential harm. So we have to take the self-report data
3 with somewhat of a grain of salt.

4 And there is a lack of sample size possible
5 because the number of these monitoring stations is probably
6 going to be fairly small because of just the expense and
7 the coordination necessary to create this kind of system.
8 So can we really say that what happened in six pharmacies
9 is going to happen in 60,000? That's an issue that we'll
10 have to deal with.

11 Now, let me say a brief word about validation
12 overall because I think the model in its entirety can be
13 talked about very quickly and very simply with respect to
14 validation. The nice thing about the model -- and it's a
15 model that human factor psychology has used for years in
16 determining the usability of products and human and
17 computer interactions and those kinds of things -- is it
18 tends to verify itself predictively. In the initial stages
19 of the model, we develop predictive expectations on what we
20 should see in the later stages. If we don't see that, we
21 can then go back and refine or revise those predictions,
22 collect more data. So the predictive validation process is
23 kind of inherent in the model.

24 As far as construct validity, the question we
25 have to ask -- and it's a question I've wanted to ask this

1 entire morning -- is what exactly are we looking for here.

2 I think what our model is designed to target, as far as a
3 construct, is error proneness. What we're looking at is
4 how prone or how vulnerable is that particular name to
5 confusion as an average statistic? When we define error
6 proneness as the construct that we're targeting, then the
7 model begins to make more sense because every step of the
8 model then can be targeted toward answering the question,
9 is this a mistake-prone name or is this not a mistake-prone
10 name? That I think is a broader question. It goes beyond
11 just the mere issues of similarity orthographically and
12 phonetically, even though that is a component, but it's a
13 broader question that may give us a more complete answer.

14 DR. GROSS: Thank you very much, Dr. Schell.

15 The next speaker is Dr. Sean Hennessy,
16 Assistant Professor, Department of Epidemiology and
17 Pharmacology in the Center for Clinical Epidemiology and
18 Biostatistics at the School of Medicine, the University of
19 Pennsylvania. Dr. Hennessy will talk about quantitative
20 evaluation of drug name safety using mock pharmacy
21 practice.

22 DR. HENNESSY: Good morning and thank you.

23 First, by way of disclosure of conflict of
24 interest, I want to point out that I recently accepted an
25 invitation to serve as an unpaid member of the Board of

1 Directors of Med Errors.

2 So I'm going to be talking about quantitative
3 evaluation of drug name safety using close-to-reality
4 pharmacy practice settings. A lot of what I'm going to be
5 presenting is similar to what we just heard from Kraig
6 Schell with the notable exception that I'm unburdened by
7 any practical experience in the area.

8 (Laughter.)

9 DR. HENNESSY: So I'm going to focus more on
10 the context in which information from such simulations can
11 be done. In Kraig's diagram, this would probably line up
12 with the movie set.

13 So first I'm going to talk about a big-picture
14 view of drug name safety. How do we improve the process by
15 making it quantitative or why might making it quantitative
16 improve it? I'll briefly go over a model for measuring the
17 error-proneness of particular drug names in a mock pharmacy
18 setting and then talk about a research agenda.

19 So an overly simplified view of drug naming as
20 it currently takes place is that there's a name. It goes
21 through some evaluation process, as we heard earlier this
22 morning. It's largely a qualitative evaluation process,
23 and then there's some outcome. Either we accept it or we
24 reject it. This is much the same process as you could use
25 either for tomato soup or for Andy Warhol's art.

1 So the question is will we derive any benefit
2 from making what is a qualitative process and depicted here
3 as a black box, not coincidentally since many of the
4 processes are not particularly well described, so they have
5 that black box feature to them. So is there any benefit to
6 making a qualitative black box process more transparent and
7 more quantitative? So let me talk about the possibilities
8 there.

9 So what might some potential benefits be of
10 injecting a quantitative aspect to this? First is that we
11 make the process more explicit and systematic. We use a
12 fuller range of available information. We have
13 transparency of data and assumptions. We acknowledge
14 places that we're uncertain, and we identify knowledge gaps
15 that then serve as areas of future research.

16 So then we need to ask the question, once we
17 have the evaluation process, do we have enough information
18 to make an accept-or-reject decision? What underlies this
19 binary decision, go/no go, or is there really a spectrum of
20 drug safety or error-proneness? And there needs to be some
21 decision as to where the threshold is set on that spectrum.

22 So maybe it's really a rating that we need to
23 have as an intermediate step between the evaluation process
24 and the outcome. Certainly the rating in the middle, which
25 is probably what I'll spend the majority of my time on,

1 should incorporate the probability of error. However, we
2 need to ask is this enough. Are all medication errors
3 created equal? There are some data that 99 percent of
4 medication errors don't result in an observable adverse
5 drug event. So should we focus on all equally, or should
6 we focus on those that are more likely than others to
7 result in an adverse drug event?

8 For example, is substituting erythromycin for
9 clarithromycin, two antibiotics with similar spectrums,
10 equally bad as confusing chlorambucil which is a
11 chemotherapeutic agent with chloramphenicol which is an
12 antibiotic?

13 So the rating may also take into account the
14 consequences of the error in addition to the probability of
15 the error. So under consequences of the error, that
16 probably has multiple components too, the first of which --
17 and I'm echoing some things that were said earlier this
18 morning, but not because I knew that they were going to be
19 said -- one of which is the probability of error of an
20 adverse event given that an error took place. And what are
21 some factors that might go into that?

22 The first includes adverse outcomes from not
23 getting the drug that was intended to have been dispensed,
24 and we can get information from that presumably from the
25 placebo-controlled trials that have been done demonstrating

1 the efficacy of the drug.

2 The probability of adverse events also depends
3 on the identity of the drug that is mistakenly substituted
4 which may be measurable empirically as I'll talk about in a
5 little while.

6 And the third factor is the frequency of
7 adverse events in recipients of people receiving the
8 substituted drug. So given the substituted drug, what's
9 the safety profile of that? And that should be known from
10 pharmacoepidemiologic data about those drugs.

11 So in this rating, we have two factors, the
12 second of which has two subfactors. So there's the
13 probability of the adverse event, and then there's also the
14 disutility of the adverse event under consequences of the
15 error.

16 Let me talk about disutility for a minute.
17 Disutility is defined as the value of avoiding a particular
18 health state which is usually expressed on a scale between
19 0 and 1. This could be measured empirically by asking
20 patients standardized questions. An example of this is
21 presented here. This is disutility for outcomes of occult
22 bacteremia going from everything to a very small disutility
23 for just having your blood drawn to a very high disutility
24 for death. I'd like to point out here that there are
25 apparently things worse than death.

1 (Laughter.)

2 DR. HENNESSY: So one possible quantitative
3 rating would be the probability of error times the
4 consequences of the error, the consequences of the error
5 being the probability of an adverse event given that an
6 error occurred, multiplied by the disutility of the adverse
7 event.

8 So then we have two axes. On the y axis, we
9 have the consequences of an error. On the x axis, we have
10 the probability of an error. You multiply those two things
11 together, you get a severity rating going from blue, not so
12 bad, to red, terrible. So you can get a bad severity
13 rating either if you have a very serious event that occurs
14 infrequently or a frequent event that's not so serious.

15 And here's Einstein discovering that time is
16 actually money.

17 All right. So then in a process we need to ask
18 the question, what settings do we perform this evaluation
19 in? We could think about doing it in any number of
20 settings: inpatient pharmacies, outpatient pharmacies,
21 physicians' offices, nursing home settings. This list can
22 go on and on.

23 So let me talk briefly about a model for
24 measurement of some of these parameters in a mock pharmacy
25 practice setting. So here's a photograph of a mock

1 pharmacy. These typically exist in schools of pharmacy,
2 although they can be built for specific purposes as well.

3 What we can hope to gain from looking at a
4 model like this would be both an empiric measurement of the
5 probability of error, as well as get insight into what the
6 consequences of the adverse event would be from knowing
7 which drugs are mistakenly dispensed for the intended drug.

8 So some of the features of the close-to-reality
9 simulated pharmacy practice include that it could be done
10 in new or existing simulated pharmacies.

11 It could be done either using per diem real
12 pharmacists or late-year pharmacy students, with the
13 tradeoff being it costs more money to pay real pharmacists
14 than it does pharmacy students, but you might get more
15 realism.

16 The test drugs that we're studying would need
17 to be listed both in the computerized drug information
18 sources that are being used in the pharmacy, as well as in
19 the computer system in which they're entering.

20 Then, of course, test drugs need to be put on
21 the pharmacy shelf.

22 We would then simulate pharmacy practice by
23 presenting prescriptions, phone prescriptions, electronic
24 prescriptions, written prescriptions, for both the real
25 drug and the test drug. As was mentioned earlier, you can

1 add prescription volume, noise, interruptions, third party
2 reimbursement issues, Muzak, irate patients, as you like.
3 The pharmacist enters the prescription into the computer
4 system and then fills it. Then we measure the rate of name
5 mixups at all stages of the filling process, as well as
6 which drug was mistakenly substituted.

7 So when using the data obtained from such
8 simulations to our formal quantitative evaluation process,
9 we need to ask for the probability of an error. Do we use
10 the measured probability of the error or do we use
11 something else like maybe the upper bound of the 95 percent
12 confidence interval? To remind you, the upper bound of the
13 95 percent confidence limit is the maximum value that is
14 statistically compatible with the data and it's a function
15 of both the study size and the measured rate, the point
16 being that if we require use of the upper bound of the
17 confidence limit, that will encourage a larger study than
18 using the point estimate.

19 Which confidence intervals do we want to use?
20 That might be subject to debate. 95 percent confidence
21 intervals are common for biomedical research. It's a
22 different context here, so we might want to think about
23 other confidence limits, and that may be based on what
24 seems reasonable going through this whole process with
25 drugs that are at least assumed to be bad, some gold

1 standard bad drugs, if there is such a thing.

2 Potential advantages versus expert opinion.

3 First, it yields empiric estimates of the error rate and of
4 which drugs are mistakenly substituted. I would put forth
5 it has better face validity. Further, the validity can be
6 tested by examining known bad drug names, if we can get a
7 group of people in a room to agree to what those are. It
8 makes the knowledge and assumptions that go into the
9 process explicit and transparent.

10 Obstacles and limitations. There are certainly
11 those. The first is the Hawthorne effect; that is, when
12 you watch people do something, they're generally better at
13 it than when you're not watching them. The way to overcome
14 that is if you do it enough, the Hawthorne effect is
15 thought to go away.

16 There are technical challenges in developing
17 movie set pharmacies and making them work also.

18 You need large sample sizes. Presumably these
19 are going to be low frequency events, and in order to
20 detect low frequency events, you need lots of repetitions.
21 That's going to be expensive.

22 Do we use such a process routinely for all new
23 drugs, or maybe do we use this as a way to validate
24 existing or improved or otherwise less costly processes?
25 And is doing so worth the added cost?

1 So now let me put forth the research agenda
2 with regard to this particular proposed model. First is
3 feasibility. Second, cost. Reliability. If we implement
4 this strategy in different settings, do we get the same
5 answers? The validity of it vis-a-vis what we believe to
6 be both known good names and known bad names. And the
7 ultimately utility of it.

8 So this is the straw man that I'm putting up
9 for discussion, and I'd be happy to take any questions.
10 Thank you.

11 DR. GROSS: Thank you, Dr. Hennessy.

12 At this point we'll entertain questions for Dr.
13 Schell and Dr. Hennessy and Dr. Hennessy's straw man.

14 (Laughter.)

15 DR. GROSS: Yes, Jackie.

16 DR. GARDNER: I'd just like to ask Dr. Hennessy
17 whether you have a recommendation about how routinely these
18 should be used, given what you've described as a fairly
19 extensive and expensive prospect. And if you only focused
20 on the 1 percent of AEs that resulted in harm, for example,
21 or targeted those, then you're looking at a big effort
22 here. Do you have some modeling recommendation for how to
23 decide what would be the most useful or cost effective way
24 to proceed with this?

25 I was thinking of your Hawthorne effect not

1 only in observation, but proceeding with an IRB which would
2 be necessary for this. Having described to everyone what
3 exactly you're doing as part of the IRB process, then you'd
4 have to wait even longer, I would think, before you saw --

5 DR. HENNESSY: Right. It's a cumbersome
6 process. Is it worth it for all drugs? That's a good
7 question. It's really a policy decision that I'll leave to
8 the group for discussion.

9 DR. LEVIN: This is just a point of
10 information. If there are no human subjects involved, why
11 is this an IRB issue?

12 DR. GARDNER: Probably because the pharmacists'
13 activities would be looked at. That would probably be the
14 stance taken.

15 DR. GROSS: Michael Cohen.

16 DR. COHEN: If you're doing it in a live
17 pharmacy, which at least one of the speakers talked about,
18 there's always a chance of an actual error, and that has
19 actually happened. We've had a recent report of a test of
20 a computer system that led to a very serious error.

21 Could I ask a couple questions?

22 DR. GROSS: Go right ahead.

23 DR. COHEN: Has anybody actually used this
24 model at this point, and is there anything in the
25 literature about it? Because I think I'd like to know more

1 about it. I see some possibilities, but I haven't actually
2 seen that used. Has anybody actually done this with
3 proposed names, not with actual products that are on the
4 market? That's the point.

5 DR. SCHELL: When you say this model, the whole
6 entire thing or --

7 DR. COHEN: The model pharmacy concept. The
8 lab is one thing, but the model pharmacy --

9 DR. SCHELL: Right. Not that I'm aware of.
10 I'm currently speaking with a school of pharmacy right now
11 about negotiating with them to use a new simulation that
12 they're building, but to my knowledge, I don't know that
13 anyone has done that.

14 DR. COHEN: I have one more question. When you
15 do this, you would use actual handwritten prescriptions,
16 but in fact, you'd need to test several handwritten
17 prescriptions from different people that actually wrote
18 that in order to make this work. So not only do you have
19 perhaps 10 different drugs, but you might have 10 different
20 actual scripts. It gets to the point where is this really
21 a real-world experiment. That's the one concern I would
22 have if you actually used a model pharmacy.

23 DR. SCHELL: And there's no question that as
24 the model gets down toward the base of the pyramid, the
25 complexity of it obviously dramatically increases. In an

1 ideal world, what we would hope is that the initial stages
2 of the model would give us some idea about what sorts of
3 script you might be more or less likely to see.

4 The other thing that I would say to that too is
5 that, as you know, more and more scripts are now coming
6 into the pharmacies electronically or with typewritten
7 words, and also there's the whole bar coding phenomenon
8 that's coming up. So I think that the model pharmacy will
9 get less complex when that becomes more of a frequent
10 occurrence.

11 DR. GROSS: Just to clarify, of the four
12 simulations described, lab simulations have been tested,
13 pharmacy schools simulations have been tested, movie set
14 simulations have not, and real pharmacy simulations have
15 been done. Is that correct?

16 DR. SCHELL: Let me say this to that. With
17 respect to our particular research and research like ours,
18 the laboratory simulation has been done and the field work
19 which would be most similar to the error monitoring
20 stations, at least a version of those -- we've done those
21 in the past. But this particular model that I presented to
22 you today in the context of drug name confusion is a
23 synthesis of several different approaches that at this
24 point is a framework model at best.

25 DR. GROSS: Any other questions? Yes, Ruth

1 Day.

2 DR. DAY: I'd just like to, first of all,
3 express regret at the passing of Tony Grasha. He had so
4 many creative ideas, and I'm pleased that Dr. Schell is
5 able to continue his collaboration nonetheless.

6 My question for him is, as you go from the
7 controlled laboratory situation to the real world, you're
8 increasing ecological validity and decreasing control, but
9 are there some controls that you can keep? For example,
10 when a pharmacist has to go and find a particular drug
11 that's a target drug, how many foils, that is to say, other
12 things on the shelves, would there be? Is that the type of
13 thing you can continue to control?

14 DR. SCHELL: Certainly. And in fact, you could
15 even create that as a manipulable variable. What I'm
16 reminded of is an experience we had with a chain in Florida
17 who had created a targeted drug shelf, so the top 25 drugs
18 that usually got misfilled, according to their records,
19 were put on a special shelf with special markings and
20 designated as different from other types of drugs that
21 could have been confused as similar to it. Now, that
22 particular intervention was not tested. It was just an
23 idea somebody had and they decided let's just do this in
24 the pharmacies. They really didn't have any idea as to
25 whether it worked well or not. So, yes, that's one way

1 that comes to mind immediately when you say that, that you
2 could test different kinds of targeting mechanisms, adjust
3 foils, et cetera.

4 DR. GROSS: Eric Holmboe. Dr. Furberg.

5 DR. FURBERG: I also worry a little bit about
6 comparability when you compare experimental settings to
7 real life. I'm particularly concerned about whether the
8 individuals you're examining know that they're being
9 tested. They always do better. We know that from other
10 settings that if you know you're being observed, you spend
11 more time, are more careful, and you end up with an under-
12 estimation of the problem.

13 DR. SCHELL: I think that's a valid concern and
14 I think where that would be best addressed would be in the
15 error monitoring stations with some sort of blind or
16 double-blind procedure. That makes it a bit more complex
17 to install and makes perhaps controlling the possibility of
18 an error escaping the pharmacy more difficult to deal with.
19 But that would be the solution to the problem.

20 Now, at the more basic levels of the model, I
21 must make this very clear. My approach to these issues is
22 slightly different than Tony's was. Tony's was very
23 applied, you know, let's do the interventions and put them
24 together right now, let's get them in the pharmacy. The
25 reason he and I complemented each other so well is that I

1 tend to be more on the basic side. I tend to be more on
2 the basic cognitive and perceptual factors that contribute
3 to confusability in a broad context that then can be
4 applied to the study of errors. So we worked very well
5 together that way.

6 That's the part of the model that I think --
7 they're going to know they're being tested, and I'm not
8 sure there's that much you can do about it.

9 DR. GROSS: Eric, did you have a question?

10 DR. HOLMBOE: No, I'm fine.

11 DR. GROSS: Louis.

12 DR. MORRIS: I had a couple questions for Dr.
13 Hennessy. The idea of moving from qualitative to
14 quantitative is very appealing, but in theory doesn't every
15 drug potentially have a consequence and a probability with
16 every other drug? So how do you go across when there may
17 be so many drugs, and have you given any thought to how you
18 might get the indices that represent the potential across
19 the whole range of drugs?

20 DR. HENNESSY: So one way to do it would be you
21 only take the drug switches that you observe empirically.
22 They're the ones that you do the calculations for and
23 assume are going to be the basis of your adverse event. So
24 if you don't observe it, you assume it doesn't happen,
25 which means that you need to do large enough studies.

1 DR. GROSS: Arthur Levin.

2 DR. LEVIN: I guess this is a question for both
3 speakers. How do you design the simulation? There's a lot
4 of range in choice in what the variables are and how you
5 weight those variables. You know, do you have more Muzak
6 and less angry customers? Is there any empirical base for
7 sort of trying to emulate what the average setting might
8 be, number one?

9 Number two, if there isn't, is that sort of a
10 gap in data collection? In other words, if we're only
11 getting reports this happened and there's very little
12 detail, should we be looking for much more detail about the
13 setting and the circumstance? I suppose that's part of the
14 RCA maybe. But it seems to me if you build a simulation
15 that purports to represent the real world, you better have
16 some real-world foundations for putting that together.

17 DR. HENNESSY: I think that's a good point. I
18 would probably do some observations in real life,
19 quantitate those factors in real life, and maybe set the
20 pharmacy at the 75th percentile of that, just as an
21 example.

22 DR. GROSS: Michael Cohen.

23 DR. COHEN: Yes, that's close to what I was
24 just going to ask. But I need to point out that the
25 pharmacy is only one area that these errors actually occur,

1 obviously. A lot of it is on the nursing unit, in the ICU
2 and the emergency room and the OR, et cetera. There are
3 different environments. There are different types of
4 patients. There are different jargons, et cetera. That
5 would have to be taken into account because some of the
6 worst errors we actually experience are in those very
7 areas.

8 DR. SCHELL: If I could, let me speak to what
9 both of our expert panelists have said and kind of piggy-
10 back on what Sean said. Obviously, no simulator is
11 perfect. Even the aircraft simulators they have in the
12 Navy and the Air Force aren't perfect. They're awfully
13 good, but they're not perfect.

14 Ideally -- and again speaking in either world
15 -- the simulation in the later stages of my model would be
16 built from data collected in the early stages of the model.
17 I know that, for instance, there's currently being work
18 done on things such as Muzak and other environmental
19 factors by a company in Canada that I'm working with right
20 now and the researchers up there who are doing good work
21 right now in figuring out what environmental conditions
22 impinge on performance and those kinds of things.

