

DEPARTMENT OF HEALTH AND HUMAN SERVICES
FOOD AND DRUG ADMINISTRATION
CENTER FOR DRUG EVALUATION AND RESEARCH

ADVISORY COMMITTEE FOR PHARMACEUTICAL SCIENCE
PHARMACOLOGY TOXICOLOGY SUBCOMMITTEE

Tuesday, June 10, 2003

8:30 a.m.

CDER Advisory Committee Conference Room
5630 Fishers Lane
Rockville, Maryland

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Andrew Brooks, Ph.D.
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William D. Pennie, Ph.D.
Kurt Jarnigan, Ph.D.
John Quackenbush, Ph.D.
William B. Mattes, Ph.D., DABT
Krishna Ghosh, Ph.D.

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David Jacobson-Kram, Ph.D., DABT
John Leighton, Ph.D.
Frank Sistare, Ph.D.
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1 P R O C E E D I N G S

2 Call to Order

3 DR. KAROL: Good morning, everybody. I
4 would like to call the meeting to order. My name
5 is Meryl Karol. I am from the University of
6 Pittsburgh and, since many of us are new to the
7 committee and the subcommittee, I would like to go
8 around the room and have everyone briefly introduce
9 themselves with their name and their affiliation.
10 We will start over there.

11 DR. LEIGHTON: My name is John Leighton.
12 I am a supervisory pharmacologist in the Division
13 of Oncology Drug Products. I am also the Associate
14 Director for Pharmacology for the Office of ODE-3.
15 I am also the co-chair with Frank for the
16 nonclinical pharmacogenomics subcommittee.

17 DR. SISTARE: I am Frank Sistare, with the
18 Office of Testing and Research in the Center for
19 Drug Evaluation and Research at the FDA.

20 DR. GOODMAN: Jay Goodman, Michigan State
21 University, Department of Pharmacology and
22 Toxicology.

23 DR. HARDISTY: Jerry Hardisty, from
24 Experimental Pathology Laboratories. I am a
25 veterinary pathologist.

1 DR. KAROL: As I said, I am Meryl Karol,
2 from the University of Pittsburgh, Department of
3 Environmental and Occupational Health.

4 DR. WATERS: Mike Waters, Assistant
5 Director for Database Development, National Center
6 for Toxicogenomics, NIEHS.

7 DR. ZACHAREWSKI: I am Tim Zacharewski. I
8 am in the Department of Biochemistry and Molecular
9 Biology in the National Food Safety and Toxicology
10 Center at Michigan State University.

11 DR. WOODCOCK: I am Janet Woodcock. I am
12 the Director of the Center for Drugs at the FDA.

13 DR. JACOBSON-KRAM: I am David
14 Jacobson-Kram. I am the Associate Director for
15 Pharm/Tox in the Office of New Drugs in CDER.

16 DR. WINKLE: I am Helen Winkle. I am the
17 Director, Office of Pharmaceutical Science in CDER.

18 DR. KAROL: Thank you very much. Now we
19 will have Kimberly tell us about the conflict of
20 interest.

21 Conflict of Interest

22 MS. TOPPER: The following announcement
23 addresses the issue of conflict of interest with
24 respect to this meeting and is made a part of the
25 record to preclude even the appearance of such at

1 the meeting.

2 The topics of this meeting are issues of
3 broad applicability. Unlike issues before a
4 committee in which a particular product is
5 discussed, issues of broader applicability involve
6 many industrial sponsors and academic institutions.

7 All special government employees have been
8 screened for their financial interests as they may
9 apply to the general topics at hand. Because they
10 have reported interests in pharmaceutical
11 companies, the Food and Drug Administration has
12 granted general matters waivers to the following
13 SGEs which permits them to participate in these
14 discussions: Dr. Meryl H. Karol, Dr. Jerry F.
15 Hardisty, Dr. Michael Waters.

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

20 In addition, Drs. Andrew Brooks, Jay
21 Goodman and Timothy Zacharewski do not require
22 general matters waivers because they do not have
23 any personal or imputed financial interests in any
24 pharmaceutical firms.

25 Because general topics impact so many

1 institutions, it is not prudent to recite all
2 potential conflicts of interest as they apply to
3 each member and consultant. FDA acknowledges that
4 there may be potential conflicts of interest, but
5 because of the general nature of the discussions
6 before the committee these potential conflicts are
7 mitigated.

8 With respect to FDA's invited guests, Drs.
9 Krishna Ghosh and John Quackenbush report that they
10 do not have a financial interest in, or
11 professional relationship with any pharmaceutical
12 company.

13 Dr. Kurt Jarnigan reports being employed
14 full-time as Vice President, Biological Sciences
15 and Chemical Genomics at Iconix Pharmaceuticals.

16 Dr. William Mattes reports being employed
17 full-time by Pfizer, Inc.

18 William Pennie is employed full-time by
19 Pfizer, Inc. and holds stock in Astra Zeneca and
20 Pfizer.

21 Dr. Roger Ulrich reports full-time
22 employment at Merck Research Laboratories and
23 holding stock in Abbott Labs.

24 In the event that the discussions involve
25 any other products or firms not already on the

1 agenda for which FDA participants have a financial
2 interest, the participant's involvement and their
3 exclusion will be noted for the record.

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

9 DR. KAROL: Thank you, Kimberly. Now
10 Helen Winkle would like to welcome everyone.

11 Welcome

12 MS. WINKLE: Good morning, everyone. It
13 is my pleasure this morning to be able to welcome
14 each of you as a member of the Pharmaceutical
15 Toxicology Subcommittee.

16 This subcommittee, which is a part of the
17 Advisory Committee for Pharmaceutical Science, is
18 important to the Center in addressing a number of
19 questions and issues that come about due to the
20 regulation of pharmaceuticals. This is one of five
21 subcommittees of the advisory committee and really
22 each one of these subcommittees has been very
23 beneficial to us in helping to address various
24 issues and concerns that we have, and helping us
25 really develop the regulatory knowledge that is

1 necessary or the regulatory understanding that is
2 necessary to maintain a strong scientific
3 underpinning to our decision-making process. So,
4 it is a really important group.

5 This is the first time the subcommittee
6 has met. We look forward to a lot of interesting
7 discussion over the years. Again, as I said, there
8 is a lot that you all can contribute to us as we
9 grapple with our decision-making processes. I
10 appreciate all of your willingness to serve on this
11 subcommittee and I especially appreciate Meryl for
12 agreeing to chair this subcommittee for us. It is
13 a big job and it will take time, and I appreciate
14 her willingness to do that. I also want to thank
15 all of the folks in the Center that helped make
16 this subcommittee a reality. This includes Dr.
17 Jacobson-Kram, Dr. Bob Osterberg and Dr. Sistare.
18 So, again, welcome. We look forward to working
19 with you. Thanks.

20 DR. KAROL: Thanks very much, Helen. Now
21 the subcommittee is going to receive its charge and
22 this will be delivered to us by David
23 Jacobson-Kram.

24 Introduction to Meeting and Charge to Subcommittee

25 DR. JACOBSON-KRAM: Thank you.

1 [Slide]

2 I am relatively new to the FDA. I think
3 this is my seventh week here, but this area is one
4 of the things that drew me to the FDA. I think
5 this is a very exciting time to be in toxicology
6 and I believe with all my heart that this is going
7 to be the future.

8 [Slide]

9 So, welcome to this meeting--the promise
10 of toxicogenomics. What do we see as the future
11 here? Using toxicogenomics, I believe we will be
12 able to identify toxic responses based on mechanism
13 of action. We will be able to identify those
14 earlier in drug development. In the process of
15 doing so, I think we will be able to use many fewer
16 animals. By doing so, we will be able to optimize
17 lead compounds early in development. We will have
18 better extrapolation from animal data to human
19 beings and ultimately, I believe, this will lead to
20 faster development of safer drugs.

21 [Slide]

22 How about the challenge of toxicogenomics?
23 Certainly the varied platforms and technologies--a
24 lot of different companies are involved; there are
25 different kinds of chips and these have to be

1 brought into some kind of uniform consistency.

2 Another big challenge is that correlations
3 of expression changes and health effects are still
4 evolving. We can document thousand and thousands
5 of changes but we don't always know what they mean.

6 Finally, since everybody is coining new
7 terms, I coined data "overlomics." This is one of
8 the challenges with this field, the amount of data
9 that it generates is overwhelming and trying to
10 bring all that together and interpret it is
11 certainly a challenge.

12 [Slide]

13 So, these are the questions for the
14 committee, the charge: Should CDER be proactive in
15 enabling the incorporation of toxicogenomics data
16 into routine pharmacological and toxicological
17 studies and in clarifying how the results should be
18 submitted to the agency?

19 [Slide]

20 What should the present and future goals
21 be for the use of the data by CDER, and what major
22 obstacles are expected for incorporating these data
23 into nonclinical regulatory studies?

24 [Slide]

25 Is it feasible, reasonable and necessary

1 for CDER to set a goal of developing an internal
2 database to capture gene expression and associated
3 phenotypic outcome data from nonclinical studies in
4 order to enhance institutional knowledge and
5 realize the data's full value?

6 [Slide]

7 Is it advisable for CDER to recommend that
8 sponsors follow one common and transparent data
9 processing protocol and statistical analysis method
10 for each platform of gene expression data but not
11 preclude sponsors from applying and sharing results
12 from additional, individually favored methods?

13 [Slide]

14 What specific advice do you have for
15 clarifying recommendations on data processing and
16 analysis, as well as data submission content and
17 format?

18 [Slide]

19 Today's program is divided into three
20 topics. The first one is overview of
21 toxicogenomics at the drug development and
22 regulatory interface, and presentations will be by
23 Drs. Woodcock, Ulrich and Pennie.

24 [Slide]

25 The second segment will be toxicogenomic

1 data quality and database issues, and the
2 presentations will be by Drs. Jarnigan, Quackenbush
3 and Ghosh.

4 [Slide]

5 The third part will be product review and
6 linking toxicogenomics data with toxicology
7 outcome, with presentations by Drs. Leighton,
8 Levin, Mattes and Rosario.

9 [Slide]

10 Frank, I guess, will mediate the questions
11 for the committee--

12 [Slide]

13 --and Dr. Karol will give us conclusions
14 and summary remarks.

15 DR. KAROL: Thanks very much, David. Now
16 I would like to have Janet Woodcock address us on
17 the concept of no regulatory impact for nonclinical
18 pharmacogenomics and toxicogenomics.

19 Topic #1 Overview of Toxicogenomics at the Drug
20 Development and Regulatory Interface
21 Concept of "No Regulatory Impact" for Nonclinical
22 Pharmacogenomics/Toxicogenomics

23 DR. WOODCOCK: Thank you and good morning.

24 [Slide]

25 What I would like to talk about this

1 morning is the whole issue of the emerging field of
2 genetic information and also proteomic information
3 and other allied types of information, and how that
4 is going to play into the regulatory review process
5 because the current regulatory review process that
6 exists does not really formally recognize or
7 incorporate this kind of information and, yet, it
8 is coming; we are starting to see results in this
9 area and so the question really does arise as to
10 how do we, as a regulatory body, get this
11 information; how do we deal with it; and also how
12 we encourage the field to develop.

13 [Slide]

14 This is really about translation of
15 innovative science to bedside medicine. This is
16 about getting candidate drugs, lead compounds
17 developed, get them through the process and to the
18 bedside. How can we use new biological science
19 that is emerging in speeding up this process?

20 [Slide]

21 Right now the new science of
22 pharmacogenomics, and increasingly these other
23 allied techniques, are applied extensively in drug
24 development. They do have the potential--I agree
25 with what was just said--to revolutionize the

1 process? Most of the data now is not seen by
2 regulatory agencies, most of the data that are
3 being generated, and partly that is out of concern
4 for what we will do with it, to be very blunt.
5 What interpretation will the regulatory agencies
6 make of these findings?

7 Therefore, I think we need an approach
8 that will enable free exchange of information, will
9 help advance the science and technology along and
10 will aid in the timely development of appropriate
11 regulatory policies to apply to this kind of
12 information. In the field of toxicogenomics we are
13 seeking your help today in developing these
14 policies.

15 [Slide]

16 Just for a brief background which I think
17 you all know so I will go through this quite
18 rapidly, but one of our problems as clinicians is
19 the tremendous variability in human response to
20 drugs. It is a huge barrier to using medicine
21 effectively in human populations because you can't
22 tell how people are going to respond.

23 [Slide]

24 There is variable effectiveness, and this
25 isn't the toxicology side so much but it really

1 will also be related to animal models. So, for
2 many drugs, if you leave aside antivirals and
3 antibiotics and things that are directed at
4 organism that aren't a human organism, the size of
5 the treatment effect that we observe in randomized
6 trials may be less than ten percent of the overall
7 outcome measure, in other words, a very small
8 amount of response. Many conclude therefore,
9 correctly I think, that the effect of the drug is
10 small, it is a very weak drug or the drug doesn't
11 work.

12 [Slide]

13 If you look at it this way, if you look at
14 a population basis, you see that you get a certain
15 response in the placebo and if you use enough power
16 in your study you can barely reach statistical
17 significance often and show that the drug is more
18 effective than placebo, but it is a very small
19 difference.

20 [Slide]

21 If you define responders though--my slides
22 in the book may not be exactly like on the screen,
23 I am sorry--but if you find responders, then you
24 can see that with the placebo you may get a little
25 bit of response but for the drug you get a small

1 population that responds very well. We have seen
2 this again and again in different areas. So, what
3 we have here is variability. Some people respond
4 to the drug and a lot of people don't respond to
5 the drug. Our problem is that we don't know in
6 advance who those people are so we have to expose a
7 lot of people to get a small population responding.

8 [Slide]

9 In the same way, we get variability in the
10 clinic in drug toxicity. If you look at drug
11 versus placebo and you look in the PDR, or
12 whatever, you see that every drug and even classes
13 of drugs have a consistent pattern of side effects
14 over the placebo. That is true for common events
15 and it is true for rare events. Some of the wide
16 effects can be attributed to the known
17 pharmacologic effects of the drug and they tend to
18 affect the population fairly uniformly, but may
19 others are considered idiosyncratic. Again, the
20 problem is we cannot predict which people are going
21 to experience these side effects or experience them
22 more severely. Therefore, currently in drug
23 development as well as in medical practice we
24 simply say oh well, this causes renal toxicity or
25 liver toxicity and that is about as far as we get

1 and we watch for it. It is very observational and
2 we really don't have a way often to say we should
3 avoid exposing this group of people because they
4 are more prone to this toxicity.

5 [Slide]

6 The good news is we think there is an
7 inherited component, a genetic component to this
8 variability in drug response. In other words, some
9 of this would be predictable if we had more
10 information.

11 I have two terms here,
12 pharmacogenomics--there is quite a bit of dispute
13 about what these terms mean so, please, this is
14 simply for the purposes of this talk. I am
15 considering pharmacogenomics to be application of
16 genome-wide RNA or DNA analyses to study
17 differences in drug actions. Pharmacogenetics, I
18 am considering as looking at the genetic basis for
19 inter-individual differences in pharmacokinetics
20 and mainly that is driven by drug metabolism
21 differences. But these two techniques can help us
22 investigate this inherited or genetic component of
23 drug variability.

24 [Slide]

25 In efficacy there are many ways to look at

1 this but there are at least three types of genetic
2 variabilities that contribute to differences in
3 effect of the drug, the beneficial effect. One is
4 the diversity of disease pathogenesis. Of course,
5 in animal models there are varying pathogenic
6 pathways or actual diseases that lead to the same
7 syndrome and often we don't have enough knowledge
8 to separate those out and we expose everyone who
9 exhibits a certain syndromic pattern. Some of them
10 respond and many of them don't respond because they
11 don't have the pathogenesis that would respond to
12 that particular intervention. So, what disease?

13 Variable drug metabolism is a very
14 important. What dose? People can have ten-fold
15 differences in plasma levels based on metabolism.
16 Right now we don't distinguish among those people.
17 We give people a couple of ranges of doses and we
18 hope they will all respond well.

19 Then, there are going to be genetically
20 based pharmacodynamic effects. This has been
21 studied, for example, in people with, say,
22 differences in the beta adrenergic receptor. In
23 people taking asthma drugs there may be genetically
24 based differences in how well they can respond to a
25 beta agonist. It has nothing to do with their

1 disease, but it has to do with other genetic
2 variability underlying the genetic variability that
3 they have, but still it may predict drug response.
4 We are looking at that for some of the cholesterol
5 lowering agents as well.

6 [Slide]

7 Drug toxicity, likewise there are genetic
8 contributions to the variability in drug toxicity.
9 One is that you may have a genetically based
10 interacting state. You may have a long QT syndrome
11 genetically, and you take a drug for some other
12 condition that prolongs QT interval and you may be
13 in trouble while the vast majority of the
14 population has no effect from that. So, you have a
15 predisposition to this toxic effect.

16 There may be differences in drug
17 metabolism just like in efficacy. So, for toxicity
18 there are some people, and we know this very well,
19 who are actually overdosed significantly by
20 standard doses of drugs based on their metabolic
21 pathways that they have.

22 Finally, there are toxicodynamic
23 interactions where you have a vulnerable subgroup.
24 Again, it has nothing to do with their disease but
25 they are simply vulnerable to some toxic effect,

1 some interaction. So, for toxicity, which is the
2 main discussion at this meeting, at the level of
3 the clinic there are genetic ways by which we could
4 predict who is going to get a toxic effect.

5 [Slide]

6 But how important are these differences?
7 That is sort of the skeptic's view. These
8 differences exist. How much of human variability,
9 for example, would be explained by genetic
10 differences? Is this worth pursuing? Well,
11 sometimes.

12 [Slide]

13 At the level of an individual a genetic
14 difference in some cases can be determinative. I
15 think this is the case both for toxic responses as
16 well as for efficacy responses. More commonly at
17 the level of an individual a genetic difference can
18 highly influence drug response. It may make you
19 much more likely to have a toxic response but not
20 100 percent, or it may make you much more likely to
21 have or not have effectiveness in the drug
22 metabolizing enzymes in your particular suite of
23 drug metabolizing enzymes. You can really predict
24 that you are getting the wrong dose or some
25 individuals will get a toxic dose based on drug

1 metabolism. So, that can be very important.

2 [Slide]

3 But we have to recognize that many
4 responses are going to be an emergent property of
5 multiple gene products that are interacting both
6 with each other and with the environment,
7 environmental factors. So, that is where we may
8 have to look at patterns. That is where proteomics
9 and other things come in because this will be more
10 of a systems issue than a single factor that is
11 determinative or highly predicted.

12 [Slide]

13 I like this pyramid, which is from Science
14 recently, which talks about the different levels if
15 we are looking at these things. At the very top is
16 the organism, the mouse or the rat or the monkey or
17 the human, and we are an interacting system of
18 many, many subsystems. When you are looking at
19 genetics you are down at the bottom; you are only
20 looking at a piece and it contributes up; the same
21 with proteomics and many of the other studies.
22 This is where the data that David was talking about
23 comes in because we have to take many snapshots of
24 the organism at many different levels to understand
25 what is really going on.

1 [Slide]

2 Currently drug development is satisfactory
3 but it is very expensive and we find out things
4 very late in drug development that would be much
5 better to find out early. We are able to determine
6 whether drugs are effective or not. I can tell you
7 that the Center for Drugs does not approve drugs
8 that are not effective anymore--

9 [Laughter]

10 --but we use a population basis. So, what
11 the public asks us today is more is this going to
12 work for me? They don't really care if a drug
13 works hypothetically in a population; they want to
14 know is this drug going to be effective for me. We
15 can't tell people that right now when we approve a
16 drug.

17 The same with drug toxicity. As you all
18 know very well; you are more expert in this than I,
19 the determination is observational. It is based on
20 exposing animals and the human is very similar. We
21 expose the human but we just don't go up to the
22 toxic doses we do in animals, and we see what
23 happens. Again, when we put that drug on the
24 market and it is being sold we can't tell a
25 patient, individual patient, you are the one; you

1 are going to get the catastrophic side effect; you
2 are going to get this bad side effect; or, you are
3 going to do just fine on this drug. We do not have
4 that kind of information. Whatever guiding
5 information we give to clinicians is very
6 crude--avoid in renal failure or something like
7 that; it is a very, very crude level. Right now
8 carcinogenic and reproductive toxicity potential of
9 the drug is based on the in vitro and animal
10 studies and, again, we do pretty well on this but
11 we can't tell people for sure.

12 [Slide]

13 What potential uses do we have for this
14 genetic information in drug development? Well,
15 David has already talked about this a little so I
16 will go through this quickly. Obviously, improving
17 candidate drug selection is very important given
18 the cost of drug development. Developing new sets
19 of biomarkers for toxic responses, first in animals
20 and then in humans, eventually with the goal of
21 minimizing animal studies and, yet, having better
22 predictability from our preclinical work. At the
23 clinical level, predicting who will respond and who
24 will have a serious side effect--this would be
25 wonderful. Also to rationalize drug dosing based

1 on the genetic substrate of the individual.

2 [Slide]

3 In sum, we can all, the biomedical
4 community in general can pull this off. We can
5 expect for the next decade or two to move from the
6 current empirical process--which is what drug
7 development right now really is; it is not a
8 mechanistic, predictive type of process--to a
9 mechanism-based, hypothesis-driven process for the
10 triumph of rational science in biology, which is
11 something we haven't really been able to achieve
12 yet. This would result in a lower cost and faster
13 process that could result in more effective and
14 less toxic drugs, albeit they would be indicated
15 for smaller groups of people because we would know
16 from people's genetic and other information who was
17 going to respond.

18 So, the potential of this is tremendous.
19 I agree with David, I have no doubt this is going
20 to happen. It is just how soon and how many bumps
21 we are going to encounter in the road. Frankly,
22 today one of the things you are going to discuss is
23 one of those bumps and how do we deal with one of
24 those obstacles effectively.

25 [Slide]

1 So, that is the question, how can this new
2 technology be smoothly integrated into the drug
3 regulatory process? How can we do that?

4 [Slide]

5 Right now our legal requirements, which
6 are driven by the Food, Drug and Cosmetic Act,
7 require that we evaluate all methods reasonably
8 applicable--this is in the new drug application--to
9 show whether or not such drug is safe for use under
10 the conditions in the proposed labeling. So, all
11 methods reasonably applicable about safety. For
12 effectiveness, that we look at adequate and
13 well-controlled trials to show that the drug will
14 have the effect it purports to have under the
15 conditions of use.

16 [Slide]

17 For the investigational new drug
18 application, the IND, there are submission
19 requirements in our regulations. They state that
20 you have to submit the pharmacology and toxicology
21 information on the basis of which the sponsor has
22 concluded that it is reasonably safe to conduct the
23 proposed clinical investigations. That is what the
24 regs say.

25 [Slide]

1 About the NDA submission the regs say that
2 for nonclinical studies you must submit studies
3 that are pertinent to possible adverse effects.
4 Obviously, when these regs were written we did not
5 know about this kind of information that we are
6 talking about today.

7 For the clinical you have to submit data
8 or information relevant to an evaluation of the
9 safety and effectiveness of the drug product. So,
10 relevant.

11 [Slide]

12 The issues that need to be resolved are
13 when and how to use developing pharmacogenetic
14 information and related information in regulatory
15 decisions. When is the information reasonably
16 applicable, pertinent or relevant to safety? That
17 is really one of the questions. And, under what
18 circumstances then is submission of this
19 information about a candidate drug to FDA needed or
20 required? Under what circumstances?

21 [Slide]

22 We have already developed somewhat of a
23 plan on this but what we are here today for you
24 help fill in some of the details I think. We
25 discussed this plan or proposal with the FDA

1 Science Board and received some endorsement, but
2 the proposal was at a very high level without
3 detail filled in.

4 What we propose to do is we will establish
5 policies on pharmacogenetic data and we will have a
6 policy on what type of data is required or not
7 required to be submitted; what type of data are
8 appropriate or not appropriate for regulatory
9 decision-making. This is the kind of information
10 the sponsors need to have.

11 [Slide]

12 What about submission requirements? I
13 have to stress we do not have a policy right now.
14 We are working on one and we will go through a
15 public process, as I will describe, but we would
16 decide whether or not submission of data were
17 required based on interpretation of the regs and
18 the statute that I quoted above. It is clear right
19 now, that without any interpretation, that any data
20 actually used in protocol decision-making in people
21 needs to be submitted. That is probably true with
22 animals too. If you are going to select animals on
23 genetic data, and so on, and manipulate them in
24 some way in the protocol, or whatever, that would
25 be obviously required.

1 In addition, it is clear and may have
2 happened, I am not sure, that sponsors may submit
3 data to FDA to bolster a claim or their scientific
4 position about something. For example, people may
5 want to explain why a finding in a certain animal
6 species is not relevant to humans and they may wish
7 to submit a variety of genetic data to show that
8 the relevant genotype, or whatever, is only within
9 that one species, or whatever. But for most
10 results, as I have here, submission not required.
11 This line is the line that we have to work on and
12 FDA is working on that.

13 [Slide]

14 The thing about submission of data, if
15 submission is not required, how is FDA going to
16 develop a knowledge base about the field? This is
17 the conundrum we are in. So, we will be requesting
18 voluntary submission of results, and this is where
19 "no regulatory impact" comes in. Results would not
20 be used in regulatory decision-making. We really
21 do need to hear about emerging results as this
22 information begins to be used routinely.

23 [Slide]

24 But how would we give this assurance?
25 When would FDA use the data for regulatory

1 decision-making? I have to stress that this is
2 sort of a working proposal that we are thinking
3 about. FDA will apply a threshold determination to
4 the data that is submitted. Okay? Data that is
5 submitted voluntarily would already be in the
6 category of "we would not use that for regulatory
7 decision-making." All right? Data submitted by a
8 sponsor to make a case, obviously we would use that
9 in regulatory decision-making; the sponsor would be
10 requesting us to use that in regulatory
11 decision-making. So, there are really three
12 categories of data that we are talking about here.

13 [Slide]

14 What we are proposing, and this is just a
15 work in progress, is that the information would
16 have to have risen to the status of being a valid
17 biomarker. In other words, when the meaning of the
18 genetic test is well understood and of known
19 predictive value, then results from testing animals
20 or patients should be submitted to FDA. In other
21 words, it would be required. That would be the
22 required submission threshold. This clearly could
23 be whether we use this for a regulatory
24 decision-making threshold because we don't use
25 information for regulatory decision-making if it

1 doesn't really have meaning yet.

2 The problem with a lot of the genetic
3 information, as you all know, is it is currently
4 being generated and we don't know what it means.
5 In a sense, we know what it means in a genetic
6 sense but we don't know what it means in a
7 predictive sense. We don't know what it will imply
8 and, therefore, we shouldn't be drawing conclusions
9 about it. Research or exploratory tests, in fact,
10 are not suitable for making decisions on safety or
11 efficacy of a drug. They are not yet suitable.

12 [Slide]

13 What we are planning to do is develop this
14 threshold and these policies using a public and
15 transparent process with advisory committee
16 oversight. While I know today the main focus of
17 the effort is to talk about the standardization,
18 and so forth and so on, this discussion today
19 before this advisory committee is what will help
20 feed into the policies as we develop them.

21 [Slide]

22 What we plan to do is publish a guidance
23 for industry that would have a decision tree for
24 the submission, what is required to be submitted,
25 and also a decision tree for whether things would

1 have regulatory impact or not, whether the data
2 would have regulatory impact. Is everybody
3 following me on that? Is that clear?