23 In the movie set and in the college of pharmacy
24 portions of the model, as Dr. Day said, we can manipulate
25 some of those things. For instance, when does music become

1 noise is a question that has to be asked. We know
2 something about that factor from human factors literature,
3 but we have not applied that basic knowledge to the
4 pharmacy setting. We would need to do that to build the
5 simulator effectively.

6 DR. GROSS: When music becomes noise is also
7 relative to the listener.

8 Eric Holmboe.

9 DR. HOLMBOE: I have a question for both of
10 you. There's been a lot of work also done in evaluating
11 physician competence using simulation, particularly
12 standardized patients. But at the same time, there's a
13 growing body of work in actually videotaping encounters.
14 And I'm thinking of the same thing with regard to
15 pharmacies and other things. Has any work been done in
16 that area where they've actually had ongoing video camera
17 type analysis and break it down more, kind of an
18 ethnographic type of study in those environments?

19 DR. SCHELL: I can only speak to the one piece
20 of work that I'm familiar with. I'm familiar with it
21 because we used it to validate our original laboratory
22 simulation where pharmacists were filmed from the beginning
23 of a script to the final production, primarily used in time
24 motion studies. Dr. Lin at the University of Cincinnati
25 has done a lot of work with shaving time off scripts and

1 looking at motion effectiveness and those kinds of things.

2 We used that work as a validation for our own process to
3 figure out whether we were able to reproduce the time it
4 took to fill a script and approximately the number of
5 errors that were being produced in those studies as well.
6 But predominantly, to my knowledge, those were used in
7 efficiency studies for the most part.

8 DR. COHEN: Can I help to answer that too?

9 DR. SCHELL: Yes.

10 DR. COHEN: There is some excellent work by
11 Flynn and Barker which was the direct observation using
12 video. So it was very revealing.

13 DR. SCHELL: Yes. Good point. Thank you. I
14 forgot about that.

15 DR. GROSS: Paul Seligman and then we're going
16 to break for lunch.

17 DR. SELIGMAN: Has there been any effort to
18 compare the ability to detect the error proneness of a
19 product in laboratory or simulated environments or more
20 real-world environments with some of the other techniques
21 that we heard about this morning using computer-based
22 orthographic and phonographic techniques or expert panels?

23 Have either you all or others had the opportunity to
24 conduct those kinds of comparisons?

25 DR. SCHELL: Not to my knowledge. Dr. Dorr may

1 know of something. Maybe Mike might know of something.
2 But from my reading of the literature, it's basically you
3 have the computer approach and then you have the non-
4 computer approach, and the twain have not met yet.

5 Ideally that's one direction I definitely want
6 to go in. In fact, one study that I'm going to do. as soon
7 as we get the drug name confusion lab constructed at ASU,
8 is construct similarity indices and then run those pairings
9 and those drug names through my perceptual task on the
10 computer to see what kind of correlations I get. Do I get
11 the kinds of proportions of errors that I should expect
12 based on similarity ratings, or am I seeing a lack of
13 correlation there? I think that would be very informative.

14 DR. GROSS: Okay. Thank you all. It's been a
15 very interesting morning. We will break now and we will
16 reconvene at a quarter of 1:00, 12:45. Thank you all.

17 (Whereupon, at 11:40 p.m., the committee was
18 recessed, to reconvene at 12:45 p.m., this same day.)

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1 AFTERNOON SESSION

2 (12:45 p.m.)

3 DR. GROSS: We will begin the open public
4 hearing. For the panel, you have a purple folder that has
5 much of the information that will be presented. Patricia
6 Staub will go first.

7 MS. STAUB: Good afternoon, ladies and
8 gentlemen. It's a pleasure to be here today on behalf of
9 Brand Institute to present to you --

10 MS. JAIN: Patricia, could we just hang on just
11 one second. There has to be a statement that's read first.
12 I apologize.

13 DR. GROSS: Before we begin, I have the
14 pleasure of reading a nice, long paragraph to you.

15 (Laughter.)

16 DR. GROSS: Both the Food and Drug
17 Administration and the public believe in a transparent
18 process for information-gathering and decision-making. To
19 ensure such transparency at the open public hearing session
20 of this advisory committee meeting, FDA believes that it is
21 important to understand the context of an individual's
22 presentation. For this reason, the FDA encourages you, the
23 open public hearing speakers, at the beginning of your
24 written or oral statement to advise the committee of any
25 financial relationship that you may have with any company

1 or any group that is likely to be impacted by the topic of
2 this meeting.

3 For example, the financial information may
4 include a company's or a group's payment of your travel,
5 lodging, or other expenses in connection with your
6 attendance at the meeting.

7 Likewise, FDA encourages you, at the beginning
8 of our statement, to advise the committee if you do not
9 have any such financial relationships.

10 If you choose not to address this issue of
11 financial relationships at the beginning of your statement,
12 it will not preclude you from speaking.

13 So the first speaker is Patricia Staub.

14 MS. STAUB: Good afternoon, ladies and
15 gentlemen, once again. It is a pleasure to be here today
16 on behalf of Brand Institute to present to you several key
17 issues and recommendations with respect to minimizing the
18 risk of confusion caused by look-alike and sound-alike
19 proprietary names for branded prescription drug products.

20 By way of introduction, I am a licensed
21 pharmacist and attorney and a former FDA employee. I am
22 currently employed as Vice President of Regulatory Affairs
23 for Brand Institute. Brand Institute is a well-known and
24 experienced international brand development company that
25 routinely conducts name confusion studies and makes risk

1 assessments in the process of developing proprietary names
2 for prescription drug products.

3 During the past five years, Brand Institute has
4 participated in the brand name development of nearly half
5 of all the prescription drug brand names approved for use
6 in the United States.

7 On behalf of both Jim Detorre of Brand
8 Institute, the CEO, and myself, I thank you for inviting us
9 here today to share with you our own best practices and
10 recommendations relative to the brand name selection
11 process. If there is time at the end of my talk, I'd also
12 like to briefly address the five questions before the
13 committee and give you our opinion on these five questions.

14 Recognition and memorability: benefits versus
15 reality. The hallmark of a successful proprietary name is
16 high brand recognition and memorability. Easily
17 recognizable and memorable names may, indeed, sell more
18 product, but strong brand names are also safer names, ones
19 that are less likely to be inadvertently confused with
20 other drugs. Therefore, we all struggle to provide safer
21 brand names that benefit both prescriber and patient by
22 decreasing the risk of medication errors associated with
23 look-alike and sound-alike names. This is no small
24 challenge today with over 17,000 brand and generic names
25 approved in the United States alone, and only 26 letters in

1 the English alphabet. Given these statistics, some
2 similarity between drug names cannot be avoided. Our
3 objective then is to avoid confusing similarities between
4 brand names.

5 When brand names are found to be likely to
6 cause confusion, one way to manage the risk of medication
7 errors is to increase a brand's recognition and
8 memorability. Some of the newer methods may involve
9 promotional campaigns around drug names after they're on
10 the market.

11 Risk management techniques. Pre-approval
12 methods of managing the risk of medication errors due to
13 brand name confusion have surfaced in the relatively recent
14 past. Regulators in the wake of the 1999 Institute of
15 Medicine report, *To Err is Human*, have increasingly sought
16 to shift the burden of risk management for brand name
17 confusion to industry.

18 Today when a pharmaceutical company proposes a
19 brand name for their soon-to-be-approved drug, the agency,
20 through DMETS, will review that name for safety. The
21 results of prescription interpretation studies which assess
22 the risk of brand name confusion and the potential for
23 patient harm have become part of industry's routine
24 activities in bringing a brand name to market. Also during
25 the pre-approval period, sponsors have started airing

1 "coming soon" ads to get the name out to the public,
2 thereby increasing recognition and memorability of new drug
3 names.

4 Proactive post-approval risk management
5 activities can be particularly useful in that initial
6 period immediately after a drug's approval when prescribers
7 may be unaware of the new drug name and the risk of
8 medication error can be high. Reminder ads as part of a
9 strong launch and targeted advertising are also employed to
10 increase name recognition. When name recognition fails and
11 confusion occurs, Dear Doctor letters informing physicians
12 of the confusion of names, the use of tall man letters to
13 accentuate differences in product names already on the
14 market can be helpful. Name withdrawal should be a last
15 resort.

16 With these thoughts in mind, we would now like
17 to share with the agency and the committee some of our own
18 best branding practices developed through our experience
19 and research at Brand Institute. We will then end with a
20 few specific recommendations that we suggest to improve the
21 regulatory review process for brand safety.

22 Best practices: multi-factorial real-world
23 approach. While generating safety signals through a
24 retrospective review of past errors can be helpful, we
25 suggest that there is no substitute for using a multi-

1 factorial approach to generate potential safety signals
2 associated with the introduction of a new proposed
3 prescription drug name. We believe that real-world testing
4 among a large sample size of currently practicing health
5 care practitioners is critical in addition to testing
6 through orthographic and phonetic analysis, expert focus
7 group review, impact review, and computer-aided research.
8 Very often in doing this extensive testing, we do uncover
9 strong signals in one category or another that causes us to
10 reject a brand name candidate before it is submitted to the
11 FDA. Our premise that this combination approach offers the
12 most comprehensive and reliable methodology for confusion
13 testing among brand names appears to be supported by our
14 relative lack of confusion over the past couple of years
15 when you compare the names that we've generated to the USP
16 drug list.

17 Although differences of opinion regarding the
18 results can still exist between regulators and sponsors,
19 even when extensive testing has been completed, the
20 inherent value of this testing is that awareness of risk is
21 identified and monitored. And risk management strategies
22 may be employed by the sponsors and the agency either prior
23 to marketing or as a condition of marketing their product
24 under their preferred brand name. Once a potential risk is
25 identified, it can be qualified and hopefully minimized.

1 Lessons learned from AERS. A retrospective
2 analysis of all reported mortality-associated medication
3 errors contained in the AERS database during a 5-year
4 period ending in 2001 was published on the CDER website.
5 Jerry Phillips' group was the author of this study which
6 looked solely at fatal medication errors, the most serious
7 consequences.

8 It is interesting to note that the confusion
9 rates of brand names were similar to the confusion rates of
10 generic names, that more written miscommunications rather
11 than oral miscommunications resulted in fatal errors, that
12 elderly patients over 60 years old in hospital settings
13 receiving injectable drugs for CNS, oncology, and
14 cardiovascular conditions were more frequent victims of
15 fatal medication errors. Most patients that died were
16 taking only one medication according to the study. These
17 potentially predisposing factors should be considered
18 possibly when assessing brand name risk: patients again
19 over the age of 60 in hospital settings receiving
20 injectable drugs and particularly patients taking
21 therapeutic categories of CNS, oncology, and CV.

22 10 percent of these medication errors were
23 fatal, of the 5,366 that were measured, and the most common
24 error was an improper dose, 40.9 percent. The wrong drug
25 was 16 percent of the time, and wrong route of

1 administration, 9.5 percent of the time. Proprietary name
2 confusion resulted in 4.8 percent of the medication errors,
3 and nonproprietary name confusion resulted in 4.1 percent
4 of the fatal medication errors. 6.7 percent were due to
5 written miscommunications and 1.7 percent of the fatal
6 errors were due to oral miscommunications. 48.6 percent of
7 the deaths occurred in patients over 60 years of age, and
8 the largest number of deaths, 26.7 percent, occurred in the
9 practice setting of a hospital. The most common dosage
10 form again in death due to medication orders was
11 injectables, 49.9 percent.

12 Benchmarking. Benchmarking is a topic where we
13 have a lot of questions from our clients. We believe that
14 benchmarking error rates in confusion studies, while
15 relevant, can also be misleading without a separate
16 evaluation of the impact on patient harm. For example,
17 even high error percentages based on potential name
18 confusion with another drug whose misadministration would
19 likely result in little or no patient harm may not be as
20 meaningful as a much smaller error rate percentage that
21 would likely result in high patient harm, for instance,
22 mistaking a diuretic for an oncology product.

23 Benchmarking, combined with impact analysis, is
24 a more useful tool for assessing risk.

25 Another misleading aspect of over-reliance on

1 benchmarking can be the fact that a certain number of
2 errors in confusion testing may be the result of
3 misspelling the new name rather than confusing the new name
4 with another drug. Misspellings alone may be harmless.

5 Overlapping characteristics. Brand name
6 similarity cannot likely be completely eliminated due to
7 the large number of approved brand names in the United
8 States. Similar or overlapping characteristics, however,
9 in combination with a similar brand name, can be important
10 additional causes of confusion, and these characteristics
11 should also be evaluated in brand name confusion studies.
12 For example, similar packaging, labeling, route of
13 administration, dosage form, concentration, strength,
14 patient settings, storage conditions, and frequency of dose
15 may make a difference between a similar brand name and a
16 confusingly similar brand name. In our brand confusion
17 studies, we prepare a chart that looks at overlapping
18 characteristics between similar sounding and looking names
19 as a factor in making our risk assessment for name
20 confusion.

21 I guess they're going to exclude modifiers from
22 this setting. So all I will say about that is that with
23 the general policy that the agency has that only one brand
24 name per product per sponsor will be approved, brand name
25 modifiers are the only way that a manufacturer can use to

1 further define new formulations of their product. Of
2 course, there are problems with modifiers that are well
3 known. In Europe prefix modifiers are sometimes used and
4 because of our international business, sometimes clients
5 would like to have prefix modifiers. This can really
6 create problems I think in the United States, and I'm glad
7 that we don't have a problem with people suggesting prefix
8 modifiers here.

9 The suffix modifiers everyone knows are
10 problems due to the fact that XL and SR have a variety of
11 meanings, depending on the drug product that you have. In
12 Europe if a drug modifier or suffix modifier doesn't have
13 the same meaning in each of the member countries, it's not
14 allowed.

15 Particularly the suffix XL I think, should be
16 noted, can be confusing with the quantity of 40 tablets,
17 since that's the Roman numeral. There are several two-
18 letter suffixes that are problematic. One-letter suffixes
19 are not allowed in not allowed in Europe, and I think that
20 they're fairly rare in the United States too. That's
21 probably a good thing because modifier drop-off is probably
22 more prone with the one-letter modifier.

23 On the subject of numerical branding, numerical
24 branding is using numbers in a single entity brand name,
25 and we highly discourage this in general since the name can

1 be confused with the strength or dosage. For instance,
2 valium-5 can look like take 5 tablets of valium and can
3 result in medication overdose. Numerical branding for
4 combination products, however, can minimize confusion and
5 improve safety in some cases but only if both ingredients
6 are listed numerically. For example, referring to Percocet
7 5, oxycodone 5 milligrams/acetaminophen 325 milligrams, by
8 only it's oxycodone number 5 can lead to the administration
9 of 5 tablets of Percocet and cause fatal patient harm.
10 However, referring to Percocet without the number 5 or only
11 using the number 5 in conjunction with the number 5/325 can
12 make clearer the dose required.

13 Trailing zeros. We agree with ISMP that
14 trailing zeros can cause confusion and that brand names
15 should never be accompanied by dosages with trailing zeros.
16 For instance, 2.50 milligrams can be interpreted as 250
17 milligrams. Leading zeros, however, do improve the absence
18 of confusion and should be always used. 0.25 milligram
19 versus .25 milligram.

20 Tall man letters. The use of capital letters
21 within a generic name to differentiate nonproprietary
22 names, acetaHEXazole and acetaZOLamide, is one risk
23 management technique that could be applied to brand names
24 in the post-marketing setting to differentiate them. This
25 has been done recently with SeroQUEL versus SaraFEM and

1 SerZONE. And that's an example of SeroQUEL's new packaging
2 that accentuates the difference between confusingly similar
3 names.

4 Bar coding, while we recognize its importance,
5 only has limited importance. It minimizes order picking
6 confusion, but does not minimize interpretive confusion.
7 Computerized order entry may minimize illegible handwriting
8 from prescribers, but it also may introduce its own set of
9 errors in picking a drug from the list. Electronic
10 solutions to these problems are not totally error-free.

11 Orthographic analysis, looking at strings of
12 letters, are instructive, but this method alone does not
13 adequately address confusion. Orthographic analysis may be
14 more helpful in real-world, handwritten prescriptions as it
15 can show the formation of certain letters may decline in
16 somewhat predictable ways such as an M bleeding into an N.

17 We also agree with DMETS that beginning drug
18 names with the letter Z can be problematic in that Z, when
19 scripted, may look like C, L, B, 2, g, y, j, or q, and
20 might sound like C, S, and X.

21 We have three recommendations for the process
22 of naming that we would like to make.

23 The first suggestion that we have -- and this
24 is really a result of some of the problems that we've
25 experienced with our clients during the process -- is that

1 tentatively approved names be made public, when they are
2 tentatively approved, via the internet so that successive
3 name candidates can test their own proposed proprietary
4 names against names that have already been tentatively
5 approved, but could potentially beat them to the market.
6 Confusion testing is only as good as the universe of names
7 that the proposed name can be tested against.

8 The second suggestion we have is that whatever
9 testing models DMETS uses from time to time, that those
10 testing methods be made transparent so that comparison
11 between the two models can be made and parallel testing of
12 names could possibly improve the accuracy of both models,
13 both the proprietary model that was being submitted to the
14 FDA and the FDA's own model that it's testing.

15 A third issue that we would like to suggest is
16 duplicate brand name exception for drugs where the brand
17 name is already widely associated with the treatment of
18 mental illness and stigma has been proven and a second drug
19 name possibly should be allowed for that compound where
20 there is a physical illness. Wellbutrin versus Zyban and
21 Prozac versus Serafem are two examples of this type of
22 exception to the normal rule of one brand name per drug per
23 sponsor. We believe that if stigma can be proven, patient
24 harm can be alleviated that may be caused by embarrassment
25 for taking a well-known mental health drug for a physical

1 condition, particularly where employer-paid prescriptions
2 are available.

3 In conclusion, there are many opportunities
4 during the name development process to safeguard against
5 medication errors caused by look-alike and sound-alike
6 proprietary names. High recognition and memorability are
7 key components of safe drug names. While post-marketing
8 risk management programs are useful, pre-marketing
9 activities are increasingly being used to anticipate and
10 identify risks before harm occurs.

11 Although predicting risk is not an exact
12 science, neither is medicine. Human error is a predictable
13 constant in any health care system. No medication error
14 prevention technology is itself error-free. A multi-
15 factorial, real-world approach to names testing to
16 prospectively identify levels of risk associated with new
17 drug names during the approval process is key.

18 We applaud the efforts of the agency in taking
19 up this difficult challenge to patient safety by creating
20 the DMETS layer of brand name review and attempting to
21 establish patterns by retrospective analysis of the AERS
22 database. While differences of opinion may still exist
23 between regulators and sponsors as to levels of acceptable
24 risks associated with a drug name, we do not see any
25 realistic substitute for comprehensive name testing in the

1 real world to assess the risk of confusion between new and
2 existing drug names. After all, the prediction of risk is
3 always based on probability and is never absolute. Real-
4 world testing allows us to observe risks that have already
5 been seen rather than to speculate on risks that may occur.

6 Thank you.

7 DR. GROSS: Thank you.

8 There are four more presenters. We would like
9 to finish these remarks by 2 o'clock. So I would ask the
10 other presenters if they could condense their presentation
11 a little bit.

12 The next speaker is Dr. Douglas Bierer from
13 Consumer Healthcare Products Association. He's Vice
14 President of Regulatory and Scientific Affairs. Thank you.

15 DR. BIERER: Thank you. Good afternoon and
16 thank you for the opportunity to present an OTC perspective
17 on sound-alike/look-alike drug names. While OTC products
18 are not the subject of this panel's conversation today, it
19 would be important to mention some comments about OTC drugs
20 since they were mentioned briefly in this morning's
21 presentations.

22 The Consumer Healthcare Products Association,
23 which was founded in 1881, is a national trade association
24 that represents the manufacturers and distributors of over-
25 the-counter drug products, and our members account for more

1 than 90 percent of the OTC products that are sold at retail
2 in the U.S. CHPA has a long working history with the FDA
3 to improve OTC labeling so that these labels are easier for
4 the consumer to both read and understand.

5 In considering the issue of drug names for OTC
6 products, it is important to stress several key differences
7 that arise from both prescription and OTC drugs. One of
8 the most important differences is how the drugs are
9 purchased. Prescription drugs are made available by
10 written or verbal order by a physician or a licensed
11 practitioner, which then, in turn, needs to be translated
12 and filled by a pharmacist.

13 OTC drugs, on the other hand, are purchased
14 directly by the consumer. Thus the OTC product package
15 must communicate all of the information the consumer needs
16 to decide if it is the right product for them. When
17 purchasing an OTC medicine, the first thing the consumer
18 sees on the store shelf is the product's principal display
19 panel.

20 As shown in this slide, in addition to the
21 brand name, the principal display panel includes other
22 important information to help consumers identify if it is
23 the appropriate product for the condition that they want to
24 treat.

25 First is a statement of identity. This

1 includes the established name, that is, the official name
2 of the drug and the general pharmacological category or its
3 intended action of the drug. It is written in layman's
4 language and must be prominent and conspicuous on the
5 package. And for those products which are combinations of
6 active ingredients, there must be a statement about the
7 principal intended action of each of the active
8 ingredients. All these elements are required on OTC
9 packages.

10 Often the principal display panel contains
11 other information such as the dose of the active ingredient
12 and perhaps a statement about a product's benefits, such as
13 it relieves or treats a certain type of ailment.

14 In addition, it may contain a flag in the upper
15 corner to alert consumers of important new information.
16 This flag was a voluntary program first initiated by CHPA
17 in 1977 to provide consumers with more information when
18 they were purchasing OTC drug products. In this case the
19 flag says "new," indicating that this is a new product. It
20 may also say "see new labeling" or "see new warning" to
21 indicate that a change has been made to the product
22 labeling on the back of the package.

23 All of this information is clearly visible at
24 the point of purchase and helps the consumer to decide if
25 this is the right product for them.

1 The next major difference is the drug facts
2 labeling. By May 2005, all OTC medications will be
3 required to use this format, and in fact, many OTC products
4 are already using them on the store shelves. Drug facts
5 standardize all the labeling on the back of the package to
6 make it easier for the consumer to read and follow the
7 label. The information appears in very clear, concise
8 consumer language. As shown on this example of a
9 chlorpheniramine product, the drugs facts includes the
10 active ingredient of the product, including the quantity of
11 each active ingredient per unit dose, the purpose of the
12 active ingredient, what the product is to be used for, any
13 warnings about the use of the product which are grouped in
14 headers to facilitate the consumer finding the information
15 and understanding the information.

16 Next, the directions, which is important to
17 mention that the directions appear after the warning signs
18 in an OTC package.

19 Finally, other information such as storage
20 conditions, and finally a list of inactive ingredients
21 listed in alphabetical order so the consumer can know what
22 is in the product that they're going to be taking.

23 Because this information is organized in
24 exactly the same way on every OTC product, this format
25 makes it easier for the consumer to find all the

1 information they need to take the product correctly and
2 safely and also when to contact a physician. It is also
3 important to note there is redundancy of the information in
4 drug facts and on the front panel, and this serves to
5 reinforce the information sent to the consumer.

6 At the 26th June meeting on drug naming and
7 also at this meeting, the agency expressed concern about
8 OTC brand name extensions in which a family of products may
9 have a similar name and may be used for different
10 conditions and may contain different active ingredients.
11 OTC brand names allow consumers to locate a family of
12 products which they have used before and that they trust.
13 OTC manufacturers confine the family of products to
14 particular therapeutic areas in order to decrease the
15 concern that consumers may take a product for one condition
16 when it really should be used for another condition.

17 It has also been suggested that brand trade
18 name extensions should not be used and that each extension
19 should have a differently named product. However, this
20 approach has potential to create more consumer confusion
21 because the consumer will be required to master separate
22 information and brand names for each product. As these
23 products are advertised in the media, the plethora of
24 different products will create confusion and make it even
25 more difficult for consumers to remember what the product

1 is to be used for and for what conditions. Brand names and
2 their line extensions do provide consumers with valuable
3 information about the products that they have used before
4 and that they have come to trust.

5 As I have illustrated, the consumer has much
6 more information than just the brand name to recognize when
7 selecting an OTC product. The uniqueness of the amount and
8 the redundancy of the information on the OTC label, when
9 compared to handwritten or oral prescriptions and
10 prescription product packages themselves, decreases the
11 reliance on the brand name and aids the consumer in making
12 the right choice about the product for the condition that
13 they want to treat.

14 Thank you for considering the views of the OTC
15 drug industry.

16 DR. GROSS: Thank you very much.

17 The next speaker is Clement Galluccio from
18 rxmarx, a division of Interbrand Wood Healthcare.

19 MR. GALLUCCIO: No slides for me today. Just I
20 guess the burden of having been involved in the validation
21 of proposed pharmaceutical trademarks for close to 15
22 years. I guess that's in opposition of being unburdened of
23 no practical experience.

24 In 1991, Interbrand Wood Healthcare and rxmarx
25 introduced the 10/10 trademark evaluation model to

1 immediate acceptance from many of the world's leading
2 pharmaceutical companies. Of the many innovations
3 introduced with the 10/10 model, paramount was the concept
4 that trademark selection was more complex than the
5 exclusive consideration of prescriber preference, but also
6 reflected the desire to select a safe name. To date, over
7 80 trademarks have been first 10/10 certified prior to
8 agency submission and subsequently introduced to the
9 marketplace, with many more presently waiting introduction.

10 To the best of our knowledge, less than 2
11 percent of trademarks validated using the 10/10 model have
12 encountered any degree of concern relative to medication
13 error. These 80 trademarks are representative of over 700
14 name validation studies, consisting of thousands of
15 proposed pharmaceutical trademarks.

16 Given the significant role that Interbrand Wood
17 Healthcare and rxmarx have served in creating and
18 validating pharmaceutical trademarks, there have been many
19 important lessons that we have learned in regard to the
20 identification of names at risk of medication error. The
21 one that we most often share with our clients in regard to
22 the certainty of our findings is the following. Regardless
23 of the methodology used to validate a pharmaceutical
24 trademark, each and every name has the potential to be
25 communicated so poorly by the prescriber or transcriber

1 that it could be potentially mistaken for another product
2 name.

3 Therefore, it stands to reason that unless
4 significant changes are made to how pharmaceutical products
5 are packaged, distributed, stored, and communicated within
6 the dispensing environment, independent of changes to
7 validate nomenclature, medication error will continue to be
8 a harsh reality for all concerned. Minimizing medication
9 error, not finding alternate methodologies to validate
10 proposed pharmaceutical trademarks, should be the primary
11 focus of the discussion. That said, it is the opinion and
12 recommendation of Interbrand Wood Healthcare and rxmarx
13 that both industry and agency should strongly consider the
14 following.

15 Grant equal time and consideration to the
16 factors other than trademark similarity that may also
17 contribute to medication error. As David Wood, CEO of
18 Interbrand Wood Healthcare, shared on June 26th, let's not
19 make trademarks the whipping boy for a system which needs
20 to pay attention to the many other things other than the
21 brand name. A good start would be to begin validating
22 nonproprietary names for safety using the same best
23 practices that have been developed for proprietary names,
24 followed by paying much closer attention to labeling,
25 packaging, and administration practices. Perhaps the

1 answer to minimizing medication error exists in creating
2 greater personalization, differentiation and security in
3 product labeling, packaging, and delivery systems as
4 opposed to creating increasingly more restrictive barriers
5 to proposed pharmaceutical trademarks.

6 Two, fund a study to provide an accounting of
7 previously identified nomenclature associated with
8 medication error over the past 10 years, as well as
9 determine present nomenclature assessment practices by
10 sponsors. We believe there exists a significant absence of
11 data relative to the actual as opposed to the perceived
12 causes of medication error. The anticipated outcome would
13 be to better understand which factors, for example, brand
14 name versus generic name, the lack of adequate legal or
15 research assessment prior to introduction, overlap of
16 dispensing profile and other dispensing factors and
17 practices, et cetera, that may have significantly
18 influenced medication error.

19 Three, in recognition of the many companies
20 within industry that have already implemented best
21 practices relative to nomenclature validation, provide
22 flexibility within whatever guidance, whatever outcome to
23 follow to allow such companies to continue in their present
24 approach until new methodologies are validated. In our
25 view the best practices for the validation of proposed

1 pharmaceutical nomenclature already exist, however, need to
2 be applied on a consistent basis by each and every sponsor.
3 In turn, agency should provide a predefined set of
4 consistent metrics relative to approval or rejection so
5 that the outcome of nomenclature validation studies is
6 predictable, for example, a proposed name misinterpreted
7 more than once for the same potential conflict is
8 automatically determined to be of high risk or higher risk.

9 High-risk candidates would then be considered for more in-
10 depth analysis, perhaps quantitative analysis or monitoring
11 programs post-launch.

12 In conclusion, we believe an inclusive approach
13 is paramount in order to provide the desired benefit to the
14 public in regard to minimizing medication error. We
15 applaud today's participants for their efforts and agree
16 that the development and selection of a pharmaceutical
17 trademark should reflect best practices relative to the
18 identification of a safe trademark. However, recent
19 advances such as the increasing use of computer-assisted
20 prescribing and dispensing tools is only one initiative
21 that supports a more comprehensive approach. These
22 advances, when combined with many of the existing best
23 practices relative to nomenclature validation, as reflected
24 in present methodologies and the recommendations I shared
25 earlier, represent the most logical resolution to

1 minimizing medication error.

2 In conclusion, beyond our statement, we have
3 released our methodologies, both proprietary and
4 nonproprietary, to the committee so we can open-source
5 these methodologies for use by all.

6 Thank you.

7 DR. GROSS: Thank you, Mr. Galluccio.

8 The next speaker is Maury Tepper III from
9 Womble, Carlyle, Sandridge & Rice.

10 MR. TEPPER: Thank you and I'll start with what
11 is a customary gesture for me: adjustment of the
12 microphone.

13 I welcome the chance to be here with you today,
14 and I do want to mention just a couple of quick things by
15 way of introduction for you. I do share one thing in
16 common with you members of the advisory committee. I am a
17 special government employee as well for the Department of
18 Commerce. I serve on the Trademark Public Advisory
19 committee for the U.S. Patent and Trademark Office. My
20 comments today will not relate to the Patent and Trademark
21 Office or its operations, but I did want to make you aware
22 of that.

23 I also, very importantly, want to note that I'm
24 pleased to see that the ACC is well represented here. As a
25 resident of North Carolina, I'm glad to see participation

1 from others who may also be traveling back to our State
2 under a weather advisory today.

3 (Laughter.)

4 MR. TEPPER: I come to you I think bringing
5 good news and hopefully some recommendations. And let me
6 just step back as one who has previously served as in-house
7 trademark counsel for a pharmaceutical company -- and
8 currently I'm in private practice representing all types of
9 clients, some in the pharmaceutical industry, some in
10 industries such as snack foods, candies, and racing
11 memorabilia -- and tell you that I think the good news here
12 is everybody in this room shares a common interest and
13 common goal. That is not always the case, but hopefully it
14 has come through today. If it hasn't, I really want to
15 emphasize I think both the FDA and sponsors are working
16 very hard here, striving to do everything that can be done
17 to find ways to minimize medication errors, to bring out
18 the safest possible products, including their trademarks.

19 I think where we may differ is in determining
20 how best to go about that and the degree to which trademark
21 analysis contributes significantly to the problem or indeed
22 may be the best solution. And I'll talk about that a bit
23 in my remarks. But I think it's important to keep in mind
24 and to understand here that at the end of the day, we're
25 all seeking the very same thing. So I think the efforts in

1 this room are laudable. I think the fact that we're all
2 working towards the same goal is encouraging and should
3 mean that we can arrive at a very workable system or
4 continue to refine that. I hope that this will be the
5 lead-in to an open dialogue.

6 It is important I think to note in looking at
7 this problem -- and I was very pleased to see some of the
8 good questions this morning -- that a lot of the
9 presentations, a lot of the data presented today start from
10 an assumption that trademarks contribute substantially to
11 medication errors. I think we would all agree that they
12 are involved and that they are a factor, but I do have to
13 reemphasize I'm not aware of any study or any way that we
14 have come about determining what a significant factor they
15 are or what their role is, if they cause the error. The
16 fact that two name pairs are similar certainly doesn't
17 automatically mean in every case that is a significant
18 contributing cause to the error.

19 I was very taken by Dr. Dorr's research this
20 morning in her presentation. For a dumb lawyer like me, it
21 was the closest I've come to understanding some of that
22 science, but it leapt out at me that in listing for you a
23 degree of name pairs that had high similarity rankings,
24 some of them were involved in errors, some of them weren't.
25 That tells us that similarity alone is not the decisive

1 factor. It is not in all cases going to tell us
2 automatically is this a problem. It is relevant. It is
3 absolutely something we need to consider, but I would
4 submit that it is simply one of many factors that need a
5 balanced approach in making a determination about the
6 safety of a name in its appropriate setting and context.

7 The other thing I think we need to be mindful
8 of -- I liked Mr. Woods' characterization that was just
9 quoted of not making trademarks the whipping boy for other
10 parts of the system -- is to be thinking about where we can
11 have the most significant impact on this problem.

12 You were shown this morning Avandia and
13 Coumadin as two names that are somewhat similar. Of
14 course, the only similarity there is in handwriting, and I
15 do have to ask the question, if we were coming to the point
16 where we're looking at trademarks as the part of the system
17 to make up for sloppy handwriting, are we really getting at
18 the problem in the best way? Are we going to have the
19 maximum on patient safety by trying to do that? That's not
20 to say we are not going to continue to strive to predict
21 and identify and address these issues and create safe
22 marks, but I think it is important to keep in mind that
23 there are probably other more significant causes that we
24 should be focused on and should be addressing as part of
25 this effort beyond trademark review, simply because

1 trademarks are prominent and are identified in each
2 situation.

3 The other thing I think is important to realize
4 here -- and this is a scientific group. Again, as a dumb
5 lawyer addressing you, I need to be careful, but at the end
6 of the day, these are subjective determinations. We would
7 love to have a validated test. We would love to have an
8 objective measure that would tell us all whether or not we
9 are going to have problems given a particular trademark. I
10 have to tell you I simply do not believe that can happen.
11 There are too many factors involved in each situation, in
12 each setting, in each combination of drugs that come into
13 play that need to be considered and need to be carefully
14 weighed and need to be looked at to allow us to simply come
15 up with a formula or any one approach that will give us
16 some prediction of error propensity.

17 All of the techniques here that have been
18 discussed this morning I think provide very useful data,
19 but it's important to keep in mind that that's all they
20 provide. They are sources of data. I don't think we have
21 any one outcome predictor here. I applaud the efforts to
22 continue to seek one, but I want to be careful here to
23 indicate that we should best view these as inputs right
24 now.

25 Another piece of good news for you I think is

1 to note -- and the question was asked -- you'll be getting
2 some additional information about this, but just the degree
3 to which trademarks are carefully screened and reviewed by
4 both pharmaceutical companies and by the FDA. I can tell
5 you as someone who works for clients in lots of industries,
6 there is no industry that even comes close to the
7 pharmaceutical industry in the care that it gives in the
8 selection and consideration of trademarks. I get lots of
9 calls from clients that are launching products next week.
10 Thankfully, those tend to be snack cakes rather than drugs.

11 Drug names are typically given very careful consideration.
12 You'll hear more, and I think you heard from Bob Lee
13 already this morning about the types of testing. But I
14 think if you really break it down and look, the types of
15 testing that FDA and that sponsors are engaging in really
16 have a lot in common. In many ways they approximate one
17 another.

18 Where I think there is a significant difference
19 is in what is being done with that data. I would propose
20 -- and my paper goes into this in some more detail that one
21 thing we need is a framework for making decisions. All of
22 these resources we've talked about this morning are best
23 viewed as providing relevant data to you, but we need some
24 framework for analyzing that data. The trademark system
25 provides that.

1 I'll apologize for anyone who heard me on June
2 26th if I sound like a broken record. This is in some ways
3 echoes my comments at that point. If anything, the
4 outcomes of that meeting solidified that belief that given
5 all of this data, that we cannot validate it and we need to
6 decide what place it should have in each situation.

7 The best test is one that can carefully look at
8 and approximate market conditions, and that is precisely
9 what the legal test for trademark availability is designed
10 to do. The likelihood of confusion test that is employed
11 by attorneys, that is employed by the Patent and Trademark
12 Office in reviewing proposed trademarks, that is employed
13 by courts in determining disputes and whether there are
14 actual conflicts is a test that is well established, well
15 defined, and yes, it is subjective, but it is a well
16 understood language for having this discussion and for
17 analyzing and balancing these factors in each situation.

18 What makes pharmaceuticals special? Certainly
19 this is a very different market than the consumer
20 marketplace. In some ways it's frightening that the
21 average consumer may go out and pay more attention and be
22 more involved in selecting their laundry detergent than in
23 receiving a medication where they in many ways turn it over
24 to the providers and the dispensers and take whatever is
25 handed to them in blind trust. We need to understand and

1 take those market conditions into account that the same
2 test provides the ability to balance those factors, to use
3 this input data about similarity orthographically and in
4 handwriting and in sound, and to consider them in a
5 framework that provides us something of a useful and
6 predictable result that gives us the basis for analyzing
7 these and for balancing the numerous factors that come into
8 play rather than seeking to emphasize one single measure.

9 I do want to come back to the important notion,
10 though, that as we are engaging in these efforts, I think
11 that the FDA has done a laudable job in bringing focus to
12 bear on the science available here and helping refine and
13 establish some of these techniques and seeing how they're
14 put to use. I think part of where we perhaps differ is
15 once that data is generated, how is a decision arrived at.
16 Attorneys are used to using a defined and documented and,
17 I'll say, reproducible test to sort of have that discussion
18 and make the analysis. FDA is looking at the same data and
19 coming to conclusions. I think anytime you're dealing in a
20 subjective area, that's natural and understandable. You
21 heard Dr. Phillips I think this morning acknowledge
22 sometimes when they have concerns, they turn out to be
23 borne out in the marketplace, sometimes they don't.

24 Again, I wish we could give you an objective
25 measure that's going to be a crystal ball for us, but I

1 think what we need to strive for is to make sure that good
2 naming practices are followed, to make sure that these
3 techniques have been employed and have been considered, and
4 then recognizing that these are subjective judgments, to
5 really carefully consider whether substituting the FDA's
6 judgment for that of a sponsor is going to substantially
7 increase or improve patient safety.

8 In many ways I submit that there are times when
9 you may increase risk by causing a sponsor close to launch
10 to have to go back and change a trademark. Typically
11 trademark reviews -- and again, I'll echo Bob Lee's
12 comments this morning -- occur at multiple stages.
13 Certainly during the creation, the sponsors are generating
14 these names and screening them internally. They're
15 conducting an analysis. They're seeking input from
16 appropriate experts. When the application is filed, the
17 trademark is again reviewed by an examiner at the Patent
18 and Trademark Office who is employing the same likelihood
19 of confusion standard. Indeed, the Patent and Trademark
20 Office and courts have both recognized a higher degree of
21 care for pharmaceutical trademarks given the significance
22 of similarity here.

23 Finally, the opposition period comes up and
24 that's when competitors also conduct the same review, step
25 in and oppose the mark if they feel there's a potential for

1 conflict or the mark is too close.

2 This process takes several years to complete,
3 and so by the time a trademark application is filed or has
4 been screened and filed, has been subject to an opposition,
5 we have a lot of eyes over that that have come to some
6 consensus that this mark does not appear likely to cause
7 confusion. To step in and have to change that mark,
8 without the time to go back through that process, in some
9 ways deprives those others of the right to review and
10 comment, forces the sponsor to make some last-minute
11 changes or determinations, and to do their best, of course,
12 in analyzing this. But I submit that we may be increasing
13 risks in some ways by causing these changes close to launch
14 and without the availability of these other reviews and
15 mechanisms and considerations that we typically would want
16 to employ.

17 I will leave my comments there in the interest
18 of your time. I have provided some answers to the
19 questions in the written material to you, but in large
20 part, I think the key answer here is we need to continue to
21 do everything we can to refine the techniques for
22 generating information to consider, but we need to keep in
23 mind that at the end of the day each of these tests can
24 only provide relevant data that we should consider. This
25 will be a subjective determination. There is a well-

1 established test that is used for making that subjective
2 determination. Trademark attorneys have expertise in doing
3 that and they attempt to balance the appropriate factors.
4 I think FDA should continue to play a role in shaping
5 practices that will provide the relevant data, should
6 provide good naming practices, should ensure that industry
7 is taking these into consideration.

8 I think FDA should be very cautious, however,
9 at substituting its subjective judgment based on a standard
10 that we do not know for that that has been arrived through
11 the likelihood of confusion analysis.

12 I also think that we need to continue to do
13 what we can to focus on the overall problem of errors,
14 understand that trademarks are a factor, but also
15 understand that efforts that may have greater impact and
16 greater significance should certainly not be overlooked in
17 the haste to squeeze tighter down on the most visible
18 aspect of the system, and that is the trademark.

19 Thank you.

20 DR. GROSS: Thank you very much.

21 The next speaker is Dr. Suzanne Coffman, who is
22 Product Manager of NDCHealth.

23 DR. COFFMAN: Thank you, Dr. Gross, members of
24 the committee, and the FDA, for the opportunity to appear
25 before you today. You should have a copy of my

1 presentation in your packet.