4 What we do when we do a guidance is we
5 will publish a draft. We hope to publish that in
6 August. Then we will have extensive public comment
7 on the draft and we will probably have a workshop
8 after that draft is published so that people can
9 react and we can have extensive input. Then we
10 will probably have more advisory committee
11 discussions about the draft. We will also
12 establish an interdisciplinary pharmacogenetics
13 review group that would provide a centralized
14 review of this information. We have a
15 carcinogenicity committee that looks at all the
16 carcinogenicity studies to provide consistency
17 across the Center. We will do the same thing for
18 this type of information so we will have a
19 centralized review and this body could also work on
20 ongoing regulatory policy development.

21 [Slide]

22 As part of today's discussion, we will be
23 working with the advisory committee and talking
24 about our work in the private sector on the
25 standardization issues. Obviously we will never be

1 able to use this information in regulatory
2 decision-making if it isn't standardized in some
3 way so we can understand what it means, one
4 platform to another. Standardization is really one
5 of the basic efforts you have to go about working
6 on when you work on various biomarkers so that you
7 know what the results in one lab mean compared to
8 another lab. As I said, we will also issue a
9 guidance, a separate guidance on the format of the
10 submission and the data, in other words, how we
11 would like to see the data, and that is going to be
12 discussed today.

13 [Slide]

14 What are some examples? These might be
15 controversial so let me say this is just the
16 working proposal and we may modify this even in the
17 guidance. What about genetic information generated
18 in animals, in toxicology studies? We don't know
19 what would be required to be submitted right now to
20 the FDA because we don't know of anything that we
21 would understand well enough that it would be
22 considered valid by a marker to be submitted. All
23 right? That is going to change over time, we all
24 hope, but that is the state we are seeing right
25 now.

1 We are definitely interested in voluntary
2 submissions and we are not seeing very many.
3 Again, as I said, to explain an animal toxicity
4 finding, that is really up to the sponsor, to
5 submit that and I think people have submitted
6 things like that.

7 [Slide]

8 We have been asked this question in
9 toxicology for animals, for cells, for people, what
10 if you are doing a screening study, an expression
11 study and you are looking across a genome and what
12 if you expose this cell, animal or person to drug
13 and you see increased expression of an oncogene
14 after drug exposure, or maybe many oncogenes?

15 Well, we have looked into that, and I hope
16 Frank talks about that a little bit or someone
17 talks about that, but we looked into that because
18 we were explicitly asked and this is the kind of
19 thing people are worried about. What we find is
20 that in some studies that have been done many
21 common drugs that are given at high dose can elicit
22 this finding in toxicity studies. Of course, these
23 proto-oncogenes weren't really put in the body to
24 cause cancer. They are used in development or
25 repair and other types of physiologic actions and,

1 naturally, they are going to be turned on after
2 injury, during development and so on. So, this
3 encapsulates I think what the sponsors are worried
4 about, that they would find something like this.
5 They would submit to the FDA and their drug would
6 never see the light of day basically. But this
7 shows, I think, the value of looking across a broad
8 range of studies, understanding what is going on
9 and having a scientific database because we are
10 able to put these fears at rest very easily simply
11 by looking at what has been done. But this question
12 will come up again and again as we start really
13 probing and finding out what is turned on when
14 animals or cells are exposed to drugs.

15 [Slide]

16 I just put this in although this is
17 clinical pharmacology. People may want to genotype
18 or phenotype trial subjects for their isoenzyme
19 polymorphism for drug metabolism. Now, in this
20 case, the value and meaning for many of the
21 isoenzymes is very well known and it is relevant to
22 assessing outliers in pharmacokinetic studies. It
23 is relevant to looking at the people who experience
24 drug toxicity and see if they were effectively
25 overdosed in the study due to their genetics. So,

1 this kind of information should be submitted to
2 FDA, should be evaluated by us. In fact, recently
3 it was put in a drug label for a drug, and should
4 probably go in more drug labels. I don't think
5 there is a lot of fear about this in the industry
6 or anywhere because we all know what this means and
7 the value of this information.

8 [Slide]

9 This, again, is a working proposal. What
10 if you gather a bunch of screening genomic data in
11 patients during a clinical trial, does that have to
12 be submitted to the FDA? Our current proposal
13 would say no. But what if you analyzed the data
14 and you saw a potential correlation with an adverse
15 event? What would FDA do? There have been very
16 exaggerated fears out there that we would say,
17 well, you can't give this drug to people who might
18 have this genotype, and so forth. How would we
19 interpret this?

20 Well, it is basically simply a potential
21 biomarker, and the way we look at those is that you
22 need a lot of evaluation in additional trials and
23 diverse populations because I think one of the
24 things that is going to happen in humans, other
25 than animals, is humans are a very outbred

1 population obviously and there is going to be
2 extensive variability in the findings. We have
3 already seen this in humans. You are laughing but
4 we are--we are becoming more outbred every day.
5 There is extensive variability in the frequency of
6 certain genotypes and, therefore, the clinical
7 impact of these findings depends on what human
8 population you study. So, simply because you find
9 it once in humans doesn't really mean a whole lot
10 except that it might be of interest.

11 [Slide]

12 In summary, I think that pharmacogenomics
13 really does hold great promise for drug development
14 and for rational therapeutics, which is really the
15 goal in the clinic, to really understand who we are
16 giving the drug to and be able to predict what the
17 effect will be. In fact, use of this technique is
18 increasing. It is actually very widespread in
19 industry right now. What we need is free and open
20 exchange of results between the industry and the
21 FDA to ensure the appropriate development of
22 regulatory policies.

23 [Slide]

24 Concerns about how the data will be used
25 by the regulators has stifled this exchange to date

1 and is continuing to. FDA will develop clear
2 policies on the use of pharmacogenomic data in
3 regulatory decision-making both for toxicology and
4 clinical. And, I think we all look forward to the
5 advances in medicine and health that these
6 techniques, I believe, are sure to bring
7 eventually.

8 I thank the committee for its work. You
9 will be making some steps today towards making this
10 come about. Thank you very much.

11 DR. KAROL: Thank you very much, Dr.
12 Woodcock. Are you available for questions from the
13 committee? Would any of the committee like to ask
14 a question?

15 [No response]

16 Thanks very much. We will move on then to
17 our next speaker and, unfortunately, Dr. Ulrich
18 isn't with us today because of the death of his
19 father. So, we will have the following speaker
20 now, and that is Dr. Pennie who will talk to us on
21 a perspective on the utility and value of
22 expression profiling data.

23 A Perspective on the Utility and Value of
24 Expression Profiling Data at the Drug Development
25 Regulatory Interface and ILSI Experiences with

1 Cross-Platform Comparisons

2 DR. PENNIE: Thank you very much.

3 [Slide]

4 It is my pleasure to speak to the
5 committee this morning, and my privilege to
6 represent a working committee organized under the
7 auspices of the ILSI Organization, which is a
8 consortium effort amongst industrial organizations,
9 academia and government to address some of the
10 technical challenges and share some of the learning
11 on these emerging technologies related to genomics
12 applications and risk assessment.

13 [Slide]

14 This committee has been in existence since
15 mid-1999. When the committee was formed, what I
16 have here is a slide of some of the challenges the
17 membership believed were facing the advancement of
18 these sciences, the first one being a lack of
19 publicly available databases to help put
20 experimental data in context; the second one being
21 a lack of validation of the available technologies;
22 a lack of comparable tools, methodologies and study
23 designs; a lack of robust and consistent tools for
24 data analysis; a lack of fundamental knowledge of
25 how gene products relate directly to toxicity and,

1 in particular, the relevance of single gene
2 changes. When I speak of genes in the context of
3 this presentation, I am talking largely about
4 genomic changes where we are measuring basically
5 the induction of gene expression or repression as a
6 consequence of a compound treatment. So, we are
7 not dealing in this committee's work at this stage
8 with a variable response which may be a result of
9 genetic variability. Certainly, the last comment
10 here, uncertainty about the regulatory environment,
11 was a comment which I think was raised quite
12 eloquently in Dr. Woodcock's presentation, and
13 certainly having a committee like this before us
14 today is an opportunity to broaden the dialogue in
15 this area.

16 [Slide]

17 So, for those of you who aren't familiar
18 with it, the ILSI Health and Environmental Sciences
19 Institute is a non-profit research and educational
20 organization which provides an international forum
21 for scientific activities. These are largely
22 experimental program-based activities. The ILSI
23 organization enjoys participation from industry,
24 primarily the drug industry, the agrochemical and
25 chemical industries and also from government and

1 academic researchers and advisors. The
2 organization runs research programs, workshops,
3 seeds databases, forms expert panels and actively
4 pursues the communication of its findings through a
5 publication strategy, and has a reputation for
6 focus and objectivity.

7 The ILSI organization is not a trade body.
8 It has specifically in its charter that it does not
9 attempt to directly influence the setting of
10 regulatory positions or policies. Instead, they
11 try and provide a basic and fundamental
12 understanding of evolving technologies for how
13 these technologies may be used.

14 [Slide]

15 As I said, the committee was formed in
16 1999. As it stands, it has a membership of around
17 30 companies, an international-based membership,
18 including government participation from labs such
19 as NIEHS, NCI, NIH, NCTR and others. We also enjoy
20 a very active participation of a group of academic
21 advisors who sit on the steering committee of the
22 organization.

23 [Slide]

24 Our objectives were to evaluate
25 experimental methodologies for measuring

1 alterations in gene expression, alterations as a
2 consequence of compound treatment. Other
3 objectives included the development of publicly
4 available data to allow the beginning of
5 discussions on relevance of findings and issues
6 around the development of databases.

7 Particularly, we charged ourselves to
8 contribute to the development of a public
9 international database linking gene expression data
10 and key biological parameters with the goal of
11 determining if known mechanisms and pathways of
12 toxicity can be associated with characteristic gene
13 expression profiles or fingerprints, as they have
14 come to be known in this field, and if the
15 information can be used as the basis for
16 mechanism-based risk assessment. So, we are
17 talking primarily about an application in a
18 preclinical setting here.

19 [Slide]

20 Here is a time-line of where the committee
21 has come from and where we are at the moment. In
22 early 2000 the committee initiated an experiment
23 program which focused on three areas of toxicology
24 for further evaluation, those being hepatotoxicity,
25 nephrotoxicity and genotoxicity. We also formed a

1 database working group to look at issues around
2 data capture, storage and transmission. We
3 initiated a collaboration on database issues with
4 the European Bioinformatics Institute early in
5 2002. You are going to hear a little bit more
6 about that initiative at the end of my talk and in
7 Dr. William Mattes' talk this afternoon.

8 Just last week, in fact, we held our first
9 public meeting on the application of genomics and
10 risk assessment, in the Washington area, and
11 invited a large number of scientists from the
12 regulatory and academic communities to join with us
13 in discussing the progress of the committee to date
14 and future opportunities for sharing of learning as
15 we move forward with these initiatives. We also
16 have an aggressive peer-reviewed publication
17 strategy which will take us through 2003 and the
18 early part of 2005.

19 [Slide]

20 Let me tell you a little bit about what
21 the actual deliverables of this committee are. The
22 program mechanism was, as I said, to organize
23 ourselves into a series of working groups to focus
24 either on experimental research in the areas of
25 hepatotoxicity, nephrotoxicity and genotoxicity or,

1 as I articulated, to begin discussions and planning
2 around contributing to an international database on
3 gene expression changes.

4 [Slide]

5 Our experimental design feature basically
6 profiling well-studied compounds in the literature
7 with known toxicity profiles and biological
8 parameters. We investigated temporal relationships
9 and the effect of dose on gene expression changes
10 and an opportunity afforded by the committee, as
11 you will see, is that given the broad membership
12 and broad access to numerous technical platforms,
13 we have the opportunity to look at some technical
14 details of the technology, including variability
15 and operating procedures that may vary from one
16 laboratory to another.

17 [Slide]

18 I have made a list of the objectives we
19 set up at the beginning of the committee's
20 activities to try to give you an understanding of
21 what our status is. For the first objective, to
22 evaluate methodologies, we have developed protocols
23 within our member labs and within the committee as
24 a whole to evaluate profiles of specific prototypic
25 toxicants. We went through an exercise of

1 distributing RNA samples to public and industry
2 labs for microarray-based gene expression analysis.
3 This allows us to consider variability that may
4 take place both in in-life studies and inter-lab
5 variability when different labs are profiling the
6 same material. We evaluated the influence of
7 specific experimental conditions on data
8 variability. These may be technical experimental
9 conditions such as the way that the apparatus is
10 set up for the experiment. Those issues are still
11 being looked at. We have utilized the outcome of
12 experiments and data analysis to stimulate
13 discussion of what the best practices may be for
14 these applications.

15 [Slide]

16 A second objective, to contribute to the
17 development of international databases linking gene
18 expression data and key biological parameters, will
19 be discussed in a little bit more detail briefly at
20 the end of my talk but also in Dr. Mattes' talk,
21 but effectively, we have been in discussion with a
22 large number of stakeholders on data formats for
23 microarray storage and transmission; building
24 database structure to include the incorporation of
25 standard toxicology endpoints in preclinical

1 studies; and a drive to make these databases and
2 the data within them available in the public domain
3 actually before 2004 but, we expect, in the course
4 of this year.

5 [Slide]

6 A third objective, this is where we start
7 to focus on risk assessment, is to determine if
8 known mechanisms and pathways of toxicity can be
9 associated with characteristic gene expression
10 profiles and if this information can be used for
11 risk assessment.

12 So, as I have said, we have developed gene
13 expression datasets on well characterized toxicants
14 and are at various stages of data mining and data
15 evaluation to characterize the mechanistic
16 information that can be gleaned from such studies.

17 [Slide]

18 I will very briefly give you an outline of
19 the three working groups, then I will try and give
20 you, for each one of them, some of the interim
21 conclusions the working groups have reached with
22 regard to the technology and its applications.

23 Our nephrotoxicity working group worked on
24 three prototypic nephrotoxicant compounds and had
25 in-life studies conducted at a single site to

1 prepare material in vivo for the analysis of these
2 compounds' effects on transcription profiles in lab
3 animals. In this case it was in rats. There were
4 eight participating labs who were involved in
5 taking the material from the in-life study,
6 preparing and analyzing it using gene expression
7 analysis technologies. These technologies,
8 including multiple technical platforms, the
9 microarrays produced by organizations such as
10 Affymetrix, Incyte, ClonTech and Phase-1 and also
11 the use of custom cDNA microarray platforms which
12 have either been generated in academia or in the
13 labs of the participant organization, and pooling
14 all this together gave the opportunity to compare
15 inter- and intra-lab variability, cross-platform
16 variability and the ability to replicate the
17 in-life study.

18 [Slide]

19 So, the interim findings were really an
20 ability to recapitulate the data on standard tox
21 endpoints for these compounds. In other words, we
22 were able to replicate what was known about the
23 more traditional tox endpoints in the rat species
24 for these compounds. Transcriptional analysis
25 yielded strong topographic specificity and some

1 mechanistic information about the mode of action of
2 the compounds.

3 Where we had individual gene expression
4 changes that were of interest to the committee, we
5 did confirmatory analysis using alternative
6 methodologies. All of these were positive and will
7 be extended to investigate potential biomarkers of
8 nephrotoxicity in preclinical species.

9 The frequency of individual animal
10 transcript changes was reduced in non-responders
11 and increased in cases of severe toxicity. In
12 other words, there was a direct linkage between the
13 magnitude of gene expression changes and the onset
14 of toxicity.

15 We, not surprisingly, found that the use
16 of pooled RNA samples may have a dilutional or
17 skewing effect on the interpretation of genetic
18 response, but at the stage these programs were
19 initiated cost was a major factor in being able to
20 take these programs forward and pooled samples were
21 analyzed in the initial stages.

22 The group has concluded that these
23 technologies have at least equal sensitivity to
24 traditional toxicology endpoints in terms of
25 detection and an enhanced opportunity to resolve

1 some mechanistic information.

2 [Slide]

3 I will move a little bit more quickly
4 through our second working group. You have the
5 tenor of how the groups are organized. The
6 hepatotox group worked on two test compounds but
7 they performed independent in-life studies to look
8 at the effect of different sources of in-life
9 material and in-life studies on data analysis.
10 They had 14 participating laboratories in the
11 analysis of the material, again performing analysis
12 on multiple technical platforms. The use of 14
13 industrial labs on two test compounds and two
14 in-life studies gave a truly unprecedented
15 opportunity to look at issues related to
16 variability.

17 [Slide]

18 Their findings were, again, the expected
19 outcome with regard to the in-life study
20 replicating what was known in the literature about
21 these two compounds. Within a given technical
22 platform, in other words, using a single microarray
23 platform such as Affymetrix, there was a high
24 degree of concordance, greater than 90 percent, in
25 the direction of the of the gene expression changes

1 across samples analyzed in different labs, but
2 lesser concordance was observed when identifying
3 probes or individual genes that were regulated
4 above or below a certain threshold for all
5 datasets, for example, a cut-off of greater than
6 4-fold to regulation. This result may be
7 attributable to differences in data capture
8 algorithms or data analysis methodologies across
9 labs.

10 Dose-related response was observed in
11 these experiments, and for one of the compounds
12 under study, methapyrilene, agreement was found
13 across all platforms with good but varying degrees
14 of congruence in the results.

15 Now, the field of data analysis for gene
16 expression changes is very much on a logarithmic
17 scale in terms of its advancement and since this
18 slide was made there have been some strides forward
19 in this particular working group in reconsidering
20 their methodology for data analysis and, in fact,
21 we believe that if you limit your data analysis to
22 genes that have a very high degree of statistical
23 rigor around the expression change within an
24 individual lab, then the cross-lab variability is
25 significantly reduced.

1 [Slide]

2 A slightly different approach was taken by
3 our genotox working group which conducted their
4 assessments in cell lines, the mouse lymphoma p53
5 null cell line and the human TK6 cell line which is
6 p 53 competent. They run their gene expression
7 profiling experiments in concert with standard
8 genotox testing regimes to look for direct-acting
9 mutagens and clastogens microarray analysis on the
10 material prepared from the cell lines and, again,
11 multiple platforms were used for the comparisons.

12 [Slide]

13 Their conclusions were that gene
14 expression changes less than 3-fold were very
15 common in all studies even at highly genotoxic
16 concentrations. So, concerns around the
17 over-sensitivity of the technology appear to be
18 unfounded, at least with the limited dataset
19 generated by this group.

20 Array technology in fact may not be as
21 sensitive an endpoint as the more standard genotox
22 testing battery which is currently in use in the
23 industries, but gene expression changes have the
24 advantage of possibly allowing us to distinguish
25 mechanistic classes of genotoxic compounds. The

1 strong push from this group is that standardization
2 of analysis and control of experimental variables,
3 as we have discussed already this morning, pose
4 challenges to data comparison and interpretation.

5 [Slide]

6 the committee-wide data findings, to
7 summarize, are that application of microarray
8 technology has all the usual sources of
9 experimental variability you would encounter in a
10 biological experiment, with the additional
11 complexity, which can come from a number of areas,
12 such as differences in the protocol for the
13 harvesting of the mRNA sample; differences in
14 protocols or conditions for the hybridization of
15 the RNA sample to the microarray platform;
16 importantly, differences in the way the genes are
17 recorded by manufacturers on their individual
18 technical platforms. In other words, gene X may
19 not equal gene X between two different technical
20 platforms--different specific nucleotide sequences
21 within probe sets across different technical
22 platforms. In other words, even if gene X on
23 platform 1 does equal gene X on platform 2, the
24 precise sequence used to make the detection may be
25 different and be subject to different hybridization

1 kinetics, for example.

2 Clearly, a big issue is that all these are
3 not made equal and there is not a direct
4 correlation for the gene sets on one manufacturer's
5 array to the gene sets on another's. It is
6 important to monitor the effect of signal to noise
7 ratios; analysis setting on the machinery used to
8 make the detection; keep a hold of false-positive
9 and false-negative rates statistically to make sure
10 you are not putting too much weight on background
11 noise in an experiment. Clearly, there are a large
12 number of different analytical tools that take the
13 raw data from these experimental platforms and
14 convert them into a subset of gene changes for
15 further investigation. There are significant
16 differences in the methodology for getting at that
17 analyzed short list that can have a fairly
18 significant effect on the interpretation of a given
19 experiment.

20 [Slide]

21 This slide I think just summarizes the
22 opportunity that was afforded to the ILSI
23 membership and, by its charter, is afforded to
24 anyone in the public community or regulatory
25 community who would like access or discussion on

1 the data. This slide basically then captures where
2 we have had an opportunity to look at variability
3 issues, be it the in-life variability, variability
4 in in vitro experiments, intra-lab platform
5 replicate variability, and so on and so forth.

6 [Slide]

7 Very briefly then, we heard this morning
8 about a data overload in genomics technologies.
9 What was once promised us as a great advantage and
10 a step forward for these technologies and the rapid
11 accumulation of very high density of information
12 turned pretty quickly into one of the biggest
13 challenges for people who dealt with the data in
14 terms of managing, storing and interpreting the
15 many, many millions of data points that can be
16 generated from even a single experiment.

17 So, in recognition of this, the ILSI
18 committee, as I said earlier, engaged in a
19 collaborative effort with the European
20 Bioinformatics Institute on building and enhancing
21 their existing ArrayExpress database platform, which
22 houses array data from multiple technical
23 platforms, is compliant with the internationally
24 regulated standard for the minimal information
25 required for a microarray experiment and,

1 importantly, has been extended to incorporation of
2 toxicology endpoint data into a microarray
3 submission. In fact, there has been the evolution
4 of a new microarray data standard, called
5 MIAME-Tox, which is the subject of one of this
6 afternoon's presentations. As I said earlier, the
7 database is largely functional. The tox component
8 of the database is expected to be rolled out to the
9 public domain sometime in the course of 2003 or
10 early 2004.

11 [Slide]

12 The complexity of such a database is hard
13 to get across to people when you are trying to
14 capture not only the data itself but the
15 experimental conditions that were used when the
16 experiment was performed, and also additional
17 biological information that is important to put the
18 transcriptional data in context. So, we have
19 within this database schema the opportunity to
20 store information on the sample pool, the way the
21 material was extracted and prepared, all the
22 experimental conditions around the generation of
23 the gene expression data and link that directly to
24 various biological endpoints, such as traditional
25 pathology, biochemistry or clinical chemistry

1 endpoints.

2 [Slide]

3 Winding down this presentation, the
4 program status for 2003 for the ILSI committee is
5 that we have completed the data analysis,
6 effectively completed the data analysis from
7 current studies. These were what we considered the
8 Phase 1 studies that we initiated in 2000. We have
9 completed an interim review and, in fact, published
10 an interim conclusions document which is available
11 from the ILSI web site.

12 We had, as I said, an invitational workshop
13 just this last week to discuss the interpretation
14 of the committee and take forward issues around the
15 application of genomic data in risk assessment. We
16 valued very much the dialogue between the
17 committee, the academic sector and various invited
18 participants from FDA and other regulatory agencies
19 and, indeed, at that meeting recognized the
20 importance of moving forward in the ILSI committee
21 of having some steerage from the FDA as to what
22 were important questions for us to answer. So, as
23 a result of discussions last week we invited Dr.
24 John Leighton to join the steering group of that
25 committee and he graciously accepted.

1 Our collaborations are to continue to
2 analyze issues of variability. We have internal
3 efforts within and across participant labs to look
4 at variability of analysis, and we are also
5 grateful for collaborations we have initiated with
6 external organizations, such as Affymetrix and
7 Rosetta Informatics, to help with consensus on the
8 important issues around the methodology for
9 analyzing data.

10 As I just showed you, the EBI database
11 continues to be supported by the ILSI committee and
12 the evolution of standards from microarray
13 expression data exchange is high on our radar for
14 important activities moving forward.

15 [Slide]

16 White papers on interim findings, as I
17 said, are available right now on the ILSI
18 organization's web site. A series of peer-reviewed
19 publications, including back-to-back publications
20 scheduled for the fall, initiated in spring 2003
21 and take place through 2004. We are in the process
22 of writing up the minutes from our invitational
23 workshop; continue to move forward with EBI and
24 ongoing discussions, such as the one we are having
25 this morning and this afternoon, on the application

1 of these methodologies to risk assessment and the
2 best practices that need to be put in place for
3 best interpretation of the data.

4 [Slide]

5 Here is my final slide. I have tried to
6 list here what I think are the opportunities that
7 are afforded to all interested parties, and
8 particularly this committee on the application of
9 genomics to mechanism-based risk assessment. I
10 think this particular committee has an
11 unprecedented opportunity to compare multiple
12 platforms analysis methodologies and inter-lab
13 variability issues. Remember, we were able on this
14 committee to harness the infrastructure of 30 or so
15 large pharmaceutical and other industry companies,
16 comparing results across multiple technical
17 platforms that no one individual organization would
18 have been able to do by themselves.

19 That has also given us the opportunity to
20 sit down with colleagues across the industry,
21 academia and the regulatory agencies to discuss
22 where we are going with improving methodologies.
23 We have the opportunity to engage database experts
24 and to seed a publicly accessible and linkable
25 database, and to ensure that such a database is

1 able to incorporate or link to toxicology
2 information.

3 What I didn't say earlier is that a key
4 issue was that that data would be transportable to
5 other databases that may evolve in the academic or
6 public sector and, as such, could be very much a
7 partnering opportunity as the data begins to evolve
8 in pockets amongst the emerging databases.

9 It has given us the opportunity to
10 contribute to discussions such as these on the
11 appropriate application of the technology and,
12 importantly, these discussions can be based on
13 shared experience rather than perception around
14 what the technology may or may not do. I think it
15 is important to promote appropriate usage in an
16 industrial setting to maximize the usage of these
17 approaches in a holistic safety assessment process.

18 Dr. Woodcock said this morning that there
19 are a number of fear factors which we have to
20 overcome to get the best usage of this technology.
21 Some of the biggest of those to overcome are
22 actually those that exist within the industries
23 themselves. Not so much fear of how regulators are
24 going to analyze the data, but really just fear of
25 doing the experiment in the first place. It is a

1 fairly standard approach in toxicology and
2 certainly in risk assessment experiments that you
3 should not conduct an experiment if you are not
4 confident you are going to be able to interpret the
5 data. You have to think harder about experimental
6 design if you find yourself in that situation. So,
7 clearly with emerging technologies such as these,
8 there is a fear within the industries that we are
9 going to generate data that we are not fully able
10 to understand and, therefore, a rather conservative
11 approach can be adopted to not do the experiment
12 and not advance the science. So, hopefully,
13 today's discussion is part of the process of trying
14 to instill courage, both in the regulators and the
15 regulated, to move these very promising
16 technologies forward.

17 So, with that, I am happy to take any
18 questions if there are any and, again, thank the
19 committee for the opportunity to come and
20 participate in the discussions today. Thank you
21 very much.

22 DR. KAROL: Thank you very much. Are
23 there questions from the committee? Yes?

24 DR. BROOKS: Talking about the
25 interactions between your working groups, you had

1 stated that at least on some level there was
2 concordance across platforms since you are using
3 multiple platforms. Any numbers or percentages
4 with respect to those platforms within the working
5 groups?

6 DR. PENNIE: It is very dependent upon how
7 you do the analysis. For example, some of the
8 early figures which we reported at the Society of
9 Toxicology meeting two meetings ago were based on a
10 less than critical assessment of the statistical
11 rigor of an experiment within an individual lab, if
12 you see what I mean. So, those were very
13 disappointing figures I think, that even what we
14 thought was a well controlled experiment may give
15 you, you know, less than 20 percent agreement in
16 the gene list for an individual experiment. But,
17 rather than give you a number right now, I would
18 say watch this space because we have some very
19 encouraging results, particularly from the
20 hepatotox group where a more rigorous analysis
21 gives a much more comforting result even with the
22 number of gene expression changes that stand up to
23 that rigorous analysis give you a much shorter gene
24 list at the end.