2 As Dr. Gross mentioned, my name is Suzanne
3 Coffman. I am a pharmacist and I am a product manager for
4 NDCHealth where my responsibilities include clinically
5 based transaction products for the pharmacy market. In the
6 interest of full disclosure, I'm also a shareholder of NDC
7 and they did pay for my travel.

8 I spoke on this topic at the joint
9 ISMP/PhRMA/FDA meeting in June. Today I'll be providing an
10 update and also just expressing NDCHealth's continued
11 interest in the topic of preventing drug name confusion
12 errors.

13 NDCHealth is a leading provider of point-of-
14 sale and information management services that add value and
15 increase the efficiency of pharmacy, pharmaceutical
16 manufacturing, hospital, and physician businesses. Two out
17 of three prescription transactions in the United States
18 travel across our network, and we are connected to 90
19 percent of retail pharmacy outlets in the U.S. We also
20 process transactions in Canada.

21 One of the services that we offer to the retail
22 pharmacy market is real-time alerts about drug name
23 confusion errors. This service is supported by a database
24 that contains all of the known look-alike/sound-alike pairs
25 that involve oral solid products that are used in the

1 retail environment. To that list, we add a likelihood
2 score, a clinical significance score, absolute dosing for
3 each drug dosage form strength that is involved in the pair
4 and also typical dosing for each form strength, and we
5 derive that from the 160 million transactions that travel
6 across our network each month.

7 We send an alert when the dose that is
8 submitted on a prescription is atypical for the drug that
9 is submitted, especially when it's typical for one of the
10 look-alike/sound-alike pairs. This does reduce name
11 confusion. Through our ability to match prescriptions and
12 to look at the follow-up prescriptions, we have identified
13 a number of changes, of course, in quantity and day supply,
14 but we've also identified several changes to the drug.
15 Some of these are known look-alike/sound-alike drugs; many
16 are not. We've had changes, for example, between sartans
17 and between ACE inhibitors which are not on the list.

18 We've also recently completed data collection
19 on a randomized controlled trial in a regional chain, 115
20 stores. Preliminary results show that pharmacy staff,
21 pharmacists' and technicians' knowledge of look-
22 alike/sound-alike pairs did improve after exposure to our
23 real-time alerts. However, even after exposure, they would
24 have only made a C if they were taking a test in pharmacy
25 school.

1 We are currently analyzing the data on the
2 actual error prevention, again using our prescription
3 matching methodology, so that we're able to tell what
4 happened after the pharmacy received our alert.

5 And we also did a survey of the pharmacists'
6 perceptions of the messages that they were receiving, and
7 while the results were admittedly a little bit mixed, they
8 were generally tending towards positive.

9 We have had two new initiatives that have come
10 out of the work that we've done so far with drug name
11 confusion. One is a potential solution for post-marketing
12 surveillance and risk management. In a manner similar to
13 that that we use today for sending alerts in real time to
14 pharmacies with dose-based rules, we could send alerts for
15 an identified pair that is of particular interest or is a
16 particular problem with other types of rules. For example,
17 if there is confusion between an antipsychotic agent and an
18 allergy agent, we could have a rule around prescriber
19 specialties such that if the antipsychotic were prescribed
20 by an allergy immunologist, that would immediately result
21 in an alert, whereas if the allergy drug were prescribed by
22 that same physician, there would be no alert.

23 We can also design a method whereby we can send
24 messages randomly. It would completely overwhelm a
25 pharmacy if you sent an alert on every single prescription

1 for a frequently prescribed drug, but if we can randomly
2 select the prescriptions for that drug that we send
3 messages on, we can still be getting the message out there
4 without having the pharmacist ignore all the messages
5 because they expect to get one.

6 Retail pharmacies are interested in this
7 service because they benefit by having errors prevented,
8 but they're more interested if they don't have to pay for
9 it.

10 Also, on the pre-marketing side, we have
11 designed a method by which we can test proposed drug names
12 in tens of thousands -- well, at least thousands. I don't
13 know about tens of -- pharmacies based on the fact that we
14 are connected to 90 percent of pharmacies in the U.S. It's
15 a real pharmacy, so you'll be testing the name in an actual
16 practice environment. You'll be testing it in context with
17 proposed strengths, and there would even be the possibility
18 to try multiple proposed strengths to test the likelihood
19 of confusion in conjunction with the strength.

20 In many ways it's similar to the methods that
21 Drs. Schell and Hennessy were proposing. I believe that
22 ours could be a little bit lower cost because it's almost
23 completely automated. There is one safety issue that we
24 don't have. We would not propose putting actual bottles of
25 a fake drug or a placebo on the shelf. We think that the

1 pharmacist just by seeing the prescription and whether or
2 not they can interpret would be enough.

3 We, of course, would send out prescriptions
4 from fake physicians and fake prescribers and follow up on
5 every single one. So we think there's absolutely no chance
6 that a prescription could be filled on a real patient and
7 take the wrong drug.

8 And we can compare the results to baseline. We
9 would perform a baseline analysis so you could compare the
10 percent that were cleanly caught and identified as a
11 nonexistent drug, the percent that require clarification,
12 and the percent that are actually interpreted as an
13 existing drug, and compare those to baseline. Of course,
14 in the case where a clarification is required or whether
15 they interpret it incorrectly, we'd be able to tell what
16 exactly it was confused with.

17 Again, retail pharmacies are interested in
18 participating in this, and they actually see the Hawthorne
19 effect as being a good thing, even though it would be a
20 confounding variable from the name confusion detection
21 side, because their perception is that if the pharmacies
22 know they're being monitored, they are more likely to have
23 better performance at all times, which is beneficial to the
24 pharmacy.

25 And in reality it would only take three to four

1 metropolitan statistical areas or the three to four largest
2 chains, and you've got 10,000 pharmacies right there. So
3 it's not an unachievable number.

4 Of course, the next frontier -- that only
5 covers retail pharmacy -- would be hospital and then
6 electronic prescribing. I think there are possibilities
7 for electronic prescribing, for prescription writing
8 systems. I haven't come up with a solution there yet. And
9 one of the issues there is that the physician initiates the
10 prescription, so there's not anything to react to. So I'm
11 still working on that one.

12 Thank you for your time.

13 DR. GROSS: Thank you very much.

14 Last but not least, Dr. Bruce Lambert from the
15 University of Illinois College of Pharmacy in Chicago. Dr.
16 Lambert.

17 DR. LAMBERT: I thought that you had forgotten
18 about me.

19 Thank you for the opportunity to address the
20 committee. Because I only have a short period of time and
21 because I addressed many of these same issues in my public
22 comments during the June 26th meeting, I'd like to direct
23 the committee's attention to my previous testimony and
24 PowerPoint presentation, both of which are available on the
25 FDA website or from me directly or in your briefing

1 materials.

2 In addition, I've submitted to the committee
3 reprints of several peer-reviewed articles published by my
4 colleagues and me during the past seven years or so.
5 Although it's not possible to summary the main findings of
6 those articles in the time allotted, each article presents
7 evidence that's directly relevant to the questions being
8 debated today. In fact, they are, to the best of my
9 knowledge, the only peer-reviewed studies that provide
10 evidence as to the validity of computer-based methods for
11 drug name screening.

12 In fact, many of the questions and issues that
13 have come up today have led to the conclusion that we just
14 don't know about X. And in many of those cases, I was
15 shaking my head because the X that we presumably just don't
16 know about was often described in one of these peer-
17 reviewed publications, especially the relationship between
18 computerized measures of similarity and performance results
19 on behavioral tests of confusion and short-term memory,
20 visual perception, and so on.

21 I want to talk a lot now about the process of
22 validation for accepting new tests by a regulatory agency.

23 To paraphrase a cliché from the domain of real estate,
24 when it comes to regulatory acceptance of new test methods,
25 there are only three issues to be concerned about and they

1 are: validation, validation, and validation.

2 Before a new testing method can be accepted by
3 a regulatory agency, it must be scientifically validated.
4 Validation alone is not enough to warrant regulatory
5 acceptance, but without validation, acceptance ought to be
6 out of the question.

7 As I prepared these remarks, it occurred to me
8 the regulatory agencies must constantly need to evaluate
9 new testing methods. I felt certain that there would be
10 standard methods for establishing the validity of newly
11 developed testing methods, but I was both right and wrong
12 about this.

13 On the one hand, there are no uniform policies
14 for the validation and regulatory acceptance of new testing
15 methods across government agencies. EPA, FDA, USDA, NIOSH,
16 and others each have their own approaches.

17 On the other hand, recognizing this lack of
18 coordination within the U.S. and internationally,
19 toxicologists and regulators from around the world have
20 worked over the last decade to develop a standard approach
21 to the validation and regulatory acceptance of new testing
22 methods. The ad hoc Interagency Coordinating Committee on
23 the Validation of Alternative Methods -- I know that's a
24 mouthful -- also known as the ICCVAM, is a U.S.
25 governmental body run out of the National Institute for

1 Environmental Health Sciences. Together with a similar
2 group in Europe and from the OECD, the ICCVAM has developed
3 clear guidelines for validation and regulatory acceptance
4 of new tests. These guidelines were developed in the
5 context of traditional toxicology with a special focus on
6 finding new alternatives to animal testing.

7 But the overall framework should apply more
8 generally to all validation and regulatory acceptance
9 situations. I strongly encourage the committee, the
10 audience, the agency to study these guidelines. They're
11 easily available on the web. Just do a Google search on
12 ICCVAM, you should find them.

13 It's my recommendation that these guidelines be
14 followed in validating and determining the acceptability of
15 new tests on the confusability of drug names. If they are
16 not accepted, I would request that the agency spell out its
17 own guidelines for validation and regulatory acceptance,
18 and I would also request the agency's rationale for not
19 adopting an existing framework that has proved to be
20 successful elsewhere and is also widely used within the
21 U.S. government.

22 I want to summarize briefly some of the
23 ICCVAM's main criteria for validation.

24 First, they define validation as a scientific
25 process designed to characterize the operational

1 characteristics, advantages and limitations of test method,
2 and to determine its reliability and relevance.

3 The criteria briefly are as follows. Now, some
4 of them apply, obviously, to toxicology, so some of the
5 vocabulary would have to be modified slightly to think
6 about what are really errors in cognition, for the most
7 part, in the context of drug names. But I'll briefly go
8 over them.

9 One, the scientific and regulatory rationale
10 for the test method, including a clear statement of its
11 proposed use, should be available.

12 Two, the relationship of the test methods
13 endpoints to the effective interest must be described.

14 Three, a detailed protocol for the test method
15 must be available and should include a description of the
16 materials needed, description of what is measured and how
17 it's measured, acceptable test performance criteria, a
18 description of how data will be analyzed, and a description
19 of the known limitations of the test, including a
20 description of the classes of materials of the test you can
21 and cannot accurately assess.

22 Next, the extent of within-test variability and
23 reproducibility of the test within and among different
24 laboratories.

25 Also, the test method's performance must have

1 been demonstrated using reference names representative of
2 the types of names to which the test method would be
3 applied and should include both known positive and known
4 negative confusing names in this context.

5 These test names should be tested under blinded
6 conditions, if at all possible.

7 Sufficient data should be provided to permit a
8 comparison of the performance of a proposed new test with
9 the test it's designed to replace. In this case the expert
10 panel is the de facto method.

11 The limitations of the method must be
12 described. For example -- that's self-explanatory. It
13 goes into more about toxicity testing here.

14 Ideally all data supporting the validity of a
15 test method should be obtained and reported in accordance
16 with good laboratory practices, which is just sound
17 scientific documentation.

18 All data supporting the assessment of the
19 validity of the test method must be available for review.

20 Detailed protocols should be readily available
21 in the public domain.

22 The methods and results should be published or
23 submitted for publication in an independent peer-reviewed
24 publication.

25 The methodology and results should have been

1 subjected to independent scientific review.

2 So those are the criteria for validation.

3 They also talk about once a test is validated,
4 how should a regulatory agency determine whether they
5 should accept the validated test because just because it's
6 validated doesn't mean it really fits or meets all the
7 needs of the regulatory agency. So briefly some of the
8 criteria for regulatory acceptance established by this
9 committee.

10 The method should have undergone independent
11 scientific peer review by disinterested persons who are
12 experts in the field, knowledgeable in the method, and
13 financially unencumbered by the outcome of the evaluation.

14 Two, there should be a detailed protocol with
15 standard operating procedures, list of operating
16 characteristics, and criteria for judging test performance
17 and results.

18 Three, data generated by the method should
19 adequately measure or predict the endpoint of interest and
20 demonstrate a linkage between either the new test and
21 existing test or the new test and effects on the target
22 population.

23 The method should generate data useful for risk
24 assessment, for hazard identification, for dose-response
25 adjustment, for exposure assessment, et cetera.

1 The specific strengths and limitations of the
2 test must be clearly identified and described.

3 The test method must be robust. It should be
4 time and cost effective. It should be one that can be
5 harmonized with similar requirements of other agencies. It
6 should be suitable for international acceptance and so on.

7 So I think these are sound criteria. The
8 report is actually a very, very illuminating one for
9 questions about validation and regulatory acceptance of new
10 tests.

11 I believe these criteria are sensible and
12 represent the consensus of an international group of
13 experts. They also have some status as policy within the
14 U.S. federal government, although individual agencies are
15 not bound by them. Again, I recommend they be adopted in
16 this context, and if they're not, I request the agency's
17 own criteria for validation and regulatory acceptance be
18 published.

19 It's worth noting, I think, that none of the
20 methods discussed here today -- none of the methods,
21 including my own, of which I am very proud, but I
22 acknowledge that none of the methods discussed here today
23 meet all of these criteria. I would argue that the methods
24 described by myself and my colleagues come closest, as
25 evidenced by the extensive validation studies published in

1 peer-reviewed journals.

2 The methods described this morning by Dr. Dorr
3 and currently being used by the FDA are likely to be sound
4 in my judgment, but they have not been validated in peer-
5 reviewed journals. To my knowledge, there's not a single
6 peer-reviewed publication providing evidence of the
7 validity of the tests being adopted by the FDA, the so-
8 called POCA method. Nor have the operational details of
9 these methods been fully disclosed, and this would violate
10 the criteria for validation as previously described.

11 I recommend that no method be accepted for
12 regulatory use until it's adequately validated in
13 accordance with the criteria set out above.

14 So that's generally the issues about validation
15 and regulatory acceptance.

16 Now I want to touch on a sort of miscellaneous
17 set of issues that have been raised today where I think I
18 might have something useful to add.

19 The first has to do with the lack of a gold
20 standard. There are many respects in which we lack the
21 gold standard if we're talking about name confusion, and in
22 order to do any sort of validation testing, we obviously
23 need a gold standard.

24 In one respect we do know what the gold
25 standard is for measuring medication errors and that is

1 direct observation of real-world medication orders,
2 dispensing, and administration. This is a method pioneered
3 by Ken Barker at Auburn University and generally is the
4 method recognized to be the gold standard method for
5 detecting medication errors. Again, direct observation of
6 real-world behavior. It's the strongest in terms of
7 ecological validity. It's obviously expensive and time-
8 consuming.

9 There are a variety of other methods which have
10 been discussed today, and I'm generally in agreement with
11 the sort of continuum of having experimental control at one
12 end in the sorts of laboratory tests that I've done and
13 having real-world ecological validity if you do direct
14 observation.

15 But another sense in which there are no gold
16 standards has to do with the USP list. Now, in my own
17 early publications I used the USP list. I sort of didn't
18 know any better at the time and it was the only evidence
19 that I was aware of. But there are very, very serious
20 problems with the USP list, and in no way should it be
21 viewed as a gold standard. In fact, I think it should be
22 viewed as what I will call an iron pyrite standard. For
23 the geologists in the room, the other word for iron pyrite
24 is fool's gold. So it's the fool's gold standard, and it
25 is so not because the people who are use it are fools, but

1 because it fools us into thinking it's a gold standard.

2 So, for example, some names appearing in
3 reporting databases are near misses and not actual errors.
4 So they're status as true positives, as gold standard,
5 truly confusing names is in doubt.

6 But much more importantly, names not appearing
7 in the reporting databases may, in fact, have been involved
8 in multiple errors but never have been reported. In this
9 case, as Donald Rumsfeld says, absence of evidence is not
10 evidence of absence. Just because a name doesn't appear in
11 a reporting database does not mean and does not even come
12 close to meaning that that name hasn't been involved in an
13 error. Ken Barker's studies comparing direct observation
14 -- and the same is true with Bates and Leape's famous
15 studies of medication errors where they compared direct
16 observation to spontaneous voluntary reporting -- indicate
17 that direct observation yields between 100 and 1,000 times
18 more errors than spontaneous reporting. So what we have in
19 the USP list is sort of the tip of the tip of the iceberg.

20 This is highly problematic because if we use
21 the USP list as a gold standard and let's say we identify a
22 pair of names that isn't on the USP list, we're going to
23 call that a false positive, but in fact there's no real
24 good justification for calling it a false positive. In
25 fact, it may have been involved in an error that was never

1 reported.

2 Similarly, if we say the name that is on the
3 list is not an error, we can't be certain that this is a
4 false negative either because of the dubious status of the
5 names that appear on these lists.

6 Related to this is the need in any sort of
7 validation testing for the proportion of truly confusing
8 names and non-confusing names to match the proportion in
9 the real world. The problem is we don't know what the
10 proportion of truly confusing names to non-confusing names
11 in the real world. But evaluations of predictive tests,
12 things like sensitivity, specificity, positive predictive
13 value, and so on, which are technical characteristics of a
14 predictive test, all depend crucially on the proportion of
15 truly confusing and non-confusing names in the population.

16 Next, we're looking at the wrong unit of
17 analysis a lot of the time, and again, I take some of the
18 blame because I myself I think used the wrong unit of
19 analysis in some of my early work. Much of the work on
20 computer methods for name screening, including my own early
21 work, has focused on pairs of names. Clearly there's a
22 certain relevance in thinking about pairs of names because
23 pairs of names are what get confused. But FDA or any other
24 screening agency must approve single names, not pairs of
25 names. So whatever criteria or screening method we use

1 must evaluate single names, not pairs of names. Methods
2 are needed, therefore, that use the single name as the unit
3 of analysis, not the pair of names. And there are lots of
4 technical reasons why this is so. I'll try to describe
5 just a couple of them.

6 Any method based on pairs of names will almost
7 necessarily have poor positive predictive value because the
8 sheer number of pairs will overwhelm the false positive
9 rate of the predictive test. That is, let's say you have
10 1,000 names in the lexicon. Well, there are roughly
11 500,000 pairs that you get from 1,000 names. If you have n
12 names, there are n times n minus 1 over 2 pairs of names.
13 So for 35,000 or however many trademark names there are,
14 you have tens of millions of pairs of names. Any false
15 positive rate above a tiny fractional false positive rate
16 will totally overwhelm a system if you have that many pairs
17 of names.

18 In addition, there's this problem that's
19 related to the pair is the wrong unit of analysis but also
20 has to do with frequency. Not nearly enough attention has
21 been paid to frequency. Frequency is a fundamental
22 mechanism of human error, but is absent from most of the
23 discussion about name confusion until very recently,
24 including in my own work until recently. There's been too
25 much focus on similarity.

1 But the problem is this. All the similarity
2 measures that have been discussed today are symmetric.
3 That is, the similarity between name A and name B is
4 exactly equal to the similarity between name B and name A.
5 The problem is errors are not symmetric. If you have a
6 common name and a rare name that are similar to one
7 another, when presented with the rare name, it's very
8 likely that you will see the common name, but when
9 presented with the common name, it's very unlikely that you
10 will claim to see the rare name. So error patterns are
11 driven by frequency, not just similarity. In fact, in my
12 experiments and in a wealth of psycholinguistic literature,
13 the frequency effect is at least an order of magnitude more
14 powerful than the similarity effects.

15 So we need to start building prescribing
16 frequency into our predictive models. This recommendation
17 alone is not trivial because there are multiple measures of
18 frequency from the government, from something like the
19 NAMCS database, from IMS, from Solutient. They don't all
20 agree with one another, and so even including prescribing
21 frequency could be complicated, not to mention we don't
22 know the prescribing frequency of a compound before it's
23 marketed, although we have some indication.

24 We have to think a lot more about non-name
25 attributes. I'm in agreement with a lot of previous

1 speakers who acknowledged that non-name attributes --
2 namely, strength, dosage form, route of administration,
3 schedule, color, shape, storage circumstances, et cetera --
4 are important contributors to errors. The exact magnitude
5 of their contribution is unknown and needs to be the focus of
6 future research.