25 DR. BROOKS: So, higher statistical rigor,

1 you think, will give you higher concordance across
2 platforms?

3 DR. PENNIE: I think it may, but also a
4 greater understanding of exactly what the
5 annotation issues across platforms are, which is
6 part of that rigor exercise. There is no point in
7 trying to compare gene X to gene X on another
8 platform if, in fact, they are not gene X.

9 DR. BROOKS: One other quick question,
10 what do you think the relative contribution of each
11 of the additional variables associated with
12 microarray data is that you had listed on that one
13 slide, in the hopes that some of them may actually
14 not be as significant and some will be more
15 significant, so we know where to focus our efforts?

16 DR. PENNIE: That is a good question. I
17 think one in particular for the Affymetrix platform
18 is the PMT setting on the detection apparatus.
19 What I think that is likely to skew the results for
20 is really borderline calls between present and
21 absent on a given microarray. In other words, you
22 will have a different size of gene expression
23 shopping list from one experiment to another but it
24 will be overlapping, and there is an area of sort
25 of noise versus signal that may be lost in an

1 inappropriately calibrated machine.

2 DR. BROOKS: From this data, do you think
3 you can do some kind of a transformation analysis
4 to assess the contribution of those sources?

5 DR. PENNIE: That is possible. In fact,
6 those and other issues were part of the
7 collaboration we engaged in with Affymetrix
8 directly to try and identify some of those sources
9 of variability.

10 DR. KAROL: Some of the anticipated
11 benefits from this technology is increased
12 sensitivity and mechanistic insight. Can you
13 comment on your findings relative to that? DR.

14 PENNIE: Mechanistic insight I think is something
15 that practitioners of this technology in an
16 industrial setting have been very confident about
17 if you run a well-designed experiment that is not
18 just generating a shopping list of gene expression
19 changes. In other words, if you believe that you
20 have a hypothesis to prove that a particular
21 toxicant may be operating through a particular
22 pathway, then you can remove some of the
23 experimental variability by using small molecule
24 inhibitors or transgenic models, for example.
25 Those are extraordinarily powerful combinations of

1 multiple technologies and have some very compelling
2 examples of an increase of the mechanistic
3 understanding of a compound's action. So, I am not
4 pouring a lot of comfort in the committee that in a
5 risk assessment sense these technologies will be
6 adding value.

7 DR. KAROL: Did you gain any mechanistic
8 insight from your studies?

9 DR. PENNIE: Indeed, we did. Actually,
10 there are a couple of manuscripts in preparation
11 and, in fact, we came up with some new mechanistic
12 insight on the particular toxicants we have had
13 under study that will be published in the
14 peer-reviewed literature.

15 DR. GOODMAN: Before getting too much into
16 the question of effect of experimental treatment,
17 could you address the issue of variability in
18 controls? How consistent are the controls, and are
19 there differences in terms of variability depending
20 on which platform is used?

21 DR. PENNIE: Yes, that is a good question.
22 So, if you compare control data with an individual
23 set of protocols performed within an individual lab
24 the results are reasonably consistent, stand up to
25 what you would expect from that kind of an

1 approach. The challenge is in comparing control
2 data from one lab to another. In fact, until we
3 get a better handle on experimental methodologies
4 and sources of variability, particularly in the
5 analysis, it is not too surprising to practitioners
6 that control data from different sources actually
7 gives a greater amount of difference than control
8 and treated within an individual lab. So, that is
9 a significant source of variability. But within an
10 individual lab control data tend to be pretty
11 tight.

12 DR. HARDISTY: When you selected your
13 compounds for this test for nephrotoxins or
14 hepatotoxins, did you have any that were not known
15 to be nephrotoxic or hepatotoxic to look for false
16 positives?

17 DR. PENNIE: Yes, that is a good question.
18 Instead of doing it that way, what we did,
19 particularly in the nephrotox study, was that we
20 harvested other tissues, other than kidney, so that
21 we would be able to look. In other words, the
22 nephrotox non-kidney tissues were used as negative
23 controls for the hepatotox experiment, if you
24 follow me. It wasn't a rational part of an
25 individual working group design but that material

1 is made available for the other groups to look at
2 different tissues than the classical site of
3 action.

4 DR. WATERS: On the slide at the top of
5 page seven you use the term topographic
6 specificity, which I think I like very much. I
7 would like for you to just expound on that
8 thinking.

9 DR. PENNIE: Okay, that one is referring
10 to the nephrotox working group. We were
11 specifically using compounds that are at a
12 different site of action in the kidney. After the
13 microarray expression experiment had been performed
14 we were able to use other technologies, such as in
15 situ hybridization to show that the changes in
16 expression were actually associated with the site
17 of toxicity.

18 DR. ZACHAREWSKI: At the meeting last week
19 there was an interesting discussion regarding
20 liability and culpability in terms of the
21 historical aspects of data reanalysis years after
22 the fact to identify that. I was wondering if
23 there was an opportunity--I will take the
24 opportunity to ask whether you have any comments
25 and see if there is any clarification for FDA

1 because I don't know if there was an opportunity
2 for FDA to respond to that as well.

3 DR. PENNIE: That is a very good question,
4 Tim. I appreciate it. I think there are two
5 challenges here. One is that as the field evolves
6 we will collect more and more data on the relevance
7 of individual transcriptional changes and have more
8 and more mechanistic understanding of various tox
9 endpoints. So, there continues to be an onus on
10 the organization that has generated the data to
11 reflect back on their findings in the light of
12 advancements in research to make sure they did not
13 observe a toxicological flag that has been
14 subsequently validated. So, that is one challenge
15 and I don't know if we will get some response from
16 our FDA colleagues or not this morning.

17 An even bigger one for me though is we
18 will just spend some time discussing how variations
19 in your analysis methodology can give you a
20 different result. So, clearly, you can analyze an
21 experiment and think you have the answer, and not
22 only can the science move on but the analytical
23 approaches can move on. So, somewhere along the
24 line you have a lot of opportunities to not be
25 picking up on what could be a potentially

1 significant finding. So, for me, this all boils
2 down to a comfort around individual genes as not
3 being an appropriate level of scrutiny for taking
4 these technologies out of context in a risk
5 assessment paradigm. If we can cross that bridge
6 and understand that we have to have a lot more meat
7 and bones to a risk assessment argument than single
8 gene expression changes, I would hope that we would
9 find ourselves in a very sensible place with regard
10 to those issues. But, certainly, comment from our
11 FDA colleagues would be extraordinarily valuable.

12 DR. WOODCOCK: Could you explain the
13 question a little more clearly because I wasn't at
14 the prior meeting?

15 DR. ZACHAREWSKI: Well, the discussion
16 centered around the fact that, you know, if company
17 A generated microarray data and they analyzed it to
18 the best of their extent at that point in time and
19 that data was then deposited within a database, ten
20 years down the road if somebody else reanalyzed
21 that data with the new technologies and the new
22 information there was discovery associated with an
23 adverse health effect, would the company now be
24 liable as a result of that and, I guess even
25 greater than that, be culpable associated with

1 that?

2 DR. WOODCOCK: Right, Well, I think there
3 are two separate trains of thought here. One is
4 sort of the regulatory train and then the other is
5 product liability, which is a much less predictable
6 and maybe science-driven process. In general, I
7 would say though if you look at drug development,
8 you are looking as positive control things we know,
9 known toxicants or whatever. We, in the course of
10 drug development--we, meaning the community
11 involved in drug development, find these things
12 because we expose animals. We are going to
13 continue to do, in other words the routine studies
14 both in animals and in humans, and we will find
15 most of these. I think the ability to predict
16 rare, catastrophic adverse events in people is
17 going to be one of the last things to happen. The
18 other kind of events we are going to find out
19 during drug development so it wouldn't be like you
20 would be clueless and you would have a drug on the
21 market and you wouldn't know, I don't think. So,
22 from a liability standpoint, you have already gone
23 through the vulnerable period, which is when you
24 are in drug development and you don't really know
25 and you are exposing humans for the first time.

1 But, of course, in the courts liability
2 has its own life and rationale and I regard this
3 issue as yet another obstacle to really integrate
4 these technologies into drug development in a
5 rational way and something we have to deal with.
6 But, again, I think the fear is greater than the
7 reality but maybe I am missing something.

8 DR. ZACHAREWSKI: I think you have
9 captured the fear aspect or the concern. It is a
10 major concern and I think as the population gets
11 balder, greater and more overweight--I am not
12 describing myself here--you know, everybody is
13 looking for that pill to sort of, you know, regain
14 and capture some youth again, and you are going to
15 find those small populations that are going to have
16 an adverse health effect. Then they are going to
17 go back and say, well, gene X went up and it is
18 associated with my neurodegenerative disease and
19 Pfizer is, you know, a deep-pocket company.

20 DR. WOODCOCK: Yes, from a clinical
21 standpoint I find that somewhat implausible. I
22 don't think from a medical-legal standpoint--I
23 mean, we have had people who have complained that
24 their coffee was too hot. But from a clinical
25 standpoint we know and put on the label most of the

1 adverse events that are associated with a drug, the
2 ones that are common; the ones that are even less
3 common. It is the very rare serious ones that we
4 may miss because they require exposure of 10,000,
5 20,000 people to observe one event.

6 Now, if you think that you are going to
7 find that through this technique soon, I think you
8 are wrong. But I understand that people fear that,
9 but I think that is a very complex, probably
10 genetic and environmental interaction usually that
11 happens and you are not going to be able to predict
12 that from even gene expression data.

13 DR. PENNIE: I think the concern that Dr.
14 Zacharewski articulated there is more between
15 companies having to do with plaintiffs rather than
16 dealing with regulatory agencies, and I think it is
17 an internal concern that organizations have to find
18 their own path through.

19 DR. WOODCOCK: I agree but I think we
20 ought to focus on what is a realistic concern. As
21 you said earlier, some of these fears--actually, I
22 am speaking scientifically, not as a regulator. I
23 think you would have a robust defense usually.

24 DR. LEIGHTON: You briefly mentioned the
25 problem about annotation and the difficulty this

1 leads to across-platform comparisons. I think this
2 may impact on the ultimate biological
3 interpretation of any results across platforms.
4 Can you comment on some of the problems with
5 annotation and a possible way forward with this
6 problem?

7 DR. PENNIE: Well, one of the main
8 problems with annotation I think, certainly for
9 toxicology, preclinical toxicology species is, you
10 know, incomplete genome coverage and the fact that
11 many arrays generated in-house or even in the
12 commercial sector, by necessity, still are not
13 identifying a lot of the genes by name and
14 certainly not by function. So, we have a large
15 number of what are called expressed sequence tag
16 identifiers on some of these microarrays which have
17 to be continually reassessed, as more genomic
18 information is made available in the public domain,
19 as to whether or not those expressed sequence tags
20 are, in fact, related to known homologs that have
21 been encountered in other species.

22 So, one of the main problems, John, I
23 think is lack of genome coverage in test species of
24 interest. But occasionally it can also be just
25 incorrect annotation that a particular species has

1 gone in 3-prime to 5-prime and so the sequence on
2 the gene is, in fact, correct in terms of the base
3 pairs but is completely inappropriate in terms of a
4 hybridization experiment. So, those kind of issues
5 we have encountered experimentally in the ILSI
6 program where we have had a completely opposite
7 gene expression change measured by one platform
8 versus another and only discovered by a lot of
9 detective work that it was an annotation error and,
10 in fact, one of the probe sets was in the wrong
11 orientation. So, there are many possible areas of
12 complexity in annotation.

13 DR. SISTARE: Bill, I am wondering if you
14 can give us a feel for do we need to prepare
15 ourselves at FDA for being able to handle data on
16 thousands of transcripts, or the concern that Tim
17 raised earlier, is it going to drive the industry
18 to look at known toxicants the way we are doing now
19 to find small subsets of biomarker tandems and then
20 just handle 10 or 20 gene transcripts at a time?
21 If that is what we are going to see at FDA, 10 or
22 20 gene transcripts at a time with very focused
23 datasets, we can do that now pretty much the way we
24 do everything else. But if we are going to be
25 seeing 10,000 gene transcripts submitted to us we

1 need to prepare ourselves for that. What is
2 coming, from your perspective? What is going on in
3 industry?

4 DR. PENNIE: Actually, that was a fairly
5 major discussion point at the ILSI open meeting
6 last week, and there was some discussion about the
7 value of submitting raw data and there weren't
8 actually very many people that were advocates of,
9 you know, sending a 20,000 gene expression list as
10 part of a submission in support of a mechanistic
11 argument for risk assessment.

12 Again, I have to stress that as far as the
13 ILSI committee is concerned, we are not in any way
14 empowered nor chartered to make suggestions on
15 regulatory policy, but it seems to me much more
16 sensible, in a risk assessment environment, to be
17 making a mechanistic argument to explain a
18 preclinical tox finding and that that should stand
19 up to a regular scientific interpretation and
20 validation using other methodologies. In those
21 cases you may only have to report the gene
22 expression changes which you consider are germane
23 to the argument you are making, but you reinforce
24 that by using appropriate methodologies or
25 functional work to further prove that that

1 mechanism is, in fact, the appropriate one.

2 In other words, I kind of danced around
3 your question a little bit, Frank, but I think a
4 combination of that kind of approach and a lot of
5 conservatism in the industry, to me and this is my
6 own personal opinion rather than the ILSI committee
7 or the organization I work for, is that I suspect
8 there is enough conservatism that you are not going
9 to be deluged by these kind of submissions until we
10 have a better internal comfort on the usage in a
11 regulatory arena, and perhaps until there is a
12 better articulation on regulatory perceptions on
13 the state of the technology.

14 DR. SISTARE: All right but, given that
15 comfort, would you foresee the future as opening of
16 the aperture and then looking at everything in an
17 experimental design, using a wide open array in
18 generating that data so that you can view
19 everything that is going on simultaneously, as
20 opposed to looking at a light here and there?

21 DR. PENNIE: My personal opinion on that
22 would be that it would be more valuable to make
23 that information available rather than to submit
24 it, in other words, to submit the facts which are
25 germane, or certainly anything that is related to

1 the argument which you are trying to make but to
2 maintain those records of the complete experiment
3 locally, like we do for other methodologies; make
4 those available for further scrutiny should the
5 technology or the regulators desire to look at a
6 complete dataset.

7 DR. SISTARE: I want to understand then
8 what you are saying, that there would be a
9 willingness to generate the data, to do the
10 experiment and to measure multiple thousands of
11 transcripts but what you are saying is the
12 indication from industry would be to submit what
13 they felt was germane.

14 That gets to the question of a lot of the
15 same terminology that Dr. Woodcock used. Using the
16 word "germane"--you know, these kinds of words are
17 very difficult to define and they are moving; they
18 are moving targets.

19 DR. PENNIE: Yes, yes, I agree. I agree.
20 But that, again, was discussed at reasonable length
21 in what I think was a very sensible and appropriate
22 discussion that was held last week. So, I think
23 moving forward, these issues have to be addressed
24 really because until they are there is not going to
25 be a significant amount of data to be quarreling

1 over.

2 DR. KAROL: Thank you very much for the
3 presentation. Well, it is time for a break so we
4 are going to take a 15-minute break and come back
5 at 10:25.

6 [Brief recess]

7 DR. KAROL: I would like to start the
8 second session with Dr. Jarnigan, who will talk to
9 us about dealing effectively with data quality
10 issues, platform differences and developing a
11 database.

12 Topic #2 Toxicogenomic Data Quality and Database
13 Issues Dealing Effectively with Data Quality
14 Issues, Platform Differences
15 and Developing a Database

16 DR. JARNIGAN: Well, thank you very much
17 for the opportunity to be here today.

18 [Slide]

19 I will try to cover several of the issues
20 that we have been discussing already this morning,
21 particularly focusing now a little bit more
22 specifically on what it might be that the agency
23 might want to see as data arrives at their site.
24 Presumably the data will arrive. I firmly believe
25 that in time it will, maybe not today, maybe not

1 this year but within the next four or five years I
2 think you will be seeing a large number of
3 submissions with fairly large chunks of data in it.

4 [Slide]

5 Of course, the vision here, the challenge
6 for us is that almost half of all the drugs that
7 fail are due to efficacy and toxicology problems.
8 Perhaps from the agency's point of view and from
9 society's point of view and patient safety point of
10 view, in this one-year period more than 20 million
11 patients were exposed to drugs that were
12 subsequently withdrawn. That is certainly a risk
13 factor for those patients. If we could do anything
14 to reduce those risk factors, it is a good thing.

15 From the industry's point of view and from
16 the agency's point of view for better new medicines
17 for humans one in ten INDs actually turns into and
18 NDA. To think about that number in a different
19 way, think about it this way, that means that all
20 of the work that has been done, and there is a huge
21 amount of work that is done prior to the time that
22 a compound arrives at the agency for an IND
23 application, you are 90 percent wrong. Nine out of
24 ten times your predictions are incorrect. So, the
25 vision here is to submit better compounds, safer

1 compounds to the agency with the belief that that
2 will improve our odds, improve the quality of
3 medicines that come out of the other end of the
4 process and ultimately, because we are spending
5 time on quality compounds, lower overall approval
6 times.

7 The solution that we, at our organization,
8 are proposing and the concepts of the agency
9 building a database of submission data include
10 bridging the genomic response of an organism,
11 bridging chemistry and genomics to broadly
12 understand a compound's effects in terms of the
13 genomic response of the organism and, as a result
14 of that, to have a better predictive power. That
15 is our vision, to have a better predictive power
16 here.

17 [Slide]

18 Before I start talking about the details
19 of some of the features that I would think are
20 necessary and my organization would think are
21 necessary to make a complete submission, let me
22 just uncover a few of the assumptions that I
23 entered into this analysis so that the background
24 is clear.

25 First off, I am assuming that the sponsor

1 is providing data to support an IND or and NDA
2 application. I haven't in most of this discussion
3 considered the fact that there may be submissions
4 without any IND or NDA supporting feature to it but
5 that could certainly happen. Today's discussion
6 will focus on support of an IND or an NDA and what
7 would be necessary.

8 I assume that the data is part of a larger
9 package and is not the sole and only evidence
10 provided to support a particular claim or a
11 particular series of claims. That is, the data, as
12 already alluded to, is an interlocking set of data,
13 this data, along with other data to contribute to
14 the claim made.

15 Furthermore, I assume that the sponsor has
16 an ongoing microarray effort, and here I am
17 limiting my discussions to gene expression
18 microarrays, not to SNIP analysis or other kinds of
19 genomic analysis of that kind, and if the sponsor
20 doesn't have an ongoing effort that they will be
21 working with a contract research organization that
22 does have an ongoing effort. I guess what I am
23 saying is that whatever the submitting
24 organization, that they aren't doing a singleton
25 experiment; that this isn't the first time they

1 have done the experiment; that their experimental
2 competency in this area is large.

3 [Slide]

4 From the agency side, I also had to think
5 about a few assumptions, and these are the
6 assumptions that I believe the agency probably has:
7 that the agency is willing to develop and train
8 their staff so that the data is meaningfully
9 interpreted and a balanced view of the
10 interpretation is made. An over-reactive view--one
11 oncogene is up--is not a view that would be well
12 tolerated by the industry and not be a view that
13 would be well tolerated by the general public
14 because it probably would kill too many compounds
15 moving forward.

16 Of course, the sponsor, and we already
17 alluded to it in Dr. Zacharewski's comments
18 earlier, the sponsor is concerned about about the
19 future liability of public disclosure as well.
20 That is certainly an issue that is in the sponsor's
21 mind, certainly an issue that would be in the
22 sponsor's mind going forward. I am not sure there
23 is anything that the agency can do about this as it
24 is more of a tort court issue but, nonetheless, it
25 is something that has to be considered and will be

1 considered very carefully by the various sponsors
2 that are submitting data.

3 I assume that the agency is able to accept
4 data in a community-defined standard format and has
5 the capability to assess its overall quality; their
6 staff is well enough trained; their staff
7 understands what the various features of the data
8 are. Furthermore, it is probably the case that
9 technologies are going to continue to develop over
10 time and that the agency will have to continue an
11 effort, a long-term ongoing effort to keep up with
12 future technologies as they come forward. We are
13 not in a static area.

14 The agency desires to deposit the
15 submitted data into an internal database for use by
16 the staff and for comparison for future
17 evaluations, so when a new application arrives they
18 may wish to look back at other compounds of similar
19 type and ask have I seen this pattern before. They
20 do this now by the use of the heads of their
21 reviewers as integrators of this kind of data but,
22 perhaps with electronic submission of all kinds of
23 data becoming more and more a reality and likely to
24 become more and more a reality, this kind of data
25 is already set up to be electronically submitted

1 and probably should be so submitted.

2 Finally, the agency understands that the
3 context of the data is very important, that
4 essentially looking at a single gene or a single
5 pair of genes perhaps isn't the best way to look at
6 such data, and it is the pattern of the response
7 and it is the context of that response in terms of
8 the other data domains, the toxicological
9 endpoints, the clinical chemistry endpoints, the
10 histopathological endpoints that also contribute to
11 one's understanding.

12 [Slide]

13 So, with that background, now let's talk
14 about how array data is different and similar to
15 traditional measurements. If we talk about a
16 sponsor submitting a single gene or half a dozen
17 different genes, how is that really different than
18 the traditional endpoint?

19 I will just start this discussion by
20 looking at a traditional endpoint. Let's talk
21 about ALT elevation. It is measured. It is
22 probably a feature of almost every IND and NDA
23 package that is submitted to the agency. We
24 certainly get data of that kind now. You evaluate
25 it by looking at the mean of the groups and the

1 fact that no single animal within the treated group
2 lies outside the control groups. You may conclude
3 then that the ALT is not significantly changed by
4 the treatment and this is consistent with good
5 hepatotoxic toxicity. That is, it has low
6 hepatotoxicity for the compound. So, how is that
7 really different for gene expression data?

8 Now suppose that we have the case of the
9 community, that is, the scientific community has
10 accepted five RNAs as indicative of a certain kind
11 of hepatotoxicity. Well, the agency and those
12 companies may well get data of the following kind
13 wherein they have the five genes measured as the
14 ratio to control, for example. They have the means
15 and the standard errors. They know that no single
16 individual treatment was outside the range of the
17 control. Would it be reasonable then to assume
18 that these RNAs are not changed? The answer is
19 probably yes. So, again, the sponsor might
20 conclude that there is no significant change and it
21 is consistent with good liver toxicity, that is,
22 low liver toxicity.

23 [Slide]

24 But microarray is different from
25 conventional measurements in some ways, the first

1 of which is that both the agency and the community
2 have a lower familiarity with the technology. It
3 is new technology. There are features that are
4 different from traditional measurements. Of
5 course, this will improve over time. Five years
6 from now this discussion probably will be much,
7 much less significant.

8 There is concern that the survey nature of
9 the data might uncover confounding factors, factors
10 that the sponsor would rather not know about or
11 that perhaps could be confounding to an
12 interpretation. The sponsor, of course, is
13 concerned by an overly reactive view. A certain
14 gene has changed, therefore, we can't go forward.
15 That may be overly reactive.

16 Of course, the agency perhaps has a
17 concern that the sponsor is missing important
18 findings, remembering that the agency may well get
19 data arriving at their site from a new therapeutic
20 class never before exposed to patients but this is
21 the fourth application in the last two years they
22 have seen. They may understand things that the
23 sponsor even doesn't understand. I already know
24 that the agency gives Greenspandian kinds of
25 comments where they say, "we think that you ought

1 to look at the kidney" as a statement. Of course,
2 you have to react to that even though you don't
3 understand why it is important that that be done
4 now.

5 Finally, I think it is very important to
6 note that there is less scientific agreement about
7 how to interpret these findings. This is an area,
8 as Bill Pennie mentioned, of logarithmic growth.
9 The methods for interpretation, the way you go
10 about these kinds of interpretations are improving
11 logarithmically right now. Pattern matching is a
12 key component of this, and this is less familiar to
13 the biological community. We are used to looking
14 at a single group of genes, a single endpoint. So,
15 it is an unusual treatment of the data for most of
16 us. Furthermore, it is different than most of our
17 training as we came along through our various
18 educational paths. It is going to take some time
19 for the community to be educated about this kind of
20 an approach, but it will happen. It will happen
21 faster than we think. I think it is penetrating
22 already and will happen even more quickly than we
23 think.

24 Finally, I would like to point out that
25 there is a perception that microarray data is lower

1 quality and noisier than our traditional
2 measurements. Certainly, five years ago or four
3 years ago that was a very true statement. Today
4 the technology has improved dramatically. The
5 quality of this data is getting to be very high
6 and, when competently executed, I believe it is
7 approaching the quality now of almost any other
8 traditional endpoint and in another five years I
9 think it will be there. So, carefully conducted
10 experiments are accurate and predictive, and they
11 will get even more so over the next several years
12 so this issue should slowly diminish.

13 [Slide]

14 Now let me just summarize what I think a
15 sponsor might want to provide to the FDA in terms
16 of a package of information for microarray data,
17 then we will go through each of the points more or
18 less one at a time. I definitely would urge that
19 the sponsor provide MIAME or MAGE-ML compliant
20 descriptions of experiments and electronic
21 submission of all data. It is not useful in this
22 context to submit data on paper--10,000
23 measurements at a time, 50 microarrays in a typical
24 submission perhaps. It is just not useful.

25 Minimum experimental design metrics

1 similar to that required for any other biological
2 experiments are a definite must. Four or five
3 years ago you could definitely find papers in the
4 literature where a single microarray comprised the
5 whole publication. It was the case where
6 scientists said, well, I am measuring 10,000
7 endpoints so I don't need to do triplicates; I
8 don't need to do multiple biological controls.
9 That is just not acceptable and shouldn't be
10 acceptable here. I don't need to tell the agency
11 how to evaluate biological data, they do it every
12 day, but we need to remind ourselves that that is
13 important.

14 The novelty of this technology requires
15 that additional quality data be submitted to
16 demonstrate the competency of the experimenter.
17 That is true for today and for the next several
18 years. Perhaps in time we won't be questioning the
19 competency of our experimenters but for the next
20 few years I certainly think that that is a
21 probable, definite thing that will have to be done.

22 I would definitely urge the sponsor to
23 provide and interpret the data in a scientific
24 style format. That way the reviewers, particularly
25 in the IND setting where they have only 30 days,

1 don't spend tons and tons of time digging through
2 mountains of data. They can go to the paper, read
3 it and then, if they have further questions, they
4 can dig again to a specific point.

5 Finally, it is very important, we found at
6 our organization, to compare to community accepted
7 RNA biomarkers and comparing to bench mark drugs
8 and toxicants is extremely valuable. It provides
9 the kind of context that you can't get through
10 other approaches. So, the interpretation needs to
11 be in the context of current drugs, failed drugs
12 and toxicants. I think that is a very important
13 feature.

14 [Slide]

15 In the next minute or two I will talk
16 about these minimal standards, a little bit about
17 the quality control data and something about this
18 scientific interpretation. So, in the next few
19 minutes the themes that I am going to delve into
20 with the quality control are constant. There will
21 be three of four different kinds of endpoints that
22 I suggest but their themes are fairly constant.

23 First, measurements versus the lab
24 historical values. Again, my assumption is that a
25 lab is running these experiments all the time and

1 could easily generate the historical data that is
2 necessary by which to compare the quality.

3 The measurements versus an external
4 standard--the agency and NIST are combining to try
5 to define a standard. Definitely, we ought to be
6 carrying these standards through with any
7 experiment that is to be submitted. To provide
8 that data and measurements versus the external
9 standard will be very important.

10 Measurements versus an internal standard.
11 All manufacturers that I am aware of provide a
12 certain number of spike-in standards to include.
13 You ought to use a few of those and include that
14 information as part of your quality control
15 measurements.

16 This is a little bit different than a
17 traditional submission to the FDA and that is, of
18 course, because of the youth or novelty of this
19 technology. You have to prove your competence at
20 doing the experiment and you need to assure the
21 competency of the experiment or you need to assure
22 that it is consistent with internal and external
23 standards and need to assure that it is consistent
24 with historical values. All of those things should
25 be possible in almost any laboratory that is doing

1 these studies routinely.

2 [Slide]

3 Now, the experiment to create a microarray
4 finding from a drug-treated animal is actually a
5 fairly complex experiment. By our count there are
6 286 steps going from a drug in a bottle to a
7 finished microarray experiment at the other end of
8 the process.

9 This pattern is similar for all the
10 different platforms. You do an in vivo experiment.
11 You isolate the RNA and you prepare a target of
12 some sort. You hybridize that. You check the
13 quality of your final product and you load it into
14 an array. Most labs will have some sort of a
15 minimal laboratory information management system
16 underlying this data generation process. So,
17 generating this historical data comparison to
18 controls, and what-not, shouldn't be a big problem.

19 But there are three or four points during
20 this process where I feel it would be very
21 important that minimal information be collected to,
22 one, prove the competency of the lab doing the
23 experiment and, two, to assure anybody else looking
24 at the data now or five years from now or ten years
25 from now that the experiment was done well. Those

1 are shown at the end of the in vivo experiment, the
2 end of the RNA preparation and then at two or three
3 different kinds of checks relating to the quality
4 of the hybridization. These points I believe are
5 independent of platform, and very similar numbers
6 could be found for all different platforms.

7 [Slide]

8 First off, just let me mention a few words
9 about the minimum experimental design just to
10 remind everybody that the minimal experimental
11 design, at least in my mind, is that you have at
12 least three treated samples; you have at least
13 three control samples; and that you carry through
14 with your process contemporaneously three of these
15 RNA standards, external RNA standards, as well as
16 carrying through all samples three spike-in RNAs as
17 a minimum. This would then impute that the minimum
18 experimental size to be submitted is nine
19 microarrays with three RNA standards in every
20 sample. So, minimum biological triplicate; minimum
21 of three untreated or mock treated vehicle
22 controls, processed contemporaneously with the
23 samples to be run; a minimum of three external
24 standard RNAs, also processed contemporaneously
25 with the samples under consideration; and a minimum

1 of three spike-in RNAs.

2 [Slide]

3 Now moving on to the RNA that is used in
4 the experiment, there are a number of different
5 procedures for preparing RNA but they all end up
6 with a product that contains 28S and 18S RNA. They
7 are present in all samples. I propose that the
8 community settle that at the very minimum the mean
9 and the standard deviation and the range for the
10 28S and 18S RNA, the amount of that and the ratio,
11 be reported and probably the traces for those
12 various RNAs that support the package of data be
13 provided. That way, ten years from now if some
14 retrospective analysis is going on and you wish to
15 understand this material the data is available. It
16 is not too much to ask most of the labs. They all
17 have this information in electronic format today so
18 adding it to the data package is not that
19 difficult.

20 I propose that this data be provided for
21 the samples in the dataset for historically similar
22 tissues or cells prepared in that lab, again
23 testifying to the lab's consistency and quality
24 over time, and that the data be provided for this
25 external RNA sample that is executed or processed

1 contemporaneously with the data.

2 [Slide]

3 Now moving on to the hybridization,
4 quality control for the hybridization, there will
5 be two different kinds. First, I propose that for
6 every microarray that is run that the array average
7 signal to background ratio be computed; the array
8 average background; the average raw signal; the log
9 dynamic range for the signal; and the average
10 signal intensity for the three spike-in RNAs,
11 minimum of three spike-in RNAs be reported, and it
12 be reported in some sort of a data table that
13 compares it to historically similar samples for
14 matched tissue type or cell type being run in the
15 lab; the historical samples averaged for the RNA
16 standard that is being run; the historical average
17 for the spike-in RNAs; for the contemporaneous
18 RNAs; and for the contemporaneously run standard.

19 With that, one can easily look at the data
20 and say it is very consistent and this lab can
21 execute a consistent experiment over a long period
22 of time. Again, I am assuming that the lab is
23 processing samples on a fairly routine basis and
24 has this information available to them.

25 [Slide]

1 The last point I would like to make about
2 the quality of the experiment has to do with the
3 internal and external consistency of the samples.
4 One of the easiest ways to measure this is to
5 measure the correlation coefficient for any pair of
6 samples in your dataset. Just assuming three, then
7 you have two pairs in your dataset and you can
8 measure the correlation coefficient versus each
9 other; versus the contemporaneous control; versus
10 the contemporaneous external RNA standard; perhaps
11 versus a historical RNA standard, again getting
12 back to the fact that the lab can do the experiment
13 consistently; and to historically similar tissues
14 or cell types. The report then for the dataset
15 provides the mean and the standard deviation, and
16 perhaps the range of the correlation coefficients
17 for those various datasets.

18 [Slide]

19 That then concludes the main quality
20 control points that I would suggest be included in
21 a submission. Now turning my attention for just a
22 minute to what might be submitted as an
23 interpretation of the findings by the sponsor, I
24 think that should be somewhat in scientific
25 literature style format. That means it starts with

1 an abstract, remembering that, particularly at the
2 IND stage, the reviewer has 30 days so they don't
3 have an infinite amount of time to review this
4 information. They need an abstract; something
5 about the significance of the experiment relative
6 to the specific application under consideration; a
7 brief methods because somewhere in that MIAME
8 submission there is a very long and detailed
9 methods and it is not necessary to make the
10 reviewer wade through that to understand what was
11 done but a brief methods should be provided here; a
12 summary of the quality evidence described earlier;
13 something about the results and a discussion of the
14 results; then conclusions relative to the specific
15 application under consideration and conclusions in
16 the context of a wide variety of other drugs,
17 standard toxicants and failed drugs that are
18 available on the market, that is, some sort of
19 comparison to an external database of some sort.
20 Of course, by providing this summary of the results
21 you are helping the agency help you. You are
22 helping them direct their attention to important
23 points in your data and providing them with some
24 understanding as you see the data.

25 [Slide]

1 So, in summary, I propose that MIAME or
2 MAGE-ML compliant descriptions be provided; a
3 minimum experimental design metrics similar to that
4 you would do for any other kind of a biological
5 experiment. Let's not treat this any differently
6 than other biological experiments. For the next
7 few years at least we need to provide additional
8 evidence that the lab is competent to perform the
9 experiment. Perhaps in time that will go away but
10 today we need that. Your interpretation of the
11 findings, and then a comparison to community
12 accepted RNA biomarkers, so appealing to whatever
13 is in the literature, and comparison to bench mark
14 drugs and toxicants. Your interpretation should
15 look outside the dataset provided.

16 [Slide]

17 Now let me talk a little bit about this
18 external dataset and how one might go about the
19 comparison, and also talk about how the agency
20 might want to build the database comprised of the
21 submissions as they come along, with the goal that
22 in time they will have a contextual view of new
23 submissions as well as a contextual view to look at
24 for things that are approved, close-failed
25 relatives in certain standards and toxicants.

1 It is my belief that the agency might want
2 to build a contextual database. Microarray
3 technology will require that we step into the
4 coming age of electronic submissions. We are still
5 getting a lot of submissions, I understand, at the
6 agency that are largely paper in nature but we will
7 be going into electronic submission and microarray
8 data is already electronic in format so it can
9 probably lead the charge here. Paper submission of
10 microarray data is not very useful. If you think
11 of a million data points on paper, it just doesn't
12 provide any interpretive context for anybody. The
13 agency is probably not going to retype that data
14 into a computer to analyze it so it has to be done.

15 I believe that this contextual database
16 will be used by the agency to better understand the
17 technology. It will be used by the agency to look
18 at the data in the context of other submissions,
19 remembering that the agency may well get data and
20 have a view on data that is not available to the
21 sponsor because new therapeutic modalities are
22 being presented to the agency that have never
23 before come along. So, they may have a view on
24 data from two or three of these that the rest of
25 the industry doesn't have. The contextual

1 database, in our experience, is highly useful to
2 provide meaning and a balance to the
3 interpretation, and I would like to illustrate the
4 point about the balance in a slide or two.

5 [Slide]

6 Before I do that though, I would like to
7 turn my attention to what will the agency do with
8 this data. Again, promoting a balanced view has
9 got to be one of the central objectives. It is
10 very easy to overreact to some single data point or
11 two or three in the data. You need to be aware of
12 what truly significant events are. The way you get
13 that awareness is by developing a community
14 consensus around what are useful RNA biomarkers,
15 and the way we get that community consensus is by
16 doing a lot of experiments. So, you need to ground
17 the analysis in the context of real-world effects
18 of drugs, failed drugs, withdrawn drugs, standards
19 and toxicants. So, a reference database is needed.

20 [Slide]

21 Such reference databases are being
22 produced and prepared now and are available. What
23 should be in one of these reference databases?
24 Well, it should contain a wide diversity of
25 successful drugs, failed drugs, toxicants and

1 standards. That is, you need to understand both
2 the pharmacology of compounds as well as their
3 toxicology. In our experience one cannot truly
4 divorce those two fields, one from another. You
5 must understand what the drug does
6 pharmacologically as well as toxicologically.

7 The database probably should include
8 multiple tissues, doses and times, and probably
9 cells in culture as well. The linkage of the
10 expression data to orthogonal data domains is very
11 important. You find a lot of good, useful new
12 insights by understanding what goes on
13 pharmacologically, including site interactions with
14 on and off target events. What happens with the
15 histopathology in animals dosed with these
16 compounds, clinical chemistry, hematology and
17 chemical structure are all useful orthogonal data
18 domains and should be present in a contextual
19 database, and in vivo and in vitro experiments so
20 that you may bridge between your in vitro findings
21 to your in vivo findings.

22 [Slide]

23 Let's just look at what the benefits of
24 using a reference database are. We have heard
25 allusion to this kind of result both in Janet's

1 talk and in Bill's talk earlier. This is data
2 taken directly from such a database looking at
3 three oncogenes. I just picked out three to look
4 at them, just for illustration, EGF-receptor,
5 cKit-oncogene and BCL2. All of these drugs cause
6 statistically significant elevations of these
7 oncogenes.

8 One single oncogene change is certainly
9 not significant. It is certainly the case that
10 these oncogenes, as Janet says, weren't put into
11 the genome to cause cancer; they are there for the
12 cell and the organ to respond to specific
13 environmental stimuli. Drugs are environmental
14 stimuli and they, therefore, cause changes in these
15 oncogenes. Elevation of one is not in itself
16 evidence of cancer. These drugs are not oncogenic
17 in general.

18 So, the context provided by such a
19 database provides a balanced view and will
20 accelerate the adoption of this technology because
21 we won't have to wait for these experiments to be
22 done as singletons in individual academic labs over
23 the next several years.

24 [Slide]

25 So, to summarize and then move on to

1 looking forward, electronic submission of the
2 data--a definite yes. Standard format--a definite
3 yes. Perhaps the agency should help the process by
4 helping devise some sort of input tool for the
5 standard data format, a better input tool than is
6 currently available. I am reminded very much of
7 what it was like to submit data to GEN Bank before
8 SCAN was available. It took hours and hours just
9 to get it into the form to be put into GEN Bank.
10 Once the SCAN tool was provided to the community it
11 went much faster. An analogous situation happened
12 with PDB a few years before that where data was
13 submitted in all sorts of formats. It was
14 impossible to database. Once an input tool was
15 developed and Brookhaven took over the job of
16 putting together a simple database it became a
17 useful tool.

18 Minimum experimental design--we can't
19 forget what we learned on how to design biological
20 experiments years ago. It is still valid in this
21 technology. New technology does not obviate those
22 needs.

23 For the next few years, perhaps
24 diminishing with time but for the next few years
25 the experimenter needs to prove their competency at

1 doing the experiment by providing additional data
2 beyond what would normally be provided with any
3 other kind of biological endpoint.

4 Sponsor's interpretation of the data I
5 think is extremely important. It should not be
6 ignored. A pile of data should not be submitted
7 without much support as a written document of some
8 sort.

9 Finally, comparison to community accepted
10 RNA biomarkers, there are some in the literature
11 already and we should definitely look at those, and
12 also comparison to bench mark drugs and toxicants,
13 withdrawn drugs and so forth.

14 [Slide]

15 So, conclusions and looking forward.
16 Microarray technology is ready to contribute to the
17 drug discovery process and to the approval process
18 today and I believe that as we start to do this we
19 will start to see improvements in our overall
20 efficacy of this process, improvements in the
21 safety of compounds that are submitted,
22 improvements, therefore, in the overall quality of
23 medicines that are being used to treat patients.

24 Simple assurances of quality are
25 definitely needed for the time being. Contextual

1 databases to allow meaningful interpretation are
2 needed and some are available. We need to develop
3 as a community a consensus around what are
4 meaningful RNA markers. This is starting to
5 happen. I think it will accelerate over the next
6 several years.

7 Again, requirements beyond normal
8 verification of data quality will diminish as
9 community sophistication improves. I will say we
10 have done a number of experiments analyzing data
11 collected over different platforms that can make
12 accurate predictions on data prepared in several
13 different platforms. The same biology is found
14 regardless. These technologies all do measure the
15 same biology and that is the critical event. That
16 is what we are after, to measure the biology and
17 understand that that biology is significant for
18 safety or for efficacy.

19 Finally, I believe and definitely know
20 that clinical applications in accessible human
21 tissues for this kind of RNA transcription
22 measurements will come and will be parts of
23 submissions very shortly to the agency.

24 [Slide]

25 So, the result of this activity--building

1 a database, providing the data in an electronic
2 format carefully controlled--will be to improve the
3 predictive power of the animal studies that are
4 undertaken and of looking at clinical samples in
5 accessible tissues. This will help realize this
6 vision to get better compounds submitted; safer
7 compounds submitted and approved; and lower the
8 overall approval time because we spend our time on
9 the best compounds. Therefore, we are addressing
10 the problems of patient exposure to drugs which are
11 subsequently withdrawn because there are fewer
12 subsequent withdrawals perhaps. It addresses the
13 problem that only one compound in ten enters and
14 IND passes an NDA test. Thank you and I will be
15 happy to take questions.

16 DR. KAROL: Thank you very much. We have
17 time for perhaps one or two questions.

18 DR. GOODMAN: I like the portion of your
19 presentation dealing with providing the information
20 in the format of a scientific interpretation. But
21 just to be a little argumentative, why do we need
22 the rest? That is, it seems to me that one way
23 that would stifle what I think is a very promising
24 technology is to, at the outset, be too
25 prescriptive as to these are the way the data will

1 be submitted; these are the types of information
2 that one wants; and maybe also to be too
3 prescriptive in terms of talking about setting up a
4 database if it will result then in driving, if you
5 will, the experiments. That is, now the data must
6 be submitted to fit the database as opposed to what
7 scientifically might be best.

8 DR. JARNIGAN: First off, I would point
9 out that if you read the MIAME and MAGE-ML
10 standards, they actually have a tremendous amount
11 of latitude built into them. They aren't overly
12 prescriptive. Perhaps I am wrong but certainly I
13 don't read them as being overly prescriptive.
14 Provision of the data as a whole, meaning all
15 10,000 genes or 20,000 genes at a time, that is an
16 issue that, as we discussed, will be difficult for
17 the community to address and I think the difficulty
18 isn't with the agency; the agency can handle this
19 problem well. The problem is the tort issue. The
20 tort issue probably has the pharmaceutical
21 companies more concerned. So, they are worried
22 about the future liability--the issue that was
23 brought up over here earlier today--the future
24 liability for something being discovered five years
25 from now or ten years from now that says you should

1 have found this ten years ago. We don't proscribe
2 it on ourselves now. I certainly know that
3 submissions arrive that have issues that ten years
4 from now are bound to be a problem but, still, it
5 is going to be something that they consider very
6 heavily.

7 To your question, I think that your
8 question is are we proscribing it too much? Will
9 this make the experiments fit into a nice, neat
10 box? I don't think the electronic submission
11 standards do demand a nice, neat box. They just
12 demand certain basic things, many of them you
13 already require of yourself for all other kinds of
14 data that you submit to the agency.

15 DR. KAROL: Thank you. I am afraid we
16 will have to move on. Thanks very much. The next
17 presentation is by Dr. Quackenbush on data
18 processing, statistics and data presentation.

19 Data Processing, Statistics and Data Presentation

20 DR. QUACKENBUSH: Thank you very much for
21 the invitation to come here.

22 [Slide]

23 My background isn't in toxicology; my
24 background really is in other areas of applications
25 for microarrays so I may not be able to address all

1 the questions specifically associated with
2 toxicology. What I am going to try to do is
3 address questions associated with data handling and
4 management and, as Frank asked me to do, try to
5 point out what some of the issues and challenges
6 are and take you, if I have time at the end,
7 through one or two examples where we have tried to
8 apply some of the lessons we have learned for
9 understanding array data.

10 I have prepared a handout for you and I
11 have already deleted a large number of those
12 slides. I tend to have too many slides always and
13 am then deleting them in the last few minutes, but
14 I haven't rearranged the order so you won't have to
15 skip through too much.

16 [Slide]

17 What I really wanted to start with in
18 looking at this problem is actually just looking at
19 the problem from the start, which is selecting the
20 appropriate platform.

21 [Slide]

22 This, in fact, can be a bit of a
23 challenge. As you know, there are two array
24 platforms. One is a resequencing-based platform
25 that developed out of the Affymetrix resequencing

1 chip in which oligos are synthesized de novo on a
2 glass substrate.

3 [Slide]

4 Then two biological samples are labeled,
5 hybridized independent arrays, scanned, relative
6 expression levels are measured, and from that
7 relative expression level measurement on two
8 independent arrays one can derive changes between a
9 query and control sample or between any two samples
10 in the experiment.

11 [Slide]

12 The alternative approach is to take DNA
13 fragments, whether PCR products or long
14 oligonucleotides, and array those on a glass
15 microscope slide using a robotic spotting device,
16 and then RNA is extracted from two different
17 samples. In this case, the RNA is labeled with
18 distinguishable fluorescent dyes, although that is
19 not always the case. Some people treat these
20 arrays also as single color assays and perform
21 independent hybridizations, but the most common
22 implementation, in fact, is to use these paired
23 samples, hybridize them to a single array; measure
24 fluorescence intensities and analyze them to
25 identify patterns of expression. The real

1 challenge, of course, is to take those patterns of
2 expression and interpret them in some kind of
3 meaningful biological context.

4 [Slide]

5 This was supposed to unfold and it really
6 didn't unfold very well at all. Somehow it got
7 rearranged in transfer. But, fundamentally, the
8 array assays start with looking at genes because
9 that is the object we want to understand. Those
10 are represented by one or more elements on the
11 array. We measure fluorescence intensity for each
12 one of these elements and from that an inferred
13 expression. We like to link that back to the gene.

14 In fact, every part in this process has
15 potential pitfalls and is problematic. One of the
16 most important is moving from spots on the array to
17 relative expression measurements. This is
18 something which I know was discussed to a certain
19 extent this morning but it is absolutely important.
20 All of the laboratory handling of the samples--how
21 you choose the samples; how you deal with them--has
22 a big effect on what you ultimately measure. In
23 fact, we are not measuring expression, we are
24 inferring expression based on fluorescence
25 intensity, which is based on hybridization, which

1 is based on relative RNA levels. So, if the
2 samples are allowed to degrade at room temperature
3 for a long time before the RNA is extracted, if the
4 RNA is degraded before it is labeled, then what you
5 see on the array expression may or may not, in
6 fact, really be the relative expression for those
7 genes.

8 The other important aspect is that what we
9 call the genes on the arrays really have to be
10 carefully defined because those genes, in fact, may
11 not be what we think they are when we look at the
12 annotated elements on the array. I will come back
13 to one or two sources of that in a minute.

14 [Slide]

15 So, there are some platform related
16 issues. One is the lack of standardization which
17 makes direct comparisons of results between
18 laboratories a challenge, not an insurmountable
19 challenge but definitely a challenge.

20 This says "lot-to-log," in fact, it should
21 say lot-to-lot variation in arrays. Lot-to-lot
22 variation in arrays can introduce artifacts and the
23 results can be dependent on either the biology or
24 on artifacts on the arrays, and that can include
25 the log-to-log variation as well as which

1 technician performed the assay, which day of the
2 week they did it, the reagent lot. So, all of
3 those have to be very carefully managed and
4 controlled to make sure that when you are actually
5 looking at an experiment what you are seeing is the
6 real variation that comes from the biology, not
7 from the fact that the arrays were done on
8 Wednesday rather than Friday when everybody was
9 ready to go home.

10 Commercial arrays provide a standard and
11 remove some of the design considerations, in
12 particular the idea of using one sample per array
13 which makes all of the experimental design much
14 easier. It presents different challenges for doing
15 analysis, but the cost is significantly greater for
16 doing these commercial arrays or using these
17 commercial platforms which drives a lot of array
18 users, particularly academic users, to use in-house
19 arrays.

20 But no matter what, one of the most
21 important things, which I tried to emphasize
22 earlier, is really the demand for a good LIMS
23 system to track every single aspect of the
24 experiment. Those have to be tracked not only to
25 report them but, in fact, to really interpret and

1 understand what you are seeing and to identify
2 potential sources of artifacts.

3 [Slide]

4 Once an array platform is selected we want
5 to move on and actually start doing array analysis.

6 [Slide]

7 There is a general strategy for doing the
8 microarray analysis. The first is to choose an
9 experimentally interesting and tractable model
10 system. To design an experiment with comparisons
11 between the appropriate variants and to include the
12 appropriate controls you have to include sufficient
13 biological replication to make good estimates,
14 which is a point that has been emphasized here
15 before. Once you have designed the experiment and
16 start doing hybridizations and collect data, that
17 data has to be effectively managed. The data then
18 has to be normalized and filtered so you can make
19 appropriate comparisons between different
20 hybridizations, different individuals, different
21 labs, different experimental protocols.

22 Then, and only then can you begin to mine
23 data to look for biologically interesting patterns
24 of expression. Then, in order to interpret those
25 patterns of expression, you would like to integrate

1 the expression data with other ancillary data,
2 including information like the genotype, the
3 phenotype, the genome, the annotation of the
4 genome, the treatments you are using, the dose, the
5 dose response, other physiological measures. In
6 fact, probably the biggest challenge is moving from
7 looking for these patterns of expression to really
8 trying to interpret what they mean based on the
9 underlying biology.

10 [Slide]

11 The first step in doing all of the data
12 analysis is actually having useful annotation on
13 the array.

14 [Slide]

15 While this may not sound like a
16 significant challenge, in fact it is. You may have
17 read that the genome has been finished yet again,
18 the human genome. That was published in April of
19 this year. Based on my definition of
20 "finished"--that we have a complete genome
21 sequence; that we understand where all the genes
22 are; we have functional assignments for those--the
23 genome is far from complete. That doesn't mean
24 that the draft human, mouse and rat genomes are not
25 useful. In fact, they are tremendously useful for

1 analyzing the data. But one thing I want to
2 emphasize is that they have to be taken with a
3 grain of salt.

4 So, we do annotation on the arrays that we
5 build in-house and for the array assays we perform
6 in-house. These are built around a series of
7 databases we call the TIGR gene index databases. I
8 am going to talk about these databases only because
9 for us the annotation process is important in
10 understanding potential pathologies that arise in
11 that annotation, important for interpreting the
12 results.

13 [Slide]

14 So, we have built these now for nearly 60
15 species. This is an example of what one of those
16 records look like. It comes from taking gene and
17 EST sequences. ESTs are still important even in
18 the realm of the complete genome because many
19 arrays have ESTs representing, including a lot of
20 the commercial arrays. So, we take the ESTs and
21 gene sequences. We assemble them. We provide
22 information about those assemblies, links to public
23 databases and information such as annotation based
24 on sequence similarity search and gene content,
25 links to other databases, in this case to the mouse

1 genome Informatics database at Jackson Labs, and
2 increasingly maps of things like the completed
3 genomes.

4 [Slide]

5 Another important element of the
6 annotation though is to try to understand the
7 functional roles that these genes play and, in
8 particular, for interpreting the results in the
9 context of the biology you are examining, being
10 able to project additional annotation and
11 classification ontologies onto the genes is
12 incredibly important.

13 So, one of the things we use are the gene
14 ontology terms or GO terms. Gene ontology is an
15 attempt to define in a rigorous fashion classes for
16 genes in three broad categories. The first is
17 molecular function; the second is biological
18 process; and the third is cellular component. So,
19 what we try to do is take each one of our array
20 elements and attach this kind of annotation which
21 allows us to place genes in broad biological
22 classes.

23 An additional attempt that we make in
24 annotating our array elements is to provide EC
25 numbers. The enzyme commission numbers allow the

1 array information to be projected back onto things
2 like metabolic pathways.

3 [Slide]

4 We are also very interested in building
5 cross-species comparison. We built a database
6 which is known as EGO, the eukaryotic gene
7 orthologues.

8 [Slide]

9 What this database attempts to do is to
10 use pair-wise comparisons between sequences to
11 identify possible orthologues requiring transitive
12 reciprocal best matches between multiple species in
13 order to define an orthologue set.

14 [Slide]

15 This has actually been very useful for
16 identifying orthologues in mammals as well as
17 across kingdoms. So, in this case what we have are
18 sort of orthologues from human, mouse, rat, zebra,
19 fish, potato, tomato, barley, beet, rice, maize.
20 In fact, even using DNA sequencing you can identify
21 these.

22 In the context of toxicology, while
23 looking at human or arabidopsis orthologues might
24 not be that interesting, really identifying the
25 human, rat or mouse orthologues is going to be

1 fundamental for interpreting a lot of the data.

2 [Slide]

3 One of the other important lessons I think
4 we have learned in looking at this data is just the
5 value of seriously questioning the annotation that
6 is provided for the genome sequence, and these are
7 just some examples I would like to show. These are
8 the official ensemble gene predictions, as well as
9 alignments to EST data from human, mouse, rat,
10 cattle and pig, the most highly sampled mammals.

11 In many instances the ensemble annotation
12 is quite good and recapitulates the gene structures
13 that you see in these other species. In other
14 cases there are ensemble annotations which have no
15 EST support despite having nearly 15 million
16 mammalian ESTs available. There are other very
17 clear examples where there is beautiful EST support
18 among multiple species or a single species but no
19 annotation.

20 So, one important lesson to learn is that
21 the genome and its annotation is only a hypothesis.
22 That hypothesis still remains to be tested. In
23 fact, one of the things I didn't emphasize at all
24 is that the assignment of gene function to many of
25 these genes is based only on sequence similarity,

1 and sequence similarity search is not an actual
2 experimental evidence.

3 We have many good examples, in particular
4 for arabidopsis where there has been a complete
5 genome duplication, where genes that have been
6 assigned exactly the same function in fact respond
7 very differently and have clearly different
8 functions. The annotation is an ongoing process in
9 biological interpretation of response to any kind
10 of challenge using array data and it is really
11 going to require careful follow-up of what that
12 annotation is.