7 There is the issue of conflict of interest. A
8 lot of money is at stake in naming decisions, both in the
9 naming companies and obviously the PhRMA sponsors. We need
10 to make sure that those doing the safety screening do not
11 have a vested interest in the outcome of the screening.
12 For example, if people who coin the names also do the
13 safety screening, they would obviously have some interest
14 in finding that the name was safe. It doesn't preclude
15 those companies from doing that screening, I should say.
16 They just need to have some safeguards in place.

17 There's this issue of public costs and private
18 benefits, which I brought up in June. Normally the FDA
19 weighs risks and benefits in drug approval decisions, but
20 here it's difficult to see how the agency would weigh risks
21 and benefits since all the risks accrue to the public, all
22 the benefits tend to accrue to the sponsor of the product.

23 Harm reduction I agree is the ultimate goal.
24 When evaluating a proposed name, we need to think not just
25 about the probability of error, but about the magnitude of

1 harm. Harm, as others have suggested is a complex function
2 of the probability of error, the number of opportunities
3 for error, the severity of each error, the probability of
4 not detecting the error, and so on and so forth. Each of
5 these components is difficult to understand because the
6 extent of harm depends on the patient status, the duration
7 of exposure, the duration without the intended medication,
8 the concomitant medications, and so on and so forth.

9 Just a matter of scope -- I said this on June
10 26th, but it's worth repeating. The best estimate which we
11 have of the actual number of name confusions in the United
12 States comes from a recent article by Flynn, Barker and
13 Carnahan in the Journal of the American Pharmacists
14 Association, and based on a direct observational study,
15 they report that the wrong drug error rate is .13 percent.
16 That is, they detected 6 wrong drug errors out of 4,481
17 observations. If you extend that to the 3 billion
18 outpatient prescriptions that are filled per year, that's
19 about 3.9 million wrong drug errors per year, or about 65
20 per pharmacy annually or about 1 per week in every pharmacy
21 in the United States.

22 Finally, I want to agree with Maury Tepper and
23 others. I agree with a lot of what Maury said, and I don't
24 just mean the part about being a dumb lawyer.

25 (Laughter.)

1 DR. LAMBERT: It's not all about names. Even
2 if we could figure out a perfect screening method for new
3 names, which we will not be able to do, I'm in total
4 agreement this is probabalistic. In the end, the decision
5 will be made by a panel of experts much like this one just
6 like in the end the decision to approve new chemical
7 entities is made by a panel of experts. In spite of the
8 thousands of pages of objective clinical trial data,
9 preclinical trial data, the decision to approve a drug is
10 eventually made by a panel of human experts. That's the
11 way it's going to be here, and it's made on a probabilistic
12 basis. That's the best we're ever going to be able to do.

13 But even if we could perfect the approval of
14 new names, we would still be stuck with the thousands of
15 names that we have, many of which seem to play a role in
16 confusion. So what are we to do about those?

17 Here I don't think there's any better authority
18 than Mike Cohen or the people at the Institute for Safe
19 Medication Practices who for years have been advocating
20 safe prescribing practices, safe medication practices,
21 which will minimize these errors regardless of the
22 confusability of names, things like putting the indication
23 on the prescription, dramatically restricting verbal
24 orders, dramatically restricting handwritten orders, using
25 computerized physician order entry, and so on and so forth.

1 So I add my voice to those who said there's a lot we can
2 do about name confusion other than getting better and
3 better predictive methods for knowing which new names will
4 be confused. While obviously I've devoted a lot of my own
5 time and effort to doing this prediction of new name
6 screening, there's a lot we can try to do to make the
7 system safer and more robust against confusion even with
8 the trademarks we've already got.

9 Thank you very much for your attention.

10 DR. GROSS: Thank you very much, Dr. Lambert.
11 These have been excellent presentations.

12 There was supposed to be one other presenter,
13 Patricia Penyak, who unfortunately was in a car accident
14 and is unable to be here, but her material that she was
15 going to present is in our handouts. So we wish her well.

16 Is there anyone else who wishes to comment
17 during the period of public comment?

18 (No response.)

19 DR. GROSS: If not, let's move on to Dr.
20 Seligman who will tell us the questions they would like us
21 to consider.

22 DR. SELIGMAN: First, let me thank both the
23 presenters this afternoon as well as this morning for, I
24 think, excellent and thoughtful presentations that I think
25 in many ways have really outlined the complexity of this

1 topic and really set the stage for what I hope will be a
2 very informative discussion this afternoon.

3 We have taken the liberty of posing five
4 questions or broad areas that we would like our advisory
5 committee to deal with this afternoon. The first one deals
6 with describing the advantages and disadvantages of
7 evaluating every proprietary drug name for potential
8 confusion versus taking a more selective risk-based
9 approach, considering as we've heard this morning, issues
10 related to consequences, probability, disutility, et
11 cetera, and whether indeed it's possible to develop an
12 approach which would allow us to triage drug names into
13 groups that may be handled differently based on these
14 potential risks.

15 The second question deals again with many of
16 the study methods that were presented today in asking the
17 advisory committee to give us an assessment of those design
18 elements of those methods that should be included in a good
19 naming practices guidance and what elements of those
20 methods should either be discounted or not considered
21 useful in developing such guidance.

22 Third, we would certainly like to hear from the
23 committee if there are, indeed, other methods that should
24 be considered in producing such good naming practices.

25 Finally, we'd be very interested in learning

1 under what circumstances field testing in a simulated
2 prescribing environment should be considered. I think it's
3 pretty clear, based on what we've heard today, that it's
4 unlikely that one method alone would be sufficient, and
5 clearly we're interested in learning what combination of
6 methods should be deployed such as behavioral testing and
7 orthographic and phonographic testing or other combinations
8 of methods.

9 Finally, we'd be interested in hearing from the
10 committee as to whether there are circumstances, if any,
11 when it might be appropriate to approve a proprietary drug
12 name contingent on either some element of a risk management
13 program being in place in the post-marketing environment.

14 With that, Mr. Chairman, I turn the discussion
15 to you.

16 DR. GROSS: Dr. Seligman, could you clarify the
17 last question? When you say approve a proprietary drug
18 name contingent on risk management program, that means that
19 for some reason the name will stick rather than trying to
20 change it or because the drug is risky and you want to have
21 a risk management program?

22 DR. SELIGMAN: No. It's basically essentially
23 allowing a name to be used knowing that there might be a
24 potential for, I guess, confusion and the degree to which
25 one might want to more carefully assess in the post-

1 marketing environment indeed whether harm occurred as a
2 result of allowing that name to proceed into the post-
3 marketing environment. Jerry, is that the interpretation?

4 MR. PHILLIPS: Yes.

5 DR. GROSS: Okay, fine. Thank you.

6 Is it the committee's pleasure to do this one
7 at a time starting with number one? Okay. Does anyone
8 want to comment on number one? Advantages and
9 disadvantages of evaluating every proprietary drug name
10 versus taking a more limited approach based on risk.

11 MS. JAIN: Well, Dr. Gross, I just want to say
12 that you had mentioned previously that you wanted the FDA
13 representatives and the PhRMA representative, Mr. Lee, to
14 produce lists of how they do their analysis in a step
15 method. I distributed the FDA version that Jerry Phillips
16 was nice enough to write up, and I've got copies for the
17 committee members from Mr. Lee as well that I'll distribute
18 at this time.

19 DR. GROSS: Okay, good.

20 Brian.

21 DR. STROM: The question is whether all drugs
22 should be screened or whether a risk approach should be
23 used. My sense is that all drugs have to be screened
24 because even if the drug itself is a low-risk drug, you
25 don't know which drugs it's going to be confused with.

1 They, in turn, may be high-risk drugs.

2 I think the place that the level of risk would
3 come into play is more related to the fifth question, that
4 if in fact the therapeutic ratio of both drugs is low so
5 that they're both relatively safe drugs, you might be more
6 willing to tolerate allowing a drug name on the market
7 despite the risk of confusion. So your threshold for a
8 decision may be different, but it's hard to imagine you
9 could not screen all names given you don't know which drugs
10 they're going to be confused with.

11 DR. GROSS: I see a lot of nodding heads on
12 Brian's response. Yes, Curt.

13 DR. FURBERG: Yes, I agree with Brian. I can
14 see a step-wise approach. You start off with screening,
15 probably very simple or simplistic.

16 The issue really is how do you define a high-
17 risk drug. That is the crux. Where do you draw the line?
18 I'm not sure I know exactly how to take a stand on that.
19 But clearly, step-wise makes a lot of sense.

20 DR. GROSS: So that's the second part of the
21 question, but for the first part, does anybody disagree
22 that all drugs should not be run through an approach?
23 Robyn.

24 MS. SHAPIRO: I don't think I disagree. I just
25 want to be sure that I'm understanding this right, and that

1 is that at the moment this happens in two different
2 spheres. One is the FDA already does that. That's the
3 practice now, and two, the whole trademark process, as we
4 heard about, also is a way of screening for this very
5 thing. Is that right?

6 DR. GROSS: No. I think that's a separate
7 issue.

8 MS. SHAPIRO: Okay.

9 DR. GROSS: We're not saying who's going to do
10 the screening. Right? Is that your question?

11 MS. SHAPIRO: No, no.

12 DR. GROSS: Paul, is your question whether the
13 FDA should do the screening or somebody should do the
14 screening?

15 DR. SELIGMAN: No, it's not a question of who.
16 It's a question of whether, whether it should be done.

17 DR. GROSS: Right. That's what I assume.
18 Okay.

19 MS. SHAPIRO: And I'm just trying to confirm on
20 the whether, not the who, that there are two systems
21 already in place doing that.

22 DR. GROSS: Okay. That does not happen to be
23 one of the questions of the five, but it's certainly
24 something that we can comment on because it's an issue
25 that's come up over and over again. If you want to discuss

1 that -- you know what? Why don't we go through the
2 questions here and then come back to that particular point
3 because it is an important issue.

4 MS. SHAPIRO: Okay.

5 DR. GROSS: So it sounds as though everyone
6 agrees that all proprietary drug names should be screened.
7 We're not specifying how.

8 Yes, Stephanie.

9 DR. CRAWFORD: Thank you. Just to clarify our
10 recommendation, would this be every drug name screened pre-
11 approval? We're not talking about retrospectively looking
12 at all existing proprietary names?

13 DR. SELIGMAN: That's correct. Pre-approval.

14 DR. GROSS: Yes, Lou.

15 DR. MORRIS: Does that include OTCs on
16 switches?

17 MR. PHILLIPS: Yes.

18 DR. MORRIS: Are they screened now?

19 MR. PHILLIPS: If they are subject of an
20 application, they are screened.

21 DR. MORRIS: So if a well-known prescription
22 drug that's on the market is switched and has the same
23 name, it has to go through new testing?

24 MR. PHILLIPS: It usually has a modifier or
25 something associated with that trade name and it will go

1 through an assessment.

2 DR. MORRIS: Oh, okay.

3 DR. GROSS: The second part -- yes, Jeff.

4 MR. BLOOM: I just wanted to add one thing. I
5 agree with that as well. I'll just add to the point that
6 even a drug that seemingly may be innocuous, we have to
7 recognize that many drugs are used in combination, and
8 whereas a drug may seem to be rather safe, but when used in
9 combination might have some other side effects or
10 interactions, I think it's very important that it all be
11 screened. I agree completely that it should be screened
12 ahead of time.

13 DR. GROSS: How about the second part of
14 question number one? Is it possible to triage the drug
15 names into groups that may be handled differently based on
16 risk? So an initial approach is a yes or a no, and if yes,
17 how? Eric.

18 DR. HOLMBOE: I think in fact what Brian said
19 earlier, it would be difficult to do that until you know
20 what it's look-alike actually is. If it turns out it's a
21 low-risk drug, but it's similar to a high-risk drug, then
22 it's hard to triage based on the single agent.

23 DR. GROSS: Yes, I agree too.

24 Does anybody else want to comment on that part?

25 Arthur.

1 DR. LEVIN: A point of clarification. There
2 are several risks here. One is risk of confusion, one is
3 the risk of toxicity. And there are probably a lot. We
4 can make a long list of risks, so we just need to be clear
5 when we talk about potential risks that we agree what we're
6 talking about.

7 DR. GROSS: Paul or Jerry, do you want to
8 comment on that?

9 DR. SELIGMAN: When we talk about risk, we're
10 pretty much talking about risks of adverse events,
11 basically the consequences, the probability, the
12 disutility, some of the things that Sean Hennessy addressed
13 this morning.

14 DR. GROSS: So it sounds as though the answer
15 is no to the second question. Anyone else want to comment?
16 Lou.

17 DR. MORRIS: Is it possible? The answer is
18 yes. But is it advisable is the question. Clearly you can
19 put drugs in categories based on the severity of the
20 adverse event, but I think the question here is is it
21 advisable to do that, and I don't know the answer.

22 DR. GROSS: Fair enough.

23 DR. STROM: Yes. To just be clear, I
24 completely agree with that. It's possible to stratify
25 based on the risk of the error with the parent drug, but

1 we're saying that in initial screening you shouldn't do
2 that because it's impossible to know what the risk is of
3 the drug it's going to be confused with because you don't
4 know yet what drug it's going to be confused with.

5 DR. GROSS: The second question then is based
6 on discussion of the study methods presented today,
7 identify the critical design elements of each method that
8 should be included in good naming practices. I'm not clear
9 on that question. I mean, we're not really going to
10 discuss the critical design elements in each of the
11 methods. Is that what you want us to address? Or did you
12 want us to say what study methods should be used in trying
13 to avoid confusion or what combination of study methods?

14 DR. SELIGMAN: I think either what methods or
15 what combination of methods, but also particularly within
16 some of those methods, were there elements of them that
17 were particularly strong or important that should be
18 emphasized in constructing good naming practices?

19 DR. GROSS: Yes. I think Dr. Lambert made a
20 very good point that there are very few that have been
21 validated except for the ones that he described. If
22 anybody disagrees with that and is aware of other
23 validations, please speak up.

24 So does anyone want to comment on that first
25 sentence? Brian.

1 DR. STROM: I wanted to make a number of
2 comments. I've been writing notes and this seems to be the
3 appropriate question to respond.

4 I think what we heard today and in June is
5 striking, that in a sense in drug names, we're equivalent
6 to a pre-FDA era in drugs. It's as if we were approving
7 drugs based on preclinical data only and no clinical data.
8 We're approving drug names here based on data that has
9 never been validated, and we don't know what the
10 interpretation of any of it is.

11 We hear, on the one hand, that industry thinks
12 it's a tiny problem. We hear, on the other hand, FDA
13 rejects a third of the ones that industry thought were a
14 non-problem. And we don't know which one is right based on
15 the available information.

16 We've heard many people talk about their best
17 practices and everybody should use best practices, but none
18 of those best practices have been validated to know that
19 any of them are in fact best practices. A lot of cutting-
20 edge, very exciting new methods that we're hearing about --
21 and I'm very interested and excited by all that, but none
22 of that has yet been evaluated.

23 So I guess my own biases would be, on one hand,
24 to be careful. I would not recommend changing a current
25 process, given we don't know what's right and what's wrong

1 with the current process. But I would recommend we don't
2 know what's right and what's wrong with the current process
3 and we need an enormous amount of work very quickly to do
4 the needed validations and to use simulations and
5 laboratory techniques and the kind of thing Sean talked
6 about and whatever as ways of trying to find out what works
7 and what doesn't. We probably shouldn't change much until
8 then because, again, we don't know that there's a major
9 problem out there. The current system with industry doing
10 it and then FDA doing it may well be fine, or at least,
11 parts of it may well be fine and you don't want to risk
12 throwing out parts that work, given we don't know what
13 works and what doesn't work.

14 DR. GROSS: Other comments? Michael.

15 DR. COHEN: I also jotted down some notes.

16 I think the expert panels, the focus groups are
17 important, and that is current practice I think for most of
18 the companies. I think it picks up the kinds of things
19 that some of the other testing may not. For example, the
20 computerized systems that we heard about today would not
21 pick up some of the prescribing-related problems like
22 stemming of a drug name, those kinds of issues that
23 sometimes cause confusion with a drug that's already
24 available.

25 I think also the value of the nurses' input and

1 unit clerk input and pharmacists' input is immeasurable.
2 True. But I think it's very important. They're likely to
3 pick all kinds of things: confusion with prescription
4 abbreviations, for example, parts of a name that might be
5 confused with a dosage form or the dose or quantity, as we
6 heard. So I'd like to see that continue.

7 The computer matching. I could see that being
8 used in conjunction with it. I mean, it is a validated
9 process. We've heard that. I think it depends largely on
10 the type of database that's used, what the database is.
11 For example, there are some databases that contain names
12 that are not really drugs on the market, and you'll get
13 printouts of that. I also --

14 DR. GROSS: Michael, I thought it was said that
15 the computerized systems have not been validated.

16 DR. COHEN: I thought that Bruce said that it
17 was. His system. Did I miss that?

18 DR. LAMBERT: Am I allowed to speak?

19 DR. GROSS: Yes. Bruce, do you want to
20 comment?

21 DR. LAMBERT: The methods that I propose and
22 have been working on for the last seven or eight years have
23 been subject to extensive validation testing. This is not
24 to say they're perfectly valid. When you subject a method
25 to extensive validation testing, what you find are both its

1 strengths and its limitations. What I argued was that the
2 methods that I have described are to my knowledge the only
3 methods for which there are peer-reviewed articles about
4 the status of their validity.

5 DR. GROSS: Yes, I know. Bruce, Bruce --

6 DR. LAMBERT: And certainly my methods, I
7 validated them against visual perception, several different
8 short-term memory tests, against the perceptions of
9 established experts, against the perceptions of lay people,
10 against databases of known errors, and so on.

11 So the methods that I propose, the bigram,
12 trigram, Edit, et cetera, are by no means perfect, but I
13 have documented in extensive detail the extent to which
14 they are valid. Those materials are in your briefing
15 packets. I sent them to the agency weeks ago, but I'm told
16 that you only received them today. So if you haven't read
17 them, I understand. They're not exactly as exciting as a
18 John Grisham novel. But these methods have been subjected
19 to extensive validation testing. It's up to your own
20 judgment as to whether you think they are valid enough for
21 use for these purposes.

22 DR. COHEN: I want to point out that I don't
23 think they can be used alone without any doubt. I think
24 they can be used in combination.

25 DR. LAMBERT: And neither do I. In all of my

1 publications, I say they shouldn't be used alone.

2 DR. GROSS: Bonnie.

3 DR. LAMBERT: I say they should be an input to
4 an expert process.

5 DR. GROSS: Bonnie.

6 DR. DORR: I just wanted to point out that
7 there is currently under peer review an article on an
8 evaluation of different techniques. One of them is ALINE.
9 Another is -- as I mentioned this morning, our best result
10 was a combination of ALINE with a bunch of other techniques
11 where we're getting high results with the caveats already
12 mentioned in my talk and also Bruce Lambert mentioned that
13 the data that you have as a gold standard -- we're having
14 problems with that. We're using USP. We did use a smaller
15 list of known error drug names that are not the USP list
16 also, and we were getting similar results.

17 And the technique itself of ALINE, outside of
18 the task of drug name matching, has indeed been validated
19 by several peer-reviewed articles. There's a Ph.D. thesis
20 on it but, again, that wasn't for the task of drug name
21 matching. Right now, within two to three weeks, we should
22 know the answer for a particular peer-reviewed article for
23 this task, and we'd like to talk more about the combination
24 of different approaches and also not just within the
25 computerized technique, but outside of that. What can we

1 combine those computerized techniques with to get what you
2 need.

3 DR. GROSS: Right. That's a separate issue.

4 DR. DORR: Because as Bruce said, you can't
5 just say it's valid for this test. Even if you say the
6 algorithms are, indeed, measurable up against each other,
7 it may not be appropriate for this task.

8 DR. GROSS: Thank you both for the
9 clarification.

10 Michael, do you want to continue?

11 DR. COHEN: Yes. Let me continue.

12 Where I think it can be valuable is if
13 something might be overlooked with the review by
14 practitioners, the group testing, et cetera, I think that
15 that can help as kind of a backup system that further
16 assures that something important is not overlooked. So
17 that's why I see this being used only in combination, not
18 by itself.

19 Then thirdly, about the model pharmacy and the
20 laboratory. I can definitely see where that could be
21 helpful post-marketing. Pre-marketing, at least at this
22 time, until we see some evidence of its value, I could see
23 a lot of problems with it, and I don't think that that
24 would be of value at this time anyway until we see it
25 actually proved for the reviews.

1 DR. GROSS: Curt?

2 DR. FURBERG: Well, it's clear that we have
3 multiple methods. They all have strengths and weaknesses,
4 and so I agree with the idea that you need to somehow
5 develop a battery.

6 My sense is that people in the field are not
7 communicating very well, and there seems to be some turf
8 issues also. We can't settle that in a hearing.