13 [Slide]

14 Another important aspect of this entire
15 problem is to try to address this cross-species
16 comparison and the cross-platform comparison
17 problem.

18 [Slide]

19 In order to do this my group built another
20 tool, that we call Resourcerer, that allows you to
21 take microarray resources and provide annotation
22 for them, including things like links to locus
23 link, links to the physical map and orthologue
24 identifications and gene ontology assignments.

25 [Slide]

1 This tool, based on having an orthologue
2 database, allows us to compute cross-species and
3 cross-platform comparisons so in this case it is a
4 cDNA clone set linked to the Affymetrix human U95A
5 array. Another important element is having access
6 to the genome sequence, in which case we can take
7 things like genetic markers and simply ask
8 questions, if we have an area of the genome that
9 has been linked to a particular response through
10 genetic mapping, can we find elements on the array
11 that will allow us to provide an intersection
12 between genetic data and expression data.

13 In the context of testing compounds this
14 may not be important; in the context of
15 understanding response it may be very important as
16 different mouse and rat strains, in fact, are known
17 to respond differently to different challenges.

18 [Slide]

19 So there are real annotation issues. The
20 first is the complete genome is incomplete. The
21 gene names are not well defined so one gene may
22 have many names. One gene may have many sequences
23 representing that gene and they may not be the same
24 sequences, and one sequence, in fact, may have many
25 names. So, looking across the aliases for each

1 gene can really be an important problem and this is
2 one place where standardization can be absolutely
3 essential and helpful in interpreting results.

4 Analysis interpretation depends on having
5 well annotated array elements and gene sets,
6 including gene names, gene ontology assignments and
7 information about pathways. Cross-species
8 comparisons also require a very careful analysis
9 and knowledge of orthologues and paralogues in
10 order to draw the correct inferences.

11 [Slide]

12 Another important area in terms of
13 applications and annotation and analysis is
14 developing appropriate tools and techniques for
15 analysis.

16 [Slide]

17 I am actually going to skip a number of
18 the slides I put in here, which is sort of
19 elementary introduction to some of the challenges,
20 but there are important steps in the entire
21 analysis process.

22 [Slide]

23 The first is choosing an appropriate
24 experimental design. In fact, in the statistics
25 community, as you probably know, there has been a

1 great deal of discussion and debate about what the
2 appropriate experimental design is and I can tell
3 you that there are important differences between
4 statistically sound designs and experimentally
5 tractable designs that aren't always addressed in
6 these debates in the literature. So, those have to
7 be addressed appropriately and carefully.

8 You perform the hybridization and generate
9 images. You analyze these images to identify genes
10 that are differentially expressed and their
11 expression levels, usually measured as
12 hybridization intensities. The data is typically
13 normalized in a variety of different ways to
14 facilitate comparisons between elements on a single
15 array and between multiple hybridizations, and then
16 we want to analyze the data to find the
17 biologically relevant patterns of expression.

18 [Slide]

19 Again, I will just mention that my group
20 builds a lot of software for addressing these
21 issues and if you would like to talk about
22 particular algorithms we can discuss them.

23 [Slide]

24 The first piece of software I showed you
25 is actually our data management software that

1 allows us to track information through the lab.

2 All this software we provide to the community with
3 source code.

4 [Slide]

5 One step in the process though which is
6 absolutely fundamental is normalizing expression
7 data. Normalization is actually important for
8 facilitating comparisons across arrays. One of the
9 simplest things you can do is to simply look self
10 versus self hybridization, compare a hybridization
11 assay to itself using either a two-color assay or
12 using multiple hybridizations across multiple chips
13 with the same sample.

14 What you would expect in an assay like
15 that is that every gene, in fact, should give you a
16 ratio of one or a log ratio of zero. In fact, you
17 know that is not true. There may be unequal
18 labeling efficiencies or hybridization or detection
19 efficiencies for the different dyes. There is, in
20 fact, inherent noise in any measurement you make
21 and there is noise in the systems that are used.
22 In fact, even when we are looking at self versus
23 self hybridizations comparing the same sample to
24 itself, we may, in fact, be seeing biologically
25 relevant differential expression if we are taking

1 two RNA extractions from the cell line drawn in two
2 different flasks in the same incubator. Not all
3 RNA is equal and handling those samples can affect
4 them.

5 So, very often when people look at this
6 kind of self versus self hybridization they are not
7 seeing what they expect because they are not
8 looking at what they expect. Normalization is a
9 process designed to bring appropriate ratios back
10 to one.

11 [Slide]

12 The technique that we use for looking at
13 two-color microarray assays is locally weighted
14 linear regression in which we try to subtract out
15 this sort of systematic curvature you see. What we
16 are looking at is the logarithm of the ratio. It
17 is really a measure of the log of the intensity on
18 the array, and we try to center that data and also
19 smooth it out. Whether doing that centering is
20 appropriate or not is, in fact, open to
21 interpretation and really depends on what the
22 biological experiment is that is under way.
23 Probably the nicest discussion of this is a recent
24 paper that appeared from Frank Holstege and his
25 group in which they looked at a situation in which

1 transcription is shut down and normalization of the
2 data, as it is typically performed, is not
3 appropriate.

4 One of the other things that is important
5 to realize is that when people talk about
6 differential expression, how they actually measure
7 that differential expression is fundamental to
8 interpreting the result and often ignores the real
9 structure in the data. So, if we look at the log
10 to the ratio and, in fact, pick a two-fold up or
11 down regulation, two-fold here is represented as a
12 log ratio of plus one or minus one. In fact, at
13 low intensities, as we approach the detection
14 threshold on the array, two-fold may be completely
15 meaningless, while at higher intensity something
16 like 1.2- or 1.3-fold may, in fact, be a
17 significant change. So, we have to be very careful
18 and very intelligent about the way in which we even
19 identify what we mean by differential expression,
20 and we have to use the appropriate tools for
21 identifying genes, including the appropriate
22 statistical tools.

23 [Slide]

24 Again, my group builds software for doing
25 some of this normalization, as well as doing data

1 analysis and we can talk about the various
2 algorithms.

3 [Slide]

4 There are some issues though. The first
5 is that there is no standard method for data
6 analysis. In part, that is tied to the fact that
7 there is no standard method for experimental
8 design. The same algorithm with a small change in
9 parameter, such as a different distance method, can
10 produce very different results when we are
11 analyzing expression data. Data normalization
12 plays a big role in identifying the differential
13 expressed genes and how you scale within and
14 between arrays can affect the results. Much of the
15 apparent disparity though that is observed in
16 microarray datasets, in fact, can be attributed to
17 differences in data analysis methods. When people
18 pick out a group of genes from one set of
19 experiments and do experiments on a different
20 platform and pick out a different set of genes and
21 they say, oh my God, they are discordant. In fact,
22 that may not be the appropriate test because how
23 you pick out that class of genes depends on the
24 assumptions, depends on the software, depends on
25 the parameters. In fact, my analysis and the

1 analysis my group has done seems to suggest that a
2 lot of that comes from the different analysis
3 methods, starting with things like image processing
4 and moving on to normalization and data mining.

5 [Slide]

6 Another important element which has been
7 discussed here at length is data reporting
8 standards so I am not going to discuss this in very
9 much detail, other than to say that I have been
10 involved in this MIAME consortium to try to define
11 standards. Really, the emerging standards are that
12 we have to report everything that is relevant to
13 the measurements that are made on the arrays.

14 [Slide]

15 The good thing I think which is motivating
16 the community to adopt these standards is that the
17 journals themselves have been asking for the
18 standards to be advanced and now most of the large,
19 high profile journals require that data be
20 submitted in a MIAME compliant fashion.

21 [Slide]

22 One of the important things I think that
23 is emerging from all of this is the development of
24 an extension of MIAME called MIAME-TOX. If you
25 want to take a look at this standard, it is going

1 to be discussed in greater detail at the upcoming
2 MGED meeting in September, in France. But,
3 clearly, implementation of all these standards is
4 going to require development of ontologies to
5 describe the experiments in more detail, the
6 analysis tools in more detail and, in fact, the
7 experimental challenges, particularly the
8 toxicological challenges in very clear,
9 well-defined detail.

10 [Slide]

11 Our software also has to be developed to
12 read and write MAGE-ML. There was a question about
13 the flexibility of sort of the openness of MIAME
14 and MAGE-ML. MIAME in fact was initially proposed
15 as a very flexible standard, in large part because
16 I think we realized within the community that the
17 standard is still being developed. In a similar
18 fashion to the MAGE-ML, the XML-based reporting
19 standard is very open to development of new
20 applications and new techniques in particular
21 extensions which will be appropriate to toxicology.

22 [Slide]

23 The public databases clearly need to be
24 extended to meet the toxicological needs or new
25 databases have to be created to include that

1 information.

2 [Slide]

3 I wanted to talk a little bit about some
4 of the science. In fact, what I am going to do is
5 I am going to skip a lot of this talking about the
6 biology, but I am going to bring up one important
7 issue.

8 [Slide]

9 The two examples I was going to show you
10 are an example of how we use genetic maps to try to
11 refine expression data; another one in which we use
12 GO terms to try to refine expression data.

13 [Slide]

14 One of the things I am going to talk about
15 very quickly is the problem of trying to predict
16 outcome since that seems to be a lot of the
17 challenge in toxicology. The problem for us is
18 that we are looking at patient samples in a cancer
19 study funded by the NCI in which we want to try to
20 use expression fingerprints as a phenotypic measure
21 for predicting things like survival, response to
22 chemotherapy and outcome.

23 [Slide]

24 The first problem we wanted to attempt to
25 address is a problem which is very simple, the

1 problem of classifying tumors. So, what we did is
2 we took a number of adenocarcinomas. We profiled
3 them on 32,000 element human arrays.

4 [Slide]

5 And, we used a variety of techniques for
6 predicting which genes would, in fact, be the most
7 appropriate for classification. The approach we
8 finally chose was one in which we used the neural
9 network and in terms of toxicology, neural networks
10 may in fact be problematic because they are black
11 boxes. In terms of doing classification though
12 they are actually quite effective because what we
13 can do is use input data, and here the input data
14 are statistically significant genes which are good
15 for separating out different tumor types and now
16 can be trained to predict the class of tumor.

17 [Slide]

18 We built a classifier that was 94 percent
19 accurate using data on cDNA arrays. Part of the
20 reason I wanted to talk about this experiment at
21 least a little bit is because what we realized we
22 needed to be able to do is to extend this
23 classifier. So, we surveyed the literature and
24 found available data that we felt we could use.
25 For a variety of reasons, the only available data

1 that was published that we felt we could use was
2 data that was collected on Affymetrix chips.

3 [Slide]

4 So, we scoured web sites. We downloaded
5 the data. We ended up with 540 tumor samples
6 representing about 95 percent of all human cancers,
7 representing 21 different tumor types.

8 [Slide]

9 The real challenge, of course, was to be
10 able to do a cross-platform comparison in which we
11 were really looking at three platforms because even
12 the two Affymetrix platforms don't have the same
13 probe sets for all of the genes on the array. If
14 you have the same gene you may, in fact, have two
15 different probe sets.

16 So, we had to do some kind of
17 cross-platform normalization. The approach we used
18 for this was actually fairly simple. On our
19 spotted arrays we compare everything to the
20 universal reference. What we did was we took these
21 Affymetrix arrays and we hybridized our universal
22 reference to those arrays and used the data on a
23 gene by gene basis to scale each one of the
24 expression levels. Having done that, we got a
25 dataset that was comparable that we could then use

1 to train this classifier and actually make tumor
2 predictions.

3 [Slide]

4 The short version of this is that at the
5 end of the day, even looking across multiple
6 platforms, we were able to build a classifier that
7 was nearly 90 percent accurate, approaching the
8 level at which a pathologist, over the course of a
9 number of tests, can actually classify these same
10 tumors. We have extended this now to look at
11 survival and to predicting outcome, and I can tell
12 you that it has been equally successful in these
13 other applications.

14 [Slide]

15 So, what are the real challenges in
16 analyzing microarray data? One is that statistical
17 significance is not necessarily the same as
18 biological significance. Having enough replicates
19 to define statistically significant results is
20 important but it is not the only thing, and one of
21 the things we have to remember when we analyze this
22 data is to look at the biology.

23 Another real challenge which I think
24 people are realizing is that if you take this
25 system and perturb it many genes change their

1 expression levels, not just one. So, in fact, a
2 very simple challenge in which you try to just
3 perturb one single pathway can produce a lot of
4 unexpected changes, and those changes may be
5 difficult to understand. One of the first
6 observations we made in tumors is that genes like
7 osteoparten change. We reported this in a paper
8 and one of the referees wrote back and said
9 obviously this data is nonsense because osteoparten
10 is a bone protein. So, really you have to be very
11 careful at how you look at these and how you
12 interpret the data in light of the annotation.

13 Multiple pathways and features in the data
14 can be revealed through different analysis methods
15 so the same dataset can show you four or five
16 different patterns, depending on how you look at it
17 and how you interpret it has to depend on biology.

18 Genes which are good for classification or
19 prognostics may, in fact, not be biologically
20 relevant in the sense that there may be some of
21 these ancillary changes that occur as you perturb
22 the system, and they may be very important for
23 making the predictions but they may not tell us
24 about the biology.

25 Finally, extracting meaning from

1 microarrays will require now software and new
2 tools, but the most important thing we need is more
3 data collected and stored in a standardized
4 fashion.

5 [Slide]

6 I am seeing that I am running over time.
7 The most important thing I think really to take out
8 of all of this is that there is still a lot of need
9 for standardization but one of the most important
10 needs we have in terms of developing statistical
11 tools and analysis tools and techniques is just
12 good data which is collected and stored in a
13 standard way.

14 So, thank you for the invitation and thank
15 you very much for the opportunity to talk here
16 today.

17 DR. KAROL: I would like to take just one
18 short question.

19 DR. WATERS: I think you accurately
20 captured the complexity of this field that we are
21 evaluating today. The question that I have, and
22 really in a way it is a comment, has to do with the
23 capture of the toxicology side of the dataset. You
24 mentioned that briefly as you went through the
25 evaluation of the various types of measurements

1 that should be made. Could you comment a bit more
2 about what you really think the importance is in
3 capturing that data. We heard in the previous
4 presentation that context was all important but we
5 didn't hear anything about what sort of toxicology
6 information must be captured with regard to the
7 microarray datasets in context.

8 DR. QUACKENBUSH: I am still learning a
9 lot about what toxicologists do and what they think
10 is important.

11 [Laughter]

12 So, for me, this has been a bit of a
13 challenge but in terms of actually interpreting the
14 data, I think what you collect has to reflect the
15 questions that you are asking. My understanding of
16 the toxicology field has to do with trying to
17 predict what the response of the organism is going
18 to be to a particular compound. So, in my view
19 some of the things that are clearly important for
20 understanding this are the compound, its structure
21 because ultimately down the road we want to do data
22 mining and what I would like to do is be able to go
23 back and say, okay, I see this response. What I
24 would like to do is know what causes that response.
25 Is it compounds that interfere, are known to

1 interfere with a certain pathway? Or, is it
2 compounds which simply have the right set of
3 aromatic rings attached as what we thought were
4 non-functional aspects or non-functional parts of
5 the molecule? So, the compound, its structure, the
6 dose, the time period or the time course
7 information, information about the animal strain,
8 genotype if it is available. I think every piece
9 of information that you have up front is going to
10 be valuable at a later date for mining this data
11 and understanding the effect.

12 DR. WATERS: And these need to be captured
13 in the database.

14 DR. QUACKENBUSH: I think they ultimately
15 need to be captured in the database. The other
16 thing which is very important, which people
17 neglect, is the need for ontologies in controlled
18 vocabularies to define these things. One of the
19 real problems with analyzing data even in our labs
20 when we started doing experiments, we sort of threw
21 things out to the anarchy of the masses and let
22 people type in their experiments. If people type
23 in cancer or people type in tumor, and if people
24 misspell tumor or use the British spelling of tumor
25 and you try to extract the data from the database

1 without knowing what all the variants are, you only
2 get a partial view of what is actually represented
3 within that database. So, having standardization
4 even at the level of experiment description and
5 compound description is fundamentally important for
6 later interpreting the data.

7 DR. KAROL: Thank you very much. We will
8 move on to our next speaker, Dr. Ghosh, and she
9 will be talking to us about fluorescent machine
10 standards and RNA reference standards.

11 Fluorescent Machine Standards and RNA Reference
12 Standards (Summary of Results from
13 the NIST Workshop)

14 DR. GHOSH: Thank you very much for giving
15 me an opportunity to come over here and update the
16 subcommittee members and all the audience members
17 on some of the efforts that we have undertaken in
18 conjunction with NIST and industry participation in
19 defining standards.

20 [Slide]

21 Some of the stuff which I will actually be
22 mentioning has already been alluded to in terms of
23 lack of standards in the gene expression area.
24 That really prompted some of the key industry
25 leaders, some of the NIST and FDA members, back in

1 2002, to get together in one of the meetings, and I
2 will be basically outlining what was outlined for
3 the group to achieve and accomplish.

4 In the second part I will cover a little
5 bit all the activities regarding the development of
6 the microarray fluorescent standard efforts and the
7 working group which has now been made up of all the
8 industry participants in terms of the fluorescent
9 standard initiative in trying to define the
10 specification of the standards.

11 The third part, of course, as we already
12 heard is in terms of the RNA standards initiative
13 and that group again assembled together. This was
14 an industry, government and several academic
15 institutions who have joined together to define
16 what that standard is, and how it would be
17 developed, and how it can help us to answer some of
18 the variabilities that we are seeing today.

19 Lastly, some of the feedback that I got
20 from NIST and I wanted to bring it to the table
21 today because there is definitely a request for an
22 active participation of FDA, requested by NIST, to
23 really help this community and this technology to
24 build some of these standards, and how FDA can
25 really make an effort and contribution in bringing

1 that to fruition. So, I am going to present that
2 request formally in front of everybody.

3 [Slide]

4 The kickoff meeting actually started in
5 2002. Fortunately, we had Frank Sistare
6 representing the FDA over there, where we had
7 defined that we should really look into two major
8 areas, one being first in the scanner area which
9 really also contributes but it was one of the
10 easiest, less challenging perception-wise which
11 people thought that we could actually accomplish.
12 To be honest, we have made some very good progress
13 in defining some of the standard needs there which
14 I can overview for the committee members here.

15 So, in terms of that particular first
16 initiative, the team got together at NIST on
17 December 10th and, in fact, basically presented
18 various practices which the microarray readers can
19 adapt and define a standard and since then every
20 month this particular working group is meeting and
21 making progress. So, I will overview some of the
22 definitions and specifications that have been laid
23 down, which NIST has now taken together and they
24 are really making that particular artifact for the
25 community which will be available for individuals

1 as a calibration standard for the scanner area.

2 The universal RNA standard, which was the
3 second objective laid out for the team--a meeting
4 was held at Stanford, in March this year, and it is
5 actually drafting a guidance document which will be
6 out for all the participants to comment on by end
7 of June.

8 The third workshop, again, was held with
9 NIST and industry leaders in respect to the
10 microarray fluorescent standard to accomplish the
11 second phase of development of the scanner
12 initiative. So, I will overview a little bit of
13 some of the final status on those.

14 [Slide]

15 In terms of the accomplishment for the
16 first group on developing an artifact,
17 specifications have been developed. Currently, we
18 are trying to define a technology which can
19 actually accomplish the specifications which have
20 been laid out by the working team. It is a little
21 bit challenging because some of the finer
22 specifications are really becoming a challenge for
23 us to accomplish because of the dyes that we have
24 defined and they have a finite life period. If a
25 standard cannot be made in a way that it can be

1 stable over a period, it really doesn't help us.
2 So, we are right now at the stage of defining a
3 technology which can really give us that stability
4 factor in the calibration standard. It is a
5 challenge but we are right now at that particular
6 stage.

7 In terms of the artifact, the draft
8 artifact is out and it has been more or less, about
9 95 percent, developed but the challenge comes on if
10 we cannot define a technology to make and
11 accomplish those, we have to go back and change
12 some of the specifications in terms of the
13 available technologies.

14 [Slide]

15 The decision in the case of the artifact
16 was that for each particular dye we will have two
17 types of artifacts in the standard manufacturing
18 area that people can use, one addressing the
19 uniformity and the signal-to-noise for the right
20 features in the scanners, and the other one will be
21 more as a limit of detection which would be
22 basically treated by the manufacturers and adopted
23 in terms of the specification definement.

24 These artifacts won't be manufactured by
25 NIST but an outside agency will work with NIST, but

1 NIST will certify and endorse it at the end of the
2 period, and that is how the whole activity has been
3 decided and it is totally supported by NIST in that
4 matter.

5 [Slide]

6 This is an outline of the preliminary
7 scanner specification decisions which the working
8 group accomplished over a period of three to four
9 months. Artifacts will be uniformly coated. There
10 will be at least two artifacts per dye. The
11 decision right now is a dye which resembles Cy-3
12 and Cy-5, and anything which can mimic those
13 particular two dyes will be the first. They won't
14 be the last but as more dyes come into the picture
15 we will be able to adapt the same principles. The
16 same technology which has been identified during
17 the first initiative can apply for the other
18 initiatives too.

19 Some of the major issues came up, whether
20 glass would be the choice feature in terms of
21 accepting as a standard and at last the committee
22 definitely decided to go with the glass. The
23 non-flatness of the glass in a microarray
24 experiment, it seems like that was one of the
25 areas, we found out, really impacts your data

1 quality, how flat the particular glass is that you
2 are choosing. And we came up with that they won't
3 exceed it than this ten micron limit because that
4 can really alter the data quality being represented
5 at the further end.

6 Various scanners right in the marketplace
7 have different issues with this particular flatness
8 of glass. Therefore, this was an alert figure
9 which prompted us that many of the home-brew type
10 of glass manufacturing may not basically understand
11 the underlying pinning of the flatness of the glass
12 and how it impacts the scanner reading, and how it
13 impacts the data quality, but it is an important
14 one.

15 The other part came in in terms of the
16 thickness of the glass, flatness and the thickness
17 of the glass, and currently this particular
18 standard which we are going to develop will really
19 keep to a one millimeter thickness. The artifact
20 which basically finally came would be a 1 by 3
21 since the major industry is facing a 1 by 3.

22 [Slide]

23 This is a picture which defines that we
24 have defined a particular area where the Affymetrix
25 chip--they would basically make a cut in the major

1 final defined artifact slide, and use that
2 particular region to calibrate their scanner.

3 So, if you look at this picture, this
4 particular artifact can be used by 10 to 12
5 available scanners available today in the
6 marketplace, and they have all actively
7 participated in finalizing this particular design
8 which is out there. This would be treated by the
9 scanner as the reading zone which helps them to
10 really scan the area, and the placement of the
11 barcodes and the placement of the backgrounds have
12 all been agreed to by all the manufacturers of the
13 scanner readers.

14 [Slide]

15 A second workshop by the same scanner
16 group was held on May 14, and the issue here was
17 what technology we have to basically adopt. The
18 Cy-3, Cy-5 are very unstable and photo bleaching
19 was one of the major issues that we observed that
20 the Cy-3, Cy-5 dyes have. Therefore, we had to
21 look into metal oxide glasses, which are less prone
22 to photo bleaching but currently all the available
23 technologies really do not help us to make a
24 particular metal oxide glass artifact which could
25 be uniformly coated or which was uniform enough to

1 help us to create this artifact standard.

2 We have engaged now Molecular Probes,
3 Evident Technologies with Crystal Technology as
4 well as the Quantum Dot Technology people to come
5 together and help us in order to define a
6 technology whereby we could basically mimic or
7 choose two dyes that we are looking for in order to
8 help us to build this particular artifact. There
9 are some experiments which have been laid down with
10 Molecular Probes. They are currently working on it
11 so it is in a development phase but very soon,
12 within the next two to three months, we are trying
13 to activate that particular activity by Molecular
14 Probes, whereby they feel there is a particular
15 dye. It is organic in nature, but it is much more
16 stable than our current Cy-5 dye where we are
17 having the biggest problem issue. So, hopefully,
18 we will be able to identify a particular technology
19 to help us meet our specification. Evident
20 Technology, I would say this is a great technology
21 to consider in terms of stability for bleaching.
22 They are the perfect technology to adopt in terms
23 of building a particular standard. Hopefully
24 again, one of the dyes, they have the material
25 available so it is not a problem. With the Cy-5 we

1 are struggling and time would be a factor but we
2 are very hopeful will we accomplish that target
3 very soon.

4 [Slide]

5 As I mentioned, these are a couple of the
6 next steps in the scanner artifact development that
7 we have to accomplish, defining some of the
8 protocols and how we view the data analysis is a
9 critical factor. It is not enough just to develop
10 an artifact. How we use it and how we interpret
11 the data is another area. For this particular
12 usage, what we are looking for is a second stage of
13 a defined protocol that every individual, not just
14 the scanner manufacturer but individuals within the
15 lab can basically use the protocol in the same
16 fashion; come up with a set of metrics which would
17 be defined. Again, technology is a big issue and
18 there is a big variation in user terminology. What
19 is uniformity? I have heard many definitions.
20 And, we need unification and understanding and
21 common consensus building in agreeing to some of
22 these terminologies and usage.

23 So, we are looking for NCCLS participation
24 in this particular last phase of activity, whereby
25 uniform protocol and terminology would be part of

1 the completion of the standardization. In fact,
2 NIST has already invited ASTM to come to the table
3 and NCCLS to come to the table. The way we might
4 work is that this working group may define the
5 protocol and get it in one of their sessions of
6 NCCLS to get some approval and understanding.

7 [Slide]

8 The next particular standards meeting
9 happened at Stanford University on March 28 and 29.
10 Again, government, industry, manufacturers and
11 microarray users all collected together and shared
12 some of their concerns, major concerns in the
13 microarray area or gene expression area and the
14 variations each one of them are facing. I will
15 very quickly actually glance through some of the
16 topics since time won't permit me to go in great
17 detail.

18 [Slide]

19 Some of the major goals of this were
20 educational, or providing a forum for everybody to
21 come and share their own methods and techniques in
22 order to define the standards for the gene
23 expression area. There were several areas where
24 people agreed and disagreed, but we wanted for all
25 of them to come to the table and actually table the

1 disagreements so that we could hear and find out
2 where some of the commonalities have to develop.

3 In fact, we were looking for a guidance
4 and how NIST could help us in this particular
5 initiative and participate since we look towards
6 them in terms of the standards development, and we
7 really need their help in order to make some
8 traceable standards, especially from a data
9 submission point of view too.

10 Requirements were laid out, like, we need
11 to define some specifications for universally
12 applied--some RNA standards which could be used
13 very effectively by IND and NDA filings initially
14 and later on as the diagnostic industry really
15 improves, it can start building some elements there
16 that could help some of the diagnosis and prognosis
17 assays which are currently being developed.

18 [Slide]

19 I wanted to take a moment to really go
20 into finer details, when we talk about gene
21 expression, what the work flow looks like and where
22 several of the standardization initiatives really
23 need to happen. At the universal RNA workshop we
24 addressed maybe some of the areas but still there
25 are some unanswered areas. Today we heard from

1 John what the annotation area and data format area
2 are going to do and provide some guidance in there.

3 But let's start from the very beginning,
4 where we talked about the sample preparation area
5 and how an RNA is extracted; how it is particularly
6 stored; what is the particular concentration of the
7 RNA which is put on the microarray chip. What
8 particular integrity of the complete RNA, before
9 even it is hybridized, how does that affect. We
10 have found that each and every element in the
11 sample preparation area is going to affect the data
12 quality. So, we do need some guidance in each and
13 every area about even the sample preparation that
14 will be important in making final conclusions or
15 calls at the end of the period.