9 So my suggestion is that the FDA appoints a
10 working group of all the experts and let them come up with
11 a recommendation of an appropriate battery that could be
12 discussed, come back to the committee, and then we can move
13 forward.

14 DR. GROSS: Ruth.

15 DR. DAY: The problem that we're having right
16 now is there are several different methods and each have
17 several different design features. Each design feature has
18 advantages and disadvantages. So if we had the list before
19 us and we had a lot of time, we could do that, and maybe
20 Curt's suggestion would be good.

21 But if we were to go down each element in each
22 method, it could be very useful. For example, an expert
23 panel. In round one, as I understand it, people
24 independently generate sound-alike or look-alike candidates
25 for a given drug name. Well, where do those come from? So

1 some of the people might just take it out of their heads,
2 out of memory, availability in memory. Some might go check
3 the PDR. Some might look at the USP database and so on and
4 so forth. You want people to be able to do whatever they
5 do because that's what they're going to do in every day
6 life. But you could document it a bit. So for each focus
7 group, after it's over or after round one is over, you
8 could get that information.

9 So a big problem in all of this is noise in the
10 data and lack of replicability. And it could be that by
11 getting more information like this, you could say, oh,
12 focus group 1 all looked up in the PDR. Focus group 2 had
13 a mix of other methods to generate and so on and so forth.

14 So especially for whatever is the first step in
15 all these processes, such as generating potential names to
16 consider -- that might be difficult -- or in the case of
17 the linguistic methods, there are other things to do first
18 like pronounceability, which I'll comment more on later.

19 DR. GROSS: Yes, Eric.

20 DR. HOLMBOE: Also, I just want to highlight
21 that it was my understanding at the beginning that your
22 hope was that in time industry actually would take a
23 greater responsibility for this. And so far, I think what
24 we've talked about is actually what you're doing. Clearly
25 the strengths and weaknesses due to that and I think we'd

1 all agree that a multi-factorial approach is probably the
2 best.

3 But I would be interested to know actually what
4 industry is doing. We haven't heard a lot about that. We
5 didn't get a lot of data, but clearly there's a big
6 disconnect. We've heard from several groups today that
7 they feel that they're doing a fair amount in this kind of
8 pre-marketing work, and yet, as we heard, you reject a
9 third of the names despite the amount of effort that
10 they're using to try to come up with a drug name even
11 before it reaches your desk, so to speak. So I think there
12 needs to be a better understanding of why we're seeing such
13 a disconnect, particularly if we're going to migrate the
14 methods back into the private sector for them to take care
15 of it instead of you doing the things you currently do.

16 The second thing I would highlight is that
17 we've heard from the epidemiologic perspective that what
18 you're really trying to look for here is a really good
19 screening test. So you're really looking for something
20 that's going to give you high sensitivity, and then how do
21 you deal with the kind of false positive rate that gets
22 generated out of that? Clearly that's another issue that
23 we haven't really brought up today, but in a sense that's
24 what we're talking about with a lot of these things that
25 we're really trying to screen. So that would be another

1 principle.

2 The finally, I'd encourage you to look at the
3 Medical Research Counsel out of Britain actually which has
4 done a very nice monograph on how to approach conflicts
5 intervention. That's what you've got here. You've got
6 multiple methods that you're using. And they provide a
7 very nice framework to think about how to move this forward
8 over time that perhaps the working group would be able to
9 use as well.

10 DR. GROSS: I wonder if Paul or Jerry might
11 comment on why the high rejection rate on the names from
12 industry when they've gone through the screening that they
13 have told us. They've told us they have gone through most
14 of the screening methods that have been described.

15 DR. SELIGMAN: I don't know the answer for
16 sure, but I'm happy to speculate because I suspect that
17 there's probably a wide diversity within industry as to the
18 kinds of techniques that they've applied. I think what you
19 heard today, if I again would venture to speculate, is
20 probably the best practices that probably are, indeed, well
21 conducted by many of the major pharmaceutical companies.

22 I don't know, Jerry, whether we have any
23 analyses that we've done on looking at those we've rejected
24 and whether there's any difference by company size or
25 generics versus proprietary names or whether there are

1 clues as to why there seems to be that disconnect.

2 DR. GROSS: Yes. I see Bob Lee's hand is
3 raised. We were going to ask him, even if he didn't raise
4 his hand, to comment.

5 MR. LEE: I thought it might be helpful to just
6 explain what it is we do do as part of our screening. A
7 lot of it initially is what you'd really call data
8 acquisition. Well, even before that, first we have to
9 generate new names. They have to be created. We can do
10 that in-house. Anybody can sit down and come up with
11 coined or arbitrary names. These are names that don't mean
12 anything, but which are pronounceable. But we usually use
13 more expert groups, branding companies who know how to do
14 that a little better, who may have been in the advertising
15 area or have other backgrounds in creativity, if you can
16 define what creativity is.

17 So they generate long lists of names that then
18 are submitted to the company, usually to a team within the
19 company that's made up of different disciplines. There are
20 so many initially, 100, 200, 300 names, that they have to
21 be narrowed down into a smaller, more manageable group for
22 extensive searching. So some are thrown out just because
23 they're not liked and some obviously have bad connotations
24 or remind people of bad things, or for a variety of
25 different reasons many of those names are just thrown out

1 from the beginning where people can spot confusion problems
2 immediately upon seeing some names.

3 But then you get down to a group of names,
4 perhaps 30 that you begin a very extensive searching
5 process on using various algorithms like the algorithms
6 we've seen although maybe not identical to Dice coefficient
7 of the kinds of letter-string systems that we've seen, or
8 the phonetic tools that we've seen today are very powerful.
9 So not necessarily those, but where you will take prefixes,
10 suffixes, letter strings and combine them in various ways
11 to try to pull out of the database that you're searching
12 other names that look similar to the one you think you want
13 to go forward with.

14 DR. SELIGMAN: Bob, do you know how common
15 these practices are within industry, and can you speculate
16 as to why there seems to be a disconnect between the
17 rejection rate of names within the FDA and your view that,
18 indeed, this work is being done very thoughtfully and
19 carefully within industry?

20 MR. LEE: Well, I think your point is actually
21 a very good one about whether or not all of the companies
22 who eventually submit names to the FDA are following these
23 practices. I'd have to say I think most of the major PhRMA
24 companies that make up the PhRMA organization are following
25 similar practices. They're not doing everything that we

1 might list, but they're doing many of them. Almost all of
2 the major PhRMA companies are doing extensive searching in
3 databases using algorithms.

4 That's not to say that there can't be improved
5 algorithms and certainly improved databases where all of
6 the factors we talked about can be accumulated in that
7 database so that they're readily available to the
8 searchers. That makes a more comprehensive review possible
9 because otherwise you have to do the trademark searching,
10 though names only, and then you have to do investigations
11 about the names that you're seeing that might be
12 confusingly similar to the ones you're going forward with.
13 You then have to do a lot of searching to find out what's
14 the dosage amount, so on and so forth.

15 Of course, getting information from front-line
16 practitioners about that is very, very helpful, but
17 sometimes it's difficult to acquire that data.

18 DR. GROSS: Arthur.

19 DR. LEVIN: Two comments. Paul, with all due
20 respect to PhRMA, I would suggest that the purpose behind
21 trademarking is not primarily safety. Trademarking, one,
22 has a legal aspect that's very powerful, and it has a
23 marketing aspect that's extremely powerful. I don't mean
24 to suggest that the safety is disregarded, but trademarking
25 is not a principle or a concept or an activity that was

1 developed in the field of safety management, risk
2 management. Number one.

3 The second thing. In a way, equally
4 interesting to the question of why this disconnect where a
5 third of the names that go through this rigorous process
6 are rejected by FDA is what about the names that FDA
7 accepts. They've gone through a rigorous process by PhRMA,
8 and then they're accepted by the FDA's rigorous process,
9 and then lo and behold, we find significant problems in
10 confusion.

11 Have we taken a look-back at those failures, so
12 to speak, and said what happened here? How did it get
13 through both of us, and what was missing in our process?
14 Because it seems to me to answer the question about what's
15 needed in terms of what sorts of combinations of processes
16 can best eliminate the problem or reduce the problem is to
17 know where the failure has been. It's like dealing with
18 error and learning from error. We go back and look at what
19 went wrong to discover how to do it right, and I think the
20 same principle should apply here.

21 DR. GROSS: Yes, doing your own RCA or FMEA.

22 Before we try to come to some conclusions on
23 question 2, let's take a look at the second sentence in
24 that question. Are there any methods that should be
25 discounted as not being -- and the key word is --

1 potentially effective. So there are some tools that we've
2 heard have not been validated but potentially they may be
3 worthwhile. Does anyone want to discount any of the
4 methods that we've heard?

5 DR. MORRIS: I wouldn't discount per se, but I
6 was struck today that I felt certain tools or certain
7 techniques were -- I was comfortable as seeing them as
8 hypothesis-generating techniques, but not confirming, and
9 yet simulations I felt I was more comfortable with at least
10 their potential. So maybe we can separate them into
11 hypothesis-generation techniques and possibly confirming
12 techniques as a means of putting them in some category.

13 DR. GROSS: Okay.

14 Yes, Michael.

15 DR. COHEN: I guess I disagree a little bit
16 with that only because, like I said before, I haven't seen
17 them proved yet, number one, and I know you'd agree with
18 that. Number two, they really do see a little complex and
19 perhaps not so practical to actually carry out for
20 trademark reviews when large numbers of names are being
21 used. They don't include all environments in which the
22 drugs are used. I don't know that they couldn't be set up.
23 All I'm saying is I think it needs a lot more work.

24 DR. MORRIS: I used the word "potentially" very
25 carefully there because I agree that because they're not

1 validated or we don't know enough about their validation,
2 I'm not comfortable saying how they should be designed, but
3 I think they have more potential for giving us better data.

4 DR. COHEN: I would say that they definitely
5 would hold promise, but it needs more work.

6 DR. GROSS: Yes. I'd like to propose as a
7 possible approach to the whole of question 2 to follow up
8 on what Curt Furberg said and that maybe the FDA could
9 appoint a small group of people to come up with maybe a
10 minimum combination of methods. Does that fit what you're
11 talking about, Curt?

12 DR. FURBERG: Yes.

13 DR. GROSS: A minimum combination of methods
14 and then if people want to supplement it with other
15 methods, fine. It's always hard whenever you take a multi-
16 faceted approach and you're picking from a menu of many
17 different methods how to pick which ones will work. There
18 aren't too many studies done in various fields where that's
19 been elucidated.

20 DR. FURBERG: But I think it's also important
21 to have broad representation. I think PhRMA should be
22 involved, should be represented on that committee.

23 DR. GROSS: Sure.

24 DR. STROM: Can I have two comments on that?

25 One is to some degree the June meeting was that in terms of

1 having groups talk to each other and with each other and
2 communicate.

3 The second, what's really needed is what you're
4 describing in terms of a work group doing it, but it needs
5 data to work with. The groups, having now talked to each
6 other in June and now presenting here, it's not clear to me
7 that a meeting yet -- I think that kind of meeting is
8 exactly what's needed after there's some data for the
9 meeting to react to because everyone can give an opinion,
10 but it's like saying I think this drug is effective because
11 in my experience it worked before the era of clinical
12 trials. Until we have some scientific data to know what
13 works and what doesn't, all we're going to hear is more
14 opinions and more expressions, best practice, without a
15 basis behind it.

16 DR. GROSS: So in the absence of enough
17 scientific data, would you like to make another proposal?

18 DR. STROM: I think there needs to be a major
19 -- well, that's why one of my suggestions before is that I
20 wouldn't change things much now yet in the way things are
21 done. I certainly wouldn't abandon what FDA is doing, in
22 terms of shifting it to industry, given a third of the
23 drugs it's getting from industry it's now rejecting. But I
24 think a major effort is needed for a large research effort
25 in order to generate data evaluating these approaches.

1 Once those data are available, that's the time to hold the
2 kind of meeting that Curt described.

3 DR. FURBERG: Yes, but you can't talk about it
4 sort of globally. We need new research direction. Who's
5 going to provide those? You need that expert group to sit
6 down and say this is what we know, this is what we don't
7 know, and then develop a plan from that.

8 DR. STROM: The people are going to provide it,
9 the researchers. There is no lack of researchers in this
10 country. And if FDA would issue, as a challenge to PhRMA,
11 RFAs to say let's evaluate the methods that are now being
12 used.

13 DR. FURBERG: I would be more in favor of a
14 coordinated effort rather than what you're talking about,
15 an isolated effort by people who have self-serving
16 interests to some extent and pursuing their own ideas. I
17 think we need to get together. All the parties should be
18 involved. We should discuss what we know and what we don't
19 know and then develop a plan.

20 DR. GROSS: Any other comments from the
21 committee? Yes, Jeff.

22 MR. BLOOM: Yes. On the Regulatory Reform
23 Committee, which I was a member of, we did have
24 recommendation 238. The reason to shift doing the safety
25 testing to industry was the recognition of the limited

1 resources of the FDA frankly, which is part of the problem
2 in this issue. The idea was that to review data from
3 sponsors who followed protocols designed to evaluate
4 potential for look-alike and sound-alike errors with
5 generic and proprietary names prior to FDA-regulated drugs
6 and use the information gathered from that name safety
7 research to improve patient safety. One of the ways you
8 would improve that is looking at Medwatch reports -- you do
9 get adverse events from naming problems and things like
10 that -- and see which ones are minimized and which ones are
11 not. You can look at those protocols and that way you'd
12 have some sort of baseline at least to start looking at
13 some systems that may be potentially beneficial for naming
14 things. The real question is the resources that you have
15 to put into this are quite limited, and that was one of the
16 reasons that we thought that would be a good approach.

17 DR. GROSS: Jackie.

18 DR. GARDNER: Along those lines, something that
19 Brian started with today about the gold standard, I think
20 at an absolute minimum -- I'm left at the end of all of
21 this discussion in not really knowing which things are
22 serious, what is the gold standard, which confusions have
23 resulted in harm as opposed to confusion, and it's
24 something that I know PhRMA raises all the time. Is there
25 a risk here?

1 So I would like to see some targeted work done
2 both in-house and maybe under an RFA about looking at some
3 of the things we've heard about. We heard that the USP
4 gold standard combines both things that have been known to
5 cause harm and things that have been just reported and
6 we're not sure or things that were caught, potential. We
7 heard from Jerry I think that it isn't exactly -- I want to
8 paraphrase, but tell me if I misunderstood what you said.
9 They don't know exactly which of the things they stopped --
10 they don't have good numbers or a clarification of which
11 things caused harm that were let go through.

12 So I guess if we could begin to clarify those
13 things as a baseline, there may be patterns buried in there
14 that would help to then direct some of the other work. It
15 may be only things that have four strings are the serious
16 ones. I don't know. But I don't feel that we have that
17 foundation to begin with about what is really potentially
18 harmful.

19 DR. GROSS: Any other comments? Arthur.

20 DR. LEVIN: I just want to caution that today's
21 near miss is tomorrow's error. So I'm cautious -- and I
22 think we were in the IOM -- about the relative weighting of
23 things that actually cause harm and things that don't. I
24 think they are different, but just because something gets
25 caught doesn't mean tomorrow it will get caught.

1 I think the problem with the gold standard,
2 with all due respect to my friend Mike, is that by relying
3 on voluntary reporting, our n's are always far from what we
4 would like them to be and to give us all of the information
5 we should have. This is not a plea for mandatory
6 reporting. I'm just saying it's a fact of life that the
7 voluntary reporting systems have not been nearly as
8 productive as we would have hoped they would be, and I
9 don't know how to address that.

10 DR. COHEN: You mean in producing numbers.

11 DR. LEVIN: Yes, in producing numbers.

12 DR. GROSS: Brian.

13 DR. STROM: I certainly agree. I think the
14 bigger problem with the spontaneous reporting system, as
15 was described before, much more than the sample size is the
16 selectivity, that you don't know what you're missing and
17 undoubtedly you're missing most of it. Overwhelmingly
18 you're missing most of it. So I'm very, very nervous about
19 using that as a gold standard for that reason.

20 On the other hand, I certainly agree that near
21 misses could well be important later, but it depends on how
22 you define them. For example, direct observation. People
23 look at these vast numbers of medication errors. Well,
24 some of those medication errors, a large number of them,
25 are things like getting a drug -- if you do direct

1 observation in the hospital, they list as a medication
2 error getting a drug 15 minutes late. I'm not worried
3 about that as a near miss, and that's not going to be a
4 disaster later for most drugs. So it is still important to
5 look at which of the medication errors matter and which are
6 the ones that don't.

7 DR. GROSS: I'm going to make a proposal here.
8 In the absence of enough data for us to make firm
9 recommendations, what would you think about recommending
10 sort of a modification of what Curt said, recommending that
11 the FDA meet with PhRMA and decide whether to maintain the
12 status quo until we have more experimental data to make
13 reasonable decisions on or whether a change should be made?

14 DR. DAY: Can you modify that to say PhRMA and
15 other groups? It's not just a PhRMA issue.

16 DR. GROSS: Yes, sure. Do you have a
17 particular group in mind?

18 DR. DAY: All the usual stakeholders are
19 potential candidates.

20 DR. GROSS: Okay.

21 Michael.

22 DR. COHEN: I think we ought to be very careful
23 with that, though, because I want to make sure that nobody
24 walks away with doing nothing. So that needs to be
25 qualified in some way. I think at least what's being done

1 now is absolutely preventing some potentially dangerous
2 names from getting on the market at all. So to do nothing
3 would be not the right way to go.

4 DR. GROSS: Wait a minute. Are you saying --

5 DR. COHEN: You said if things should stay the
6 same, status quo, or not.

7 DR. GROSS: Right.

8 DR. COHEN: So I say qualify it by saying you
9 don't want to go back to doing nothing.

10 DR. GROSS: Well, no, we don't. We're not
11 doing nothing now.

12 DR. COHEN: Correct, but the way it was stated
13 I think left the impression, at least for me, that one of
14 the decisions could be we would do nothing.

15 DR. GROSS: No, no. That wasn't what I meant
16 to imply.

17 Brian.

18 DR. STROM: Yes. I would suggest a
19 modification of it. I'm not comfortable with the way you
20 worded it in the sense of I don't see how FDA could meet
21 with PhRMA and decide whether or not to make a change,
22 again without any data. Without any data, I don't see
23 there's a reason to make a change. I would suggest that
24 FDA should be meeting with PhRMA and other relevant
25 stakeholders to decide what data are needed in order to

1 decide and design a plan to gather those data.

2 DR. FURBERG: And bring it back here.

3 DR. GROSS: That's fine.

4 Yes.

5 DR. CRAWFORD: Thanks. I would like to echo
6 what Brian just said because with the handwriting problems,
7 I had to look a few times.

8 (Laughter.)

9 DR. CRAWFORD: I do appreciate the analysis of
10 the processes presented both by the agency and the PhRMA
11 representative. What I didn't see on the FDA steps was
12 interaction with the sponsor. What I didn't see on the
13 sponsor's steps was interaction with the FDA. So I'm
14 wondering as part of the process, at some point if the
15 proposed nomenclature is problematic for FDA, is there a
16 step whereby the FDA interacts with the sponsors and is the
17 sponsor given the opportunity to present safety
18 information, a similar level of validation as you do with
19 all the other benefit-to-risk safety data presented in an
20 application. And if that is not done, then is it just a
21 second-choice name or what happens?

22 DR. GROSS: Jerry.

23 MR. PHILLIPS: The process is reconciled at the
24 end of the day when they're given a choice of either coming
25 back with another name or coming back with persuasive

1 evidence. So a sponsor has the ability to go out and do a
2 study or provide us the data to persuade us to change our
3 opinion. So the sponsor always has that ability to
4 persuade us to change our mind or to submit another name
5 for review.

6 DR. FURBERG: But, Jerry, before you get to
7 that stage, before you reject it, you need to sit down
8 before the name is submitted almost to agree on the plan
9 how you find out about this name confusion.

10 DR. GROSS: Yes. I think we could spend the
11 rest of the day and the week debating this issue, and the
12 reason we're debating is because we don't have the data we
13 need to make a reasonable recommendation.

14 So, Brian, do you want to restate your version
15 of everybody else's version, if you can remember?

16 (Laughter.)

17 DR. STROM: I guess my recommendation would be
18 that the current process not be changed on both sides, the
19 FDA or industry, absent data to the contrary, but that
20 we're not affirming that it is the correct process. Our
21 recommendation is that PhRMA, FDA, and all the relevant
22 stakeholders meet to discuss what data are needed in order
23 to, in fact, find out which approaches are correct and to
24 develop a mechanism for generating those data.

25 DR. GROSS: Okay. I hope nobody wants to amend

1 that.

2 (Laughter.)

3 DR. FURBERG: And bring it back here.

4 DR. GROSS: And bring it back here. Accepted.

5 All in favor, raise your hands, please.

6 (A show of hands.)

7 DR. GROSS: Thank you. That was a tough one.

8 The next one hopefully will be a little bit
9 easier. Are there any other methods that were not
10 discussed today that you think should be considered? Ruth?

11 DR. DAY: I'd like to suggest a method which is
12 quick, easy, cheap, and I think very valuable. It is
13 pronunciation screening in a systematic way. A lot of the
14 methods we've heard about today assume that a drug name has
15 a pronunciation. In fact, drug names often have
16 alternative pronunciations. We've heard today quinine,
17 quinine, quinine. We heard about Novicar, a made-up name.
18 It could also be Novicar. It could be a lot different
19 things. And does it matter? As the old song said, you say
20 Arava, I say Arava, but it doesn't make any difference
21 because we understand each other. That's a case where
22 perhaps it doesn't make a difference.

23 However, there are many cases where the
24 pronunciations that people give, when they first see a drug
25 name, are wildly different. So for amoxicillin you can get

1 amoxicillin. For clonazepam, you can get clonazepam,
2 clonazepam, clonazepam, clonazepam, et cetera. You can get
3 wild variations. So how do we know what the effective
4 pairs are to be worrying about in the first place.

5 So I'm concerned that the horse has gotten out
6 of the barn in a lot of these methods before the
7 appropriate phonetic cart has been attached. We don't know
8 then how --

9 DR. GROSS: Or that there are a lot of other
10 horses in the barn that we haven't seen yet.

11 (Laughter.)

12 DR. DAY: Not only are there other horses in
13 the barn, but we don't know which ones to be comparing. So
14 this can account for the incidence of both false positives
15 and false negatives. So we may be identifying "problem"
16 pairs by linguistic methods, where in fact psycholinguistic
17 methods where people would pronounce in advance would say,
18 no, people aren't going to be confusing those. Also, false
19 negatives where we think a pair is okay, but in fact, the
20 way people pronounce them would make it not an okay pair.

21 So a very simple task. A person sees a drug
22 name and says it out loud. Of course, you have a bunch of
23 different ones that you present. The main dependent
24 variable is agreement and the different pronunciations that
25 are given, and I'll come back to that in a moment. Also

1 speed of naming and the number of attempts to repronounce
2 and change one's mind about how it's said. So on the
3 agreement side, a given drug name -- does it only have one
4 pronunciation, and does everybody agree? That would be
5 great. Go ahead. But if it has multiple ones, what is the
6 probability of each one? So if it has two, but one is 95
7 percent and one is 5 percent, that's different from if you
8 have a 40/40 and then some dribbling off. So the overall
9 frequency distribution of pronunciations can be very
10 informative.

11 Once you have this set of data, you can then
12 look at the effects on both other cognitive tasks and on
13 behavior. For cognitive tasks, free recall. What were the
14 names of the drugs you just saw? Can people even say them
15 or remember them? Or give a recognition task. Show them
16 one at a time and say is this one of the drugs you just saw
17 or not, and then you can put in potential confusable pairs
18 and so forth.

19 So very quickly, the advantages and
20 disadvantages of this very quick little thing are the
21 following. The advantages are it can be very quick. You
22 can do an effective experiment or test in even 5 to 10
23 minutes, depending upon what you include in it and so on.
24 It's easy to do. It's inexpensive. The data are
25 quantitative. They are easy to replicate. The data are

1 objective. It's easy to understand the results. It's easy
2 to apply them in a variety of ways, and this approach may
3 well reduce the noise in all the data of all these other
4 methods. So when one of the wonderful linguistic analyses
5 that makes great sense from a linguistics and computational
6 standpoint does not identify or has some kind of problem,
7 it might be because of pronunciation alternatives.

8 Also, with the outcomes of these studies, we
9 can determine pairs are then likely to be confusable, and
10 the probable pairs or likely pairs are likely to change
11 relative to what we have now. And building on something
12 that Bruce Lambert said, this is also a way to evaluate a
13 single drug name before you start looking at any pairs.

14 Of course, there are limitations. Every method
15 has limitations. It only is addressing the sound-alike
16 problem. It cannot stand alone, obviously. And it's only
17 really for initial screening. But it could be used later
18 on as well as new products start coming on the market and
19 maybe they come in through some route and they're there so
20 that a sponsor could launch a risk management approach
21 based on something that happened. So it could be a TV ad.
22 I say Arava, you say Arava, but together we agree that it
23 works. I don't know. Whatever it would be. But some kind
24 of approach could be taken then to handle things that come
25 up.

1 On the sponsor's side, you can then reduce that
2 tremendously long list of 100 to 200 to 300 names that you
3 generate right away by looking at the pronunciation data in
4 a systematic way, not in expert groups sitting around and
5 doing it because I think we need to have a variety of
6 different participants in such tasks from the health care
7 professionals, the doctors, pharmacists, nurses, and the
8 lay public, the patients and the caregivers and so on, to
9 see the variety of namings that would happen.

10 On the linguistic models, they could then
11 perhaps start with more realistic phonetic transcriptions,
12 as Dr. Dorr admitted this morning or acknowledged, but also
13 they might discover new variables that need to be taken
14 into account. I didn't hear anything today about analyses
15 about syllabicity. How many syllables and where are the
16 syllables segmented and the stress and intonation contours
17 of how you say something? So the stressed and louder and
18 higher-pitched syllable is then the one perhaps going to be
19 more likely to be confused with other things.

20 For regulators, the advantages of having
21 something like this are that they could replicate using the
22 exact same methods within one day on these things, and they
23 could then have standardized methods across all of those
24 people who want to do some kind of testing.

25 So, in conclusion, whether there is a screening

1 test or not for pronunciation or pronounceability, it is an
2 essential ingredient in all this and could be responsible
3 for some of the problems across the methods.

4 DR. GROSS: Ruth, thank you very much. We
5 expect to see the results of your study published in a
6 peer-reviewed journal soon.

7 (Laughter.)

8 DR. GROSS: Yes, Lou.

9 DR. MORRIS: Yes.

10 DR. GROSS: I think it was a very good
11 suggestion, Ruth.

12 DR. MORRIS: I'm not totally comfortable that
13 we really understand the root cause of sound-alike/look-
14 alike problems. We're making an assumption that there's a
15 problem in the communication between the doctor and the
16 pharmacist per se.

17 I was struck with something Jerry presented
18 that there are actually a lot of problems with doctors
19 writing the wrong name, and I think there may be memory
20 retrieval problems that doctors have recalling the wrong
21 name. I guess what I'm suggesting is as part of this
22 research that we're suggesting, as we understand these root
23 causes better, there may need to be different methodologies
24 in the future and that we should not make the assumption
25 that we really understand what's causing these problems.

1 DR. GROSS: Any other comments? If not, we'll
2 draw number 3 to a close. Okay, Brian. Robyn, do you want
3 to go first?

4 MS. SHAPIRO: I just want to say that I agree
5 and that the first thing I said this morning I feel no
6 better about at the end of the day, and that is, that we're
7 accepting an assumption about cause and effect that I don't
8 feel comfortable that we can prove. Until we have our arms
9 around that better, I don't think we could possibly answer,
10 for example, question 5.

11 DR. GROSS: Well, you're going to get the last
12 word and create a new question that we'll have to answer.

13 Brian.

14 DR. STROM: Three comments. One is as one
15 additional thing I think we should do and which I think
16 very much follows up on the comments that have just been
17 made is the root cause analyses of the drugs that got into
18 trouble with names even after the current process is over,
19 as was suggested before.

20 Second is a caveat. There's been a lot of
21 discussion about computerized order entry as the solution.
22 We actually have data we haven't published yet of enormous
23 numbers of errors introduced by computer order entry. So
24 it is very far from a panacea. It solves the handwriting
25 problem, but it introduces many, many other kind of

1 problems. So people should just be careful.

2 Third -- and this is in some ways is the
3 opposite of Ruth's suggestion, which was obviously very
4 well thought out and thought through, and where this is
5 sort of seat of the pants, but it never stopped me from
6 talking anyway. I wonder if you could take advantage --
7 this is not screening before marketing but after marketing,
8 perhaps as part of risk management programs, perhaps just
9 from a validation point of view -- using databases. For
10 example, Avandia/Coumadin. One of the key questions that
11 we've been struggling with today is how common are these
12 problems. How much of a problem are they really? How many
13 times do we see diabetics who get a single prescription of
14 Coumadin in a database on the market or using claims data?
15 Or how often do you have somebody who doesn't have
16 diabetes, who is on no other diabetes drugs, who's on
17 longstanding Coumadin, who gets a single prescription for
18 Avandia? Those kinds of analyses would be easy to do and,
19 in selected situations like that, could be used as a gold
20 standard to try validate the kind of things that we've been
21 talking about. It wouldn't work in many situations, but it
22 would work in one like that.

23 DR. GROSS: Thank you all very much. We're
24 through the first three questions. We'll reconvene at 3:15
25 to do the last two questions, plus a question yet-created

1 by Robyn.

2 (Recess.)

3 DR. GROSS: Thank you all. We're a few minutes
4 late in getting started. The weather is approaching, so
5 why don't we reconvene and let's begin with question 4.

6 I will read question 4 to you. Under what
7 circumstances should a field test in a simulated
8 prescribing environment be recommended? Is any one method
9 alone sufficient as a screening tool, or should a
10 combination of methods routinely be employed, such as
11 behavioral testing and orthographic/phonographic testing?

12 We actually discussed much of this question
13 previously. Does anybody have any additional comments that
14 they want to make on this? Brian. I never would have
15 guessed.

16 (Laughter.)

17 DR. STROM: I just want to go one step further
18 and agree with what Mike was saying that I think the field
19 test is an enormously useful idea but should not be
20 required yet and should not be uniform. I think it needs
21 to be evaluated and tested. To me I think it is probably
22 the gold standard that should be used in evaluating the
23 others and ultimately will be too impractical and too
24 expensive to be used uniformly. So the answer to the
25 question of under what situation should a field test be

1 done, I would say as part of validation efforts.

2 DR. GROSS: Thank you.

3 Eric.

4 DR. HOLMBOE: The only other thing I would add
5 is I know that the FDA is currently doing something along
6 those lines. It's listed as number 3.

7 I had some concerns about that just because of
8 the numbers of people involved, the fact that there may be
9 a bias there to begin with because you're intra-agency. So
10 if you're going to continue that, I'd just really encourage
11 you to look at that very carefully given you have a small
12 n, and it gets back to Dr. Lambert's point that if you have
13 a low frequency of events for certain drugs and you're
14 dealing with only a small number of physicians
15 participating, you might get into trouble.

16 DR. GROSS: Anybody else have any comments?

17 DR. MORRIS: Yes, just definitional. When I
18 think of a field test, I think of a very, very big sample,
19 but if you mean a simulated environment, that's not a -- as
20 long as that's not ruled out, small samples of 50 or 100
21 pharmacists or doctors is reasonable and I think gives some
22 sense of data, not just qualitative information. I would
23 encourage that, but I agree, if we get into large amounts
24 of money, then we're not there yet.

25 DR. GROSS: So there is a definitional problem

1 in what a field test means for the first part of the
2 question.

3 For the second part of the question, from the
4 earlier discussion I sense that the committee would agree a
5 combination of methods, but it's hard for us at this point
6 to define what should be in the combination. Is that fair
7 enough? Okay.

8 Number 5. Yes, Lou.

9 DR. MORRIS: I'm pretty comfortable even at
10 this point in saying that some combination of methods is
11 going to be necessary. The idea that any single method is
12 sufficient, given that we don't even know what the problem
13 is -- I'm pretty comfortable that we're going to need a
14 multi-factorial approach.

15 DR. GROSS: Yes, I think that's certainly the
16 sense of the committee. Does anybody disagree with that?

17 (No response.)

18 DR. GROSS: Okay, fine.

19 Number 5. Describe the circumstances, if any,
20 when it would be appropriate to approve a proprietary drug
21 name. And I'll add for clarification that may cause some
22 confusion, but it should be added "with a risk management
23 program." Is that paraphrasing it right, Paul?

24 DR. SELIGMAN: Yes.

25 DR. GROSS: Comments? Arthur.

1 DR. LEVIN: When would that occur? Only if
2 there was a breakthrough drug or something like that with
3 the company refusing to -- I mean, you guys have the last
4 word. Right? I'm just trying to sort of figure out when
5 would that happen.

6 MR. PHILLIPS: There have been occasions where
7 we reached an approval stage. Let's just say that we get
8 to the final minute of an approval and we realize that we
9 observe something now that we didn't think about. So we
10 don't want to hold up the approval. We're not 100 percent
11 sure that this error is going to occur. We have some
12 doubts and the sponsor is willing to undergo a risk
13 management program to address that concern, whatever that
14 is. It is definitely associated with the name. So it may
15 be that you have to do some extensive monitoring. It may
16 have to do with setting up a surveillance system,
17 educational campaigns, et cetera, anything that is a
18 component of a risk management plan.

19 DR. GROSS: But wouldn't this be a place where
20 you might want to do field testing to decide whether or not
21 this was going to be an issue or not and then make a
22 decision?

23 MR. PHILLIPS: Well, we would have put it
24 through our analysis at FDA and maybe, one, there may be a
25 difference of opinion internally at the FDA that might say

1 yes, we see your point, but we want to go ahead and issue
2 the approval with a risk management plan. So maybe DMETS
3 had a recommendation. The office, on the final approval,
4 decides to go ahead and let it go with a risk management
5 plan. So FDA has agreed to do this.

6 DR. GROSS: So it would be a post-approval --

7 MR. PHILLIPS: It's a pre-marketing agreement
8 to institute a risk management plan post-marketing.

9 DR. GROSS: Curt.

10 DR. FURBERG: But how do we know that that risk
11 management plan will work? In order to document its value,
12 you have to spend a lot of time figuring out. So I'm not
13 sure this is the solution. It makes me very nervous.

14 The only situation I can see is if you have two
15 approved drugs and you find out after the fact that you
16 have a problem. Before you would remove a name or change a
17 name, you can say, well, the option is to come up with a
18 risk management. That's the only situation I can think of.

19 DR. GROSS: Eric.

20 DR. HOLMBOE: That's exactly what I was going
21 to say. Just, I want to second what Dr. Furberg said.

22 DR. GROSS: Okay.

23 Michael.

24 DR. COHEN: Perhaps this is where the
25 laboratory and the model pharmacy might come in where they

1 could actually test in a controlled environment whether or
2 not various measures that are being suggested -- other than
3 the monitoring. For example, we've heard about tall man
4 letters that help to differentiate one mark from another by
5 enhancing the unique letter characters or the background of
6 those unique letter characters, for example. That might
7 work. There's some evidence that it does from Dr. Grasha's
8 studies. There are other things that could be done.
9 Another one was pre-market advertising, "coming soon" to
10 help educate practitioners. So we just don't know how
11 effective they are necessarily. That's the problem, but I
12 could see where you could have a risk management plan
13 approved for these rare cases, but exactly what they should
14 be I guess we don't know at this point.

15 DR. GROSS: Yes. Jerry described some cases.
16 Does anybody here have some other circumstances where they
17 think this might need to be invoked? Lou.

18 DR. MORRIS: I was struck this morning, Jerry,
19 when you said you reject a third of the names and then
20 there's another class of drugs that you feel uncomfortable
21 about. What percent do you actually feel comfortable
22 about?

23 (Laughter.)

24 MR. PHILLIPS: No. I wouldn't categorize it
25 that way. Out of that third, there might be some where we

1 have a difference of opinion on the objections.

2 DR. MORRIS: Okay. So what percent is it
3 unanimous? Let me do it that way.

4 MR. PHILLIPS: We still reject a third. Okay?

5 DR. MORRIS: Yes.

6 MR. PHILLIPS: And for the most part, I would
7 say probably 90 to 95 percent of those rejections are
8 accepted by the reviewing divisions and are relayed back to
9 the sponsors. The sponsor still can argue with us about
10 whether we are correct or not. So you get into a
11 discussion with the sponsors which may at this point bring
12 up a risk management plan as a means to manage a perceived
13 risk.

14 DR. MORRIS: Okay. So you're saying of the
15 third that you would have rejected a small percentage, they
16 come back and propose what if we do this risk management
17 program. So that's the circumstances.

18 MR. PHILLIPS: That's the circumstances
19 behind --

20 DR. MORRIS: It brings you up to a comfort
21 level that you feel that it would be safe for the drug to
22 be in the marketplace.

23 Does the risk management plan you're proposing
24 also have an evaluation component or just have an
25 evaluation component?

1 MR. PHILLIPS: Oftentimes we're very interested
2 in learning the outcomes and whether they're effective or
3 not. So that is discussed with the sponsor.

4 DR. GROSS: Can you give us any examples,
5 Jerry, where this has occurred in the past with approved
6 drugs? Or is this a theoretical thing?

7 MR. PHILLIPS: It's not theoretical, but I'm
8 not sure I feel comfortable talking about it right now.

9 DR. GROSS: Okay, fine. I understand.
10 Brian.

11 DR. STROM: I want to go back. I strongly
12 agree with Curt's comment, and I think it's important we
13 keep focused on that. The purpose of risk management plans
14 normally is to say a drug that has real benefit on one side
15 but it has a risk, you try to reduce the risk or increase
16 the benefit because the risk/benefit balance is a close
17 call and a risk management plan would improve that close
18 call.

19 We're not talking about a drug here. We're
20 talking about a drug name. There's no public health
21 benefit in having a drug name available versus another drug
22 name. So to me the only reason one would ever do that
23 would be exactly as Curt said, if in fact the drug is
24 already on the market and there are side effects from a
25 patient point of view of removing a drug name that is

1 already available.

2 I think the situations Jerry is describing I
3 see as something different. I see it as a situation where
4 you don't know as an agency that you want to reject it.
5 There's not adequate data and you've decided you're going
6 to generate some of the data after marketing instead of
7 before marketing in order to get the answer. If there were
8 better methods before marketing, simulations or laboratory
9 or otherwise, you would generate those data before
10 marketing.

11 But that's different from saying you have a
12 concern about a drug name. I don't see why in the world
13 from a public health point of view pre-marketing you would
14 ever allow that drug name on the market. There's no
15 positive to counterbalance the risk.

16 DR. GROSS: Let me ask the committee. Can we
17 specifically answer this question or not? Can we describe
18 circumstances in which this would occur? Jeff.

19 MR. BLOOM: I seem to recall in reading the
20 review materials -- and I would certainly agree with it --
21 that the one circumstance that I could see where it could
22 occur if there is a breakthrough drug that is meeting an
23 unmet need where there is not any existing therapy for a
24 serious or life-threatening condition. That's the only
25 circumstance that comes to mind.

1 DR. STROM: But you change the name. You could
2 still have the drug available.

3 MR. BLOOM: Yes. Absolutely. I agree with
4 that, but I wouldn't want it to be held up because of a
5 drug name, of course.

6 DR. GROSS: Michael.

7 DR. COHEN: I just have to say I think it's not
8 so easy to say just change the name. There's a lot that
9 goes behind that. We've heard that today too. And it
10 might delay the drug by three months or six months or maybe
11 even longer for all we know. I don't know everything the
12 trademark attorneys know, but I'm sure they might run into
13 situations like that. So I could see a public health
14 benefit of an occasional use, a rare use of a risk
15 management program.

16 DR. GROSS: Jerry, do you want to help us out
17 on this? Give us some examples of circumstances.

18 MR. PHILLIPS: I'm going to give you another
19 example. I'm not going to name the drug product, but the
20 circumstance was a similarity with a trademark in which the
21 product was no longer marketed in the United States, but
22 was widely available in reference textbooks and in the
23 literature. So within the practice setting, there was a
24 wide recognition of this name, although it wasn't
25 available. So there was an argument made. The risk

1 management plan included going and cleaning up those
2 reference texts. It's hard to change reference textbooks
3 that sit on our shelves.

4 (Laughter.)

5 MR. PHILLIPS: So it's an interesting argument.
6 This is an example of how do you weigh the risk and the
7 benefits.

8 DR. GROSS: You mean you're good, but you're
9 not God.

10 (Laughter.)

11 DR. GROSS: Ruth.

12 DR. DAY: As I understand it, the FDA
13 encourages sponsors to have backup names, and if the backup
14 names went through all of the same processes that the lead
15 name did, then we wouldn't have to wait for 3 to 6 months
16 to switch. We'd have a backup name which was as good in
17 many respects. Right?

18 DR. STROM: Plus developing a risk management
19 plan probably wouldn't take any shorter time than testing a
20 new name.