16 For the manufacturers in the array
17 fabrication a lot of quality control issues most
18 probably are there, but it needs to be well
19 understood with an idea of how it is going to
20 impact the data quality at the end when we are
21 doing just the data analysis. As we go through
22 this work flow process we are accumulating all the
23 errors as we are going through.

24 The effect of labeling is another part,
25 how well we have labeled? What is the optimum

1 percentage of labeling that is required to give the
2 optimum output? How balanced are the channels? We
3 already know there are environmental effects when
4 you work with labeled samples. How are we really
5 taking precautions? What is the time period? What
6 is the protocol? They need some standardization in
7 the labeling and hybridization area.

8 People use different protocols in the
9 hybridization, and they do have an impact on how we
10 get the data at the end point. So, what is the
11 particular hybridization protocol? How stringent
12 is it? How well will it hybridize? Those are some
13 of the factors--what is the cross-reactivity of the
14 probes, and how does it affect the data
15 manipulation at the end? We need to understand
16 those factors.

17 I already talked about the scanning area,
18 and I think the movement we have started with
19 defining the standardization effect, it would take
20 care of most of the scanning zone which is most
21 promising. Then, coming to the probe area and John
22 has mentioned a lot of these areas. Sequence
23 homology, clone specifications and the noise, and
24 cross-reactivity are some of the other issues that
25 need to be developed and, again, we need some

1 standardization to be developed and put into place
2 in order to have more reliable data.

3 [Slide]

4 I have talked about this, generalized work
5 flow area. In terms of this particular Stanford
6 meeting, we addressed the two technologies, the PCR
7 technology as well as the microarray technology, in
8 trying to establish a standard which can really
9 help all the technologies. This is the common,
10 general outline of the work flow which came out in
11 terms of discussion. As we see, there are very
12 generic commonalities between the two and
13 standardization needs.

14 [Slide]

15 So, session one of our universal
16 microarray standards--actually, Frank Sistare was
17 our session chair and he really helped us to bring
18 an understanding from a diagnostic perspective,
19 what some of the standardization needs are. Maria
20 Chen, from FDA, in fact, presented some early views
21 on what we need to accomplish if we are really
22 looking into some IND submissions. Again,
23 standards were something which really popped up,
24 that we need to develop them in order to make some
25 relevant contribution or meaningful contribution.

1 Carol Thompson, from the Pharmacology
2 Department, basically, she presented her teams and
3 one of the projects that they are going to initiate
4 in terms of standardization with various platforms
5 and with mixed tissue samples in order to
6 understand the toxicology effects across standards,
7 and what type of standardization might be helpful
8 in terms of protocols and interpretations. Data
9 understanding was one of the areas that she talked
10 about.

11 Some of the areas in terms of
12 bio-international standards were brought by Merck.
13 Roland Stoughton, in fact, talked about some
14 guidelines, again, needing to be developed in terms
15 of how data interpretation in the diagnosis and
16 prognosis areas are made; how we create different
17 standards. So, a general flavor was that for each
18 application we might need to look into different
19 types of standardization, but universal standards
20 at the end of the workshop basically came out by
21 two general guidelines of having an external
22 standard and an internal standard.

23 [Slide]

24 I wanted to bring this experimental design
25 which was put forth by Brenda Weiss, from the

1 NIEHS, whereby basically they have taken about five
2 or six different platforms which are participating
3 in that particular consortium.

4 [Slide]

5 The data outcome basically comes from the
6 array platform and different labs and array to
7 array variability trends from the maximum in terms
8 of data variation. So, these results, which were
9 shared, really made it very clear that unless we
10 address the standardization needs very soon and
11 early on with some really good participation from
12 every segment, we will still be struggling to make
13 some meaning out of this particular technology.

14 [Slide]

15 This is the one which was presented by
16 Carol Thompson, from FDA, where standards for
17 toxicogenomic studies basically would be using
18 bench mark genes within the mixed tissue samples.
19 Currently, that activity has already started and
20 Frank has been actively engaging various industry
21 participants, as well as academic participants, to
22 really contribute to this particular project.
23 Hopefully, some of the expected initial outcomes of
24 this particular activity would be to identify some
25 of the probes that can perform similarly across the

1 platforms. Unless we do that activity, building
2 any databases with only one type of data may not be
3 sufficient. It would be incomplete.

4 Determining the normal range of false
5 positive and negative would be another objective of
6 this, and lab to lab variance. Again, without some
7 universal standards being developed, we will see a
8 lot of variation, as being observed already by the
9 NIEHS consortium, reported by Brenda. Ultimately,
10 hopefully, this particular publication will be
11 available with the findings which will help all of
12 us to understand where we have to focus our energy.

13 [Slide]

14 The second session during our RNA
15 development session was basically targeted towards
16 defining some of the metrics that each of the
17 microarray platform users needs to acquaint
18 themselves with. These may not be just platform
19 specific. We may need to define some metrics and
20 RNA input sample which goes in an microarray. Some
21 of those thoughts were basically--

22 [Slide]

23 --this particular slide shows that even
24 procurement of RNA, when we are getting it from
25 different sources, has impacted the data quality.

1 So, procurement, the source of a participant RNA,
2 the tissue samples, isolation methods, temperature,
3 storage, all have contributed to data quality at
4 the end. This was a great slide, presented by
5 Ambion, known experts in RNA. They spent a fair
6 amount of time in digging deeper into the issues of
7 RNA and how they have basically contributed. So, I
8 think the metric definition part, which we have
9 already laid out from a platform perspective, was
10 good enough but now we feel that that is just not
11 enough. We now have to extend it into defining
12 some metrics, even RNA quality which is right at
13 the beginning, and we are seeing some results
14 coming out on how they have been impacting the data
15 results at the back end. So, unless we define some
16 good controls and some good specifications right at
17 the beginning for a particular platform to address,
18 we may not be able to interpret our data very
19 meaningfully at the end of the experiment.

20 [Slide]

21 Going back, some of the teams from the
22 universal RNA workshop came out with multiple
23 sources of data variability from different
24 technologies, from different probes and primers
25 used by different platforms, different

1 laboratories, sample types and extraction methods.
2 And, we heard it coming from every angle, wherever
3 we looked into.

4 There was a great difficulty of sharing
5 data between the platforms, and we have heard that
6 today also. MIAME is a definite, very good start
7 and it is being extended to the tox area. But we
8 need to do more about the annotation problems.
9 Unless we address the annotation issues through
10 some work groups and common understanding, we will
11 still be struggling to make some valuable,
12 meaningful data interpretation.

13 Standards and methods for labs, which was
14 actually very well presented, why GLP practices
15 have always been treated as one of the areas of
16 keen interest, we need to look into those and how
17 each of the labs were producing these data; how
18 they are standardizing their activities around
19 different metrics; and how we refine our methods.
20 That is another area I think we need to start
21 looking into more to define and bring some
22 consistency in our data interpretation.

23 [Slide]

24 A very interesting factor came in, which
25 was RNA quality index. That is gaining some

1 momentum also now. We would eventually like to
2 define some RNA quality index as a factor which
3 would be treated as one of the standards as input
4 quality RNA factor. If we have to define some of
5 the metrics, maybe these are some of the proposed
6 metrics which are being considered that can really
7 make--that the metrics, when we need to define an
8 RNA standard, we define it with particular metrics
9 and eventually these can form our data submission
10 pipeline.

11 [Slide]

12 So, what a good standard should be--John
13 had actually presented the slide at our universal
14 RNA standards workshop--what it should do. It
15 definitely should be something that could be used
16 by a platform over time, compare between the
17 different platforms; should be consistent enough,
18 therefore, some of the concerns of using biological
19 samples as a universal standard were basically
20 thought through and we couldn't find the number
21 three parameter, that it has to be consistent over
22 time. We thought that most probably we might have
23 to go to synthetic model having all the biological
24 characteristics for that standard so that
25 consistency can be maintained over time. We should

1 have a well-defined protocol. That was definitely
2 one of the themes that ran across and people agreed
3 that a defined protocol needs to come out through
4 that activity. And, we must be able to make both
5 absolute and relative measurements using this
6 particular standard. It should not just be
7 confined to use in the gene expression but QRT-PCR
8 technology should be able to use that.

9 [Slide]

10 What are some of the microarray
11 performance characteristics? From a design and
12 fabrication point of view, platform types. The
13 surface types which are used by fabrication and a
14 manufacturer may impact in terms of data quality;
15 understanding each and every aspect of the surface
16 types. Composition and spatial layouts, a number
17 of replicates identifying that particular array can
18 be some of the very good requirements that can be
19 laid out during submission of data. In terms of
20 the spot elements on a microarray, clones,
21 sequence, primers, probe lengths, gene name, etc.,
22 can basically be added to the list of spot element
23 definition. Built-in controls, which are the
24 housekeeping genes for the controls defined by an
25 array manufacturer, can be defined in terms of

1 requirements.

2 Again, in the microarray controls area,
3 use of internal controls, which can be synthetic
4 housekeeping genes; pooled RNA from sample cell
5 lines or pooled RNA from test samples; and RNA and
6 oligonucleotides from plants and bacteria can also
7 form microarray controls. But these were some of
8 the controls that we saw came out of the meeting
9 that individuals presented.

10 So, there is a lot of different variation
11 where people have been working. Because
12 availability of a standard is missing, people have
13 been trying to use some of the internal controls
14 but it seems like it comes that we do now have to
15 come up with a unified defined protocol for all
16 this.

17 So, standards are required for several
18 purposes. This was the proposed workshop
19 recommendation, that periodic laboratory
20 proficiency testing can be used for platform
21 performance validation and baseline monitoring;
22 cross-platform performance validations and
23 inter-laboratory performance validation. These are
24 some of the themes that would be basically
25 addressed as we define the external standard

1 through this work group.

2 A consistent definition of terminology,
3 which was pretty varied, and through the guidance
4 document this particular definition of terminology
5 part would be addressed so we can define a
6 consensus for how we can define the terminology.
7 Finally, the consensus of the attendees at the end
8 of the session was that there has to be an external
9 synthetic RNA standard reference and an internal
10 RNA standard reference which would be treated as a
11 spiking control.

12 [Slide]

13 These were the two particular standards
14 which were defined by the work group. The
15 definitions and the specifications of the RNA
16 standards are coming out, as I said, in a guidance
17 document which will help us. In terms of the
18 reference method, we most probably again have to
19 engage external agencies, like NCCLS and ASTM, to
20 work with NIST in order to define the reference
21 standard method.

22 [Slide]

23 I want to go to my last slide. Here are
24 some of the open questions which came up at the end
25 of the session. NIST had taken up this particular

1 initiative to define the specification for the work
2 group but the next phase of execution and
3 implementation plan, they are really requesting FDA
4 to come to the table and define their requirements,
5 and they are proposing a partnership model with the
6 industry to take place in order to execute it. So,
7 I wanted to formally place that requirement, as per
8 my discussion with NIST on Friday where they made
9 this requirement. They are ready to come and sit
10 with FDA and take the requirements from FDA so that
11 they can work to a particular objective which will
12 help FDA to accept the data. That would be the
13 next step. Frank has really been helping this
14 particular activity and bringing all the feedback
15 to the table to help really guide us on what should
16 be our next step and how we should address that.

17 With that, I will address any questions if
18 the committee has any questions.

19 DR. KAROL: We will just take one
20 question.

21 DR. ZACHAREWSKI: In the open questions
22 you said that the guidance document was going to be
23 published by the end of June, 2003. That is in a
24 couple of weeks. Is that still on schedule?

25 DR. GHOSH: Right, it is on schedule. It

1 is written up. It is waiting to go to the session
2 chairs, and John Quackenbush was one of our session
3 chairs and Frank was one of the session chairs. We
4 have two other session chairs who need to review
5 the document and give their comments in terms of
6 completion.

7 DR. ZACHAREWSKI: And where will that be
8 published?

9 DR. GHOSH: It will be published by NIST
10 actually.

11 DR. ZACHAREWSKI: How will it be
12 available?

13 DR. GHOSH: All the activities of the
14 standards workshop are currently available on the
15 NIST web site. So, this particular guidance
16 document will eventually go up on the NIST web
17 site.

18 DR. KAROL: Thank you very much. We
19 appreciate your presentation. In order to be able
20 to fit adequate discussion and the open public
21 hearing, we are going to change our agenda just a
22 bit. We are going to break for lunch now and
23 reconvene at one o'clock after lunch.

24 [Whereupon, at 12:15 p.m., the proceedings
25 were recessed until 1:00 p.m.]

1 A F T E R N O O N P R O C E E D I N G S

2 DR. KAROL: I would like to start the
3 afternoon session. First is the open public
4 hearing but there is no one scheduled to speak so
5 let's move on to Dr. Leighton, who is going to talk
6 about the CDER IND/NDA reviews.

7 Topic #3 CDER FDA Product Review and Linking
8 Toxicogenomics Data with Toxicology Outcome
9 CDER IND/NDA Reviews - Guidance, the Common
10 Technical Document and Good Review Practice

11 DR. LEIGHTON: Good afternoon.

12 [Slide]

13 I will spend the next few minutes
14 providing a general overview of the CDER IND/NDA
15 review process and describe the nonclinical studies
16 that are usually submitted to support these
17 applications. I will also spend some time
18 discussing the role of FDA and INCH guidance in the
19 review process; a slide on the common technical
20 document, as well as the CDER pharmacology good
21 review practice.

22 The purpose of my presentation is to
23 present to you the current review practice and to
24 introduce a possible future role of
25 pharmacogenomics in safety assessment, and this is

1 not intended to be a complete discussion of the
2 review process.

3 [Slide]

4 The review team for any IND and NDA
5 consists of the professionals shown on this slide.
6 It includes project managers that are the first,
7 and sometimes the only contact that a sponsor has
8 with the division; medical officers;
9 pharmacologists, toxicologists; chemists that
10 examine the manufacturing process; and clinical
11 pharmacokineticists and statisticians. Now, the
12 first four disciplines are primarily involved in
13 the initial IND review. Clinical
14 pharmacokineticists and statisticians are brought
15 into the review process on an ongoing basis as
16 needed.

17 [Slide]

18 The nonclinical studies usually submitted
19 to support an IND and NDA are shown on this slide,
20 including studies on the mechanism of action, such
21 as pharmacodynamics and pharmacology studies;
22 studies on pharmacokinetics, including absorption,
23 distribution, metabolism and excretion; safety
24 pharmacology studies which are studies that provide
25 an evaluation of vital organ function, in specific,

1 cardiovascular, central nervous system and
2 respiratory function; general toxicology studies
3 that provide the pivotal safety data for an initial
4 IND. Genetic toxicity, reproductive toxicity and
5 carcinogenicity studies are also provided.

6 [Slide]

7 The goals of nonclinical IND studies are
8 primarily at the initial stages, number one, to
9 identify an appropriate start dose; secondly, to
10 identify organ toxicities and their reversibility;
11 and third, to guide dosing regimens and escalation
12 schemes.

13 [Slide]

14 Pharmacology studies--pharmacologic
15 activity as determined by in vitro and in vivo
16 animal models, and nonclinical studies are
17 generally considered of low relevance to the
18 current safety assessment as provided in the IND
19 and efficacy studies in the NDA, which is primarily
20 determined by Phase III clinical data. Therefore,
21 for this reason, summary reports, without
22 individual animal records or individual study
23 results, usually suffice for reporting requirements
24 for pharmacology studies.

25 [Slide]

1 However, toxicology studies provide the
2 pivotal information for the initial safety
3 assessments, as well as the start dose decision.
4 Ideally, toxicology studies should mimic the
5 schedule, duration, formulation and route as that
6 proposed for the clinical trial. They should
7 conform to standard toxicology protocols and should
8 be conducting according to good laboratory
9 practices, or GLPs, as identified by Code of
10 Federal Regulations, Section 21, Part 58, or 21
11 CFR, Part 58.

12 [Slide]

13 To support an initial IND what should be
14 provided? An integrated summary of the
15 pharmacology/toxicology data should be provided.
16 Unlike that I described earlier for pharmacology
17 data, a full tabulation for each toxicology study,
18 including individual animal data, should be
19 provided to the review divisions in order to
20 support the safety of a proposed clinical trial.

21 How can pharmacogenomic data be
22 incorporated into the initial IND safety
23 assessment? Well, perhaps this data can be used to
24 assist in the selection of a start dose, a choice
25 of a relevant species for additional long-term

1 studies, or to identify biomarkers for future
2 clinical evaluation.

3 [Slide]

4 Not all toxicology studies need to be
5 provided with the initial IND. It is an ongoing
6 process that should be conducted concurrently with
7 clinical develop. So, some of the studies that may
8 be provided, and this depends to some extent upon
9 the intended indication for the drug--some of the
10 studies that could be provided at a later date
11 include long-term toxicity studies. The genetic
12 toxicology panel should be completed if it hasn't
13 been completed by the initial IND. Reproductive
14 toxicology studies should be provided, and
15 carcinogenicity studies should be provided if the
16 indication and the treatment warrants them.

17 So, how can pharmacogenomic data assist at
18 this stage? Possibly by decreasing the study
19 length. For example, carcinogenicity study
20 standard is usually a two-year rodent bioassay.
21 Perhaps now, with additional pharmacogenomic data,
22 studies can be conducted in a shorter duration,
23 perhaps six months. Improve assessment of organ
24 toxicity in terms of clinical relevance, and
25 provide mechanistic explanation of toxicity.

1 I would like to emphasize that at least
2 initially it is unlikely that pharmacogenomic data
3 will replace the standard assessment. For example,
4 in general toxicity studies there is usually
5 provided histopathological evaluation of over 50
6 tissues. Most pharmacogenomic studies only look at
7 one, two or maybe even a handful of tissues. So,
8 it is unlikely that the data will be of sufficient
9 extent to supplant our traditional general tox
10 environment.

11 In addition, one other point is that the
12 animals often die in the middle of the night. It
13 is very inconvenient and you may get a lot of
14 tissue autolysis and with the issue of RNA
15 standards being critical, how will this RNA look in
16 the morning when the animals are finally found and
17 the tissue is extracted? So, the cause of death
18 may not be amenable to understanding by genomic
19 analysis.

20 [Slide]

21 What is the role of FDA guidance in the
22 review process? ICH stands for International
23 Conference of Harmonization. FDA/ICH guidances
24 represent the current thinking of the agency.
25 These are recommendations, not requirements. And

1 FDA guidance can either be drafts, which is for
2 comment purposes only, or final documents. So, it
3 is a step-wise process where the agency can get the
4 input of outside experts. Guidances are available
5 on the CDER web site.

6 [Slide]

7 Some of the FDA/ICH guidances, on the
8 left-hand side are process-driven guidances. These
9 include things like guidances on how to submit an
10 IND; how to select an appropriate start dose; how
11 to design an appropriate study for acute toxicity
12 testing; and how to submit an electronic NDA. On
13 the right-hand side are some guidances, and this is
14 not a complete list but some of these guidances
15 that are available include some more
16 scientific-based guidances, including guidances on
17 carcinogenicity dose selection; genetic toxicity;
18 reproductive toxicity; photo safety testing;
19 immunotox; and biotechnology.

20 [Slide]

21 One of the guidance documents that are
22 available is the common technical document. This
23 is a guidance that describes a harmonized format
24 for technical documentation for registration in all
25 three regions. By the three regions I mean United

1 States, the European Union and Japan. This is for
2 registration so this would be for the NDA stage.
3 It consists of five modules. Modules two through
4 five are common to all regions. Module one would
5 be region specific. The purpose of the common
6 technical document is to reduce the time and the
7 resources used to compile a registration document.
8 It is intended to be used with other ICH and agency
9 guidances and to allow for regional specific
10 summaries.

11 [Slide]

12 In an effort for transparency, the
13 pharmacologists have developed what is called the
14 good review practice. This is a guidance for
15 reviewers and provides for a standard review
16 format. It is an internal review format for the
17 IND and NDA primary pharmacology reviews.

18 The purpose of this good review practice
19 is to provide for standardization of reviews across
20 divisions to ensure that important information is
21 capture in all reviews, and it allows for continued
22 assessment of an IND. It is consistent with the
23 common technical document that is available at the
24 web site at the bottom of the page.

25 [Slide]

1 Some of the information that is collected
2 in a good review practice, currently collected as
3 part of a general toxicology study review, includes
4 the information shown on this slide. It evaluates
5 mortality, clinical signs, body weight, food
6 consumption, ophthalmoscopy, electrocardiography,
7 hematology, clinical chemistry, urinalysis
8 parameters, organ weights, gross pathology,
9 histopathology and toxicokinetics when they are
10 available.

11 [Slide]

12 In summary, there is a different
13 submission format provided for pivotal safety data,
14 in other words your toxicology data, relative to
15 pharmacology data. We have developed good review
16 practices for the evaluation and capture of data in
17 order to provide consistency among review divisions
18 and to increase transparency. Good review
19 practices, if they are developed for
20 pharmacogenomic data, will need to consider the
21 interdisciplinary review of pharmacogenomic data
22 that was discussed earlier by Dr. Woodcock. It is
23 my belief that pharmacogenomic data will play an
24 important role in the safety assessment in future
25 INDs and NDAs. Thank you.

1 DR. KAROL: Thank you very much. We will
2 have questions at the end of this session, after
3 the four speakers, so we will move right on to the
4 second speaker. This is Dr. Levin who will talk
5 about electronic submissions guidance, CDISC and
6 HL-7.

7 Electronic Submissions Guidance, CDISC and HL-7

8 DR. LEVIN: I am going to be talking about
9 some of our standards development and
10 implementation at FDA.

11 [Slide]

12 I am going to go over some of the
13 standards organizations that we work with at the
14 FDA, the FDA Data Council inside the FDA but then
15 there are these four other organizations I will be
16 covering. I would like you just to concentrate on
17 these four organizations, right here, and see if
18 you can find a pattern in all those initials and
19 see what the next organization should be after this
20 one.

21 [Laughter]

22 I will go through what all those
23 abbreviations stand for. I have three initiatives
24 here but I understand we are a little pressed for
25 time so I am going to go over two initiatives, the

1 clinical and nonclinical study data standards and
2 the annotated ECG waveform data standard. I will
3 describe why those things are important here.

4 [Slide]

5 We deal with a number of different
6 standards development organizations inside the
7 government, accredited standards development
8 organizations and a variety of other standards
9 organizations that are not accredited.

10 Inside the government we have the FDA Data
11 Council. We also work with a group called
12 Consolidated Health Informatics. For accredited
13 standards development organizations we work with
14 Health Level 7, which is accredited by the American
15 National Standards Institute, and then two other
16 standards groups that we are working on with ICH.

17 [Slide]

18 The FDA Data Council is what we have
19 formed inside the FDA to try to standardize across
20 our various centers. We have the Center for Foods,
21 Drugs, Devices, Biologics and Veterinary Medicine
22 so we try to standardize across these different
23 groups to have standards that are common in the
24 FDA. We have representatives from all the various
25 centers as well as the different offices and the

1 Office of the Commissioner. This group is involved
2 with the national and international standards
3 development.

4 [Slide]

5 Here, in this group, we coordinate the
6 standards development. We get information that is
7 coming from different centers or offices where they
8 want to have data or terminology standards. We
9 form expert working groups within the FDA, work on
10 the standards, work with standards development
11 organizations if there are already standards
12 created or, if we create our own standards we try
13 to bring them to a standards development
14 organization, like HL-7.

15 [Slide]

16 There is another group we work with, the
17 Consolidated Health Informatics. This is a group
18 that is part of the President's eGov initiatives
19 and it is to set the standards for inter-agency
20 use. There are three major partners in this
21 organization, Department of HHS, Department of
22 Defense and the VA. So, those are our three major
23 partners in this and what they are trying to do is
24 set standards that can be used across the different
25 agencies in health care. This was started because

1 the Department of Defense and VA were trying to
2 exchange information and were unable to because
3 they use different terminology and they said we are
4 going to use the same terminology and form this
5 group. All the government agencies that deal with
6 health care are involved with this group.

7 They have set five standards so far. One
8 is to use HL-7, Health Level 7, for messaging
9 standards. The other is to use logical
10 observations, identifiers, names and codes, LOINC,
11 for lab test standards, and use DICOM for
12 transmission of images, and the National Council of
13 Prescription Drug Products for prescription
14 messages and IEEE for ECG monitoring messages. So,
15 these are some of the standards that they have.
16 These are the first five. They have now listed 24
17 different standards groups that they want to
18 establish and they are moving forward on that.
19 Once these standards are established, that means
20 these government agencies will use these standards
21 for exchange of information. The first two are
22 important to the FDA, the other three are more
23 related to agencies involved directly with health
24 care but there are other standards that will be
25 coming forward that will be important for us when

1 we are dealing with research and the other things
2 that we deal with as we interact with drug
3 companies and investigators.

4 [Slide]

5 Health Level 7 is an ANSI accredited
6 standards development organization. They are an
7 international group. They have open membership.
8 They follow all the procedures laid out by ANSI so
9 that their standards are accredited and they can be
10 accredited by ANSI or ISO. They are involved with
11 standards development activities in the government.
12 They were involved with the Health Insurance
13 Portability and Accountability Act which provides
14 standards for exchange of insurance information and
15 prescription drug information. They are involved
16 with the national health information infrastructure
17 which is to develop standards so health care groups
18 can communicate information. They are labeled as
19 the standard message for the Consolidated Health
20 Informatics group.

21 FDA is part of the Health Level 7. We are
22 on the clinical research information management
23 technical committee in Health Level 7, and this is
24 where standards that are of interest to the FDA
25 would go for accreditation. So, we take our

1 standards to the HL-7 group and we have taken a
2 number of standards there for development and
3 subsequent ANSI accreditation. We are also
4 involved with the vocabulary technical committee
5 where terminology standards are being looked at.
6 Since there is a lot of government involvement in
7 Health Level 7. We are involved in the government
8 special interest group which includes groups like
9 the Department of Defense, VA, CDC and NIH.

10 [Slide]

11 John was just talking about ICH. We are
12 involved with that. There is the common technical
13 document, as he was describing, as well as some
14 terminology through ICH. There is something called
15 MedDRA, which is terminology for describing adverse
16 events, and we are using that for exchange of
17 individual case safety report information.

18 [Slide]

19 Finally, there is a group called CDISC,
20 the Clinical Data Interchange Standard Consortium.
21 This group is an open group. Though they are not
22 accredited, they joined HL-7 so they are involved
23 with HL-7 as well. There are representatives in
24 this group from vendors, pharmaceutical companies,
25 industry consultants and government agencies. They

1 are trying to develop standards for clinical trial
2 data between pharmaceutical partners and between
3 the pharmaceutical companies and regulatory
4 authorities. They have set forth a standard, what
5 they call a submission data model for submitting
6 clinical data, research data to the FDA.

7 [Slide]

8 These are the standard initiatives that we
9 have brought forward, that we are working on right
10 now. There is one for electronic submissions of
11 applications; study reports; structured protocols;
12 a standard for product labeling; a standard for
13 individual case safety reports; electronic
14 MedWatch; stability data; annotated ECG waveform
15 data; and study data.

16 [Slide]

17 Now I will just briefly go over two of our
18 standards. One is the one for clinical and animal
19 study data. The clinical study data comes from the
20 CDISC group. The animal study data we are working
21 on is a separate group but it was facilitated by
22 the CDISC group and this has been following the
23 same basic standard that was worked out with the
24 clinical standard, which I will go over.