21 DR. GROSS: I'm getting the sense from the
22 committee that it's hard to commit on this and maybe we
23 should just say there may be circumstances in which this
24 arises. It's hard for us to define them and if you feel
25 you need to have a risk management plan and you have to go

1 through with the name and there's no possibility of
2 changing the name at that point, then you have to do it.

3 MR. PHILLIPS: I think there's always the
4 possibility of changing the name or approving the
5 application without a name. But that presents its own
6 problems for the sponsor for marketing the drug product.

7 DR. GROSS: So how does the committee want to
8 deal with this question? How do you want to answer the
9 question? Jackie.

10 DR. GARDNER: Perhaps in two parts. With
11 respect to a post-approval situation that's been described
12 here, I think that, as Brian defined it and Curt, if you're
13 in a post-marketing situation, then we clearly could see a
14 pause, a hiatus, while a risk management program is being
15 developed before firm action is taken and, as Michael said,
16 evaluate alternatives for the risk management program.

17 So in an after-market situation, a post-
18 marketing situation, I think there are many circumstances
19 in which it would be appropriate. Pre-marketing I have
20 less confidence.

21 DR. GROSS: Jerry, does that answer the
22 question? Paul?

23 MR. PHILLIPS: Yes.

24 DR. SELIGMAN: Yes.

25 DR. GROSS: As best we can. It's tough.

1 Robyn, question number 6.

2 MS. SHAPIRO: Okay. Here's question number 6.

3 You're not going to like it.

4 To develop an approach to address the risk of
5 harm related to look-alike/sound-alike drugs, is it
6 possible -- and if so, is it advisable -- so two parts --
7 to pursue research or acquire data that will more precisely
8 identify causative factors in such harm? That's my
9 question.

10 VOICES: Yes.

11 MS. SHAPIRO: Then why aren't we talking about
12 doing that before we get to all these other questions?

13 DR. STROM: We are.

14 MS. SHAPIRO: Did that whole proposal include
15 collecting that kind of data?

16 DR. STROM: Yes.

17 MS. SHAPIRO: Wonderful, great. I'm happy now.

18 DR. GROSS: Lou.

19 DR. MORRIS: I disagree. I think what you were
20 talking about, Brian, was validation processes.

21 MS. SHAPIRO: That's what I thought.

22 DR. MORRIS: And what Robyn is saying is
23 causative factors for medication errors per se at a much
24 more specific level, and I'm with her. I think that that's
25 another research agenda that we should recommend.

1 MS. SHAPIRO: I don't know, although Curt is
2 helping me along with my thinking here, how you can do any
3 of this without doing that.

4 DR. GROSS: Ruth.

5 DR. DAY: Michael Cohen gave us an example,
6 Robyn, which I think might help out, and that is that there
7 were cases where there were two drug names on the market
8 and there were a lot of errors being tracked. One drug
9 name was withdrawn and a new name was given and there were
10 no longer those kinds of errors.

11 MS. SHAPIRO: That's an example. That's great.

12 DR. DAY: It's not the whole answer. It's a
13 tiny part of it, but it can't be overlooked.

14 MS. SHAPIRO: That's why my question
15 acknowledges that closely related names or names that sound
16 alike are related to harm. I think that we can assume
17 that. It's a factor. But if we want to do a risk
18 management approach --

19 DR. GROSS: I thought that was your question,
20 what you're assuming.

21 MS. SHAPIRO: No. Part of the question is to
22 develop an approach to address the risk of harm related to
23 look-alike/sound-alike drugs. The assumption is that there
24 is some. Is it possible, and if so, advisable, to pursue
25 research that will more precisely identify causative

1 factors in such harm, that is, in harm that is related to
2 look-alike/sound-alike drugs? So the assumption is that
3 there's some and the desire is to drill deeper to find out,
4 well, does that vary depending on whether we're looking at
5 handwritten as opposed to verbal, does that vary depending
6 on whether we have vast differences in dosages or
7 administration routes. Let's get more precise in the
8 factors involved so that we can be better in the risk
9 management approach.

10 DR. GROSS: Yes, I think some of that has been
11 done and a lot is still in progress.

12 MS. SHAPIRO: Good.

13 DR. GROSS: Arthur?

14 DR. LEVIN: It seems to me that the
15 presentations we had on labs offer an opportunity to get at
16 that because in a controlled situation, you can vary the
17 variables and get a better understanding of the things
18 you're asking about probably more quickly and less
19 expensively than sort of going out and doing RFAs. I don't
20 know. It might be a chance to have a down and dirty
21 opportunity to get a little better handle on how all the
22 variables play out in this.

23 MS. SHAPIRO: In a pharmacy, but I've seen a
24 lot of errors that don't happen in a pharmacy that are
25 terrible.

1 DR. GROSS: Brian.

2 DR. STROM: Yes. You're broadening the
3 question to medication errors in general which clearly is
4 appropriate and needs to be done, but realize it's a whole
5 other field. The focus of today was on the name because
6 that's what FDA regulates. But ARC, for example, has a
7 close to \$60 million a year budget studying patient safety
8 issues. A substantial amount of that focuses on medication
9 errors, and there's a lot of research underway. For
10 example, at one of the centers for patient safety, we have
11 studies underway looking at sleep issues, looking at things
12 that determine, in an in-hospital setting looking at
13 patients making errors from an adherence point of view.
14 There's lots and lots of low-hanging fruit about why is it
15 that there are medication errors. It's very clear that
16 name confusion is a small part of it.

17 MS. SHAPIRO: But I think that I'm looking at a
18 subset of that universe, and that is, if we take only the
19 subset of look-alike/sound-alike, are there other factors?
20 Again, if our task is to have a risk management approach
21 that makes sense or, even before that, to determine whether
22 we need one, then take that subset and look at other things
23 so that we can be more sophisticated in making
24 recommendations.

25 DR. GROSS: Louis.

1 DR. MORRIS: Again, I'm with Robyn. Just take
2 a cognitive psychology look at this. Is it a pattern
3 recognition problem, a pharmacist not looking long enough
4 and hard enough, and if they did, would they then see it?
5 Or is it not just the way the letters are formed, but is it
6 some other aspect of the way they search their memory?
7 There are lots of very specific issues that could help us
8 understand the problem better. I asked Mike before. There
9 are lots of problems here. We don't know that we know them
10 all, and if we did know them, we don't know how much they
11 contribute. So I think if we just stepped back and said,
12 okay, what is the specific problem and understood that
13 better, I'd be a lot more comfortable.

14 DR. GROSS: I think these comments are very
15 important. I think they're a little bit beyond the scope
16 of the questions. One of the panelists brought up to me,
17 as far as question number 2 is concerned, how will we find
18 out what's been decided? Can this advisory committee get a
19 report back in three to six months as to what was decided
20 about what study methods will be used as a minimum
21 combination, and how will the other study methods be
22 handled as far as proposals for future studies? What do
23 you think, Paul? Can we get an answer? Can you just give
24 us a follow-up in a few months as to what's going on?

25 DR. SELIGMAN: I'm happy to give you a follow-

1 up.

2 The challenge for us always is how to develop
3 good practice in the context of an evolving science where
4 there are people who are being injured or harmed and the
5 degree to which we can foster best practice as we are
6 developing the best science. This, of course, is the
7 challenge to us. We're certainly happy to do our best to
8 look at the data that are out there. We've done that in
9 large measure already. The challenge that we face is, at
10 least at this point in time, how to create practices -- we
11 think internally within our own organization, we are doing
12 I think the best practice we can in involving experts,
13 using computational software, engaging in simulations to
14 try to best understand where problems might occur with
15 names, drawing on the best that's available within the
16 current literature.

17 As I indicated, our ultimate goal is to try to,
18 to the degree we can, level the playing field and ensure
19 that industry is taking these approaches and looking at
20 trade names beyond just their commercial value and trade
21 name, but also to incorporate principles of safety and
22 consideration of safety in those processes. At the end of
23 the day, can we create a guidance based on what we know
24 about the data to date in a way that will at least foster
25 and improve the way all sponsors look at names that they

1 submit to us at the agency for review and create processes
2 that allow some consistency so that sponsors will know the
3 basis for which we make decisions about either accepting or
4 rejecting such names.

5 DR. GROSS: Are there any other issues you
6 wanted us to deal with today?

7 DR. SELIGMAN: Not that I'm aware of, no.

8 DR. GROSS: Brian.

9 DR. STROM: Just in comment to one of the
10 things you're saying, Paul. I'm interested in the rest of
11 the committee's comments on this, but my sense is it's
12 premature to issue a guidance because we don't know what
13 the best practices are is what I was hearing. I don't know
14 if other people feel the same, or maybe I'm
15 misunderstanding what a guidance is.

16 DR. GROSS: Yes. I guess, as happens to much
17 in medicine where there aren't randomized controlled
18 trials, decisions still have to be made. My sense is
19 that's the position that they're in. Given what we know
20 now, what are the recommendations they can make.

21 DR. STROM: Absolutely, but that's different
22 from putting it in a guidance which I would think should be
23 data-based. That's what I'm saying. I'm not saying you
24 should change. I think doing what you're doing is on
25 target. The new advances you're incorporating, I think all

1 that makes enormous sense. I think that's different from
2 codifying it absent the data to know it's the correct
3 thing.

4 DR. GROSS: Michael.

5 DR. COHEN: Peter, we spoke before about having
6 FDA get together with PhRMA and other stakeholders. Could
7 we set something now or at least set an expectation that
8 that take place within the next 3 to 6 months and that
9 there be a report back to this committee by perhaps the
10 next 9 to 12 months at least?

11 DR. GROSS: I thought Paul said that he would
12 do that.

13 DR. SELIGMAN: I guess the question is what's
14 the nature of the feedback that you're looking for in this
15 report. What are the questions that you're asking us to
16 answer in getting together with PhRMA and other
17 stakeholders? What are your expectations in terms of what
18 we can produce in the next 6 months?

19 DR. GROSS: Arthur.

20 DR. LEVIN: I would agree with Paul's confusion
21 about expectation because we've said get together, but
22 we've also said get together so that you can start planning
23 out the research agenda to move this along to a place where
24 we feel is evidence with which to go out with a guidance.
25 And that's going to take longer. I mean, just to know that

1 in 6 months you're going to get together with the
2 stakeholders, great, but it's not going to move this much
3 further. It's going to be more time. By saying this,
4 we're delaying the process, and that's just the reality.
5 We're not going to get a quick fix on this. The evidence
6 base does not yet exist to make us comfortable to set up
7 standards or criteria to form a guidance to give to
8 industry to say this is what we'd like you to follow, and
9 if you follow this, you'll be okay. We're not there yet
10 and it's going to take not 3 to 6 months, but probably at
11 least 12 to 24 months to get there.

12 DR. GROSS: Yes. My suspicion is to have an
13 adequate evidence base to make recommendations on where
14 each recommendation is based on good, solid scientific
15 evidence, it will take a few years. In the meantime, drugs
16 are still being approved. So some decision has to be made
17 as to what methods will be used to clear those drugs to
18 avoid confusion with other drugs. Again, we're in that
19 scientific limbo where we don't have the evidence to make
20 the kind of decisions we want to make but yet decisions
21 have to be made.

22 DR. SELIGMAN: I also struggle a little bit
23 with the kind of evidence that we would be looking for at
24 the end of the day and would be actually interested in
25 hearing from members of the panel as to what evidence we

1 might be looking for.

2 DR. MORRIS: But that's the purpose of this
3 process we're suggesting. Eventually FDA is going to call
4 for evidence in support of drug names, but we're saying we
5 don't know what that evidence should look like. So as the
6 first step in the process, because we don't know which of
7 these methodologies or any other methodology might be the
8 best evidence or combination of evidence, why not start a
9 public process with PhRMA to decide, based on validation,
10 what that evidence should be? What we're asking for is,
11 rather than it just being a consensus process, that there
12 actually be science underlying the type of evidence that
13 you will eventually get and you go through this process of
14 learning about what's the most valid methods before you ask
15 for them.

16 DR. SELIGMAN: But I would argue that there's
17 science, for instance, behind the computational searches,
18 that these are indeed well-validated methods. Ultimately
19 at the end of the day, somebody is going to have to look at
20 that ranking of things that either look alike or sound
21 alike and make some decisions based on input from expert
22 panels which again I think can be constructed in a way that
23 are well defined even though there are I think some
24 significant issues regarding the validation of those.
25 Similarly, one can go through a process, as we do, of

1 written and verbal Rxs and define that process very
2 carefully.

3 But I guess we can do a lot of what I would
4 call sort of internal validation of these techniques. The
5 problem for us is how to externally validate them, to know
6 that information that is generated out of each one of these
7 components or the ultimate risk assessment, indeed, does
8 have its intended impact of essentially preventing a name
9 confusion.

10 DR. GROSS: Curt.

11 DR. FURBERG: My sense is that we have three
12 silos. We have the FDA addressing the problem. We have
13 the industry and then academia, and there's very poor
14 communication between the three groups. Even within a silo
15 there's a problem. You just heard about the pharmaceutical
16 industry, that some companies are doing a lot and others
17 are doing probably very little.

18 So I think what we need to do is to set up a
19 situation. We can have a dialogue about what is being done
20 right now and what are the lessons learned, what is working
21 and what is not working. So focus on two things: one, on
22 the knowledge we have and even take advantage of the FDA
23 database, the 100 cases disapproved. We can learn from it.
24 What are the patterns in that that we can learn from.

25 So that's what I think a meeting could do,

1 bring in the parties, have a good discussion about what we
2 know and what we have learned, some further analyses, and
3 then in addition, talk about the process. I'm not sure the
4 process is well defined. You get names submitted to you
5 and you review them, but maybe there should be something
6 happening earlier than that. Maybe they should come to you
7 and talk about this is how we're going to go about
8 evaluating name confusion, and you need to have some
9 guidance to them, what is it that they should do, what
10 would speed up the process and make it more acceptable to
11 you.

12 This lack of communication I find a little bit
13 troubling, and that's why I suggested just get people
14 together in a room and let them talk and you'll come up
15 with something. Based on that, you may be able to, on
16 existing evidence, come up with guidelines that could be
17 refined, and I'm sure there will be areas or gaps. The
18 other outcome would be even to learn what are the gaps and
19 see what is essential that we focus on in the future.

20 DR. GROSS: Brian.

21 DR. STROM: I still think the conversation is
22 necessary but not sufficient and you're not going to be
23 able to put people in the room together and have them come
24 up with a scientifically reasonable decision because
25 there's no data underlying it. We've had two of those

1 meetings. We've proven that.

2 I think, Paul, you talked about there's science
3 underlying the computerization. I think that's a perfect
4 model. That's analogous to there's science, physiology,
5 and preclinical data underlying why a drug might work and
6 be safe, but yet we test it in people to find out and drugs
7 don't survive their testing in people. The science that
8 exists now is process-based science. What isn't there is
9 outcomes-based science. There are lots of different ways
10 you could generate it ranging from looking at drug names
11 that failed in the past, looking at drug names that Jerry
12 has rejected that industry has passed, doing some of the
13 mock pharmacy or the laboratory kind of approaches.

14 We need outcomes-based data to validate what
15 works and what doesn't work because the chances are there's
16 a significant amount of what's being done now, which is
17 fine, and there's a significant amount of what's being done
18 now which is wasted effort. Get rid of the wasted effort.
19 Require the stuff that's fine and add other things that are
20 useful.

21 But you're not going to be able to know any of
22 that without looking at gold standards -- or at least
23 silver standards. There are no gold standards in the
24 field, but at least silver standards as opposed to the
25 fool's gold as the gold standard. We need to test all of

1 the methods that are now being used against some at least
2 silver standard or group of silver standards, given none of
3 them are gold standards. Until that's done, how can you
4 codify requirements for what should be best practices? We
5 don't know what the best practices are.

6 DR. FURBERG: Yes, but Brian, I don't think you
7 can make progress by having another session or two of show
8 and tell.

9 DR. STROM: I strongly agree.

10 DR. FURBERG: You need to get people together
11 and define the issues and, maybe with you as one of the
12 moderators, make sure that they stay on track and address
13 the real issues.

14 DR. STROM: I agree, but the issue of that
15 getting together isn't what's the best way to do it because
16 then we're just going to have another show and tell. The
17 purpose of the getting together is what is the research
18 that needs to be done and who's going to come up with the
19 money and who's going to fund it and what's the process and
20 ideally come up with a joint process that everyone will be
21 comfortable with which will validate or not the approaches
22 that --

23 DR. FURBERG: But I see that as step two, sort
24 of the future, what do you do. Right now, let's see what
25 we have.

1 DR. GROSS: I'm hearing two different things.
2 I'm hearing the science is insufficient to make a
3 recommendation, and I think everybody seems to agree with
4 that. But what's the corollary? The corollary status quo
5 or what does the group think?

6 DR. STROM: It's status quo until we generate
7 more science, and the priority should be in generating more
8 science.

9 MS. SHAPIRO: Outcomes-based.

10 DR. STROM: Outcomes-based, yes.

11 DR. GROSS: Lou.

12 DR. MORRIS: There's another thing that we can
13 recommend and that is that rather than being specific on
14 what to request from the industry, that FDA, as part of
15 this process, ask for some evidence from the industry at
16 their choosing and that part of this time that we're
17 spending validating, FDA can also be spending the time kind
18 of internally validating industry evidence, and that there
19 should be some requirement for some form of evidence. But
20 what form it should be ultimately again is like a year-and-
21 a-half out before we put a final guidance, but there will
22 be this evaluation period for gathering new data and
23 evaluating existing data that industry is already gathering
24 but not submitting.

25 DR. LEVIN: A couple of things. One is I'm

1 comforted by comments from Michael and others that things
2 are much better today than they were. By talking status
3 quo, it's not the worst possible scenario. This is an
4 issue. It's an issue people are concerned with, an issue
5 people are working on, and there's a lot of room to grow.
6 But things are being done.

7 I just want to sort of do a mea culpa from the
8 IOM Committee perspective, that when we set a goal of error
9 reduction and we tried to put some meat on the bones of the
10 To Err is Human report, we thought it was incumbent on us
11 to pick some concrete steps that could be taken right away.
12 I guess perhaps we were delusional in thinking that this
13 was a simple step, that we could suggest that it could
14 happen right away, which was to get rid of this issue of
15 sound-alike and look-alike drug names. Clearly, it is a
16 complex issue and not so easy to resolve. So I want to
17 sort of take partial responsibility for pushing this issue
18 forward in a way that I think did not fully anticipate the
19 difficulties in even something this well-focused.

20 I would again like to urge a reexamination of
21 where things went wrong with this process, in other words,
22 taking a look at where everything passed through the screen
23 and got out there and all hell broke loose, and what was
24 everybody thinking, both PhRMA and FDA, and maybe learning
25 from the mistakes and using that as sort of a down and

1 dirty way to get much more focus on where we need to be
2 looking.

3 The second thing I'd like to urge is the lab
4 approaches, again, being able to, I would suggest, produce
5 some very quick notions about a lot of things, including
6 your concerns about what are all these factors that
7 contribute, and if we can't weight them, how do we know how
8 to react to the problem.

9 DR. STROM: Peter, I had two related comments.
10 One is to clarify a point I made. When I say
11 status quo, I don't mean freeze in place. What's very
12 clear is FDA is doing a lot of neat stuff, and by status
13 quo, I mean keep doing that neat stuff and keep advancing
14 the science as you're doing and the public health will
15 improve accordingly. But don't put into codification
16 something until we know what's correct or not.

17 I think using lab approaches makes enormous
18 sense in validation, and I guess one of the things we
19 didn't talk about before, in talking about prioritizing
20 high-risk/low-risk drugs, is I would go to the high-risk
21 drugs to be the drugs that you use those lab approaches in
22 as part of those validation tests.

23 DR. GROSS: Michael Cohen, any comments?

24 DR. COHEN: No.

25 DR. GROSS: Stephanie.

1 DR. CRAWFORD: Thank you. Again, I'm piggy-
2 backing on Dr. Strom's comments. I applaud the efforts
3 that the FDA has done. I think the multi-faceted approach
4 is certainly a phenomenal step in the correct direction.

5 As I assimilate some of the comments that were
6 made by the speakers earlier this morning, something that
7 came up on more than one occasion was concern about was the
8 lack of transparency. So I think perhaps the agency needs
9 to better articulate to the audiences exactly how it is
10 determined which of the programs is used, what goes into
11 evaluating, exactly what processes are used because I think
12 that adds to the discomfort when it's not there, and also
13 perhaps some people think it's not comprehensive enough in
14 looking at all the alternatives. But otherwise I think
15 these steps are in the right direction.

16 DR. GROSS: Does anybody have any other
17 comments?

18 (No response.)

19 DR. GROSS: If not, then the meeting is
20 adjourned.

21 (Whereupon, at 3:58 p.m., the committee was
22 adjourned.)

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