25 What I am going to talk about is a

1 standard that is based on the CDISC version three,
2 and this is available on their web site as
3 CDISC.org if you want to find out more information
4 about that. The standard development is divided
5 into two parts. One is the submission data model
6 and the second part is terminology. What I am
7 going to describe now is just the part we are
8 working on now, the data model, not the terminology
9 which we haven't really gotten into. What we are
10 working on also is standardization procedures,
11 including the development of specific analysis
12 tools and a data repository for this type of data.

13 [Slide]

14 The CDISC version three data model divides
15 a study into a collection of observations, and
16 there are three types of observations,
17 interventions which are therapeutic or experimental
18 treatments; events, which are incidences that are
19 independent of the planned study observations, for
20 example adverse reactions; and findings, which are
21 observations resulting from planned evaluations to
22 address specific questions.

23 [Slide]

24 Each observation is characterized by a set
25 of descriptive variables. There is a topic

1 variable which identifies the focus of the
2 observation. There are identifiers which identify
3 the subject or the study uniquely. There are
4 timing variables that describe the start and end of
5 an observation. There are qualifiers that describe
6 the trait of an observation.

7 [Slide]

8 Here is an example of an observation in a
9 clinical trial. This would be the topic of the
10 observation. The identifier, subject 101A is the
11 identifier. Starting on study day six would be an
12 example of the timing variable, and that it was
13 mild would be an example of the qualifier. There
14 is a series of these variables to describe the
15 different topics, identifiers, timing variables and
16 qualifiers. So, this is what the model consists
17 of, a series of these descriptive variables to
18 describe observations.

19 [Slide]

20 The other standard that we are working on
21 that might be relevant to this discussion is the
22 annotated ECG waveform data standard. This
23 standard is also brought through HL-7 and is based
24 on their reference information model, and is an XML
25 file.

1 The interesting part about this data is it
2 represents the digital ECG with all the annotations
3 that the company would put on the ECG--where the P
4 wave starts, the QT interval duration and things
5 along those lines. But it is a large amount of
6 data since it records every point along the line of
7 the ECG. It really was started off as a correlated
8 data standard or way to transport correlated
9 clinical data or study data. So, when we looked at
10 this model, since it is transporting a tremendous
11 amount of information that is correlated, this
12 might be something that might be useful for the
13 data that we are discussing here.

14 This data, along with the clinical data
15 model, are two things that we would have to
16 coordinate as we are working with our data
17 standards so that whatever way we decide on
18 transporting this information is related to a
19 standard that is coordinated with everything else
20 that we are doing, and we would like to take it
21 through the different standards groups so that we
22 are coordinated with the other parts of the
23 research community. Thank you.

24 DR. KAROL: Thank you very much. We will
25 move right on to Dr. Mattes, who will tell us about

1 MIAME-Tox.

2 MIAME-Tox

3 DR. MATTES: In truth, I am going to be
4 talking about MIAME-Tox in context of a larger
5 issue, much of which has been covered before and I
6 am probably going to rehash quite a bit but I will
7 try and make that fast.

8 [Slide]

9 The larger issue is that of the ILSI-EBI
10 collaboration which has been a learning experience
11 for both of us in terms of handling toxicogenomic
12 data.

13 [Slide]

14 Again, I am going to kind of come at a
15 pretty high level and talk about why we need a
16 database, why it is essential; how we envision that
17 it is going to be developed; what are the issues;
18 and who is involved, particularly the ILSI-EBI
19 collaboration.

20 [Slide]

21 Just to reiterate kind of one of the
22 issues which I think is the most significant issue,
23 and the most significant issue is how we were
24 trained X number of years ago, even maybe five, ten
25 years ago, to think about biology. In fact, we

1 were trained as graduate students and post docs to
2 look at one tree at a time, focus down and analyze
3 it and write up your thesis along those lines.

4 [Slide]

5 "Omic" biology--genomics, proteomics,
6 whatever, really, unfortunately or fortunately, or
7 whatever, the characteristic is looking at the
8 forest and mountains, the big landscapes and trying
9 to discern from that what is going on. Yes, things
10 do happen in individual trees but the data can't be
11 addressed at that level. So, the way forward is
12 really with informatics. Quite frankly, I think it
13 forms a stumbling block for most people and it is
14 very hard to fully integrate your thinking along
15 the lines of informatics as the way forward.

16 [Slide]

17 Again to reiterate why you need to handle
18 this sort of data in a database, if you think about
19 the traditional endpoints that are accumulated per
20 animal it is, you know, dozens, whereas genomic
21 endpoints in any given animal is going to be
22 thousands.

23 [Slide]

24 But there are other issues, and there are
25 other significant issues that can only be addressed

1 at an informatics level. One is the influence of
2 the technology. I have spent a fair amount of my
3 time getting hung up on the informatics of sequence
4 analysis and I am passionate about that because it
5 really influences the endpoints, the measures you
6 are getting.

7 [Slide]

8 I give as an example that many genes are
9 alternatively spiced and these events are not
10 usually unambiguously detected by microarray.

11 [Slide]

12 I give as an example a classic one, which
13 gives the all too famous UGT1 gene which consists,
14 when it is spliced, of five axons that are spliced
15 together but there are six alternative axons which
16 result in six different proteins from this one
17 gene, if you will. Yet, when you think of array
18 technology most arrays are going to be targeting
19 the 3-prime UTR that is just sort of
20 technologically driven. So, all too commonly you
21 may think you are measuring one sequence but, in
22 fact, you may be measuring something else.

23 [Slide]

24 On another level, for most cDNA arrays you
25 have to address the issue of whether or not the

1 probe may hybridize to more than one sequence, and
2 the bottom line is that you have to have a database
3 that captures the probe sequence to resolve the
4 discrepancies between array platforms at the level
5 of sequence. There is just no way it is going to
6 be done manually.

7 [Slide]

8 How are we going to develop the databases?
9 The efforts that have already been put forward were
10 organized by what is called the Microarray Gene
11 Expression Data Society, or MGED. They have come
12 up with a number of key concepts. The first is
13 this MIAME, the minimum information about a
14 microarray experiment. I have quoted from the MGED
15 web site how they describe that but it is
16 essentially what should go into the database; what
17 is the minimum information u need to be able to
18 make sense out of the results.

19 [Slide]

20 The basic areas that are covered in this
21 are the experimental design, samples used, the
22 extract preparation, labeling, the hybridization
23 procedures and parameters, measurement data and
24 specs and the array design. Now, truth be told,
25 all of this is focused around the original MGED and

1 MIAME focus which was not toxicology. It was more
2 looking at array experiments that would come with
3 kind of a minimal amount of biological descriptors.

4 [Slide]

5 The MGED Society also came up with MAGE,
6 and I should say MAGE-ML. Under MAGE there is more
7 than just MAGE-ML. These are the programming
8 conventions and the data structures to be able to
9 communicate the data. So, you have a MAGE-OM, the
10 object model for the data. Then you have a markup
11 language which allows the exchange of the data from
12 one database to another. So, really what MAGE is
13 about is structuring your data and structuring a
14 way to communicate your data such that, quite
15 frankly, as long as you have a MIAME compliant
16 database it doesn't matter whether or not you use
17 your database or somebody else's database, the data
18 should be able to transfer seamlessly back and
19 forth.

20 [Slide]

21 Finally, under the MGED Society--not
22 finally, there is another point but under the MGED
23 Society is an ontology working group which is
24 striving to provide a vocabulary that will
25 communicate the information about a particular

1 topic, in this case microarrays, but it is also not
2 just communicating the knowledge but allowing its
3 interpretation and use by computers. That is an
4 important point because when we say, in the example
5 that was given earlier using two different
6 spellings for tumor, the British and the American,
7 anyone in the room would understand what that is
8 but, one, if the computer wasn't trained to
9 recognize the synonyms or there was only one way
10 forward on that, one of those would cause serious
11 problems. So, it is not just communication from
12 person to person; it is communication from computer
13 to computer in a way that the computer can make
14 sense out of it. So, if you do have an ontology
15 that has standard terms, what you allow are
16 structured queries and unambiguous descriptions of
17 experiments.

18 [Slide]

19 John Quackenbush is a representative from
20 this angle of the MGED Society. There is a data
21 transformation and normalization working group
22 which is striving to establish standards for
23 recording how the microarray data is transformed
24 and normalized.

25 [Slide]

1 So, what about toxicogenomic databases?

2 What are the issues here? Well, first I want to
3 throw out an overview where the ILSI effort is.

4 Again, you have probably heard some of this but
5 just as a recap, in the genotoxicity group there
6 are upwards of 10 array platforms, 11 compounds
7 with two time points and up to 10 doses per
8 compound--it is fair to say, a fair number of
9 arrays. Nephrotoxicity group, six array platforms,
10 three compounds, a total of 260 animals. Suffice
11 it to say that 260 animals means that there are at
12 least that number of array data points in there.

13 [Slide]

14 In the hepatotoxicity group they used
15 about eight platforms, two compounds, a total of
16 144 animals. In this case, those 144 were split
17 into two in-life studies per compound. Now, for
18 all of the groups there was analysis of each sample
19 at multiple sites. So, the ILSI effort really
20 represented I think a microcosm of the kinds of
21 issues that are going to be confronted when folks
22 try to pool data together from multiple sources.

23 [Slide]

24 One of the issues going into this we
25 really fully unappreciated was that MAGE, MIAME or

1 MGED ontologies just did not address the
2 traditional toxicology endpoints, the issue of
3 organ weights, clinical pathology, histopathology
4 and the like. That was not specified in the
5 original MIAME document or the MAGE-ML. So, that
6 became an issue for ILSI and EBI to address.

7 [Slide]

8 Likewise, another issue is that these tox
9 endpoints are standardized in nomenclature. We
10 have heard that referred to before. I have dug up
11 at least two types of nomenclature for clinical
12 pathology and chemistry. Under histopathology,
13 this is at least the length of the list and who
14 knows there are groups using their own customized
15 list as well. For putting together the ILSI-EBI
16 database we chose to work with the IUPAC
17 designation for clinical pathology and we borrowed,
18 if not stole, liberally from the NTPs TDMS
19 pathology code database.

20 [Slide]

21 I keep referring to the ILSI-EBI effort
22 but I think it is important to remember that it is
23 not occurring in a vacuum, nor is there a lack of
24 other players out there. A number of private
25 companies have put together toxicogenomic databases

1 with a variety of different foci. Genelogic,
2 Iconix and Curagen are the main players in this.
3 Tim Zacharewski's lab at Michigan State has
4 published a database structure that is designed to
5 handle toxicogenomic data. It is called dbZach.
6 Mike Waters' group at the NIEHS is putting together
7 a database referred to as CEBS, which is Chemical
8 Effects in Biological Systems. NCTR has also
9 developed a structure to capture array data, called
10 ArrayTrack, and last on the list is the effort that
11 ILSI partnered with EBI.

12 [Slide]

13 The collaboration came out of one of
14 ILSI-HESI's goals as far as the genomics
15 subcommittee. That was the establishment of a
16 database for toxicogenomics data. Indeed, these
17 three bullet points are the ones that we were
18 charged, in the database working group, to push
19 forward on. Importantly, and I think this is an
20 important point, we wanted the database to be able
21 to interrogate the gene array data and integrate it
22 with genomic experimental and toxicological
23 domains. That would gain knowledge of links
24 between gene experiments changes and toxicological
25 endpoints. This is a key point because I would

1 venture to say that while you have heard
2 discussions and often hear discussions of people
3 looking at array data and saying I see a correlate
4 with a biological endpoint, usually that
5 correlation is made, quite frankly, sort of by
6 human intuition, in other words, at the high dose
7 group I saw certain histopathological effect and I
8 see the gene changes so, therefore, there is a
9 correlate. Or, let's say a particular group had on
10 the whole an elevated ALT level and that correlated
11 with on the whole the gene changes we saw for that
12 group.

13 What we are trying to drive to here is to
14 be able to do that kind of correlation on a
15 statistical, electronic and individual animal basis
16 within the database. So, the thrust of it and the
17 challenge is a little bit beyond that essentially
18 intuitive approach to those correlations. It is an
19 approach that would get you to answering certain
20 questions. I will get to that in just a minute
21 because I just want to mention some of the issues
22 that we have in the collaboration.

23 [Slide]

24 We needed to provide a way to integrate
25 the different domains. We needed to control the

1 annotation. Of course, you need to centralize the
2 information. You need to improve the array
3 annotations as genome assemblies are released and
4 improved, and allow data comparison. That gets to
5 the point that you want to be able to go and
6 compare data from different domains.

7 [Slide]

8 I think my point here is just simply that
9 we needed to get internally consistent data to be
10 able to run these complex queries and, yet, we had
11 data emanating from several different sites.

12 [Slide]

13 Here is the meat of the question, a simple
14 question, does gene X expression go up after
15 treatment with compound Y with biological endpoint
16 Z in experiments from ILSI members A and B? That
17 is relatively easy to ask. You look at gene X, you
18 look at biological endpoint Z and, you look at
19 compound Y, and you look at a couple of datasets.

20 However, it is not a simple question. One
21 that you can only address with the databases, is
22 one which follows: Which are the most reproducible
23 gene expression changes for all the experiments on
24 the array with biological endpoint X, and which
25 functional category do these genes belong to and

1 which are the human homologues? That is a
2 challenge and it simply requires you to have a
3 robust database where the data is captured in a
4 standardization way and mapped on the sequence
5 level.

6 [Slide]

7 Which brings me, since I am talking about
8 standardization, to MIAME-Tox. MIAME-Tox is simply
9 an international effort to share expertise,
10 encourage harmonization and promote a
11 standardization initiative. So, with the central
12 theme being toxicogenomics, this represents an
13 alliance between ILSI-HESI, EMBL-EBI and, quite
14 frankly, Mike Waters' group at the NIEHS, at the
15 National center for Toxicogenomics. It has been an
16 extremely fruitful effort so far and I would say
17 that this is a party that is growing and we are
18 encouraging folks to join in.

19 [Slide]

20 These are the objectives. The first is to
21 come up with standard contextual information. That
22 is, put together a worldwide scientific consensus
23 on what is the minimal information or descriptors
24 you need for array-based toxicogenomics
25 experiments.

1 Another objective is that of data
2 harmonization, how you encourage use of controlled
3 vocabularies for the toxicological assessments.
4 Another objective is to push for data integration
5 and data sharing so that you can link data within a
6 study or several studies from an institution and
7 exchange datasets among institutions. Finally, to
8 set up a structure for data storage that will allow
9 the development of data management software and
10 databases. Right now, the two that we are talking
11 about in development are ArrayExpress at the EBI
12 and CEBS at the NIEHS National Center for
13 Toxicogenomics.

14 [Slide]

15 There is a document out there to promote
16 standard contextual information. It is trying to
17 define the core common to most experiments. It is
18 designed to promote data harmonization, capture and
19 communication. Along those lines, in terms of this
20 harmonization and communication, it is worth
21 remembering that MIAME-Tox is based upon the same
22 structure that MIAME has. However, MIAME-Tox
23 document really is a focus on the toxicological
24 domain, the sample treatment and conventional
25 toxicology information as it is integrated with the

1 microarray information.

2 [Slide]

3 You can look at this document at either
4 the MGED Society web site of the ILSI-HESI web
5 site, and it is really out there for circulation,
6 for review and for comments. The MIAME-Tox group
7 is working closely with the MGED working groups, in
8 particular the ontology working group, with the
9 thrust of trying to develop controlled
10 vocabularies.

11 [Slide]

12 In our hands, really what we were
13 confronted with for this controlling data and
14 controlling the structure and nomenclature was to
15 look at data input as a key step. So, with the
16 charge of capturing data in a standard manner, EBI
17 developed what they call the Tox-MIAMEExpress. This
18 is used to store information domains in a database,
19 the ArrayExpress database, and allow comparing
20 queries across and within domains.

21 [Slide]

22 I am going to kind of quickly go through
23 some Tox-MIAMEExpress web shots because I think to
24 take a look at this gives you some sense of how the
25 data is organized, how it is going in. First you

1 have a protocol submission which really covers not
2 just the microarray experiments but, obviously in
3 the case of toxicology now the conventional
4 toxicology tests. So, you can see here are the
5 kinds of protocols that you can submit. Obviously,
6 once you submit one you can refer to it for any
7 experiments that use that protocol.

8 Then you move on to the array design
9 submission which is important because these are the
10 procedures that format the array design into
11 something that EBI database can use to refer from
12 one array to another. It also sets up a set of
13 procedures to re-annotate or update your array
14 designs via link to sequence data at EBI.

15 The experiment submission is now actually
16 the meat and potatoes of it where, first, you are
17 going to submit the experimental design, some of
18 the information about quality controls and,
19 finally, the samples. Quite frankly, the samples
20 are your individual animals.

21 The point that follows is to submit
22 toxicological endpoints, what sort of extracts you
23 make from individual tissues, what sort of labeled
24 extracts are going to be used for microarray data
25 and finally the hybridizations that are used for

1 the microarray data.

2 [Slide]

3 This gives you a screen shot of the data
4 that we have been entering into it. Obviously, you
5 can get a flavor for what kind of data is captured,
6 how it is captured. The drop-down menus allow
7 control of the vocabulary. I venture to say, after
8 working through this personally, it is a work in
9 progress. It captures a great deal and represents
10 I think a fantastic starting point but it is
11 something that I encourage everyone in the
12 audience, and anyone out there, to offer input on.

13 [Slide]

14 Here is an example of data entry for
15 clinical pathology. The challenge, of course, as
16 we have found in our own hands, is if you have
17 collected the data in different units and you have
18 to convert them.

19 [Slide]

20 These are the sorts of clinical
21 observations that are collected.

22 [Slide]

23 I would like to add something to this
24 slide, and that is some of the future directions
25 but first I want to say where we are with the

1 status. I have shown you the interface and the
2 infrastructure that is already in place. I have
3 alluded to the fact that it is not as if it is
4 fixed or immutable at this point. We are putting
5 data into it. It is not complete yet but we
6 envision that probably in the next quarter or so.

7 There are some key important points I want
8 to mention in terms of future development.

9 Certainly what I have alluded to is developing the
10 tools that will query across different domains.

11 That is not listed in this slide but it is
12 definitely something that we are looking to work
13 with EBI on. Finally, a key point in further
14 development is working towards automated data
15 upload or electronic data upload of toxicological
16 data. That is, if it is already collected in an
17 in-house electronic database, how can we transfer
18 that data seamlessly using an electronic upload?

19 [Slide]

20 I would like to end with some mention of
21 the guilty parties. Certainly, the Microarray
22 Informatics team at EBI and Alvis Brazma is the
23 MGED Society president and really I would say one
24 of the MIAME proponents. Susanna Sansone has been
25 our key contact at EBI and responsible for really

1 all the progress you have seen in the database
2 there, with Philippe Rocca-Serra helping her in
3 putting that together. I don't have Mike Waters'
4 name here but I should because he has been an
5 invaluable help in contact at the NIEHS. Of
6 course, the rest of the EBI steering committee has
7 been an important player and, finally, certainly
8 the genomics committee. With that, I thank you and
9 will take questions.

10 DR. KAROL: We will take questions right
11 after the next speaker. So, our last speaker in
12 this session is Lilliam Rosario, who will talk to
13 us about CDER FDA initiatives.

14 CDER FDA Initiatives

15 DR. ROSARIO: Good afternoon.

16 [Slide]

17 My presentation today will basically
18 address four main initiatives that CDER has
19 undertaken so far in an attempt to better
20 understand the field of pharmacogenomics and to
21 anticipate regulatory considerations stemming from
22 the rapidly evolving field of toxicogenomics.

23 [Slide]

24 So, what I would like to do is tell you
25 about the formation of the nonclinical

1 pharmacogenomics subcommittee. I also would like
2 you to know about some of the regulatory research
3 lab-based initiatives currently going on stemming
4 from the Office of Testing and Research. I also
5 would like to tell you about ongoing collaborations
6 with Iconix Pharmaceuticals, the developers of a
7 drug matrix of microarray data linked to tox
8 parameters and, finally, our collaboration with
9 Expression Analysis to come up with a mock
10 submission of microarray data provided by Schering
11 Plough.

12 [Slide]

13 First I would like to tell you about the
14 nonclinical pharmacogenomic subcommittee. The
15 subcommittee is part of the pharm/tox coordinating
16 committee and has been founded to address the
17 rapidly developing field of pharmacogenomics. The
18 goals of this committee are to recommend standards
19 for the submission and review of nonclinical
20 pharmacogenomics and toxicogenomic datasets to
21 develop an internal consensus regarding the added
22 value, the best interpretations in drug development
23 and regulatory review implications of this type of
24 nonclinical data, and to develop Center expertise
25 and an appropriate infrastructure to support the

1 review of these types of data. I also should note
2 that the objectives of this committee may continue
3 to evolve with time to include, for example,
4 proteomics and metabonomics.

5 [Slide]

6 The membership of this committee is
7 intended to be very broad and currently it has
8 participants from all the different ODEs, the
9 Office of Testing and Research as well as the
10 Center for Biologics.

11 [Slide]

12 The functions of the subcommittee are to
13 interface with other CDER review disciplines, such
14 as the clinicians and the statisticians, and other
15 centers within the agency in recommending review
16 standards. It is also to develop specific
17 initiatives to keep committee members abreast of
18 the latest developments; to assist other
19 submissions and center groups in developing
20 educational opts in pharmacogenomics and
21 toxicogenomics; to provide forums for communication
22 to regulated industry; to obtain external expertise
23 to evaluate the scientific developments, as well as
24 to provide internal expertise in evaluating
25 nonclinical data submissions that contain

1 pharmacogenomic or toxicogenomic information.

2 [Slide]

3 This committee was formed last August and
4 it has been extremely active since then. So far it
5 has contributed input to CDER mg concerning
6 research information package and no regulatory
7 impact, as you heard from Dr. Woodcock this
8 morning. It has contributed to the nonclinical
9 section of the CDER draft guidance on
10 pharmacogenomics and pharmacogenetics, and
11 initiated process toward the development of a draft
12 guidance on the content and format of nonclinical
13 pharmacogenomic data submissions, and this is one
14 of the reasons why we are gathered here today.

15 It is currently actively participating in
16 collaboration with Iconix Pharmaceuticals, and I
17 will tell you a little bit about that collaboration
18 further on, and participates in the collaboration
19 with Expression Analysis and Schering Plough. So,
20 as you can see, this subcommittee has poised itself
21 to really serve as an interface within the agency
22 to provide internal expertise and to seek out
23 expertise from outside collaborators.

24 [Slide]

25 I would also like to tell you about some

1 of the regulatory research lab-based initiatives.
2 These are aimed at really getting the technological
3 part of microarray data to bring it into regulatory
4 practice.

5 [Slide]

6 It has done so by an early active
7 participation in the ILSI collaborations, and this
8 will be nephrotoxicity and genotoxicity working
9 groups; collaborations with Affymetrix and Rosetta,
10 and this will be with the cardiotoxicity focus;
11 also collaborations with NCTR and Schering Plough.

12 [Slide]

13 As was mentioned before, these lab-based
14 initiatives are trying to get a handle on all the
15 technology issues. For example, genome scale
16 expression data submitted to the agency could be
17 generated from a variety of microarray platforms,
18 and these platforms can be from oligonucleotide or
19 cDNA-based arrays, numerous commercial platforms as
20 well as in-house custom arrays. So, one of the big
21 questions is can a standard be developed that would
22 help assure the FDA of the biological truth, that
23 is, the biological truth independent of a platform
24 and site or processing?

25 [Slide]

1 As you briefly heard from Dr. Ghosh, there
2 is an ongoing project through the FDA Office of
3 Science and Health Coordination. It has funded a
4 collaborative project to evaluate performance
5 standards and statistical software for regulatory
6 toxicogenomic studies. This study as a laboratory
7 component that is headed by Drs. Thompson and
8 Fuscoe from CDER and NCTR respectively. It has a
9 laboratory component with outside collaborators
10 that include Rosetta, Agilent, NIEHS, Amgen, Iconix
11 and Affymetrix, and it has a statistical component
12 that is being provided by FDA centers.

13 [Slide]

14 The goal of this project is to generate
15 and evaluate a complex mixed tissue standard's
16 utility for assessing platform features. What will
17 be assessed in this case will be to assure that
18 there are no manufacturing defects; that there is
19 insignificant platform lot-to-lot variability; to
20 assess the integrity of feature location; to ensure
21 that there is unambiguous consensus sequence
22 annotation; and a lack of cross-contamination in
23 tiled probe features.

24 [Slide]

25 The standard will also serve to assess

1 experimental performance. I won't go through all
2 these points but just tell you that these will be
3 aimed at assuring that the biological conclusions
4 are independent of the platform and represent the
5 biological truth.

6 [Slide]

7 Again as Dr. Ghosh mentioned earlier, the
8 proposed steps for testing the feasibility of a
9 mixed tissue standard is by using bench mark genes,
10 in this case to identify tissue-selective, low
11 variance housekeeping genes from control animal
12 data in large databases, and to select the tissues
13 with most consistent expression among control
14 animals and most coverage of the probes.

15 [Slide]

16 As you can see, we also have a laboratory
17 component that is trying to sort out the
18 technological issues in order to bring this new
19 technology into regulatory practice.

20 [Slide]

21 I briefly want to tell you a little bit
22 about our collaboration with Iconix
23 Pharmaceuticals. Iconix Pharmaceuticals are the
24 developers of the DrugMatrix that contains
25 microarray data that is linked electronically to

1 toxicology and pharmacology endpoints. So far,
2 Iconix has provided research access to the
3 DrugMatrix system for evaluation purposes to train
4 members of the subcommittee.

5 We visited their facility back in January
6 and they provided some training, and continue to
7 provide support and understanding in working with
8 their database. They have provided us with
9 hands-on experience using a chemogenomic data and
10 tools, including the application of molecular
11 toxicology markers to predict drug actions. Also,
12 we got first-hand experience with a very large
13 dataset linked to traditional toxicology outcomes.
14 The importance of this is to know that we are going
15 to be developing guidance in terms of the optimal
16 and minimal content and format for the submission
17 of microarray data, and looking at this database
18 has definitely provided us with a very, very good
19 experience as to how they look and the things that
20 we should consider important. So, as I mentioned,
21 Iconix continues to provide training and support in
22 the area of QA/QC, as Kurt mentioned this morning,
23 and analysis of the data across multiple gene
24 microarray product platforms, and the derivation
25 and validation of markers or toxicity and mechanism

1 from integrated chemogenomic datasets.

2 [Slide]

3 Finally, I would like to tell you about a
4 collaboration with Expression Analysis and Schering
5 Plough. This is to develop a mock submission of
6 microarray data, and the data will be provided by
7 Schering Plough.

8 [Slide]

9 The objectives are to provide a suitable
10 framework in which to augment, reduce or further
11 define a potential list of recommendations; to
12 contribute to the development of consensus around
13 the specific elements of applicable recommendations
14 within the context of a mock submission; and to
15 contribute to building and refining a process in
16 which microarray data may be submitted to the FDA.

17 [Slide]

18 We met with Expression Analysis back in
19 May for concept definition and refinement of scope.
20 We are expecting a pilot submission in July and a
21 completed mock submission by October. This should
22 give us a very good experience as to the details
23 that we need to sort out in order to receive
24 microarray data.

25 [Slide]

1 The areas to be addressed during this
2 process of receiving this mock submission of
3 microarray data are laboratory infrastructure, data
4 management, study-specific array performance,
5 experimental design, pre-processing and statistical
6 analysis methods, as well as the interpretation of
7 the results.

8 [Slide]

9 For the purpose of this presentation I
10 just want to focus on the data management aspect.
11 It is to attempt to sort out things like data files
12 and file structures, the variables and their
13 definitions, and how to link all this information
14 or microarray data to other databases such as
15 histopathology or clinical chemistry.

16 [Slide]

17 I should tell you that the first thing we
18 want to do is just to look at the infrastructure
19 that is currently in place. What we did was we
20 looked at what we have. There is a guidance that
21 was published in January of 1999 providing
22 regulatory submissions in electronic format.
23 Specifically, this guidance says that animal line
24 listings can be submitted as datasets. So, animal
25 line listings that you would provide on paper or in

1 PDF format may be provided as datasets. So, each
2 domain should be provided as a single dataset.

3 [Slide]

4 The guidance goes ahead and gives a list
5 of recommendations. I won't go into a lot of
6 detail, but just to mention some of the salient
7 points, such as each dataset should be provided as
8 a SAS transport file. The size should be less than
9 25 MB per file, not compressed. There are some
10 specifications about the data variable names and
11 the description of these data variables and the
12 labels. Data elements should be defined in
13 definition tables. Each animal should be
14 identified standard a single, unique number for all
15 the datasets in the entire application. The
16 variable names and codes should be consistent
17 across the studies, and the duration of treatment
18 should be provided based on the start of the study
19 treatment.

20 [Slide]

21 This is an example of a dataset and data
22 elements as stated in the guidance. What I would
23 just like to point out is some of this--variable
24 name and it is stated that it should be eight
25 characters. The label should be very descriptive

1 of the variable. For example, here, lab test is
2 the name of the variable and it would include any
3 other variable, such as clinical, chemical or
4 hematology or clinical science.

5 [Slide]

6 This is an example that tells you what the
7 histopathology table should look like. For
8 example, the name of the organ and then the
9 different findings, macroscopic findings and
10 microscopic findings, should be defined after that.

11 [Slide]

12 So, we have something in place in order to
13 submit datasets electronically. However, so far
14 this does not include anything on how to submit
15 microarray data. However, back in January there
16 was a notice in the Federal Register on a pilot
17 project for nonclinical datasets. Dr. Randy Levin
18 actually told us a little bit about the CDISC
19 project. This pilot project is part of an effort
20 to improve the process for submitting nonclinical
21 data. Eventually, FDA expects to recommend
22 detailed data standards for the submission of
23 nonclinical data.

24 The FDA received recommendations for a
25 standard presentation of certain clinical data from

1 the CDISC and CDISC is currently facilitating the
2 work on similar standards for nonclinical datasets.
3 So, now what we have is some infrastructure and we
4 have an initiative going on, which just points out
5 that this is a very opportune time to try to get
6 these issues resolved.

7 [Slide]

8 So, what we did, we went ahead and
9 compared our current infrastructure to some of the
10 mechanisms being proposed outside. So, we compared
11 the CDER guidance to the MIAME-Tox proposal. I
12 should mention that this is by no means an
13 exhaustive comparison but it is just to point out
14 and highlight some of the similarities and
15 disparities that we currently have, again
16 emphasizing that this just points out that it is an
17 opportune time to try to get these issues resolved
18 and addressed.

19 For example, the CDER guidance paradigm
20 appears more comprehensive with less restrictive
21 vocabulary. For example, the CDER proposal treats
22 LABTEST as a variable, while the MIAME-Tox proposes
23 a field for each possible clinical chemistry test.

24 Again, what this really tells us is that
25 the CDER guidance is actually more malleable and at

1 this point will be able to accept MIAME-Tox
2 formatted data. So, if there was consensus that
3 this would be the best way to get the data
4 formatted, then the agency will be able to accept
5 such data.

6 The MIAME-Tox collects information on in
7 vitro experiments, whereas the agency generally
8 does not receive line listing for pharmacology
9 data. This goes back to what Dr. Leighton was
10 telling us about a little bit earlier, that the
11 requirements for the submission of data that is
12 pharmacology and toxicology are different. For
13 example, line listings are required for toxicology
14 data and are not for pharmacology. Thus, the CDER
15 guidance currently doesn't have a mechanism to
16 accept pharmacology data because it is typically
17 not submitted as line listings.

18 On the other hand, in a typical toxicology
19 study you generally have pharmacokinetic
20 assessments and MIAME-Tox at this point does not
21 collect information on drug plasma levels. So,
22 these are just some of the differences, very
23 overall differences and similarities but mainly
24 what it points out, again, is that now that we have
25 initiatives going to standardize the nonclinical

1 terminology, as well as initiatives to figure out
2 the best way to collect a standardized
3 database--that this will be the best time to try to
4 get those two things together and make them
5 compatible.

6 [Slide]

7 I am just going to mention some
8 considerations for the submission of microarray
9 data. Based on what I just told you, it seems that
10 it would be useful to have sponsors provide
11 annotations to nonclinical data containing array
12 information by following a guidance-compliant
13 format. That would be with the disclaimer that the
14 guidance may have to be extended to include how the
15 array data may be submitted.

16 This is, again, something to consider,
17 that is, to include the following files. So, the
18 raw data files post image analysis, and in the case
19 of the Affymetrix array data that would be the CEL
20 and the CHP files, linked by animal identifier; and
21 to include a summary report to describe any
22 normalizations, data processing, and/or statistical
23 analysis, basically how conclusions were derived.

24 [Slide]

25 Let me tell you a little bit about the

1 thinking behind perhaps having sponsors submit
2 these raw data files post image analysis. Here is
3 a table that presents what these files mean,
4 particularly for the Affymetrix data. For example,
5 in this case we would perhaps be asking the sponsor
6 to submit the CEL file, which basically can be used
7 to reanalyze data with different expression
8 algorithms but it basically gives it to you
9 readable in any type of text editor. So, you would
10 have to be able to generate data tables that would
11 be suitable for review purposes. The CHP file in
12 this case would quantify and qualify the transcript
13 and its relative expression level.

14 So, the question is how about this DAT
15 file? It is 40 MB. It is raw data. At this point
16 we are leaning not towards the submission of this
17 specific file. Some people argue that one of the
18 reasons why you might want to have the DAT file is
19 because you would be able to address issues such as
20 this.

21 [Slide]

22 As you can see here, this just shows a
23 defect in this chip, and by looking at this image
24 you would be able to assess that. However, I think
25 we can probably come up with some other ways in

1 which you can get this information without having a
2 40 MB file submitted to the agency, perhaps a
3 picture in a PDA format or just the information
4 from the CEL file, or come up with some QA/QC
5 matrix that would allow us to determine the
6 appropriateness of the experimental setup, in this
7 case the chip integrity.

8 [Slide]

9 This is just to give you an example of
10 what a probe detection report would look like
11 coming from a CHP file. Again, since this will be
12 able to be modified in any text editor, the tables
13 might look different depending on how the sponsor
14 would like them to look.

15 [Slide]

16 So, these are suggestions for submission
17 of array data. By evaluating several submissions
18 we can gain understanding of the fields and issues
19 that need to be reconciled for database purposes.
20 This proposal works with the current guidance. It
21 does not create any additional burden for the
22 sponsor and leaves the possibility of an in-house
23 database creation.

24 [Slide]

25 With this mock submission data, what we

1 are trying to do is sort out the details as to how
2 the data should be submitted, what it should look
3 like, and it also would give us an idea of the
4 things that we need to consider in order to have
5 the best infrastructure to receive this data.

6 I hope that with this presentation I have
7 given you a flavor as to the main initiatives that
8 are currently going on here, in CDER, in order to
9 prepare ourselves to really understand the field of
10 pharmacogenomics and the regulatory considerations
11 stemming from the development of toxicologies.
12 Thank you.

13 DR. KAROL: Thank you very much. What we
14 will do now is have questions for any of the
15 presenters, then at 2:30 I am going to turn the
16 session over to Dr. Sistare for him to ask
17 questions of the panel. So, now any of the papers
18 are open for questions. Yes?

19 DR. SISTARE: A question for Bill Mattes.
20 Bill, one of the fields that didn't come across on
21 one of the visuals that you had was histopathology.
22 What is the current thinking? What is the current
23 status really of the MIAME-Tox menu and choices
24 with respect to being able to pick and choose the
25 descriptors you need for the histopathology? Is it

1 felt it is robust enough, it is adequate? Do you
2 feel that you have got the consensus of the
3 pathology community and professional societies? Is
4 there some work that needs to be done there to sort
5 of get a better feel that we have the consensus; we
6 have what we need at this point in time?

7 DR. MATTES: No and yes. No, you didn't
8 see the histopathology. I was trying to keep
9 slides to a minimum and it is always a question
10 what you put in and what you leave out. In the
11 case of histopathology, that was an interesting
12 dynamic we went through. We had considerable
13 debate on what to do. Histopath was obviously
14 collected at numerous sites originally, yet, when
15 we sort of met as a group to discuss how to handle
16 this--we had Roger Brown from GlaxoSmithKline sort
17 of enlighten us, those of us who had not been so up
18 close and personal with pathologists. He
19 enlightened us that, you know, if you have two
20 pathologists you will have three different opinions
21 so he encouraged us to take the approach of having
22 all of the data reread by one pathologist.

23 So, what we did, we were having Peter Mann
24 at EPO read it and capture it in an EXCEL
25 spreadsheet. It has drop-down menus and controlled

1 vocabulary. He kind of agreed to it and the
2 nomenclature was basically ripped off from NPT.
3 So, we are in the latter stages of capturing that
4 data. There is good and bad to this approach. The
5 good is that for this particular dataset we will at
6 least have consistent histopath. We haven't
7 entertained the thought of trying to see how that
8 correlates with the previous histopath that was
9 done, obviously not collected electronically, but
10 that is the status.

11 Now, in terms of how does this jive with
12 the rest of the histopath community, you know, I
13 certainly don't want to die on that hill. I know
14 that is a tall order, to harmonize that
15 nomenclature. I am hoping that in this exercise we
16 might be catalyzing some movement along those
17 lines. As I say, the other thing would be to
18 capture all the separate histopath readings that
19 were done in the individual companies and sort of
20 run an "ooh, what did you think" comparison. But
21 for the purposes of this dataset we had one
22 pathologist read it, or we are having one
23 pathologist read it and that nomenclature is pretty
24 similar to the NTP.

25 DR. BROOKS: I have a question for Kurt

1 Jarnigan. A number of the speakers spoke to the
2 importance of experimental design and I think for
3 this technology or most genomics-based technology
4 that is critical. However, you were the only
5 person that provided a number as far as replicates
6 in experimental design goes, and I was wondering if
7 you could go into more detail with respect to your
8 biological replicates of three and whether or not
9 that is something that should be limited to in
10 vitro studies or can be expanded to in vivo
11 studies, and I guess speak to how you arrived at
12 that number and expand on that a little, please.

13 DR. JARNIGAN: Those were designed to be
14 minimum study sizes. Those are the minimums that
15 we find useful, mostly because that is the minimum
16 you can do any useful statistics on.

17 DR. BROOKS: But let's say you are looking
18 at human tissue, still a minimum of three
19 irrespective of the control for genetic diversity
20 and some of the other factors in your models?

21 DR. JARNIGAN: Well, a minimum of three
22 but, yes, probably in those settings--I can only
23 speculate as I have no personal experience with
24 human tissues derived from patient samples, but I
25 would speculate that you would need more than three

1 to derive any statistical power of any kind in that
2 setting. But for the case of animal studies, which
3 we have done a lot of, I can say that three is
4 very, very good and in a good lab with careful
5 quality control it would be adequate to cover most
6 major toxicological and pharmacological findings.
7 Clearly, for some of the more idiosyncratic
8 findings, yes, you will need more than three to
9 cover those and in some specific experimental case
10 you probably would need more. But for your average
11 run-of-the-mill toxicological findings or the
12 average run-of-the-mill pharmacological findings
13 three will do if the experiment is done carefully.

14 DR. BROOKS: Do you find that increasing
15 your number of replicates will increase your
16 sensitivity depending on what you are looking at?
17 Or, does it not make a difference at this point?

18 DR. JARNIGAN: We have only examined
19 between three and six, to answer that question. I
20 haven't gone beyond six but it looks like we are
21 approaching an asymptote pretty quickly and beyond
22 six you don't really get much additional
23 sensitivity. In theory, it is a square root kind
24 of function so you quickly get to a point of
25 diminishing returns in that kind of a situation.

1 DR. QUACKENBUSH: If I could actually add
2 to that, I think part of the answer to your
3 question depends on what the goal of the experiment
4 is and how you want to do it. There are actually
5 two places in the literature where you can find
6 discussions of this to some extent. One is a paper
7 published by Gary Churchill in CHPing Forecast
8 Supplement to Nature Genetics where he talks about
9 the value of biological replication. Probably a
10 better reference is a paper by Rich Simon. I don't
11 have the journal citation at my fingertips right
12 now. [Simon et al., Genetic Epidemiology,
13 23:23-36, 2002] I can pull it up on a laptop if you
14 like, but he actually introduces a power
15 calculation for microarray experiments where he
16 goes through and looks at the level of sensitivity
17 you want to approach and the degree of biological
18 replication that you need as a function of the
19 variability in your assay.

20 So, while I think three is a good starting
21 point, you really have to be much more careful and
22 much more proactive about doing the up front work
23 to estimate what the inherent variability is before
24 you decide on a certain level of replication to
25 reach a certain goal in sensitivity.

1 DR. BROOKS: So, one could establish a
2 guideline based on the question or the model as to
3 how many replicates would be acceptable for a study
4 so you could properly evaluate the data.

5 DR. QUACKENBUSH: Exactly. I think what
6 you need to do is look at these power calculations
7 and sort of validate them, and then use that as a
8 standard.

9 DR. BUSH: I guess what I was getting at
10 is there need to be multiple different things;
11 there can't just be one design.

12 DR. KAROL: John, is that reference on
13 your slide? This might be a very good time to
14 announce that all of the slides will be posted to
15 the web site so that it will be on the web site,
16 John. There is no need to get it now.

17 DR. QUACKENBUSH: It wasn't actually
18 there.

19 DR. ZACHAREWSKI: While we are waiting for
20 that, I was wondering if I could ask Dr. Rosario to
21 talk more about the Schering Plough collaboration.
22 Is the source of the data part of the ILSI-HESI
23 effort or is this a separate effort altogether?

24 DR. ROSARIO: No, it is a separate effort.
25 The data provided by Schering Plough is not from

1 the ILSI effort. It is an independent dataset from
2 a compound and they have some microarray data
3 linked to toxicology parameters but it is just an
4 independent dataset.

5 DR. ZACHAREWSKI: So, it is not just the
6 microarray data, it would be microarray data and
7 all the other supporting IND data that is typically
8 submitted?

9 DR. ROSARIO: No, no, no. I think not in
10 the context of an IND; it is independent of that.
11 It is microarray array linked to some toxicology
12 parameters, but not within the context of a pooled
13 IND. Basically, the point of that is to sort out
14 exactly how the data should look, what components
15 should be submitted and, you know, sort out
16 variable names and the details of are we able to
17 actually receive the data with our infrastructure,
18 and things like that.

19 DR. ZACHAREWSKI: So, there will be, like,
20 clin chem and histopathology and all the other
21 nasties and goodies?

22 DR. ROSARIO: Yes.

23 DR. ZACHAREWSKI: So, will there be a
24 report about that?

25 DR. ROSARIO: Sorry, will there be a what?

1 DR. ZACHAREWSKI: A report.

2 DR. ROSARIO: Yes. I didn't go through
3 all the different statements in terms of the
4 deliverables. We have a report that should be
5 submitted, yes.

6 DR. LEIGHTON: With regard to the question
7 of variability, I think it is interesting or
8 instructive to point out that about three years ago
9 there was a very important paper, I believe, in
10 Cell by Yu, et al. from Rosetta Informatics where
11 they were looking at microarray data from a
12 particular strain of yeast that they were
13 experimenting on. In order to make sense of their
14 experiments and get a handle on variability--this
15 is in one laboratory with one sub-strain of
16 yeast--they did something like 50 or 52 controlled
17 cultures to get a handle on variability. Then,
18 once they were able to identify about 80 or 90
19 genes that varied tremendously in their controls
20 and tuned these out, they then were then able to
21 make sense of their experiments. So, I have become
22 a little concerned actually when people talk about
23 maybe three as the number for mammalian studies.

24 DR. JACOBSON-KRAM: One of the issues that
25 appears to be quite controversial is the issue of

1 whether or not studies need to be conducted under
2 good laboratory practices. So, I would like to
3 perhaps discuss this topic and say that any data
4 that is conducted as part of an initial safety
5 assessment, if it is pivotal data, then that should
6 be conducted under GLPs and all other data do not
7 need to be so conducted. We heard a lot about data
8 integrity, data quality going on. It seems to me
9 that good laboratory practices could help this
10 process. I would like to perhaps throw this out
11 for a question for discussion.

12 DR. KAROL: Any response to that?

13 DR. JACOBSON-KRAM: Has any vendor tried
14 to validate their system for GLP? I would be
15 pretty surprised. Kurt, do you know anything?

16 DR. ZACHAREWSKI: Kurt, were your studies
17 run under GLP?

18 DR. JARNIGAN: No.

19 DR. SISTARE: I would just mention that
20 the Expression Analysis does perform this function
21 as a service for sponsors, and they are striving
22 toward that end. We are actually trying to hold
23 them back a little bit, saying we don't have to
24 achieve GLP status at this point in time. But they
25 are striving to get there. So, I am seeing efforts

1 in that direction to do that, but for our purposes,
2 we indicated we don't have to achieve GLP status
3 here. You can specify however you want to the
4 first part, the laboratory parameters that they are
5 following, but they are doing things GLP-like.

6 DR. KAROL: Are there any other questions?
7 If not, I would like to turn it over to Frank.

8 Questions to the Subcommittee

9 DR. SISTARE: We have had a pretty full
10 day. Our attempt, our goal here today was to bring
11 all the committee members up to speed, up to the
12 same level playing field and, at the same time,
13 speak to our outside constituency as well. What we
14 have here is an opportunity to get open public
15 discussion, open public transparency with respect
16 to where the agency is at this point in time in our
17 thinking and in our goal setting.

18 I think as you can see from what we have
19 done today, we have brought everybody up to speed
20 with respect to where the experts out there in the
21 real-world are in terms of the technology
22 providers, in terms of trying to develop standards,
23 in terms of sponsors, how they are using the data.
24 We have heard excellent discussions from within the
25 agency on what we are trying to do to adhere to

1 existing standards with respect to electronic data
2 submissions, the kind of playing field boundaries
3 we have to stay within so we don't have to start
4 all over from scratch and create something that
5 creates a lot of havoc in the field. And, we have
6 brought you up to speed with respect to everything
7 we are doing internally as well.

8 We don't want to be perceived as being way
9 out there and trying to force a future. What we
10 want to be perceived as is as enabling and allowing
11 whatever the best future is for all of us to evolve
12 and to do things a better way. So, that is really
13 what we are trying to do here. FDA's goal is to
14 work as compatibly as we can with our constituency
15 out there. Our constituency is both the American
16 public in terms of assuring the best drugs get to
17 the marketplace, as well as the sponsors who we are
18 highly dependent on to develop these drugs and to
19 bring these drugs to market. So, they are as much
20 our constituency as the American public. We want
21 to work as closely as the regs allow us to, to
22 enable some preferred future and we have to define
23 what that preferred future is.

24 With that in context, I want to pose these
25 questions. I am just going to go through all of

1 them, all three of them. We have an hour for
2 discussion and I think the rules are that only the
3 people at the table can comment on these questions.
4 I apologize to those in the audience but these are
5 the playing rules. So, I will invite a lively
6 discussion from all the participants on the
7 committee here. I will go through the questions
8 and I will just invite all of the participants on
9 the committee to dive in on any particular question
10 that excites them the most but let's try to cover
11 them all if we can.

12 While most data from genome-scale gene
13 expression experiments are incompletely understood,
14 at the same time much of these data are considered
15 valuable. I think each and every day, as we have
16 heard, there is exponential growth in the
17 realization of the value of the measurements of
18 these transcripts. So, it is a rapidly growing
19 curve that we are on. Reluctance, however, has
20 been expressed in incorporating these endpoints
21 into routine pharmacological and toxicological
22 investigations.

23 The questions are, should the FDA, Center
24 for Drug Evaluation and Research in particular, be
25 proactive at this time in enabling the

1 incorporation of such study data into nonclinical
2 phases of drug development and in clarifying how
3 the results should be submitted to the agency?
4 What should present and future goals be for use of
5 the data by CDER? What major obstacles are
6 expected from incorporating these data into
7 nonclinical regulatory studies?

8 Second question, concerns have been raised
9 about gene expression data reproducibility across
10 laboratories, across platforms and technologies and
11 over the volume of data generated from each
12 experiment. First of all, is it feasible,
13 secondly, reasonable and, third, necessary for CDER
14 to set a goal of developing an internal database to
15 capture gene expression and associated phenotypic
16 outcome data from nonclinical studies in order to
17 enhance institutional knowledge and realize the
18 data's full value?

19 We have had a few submissions of
20 microarray data. They have come to us in paper
21 format. I think we have heard a number of speakers
22 today indicate that that is a pretty difficult way
23 to get any really useful information out of the
24 full dataset. So, the question is should the data
25 come to us electronically in a format that we can

1 archive and use and learn from?

2 The third question is concerns have been
3 expressed over reanalysis and re-interpretation of
4 large gene expression datasets. You heard Lilliam
5 say that the CEL file would be a nice file to be
6 submitted. The CEL file does allow reanalysis of
7 the data. Affymetrix data analysis has gone
8 through an evolution from a number of different
9 ways of doing that and we see publications coming
10 out at least once or twice a year on another way of
11 analyzing data. So, if the CEL files are
12 submitted, that would allow that kind of a process.

13 Is it advisable for CDER to recommend that
14 sponsors follow one common and transparent data
15 processing protocol and statistical analysis method
16 for each platform of gene expression data that
17 would be submitted but, at the same time, not
18 preclude sponsors from applying and sharing results
19 from additional individually favored methods? This
20 would at least allow one beginning, starting level
21 playing field.

22 What specific advice do you have to us for
23 clarifying recommendations on data processing and
24 analysis, as well as data submission content and
25 format? Our goal over the next six, seven, eight

1 months is to take your advice and to work from this
2 as well as our experience from the mock submission
3 data and from our own experience from working with
4 gene expression data to come up with a draft
5 guidance that will be used as a template, if you
6 will, for sponsors who choose to--we are not in any
7 way specifying that sponsors have to generate
8 microarray data, but if they choose to generate
9 data and as upper management works out the details
10 of whether data need to be submitted or not; if the
11 data need to be submitted, whether it goes into--I
12 will use the words safe harbor, I am not supposed
13 to use that word--safe harbor or non-safe harbor.
14 The question is how should the data be submitted to
15 us.

16 So, we are not going to focus on those
17 bigger issues that will be worked out in dialogue
18 with PhRMA and will be handled at a much higher
19 level, but the technical issues of how the data
20 could and should be submitted to us is really what
21 we hope to clarify for those sponsors who choose to
22 and wish to submit their data to us.

23 So, I leave those questions out there for
24 people to dialogue on. I guess I should just step
25 back and just let you dialogue.

1 DR. GOODMAN: Well, I would first like to
2 say, Frank, I congratulate you and your colleagues
3 here in terms of wanting to be proactive. It is
4 very, very important. But I think that I would
5 like to make just four points.

6 I think that toxicogenomics has a bright
7 future, but I think that there is a possibility to
8 short-circuit this by being too prescriptive at an
9 early time and we are, indeed, at an early time.

10 My suggestions would be to permit sponsors
11 to supply their data as they would write a paper
12 for a high quality journal and allow each to do it,
13 and do it in a scientifically solid, comprehensive
14 and defensible fashion. I would not move to set
15 standards at this time. I would try to shy away
16 from fixing in stone a database now because I am
17 concerned that fixing the database now could then
18 limit the ability to be expansive in terms of the
19 experiments because the experiments may then be
20 done to fit the database rather than following the
21 science.

22 The other thing that I frankly find a
23 little bit disturbing from the speakers and from my
24 general reading is that in the majority there seems
25 to be a tendency, although no one explicitly said

1 this, that the larger the number of genes on the
2 array the better and if someone has 15,000 someone
3 should try for 20,000 or 25,000 or 30,000. With
4 all of the difficulties we see in terms of analysis
5 and reproducibility etc., maybe there should be
6 some encouragement to focus on smaller subsets of
7 genes and, in a sense, to start walking before we
8 start running. Thank you.

9 DR. KAROL: Tim?

10 DR. ZACHAREWSKI: I would like to disagree
11 with my esteemed colleague. I think it is
12 important to provide guidance and that those
13 guidelines can change as we become more
14 knowledgeable in terms of the structure and the
15 format of the data. I think that if it is 15,000
16 genes or 30,000 genes it doesn't make that much
17 difference in terms of the analysis.

18 Interpretation is a different story and
19 what I would really encourage is that with these
20 mock submissions it comes as close to the other
21 required information as possible being provided as
22 well because I think it is going to be that other
23 supportive toxicological data that is going to put
24 that gene expression data into perspective, into
25 biological context. That is key. It will not only

1 help in terms of making sure you are not chasing
2 insignificant changes in gene experiments, but it
3 will also have significance in terms of providing
4 some kind of direction of what are the significant
5 changes in gene expression and, as NIH likes to
6 call it, phenotypically anchor those changes as
7 well.

8 I can't remember what other point I wanted
9 to disagree with. Do you want to share that again?

10 DR. GOODMAN: Just leave it as a general
11 disagreement.

12 DR. ZACHAREWSKI: Yes, we will continue
13 this on the plane home.

14 DR. HARDISTY: I feel that the FDA should
15 be proactive in any initiative like this. My
16 concern is that it may be a little bit premature to
17 incorporate these into routine nonclinical studies
18 and make them a requirement. I hear there is a lot
19 of need for standardization in the way the tests
20 are run, the protocols, the nomenclature. So, it
21 seems like it is very early in the process and it
22 may be that on a drug by drug or class of drug
23 basis that data may be very useful in helping in
24 risk assessment, but in most instances it is going
25 to be part of the evidence to support an overall

1 decision based on more standard toxicity studies.

2 I think though that this is the time for
3 FDA to get involved in it when it is early in the
4 process so that you can help lead it. Right now I
5 see that there are two or three groups almost
6 progressing in parallel and there is a lot of
7 overlap between those groups in nomenclature,
8 protocols and things like that. It is going to be
9 important to have some coordination between those
10 groups.

11 I just might mention a little bit about
12 nomenclature as a pathologist. It seems like there
13 is a lot of discussion about pathology
14 nomenclature. I realize that on this first study
15 one pathologist is going to reread all the
16 important target tissues. It may be a little
17 impractical down the road if studies are submitted
18 to the FDA to have one pathologist reread all the
19 important target tissues. Now, if you do have one
20 pathologist and he uses one set of nomenclature
21 such as that Dr. Mann is going to use the TCMS
22 nomenclature, the TCMS nomenclature in Dr. Mann's
23 hands will be fine but it is a list of words; it is
24 not a list of definitions. So, another pathologist
25 can use that same list of words and define them

1 more in line of his thinking as far as those words
2 go. So, I think that before we decide on which
3 nomenclature is accepted or is used, it may be good
4 to get a group like the Society of Toxicologic
5 Pathology or them in conjunction with maybe the
6 Society of Toxicology to look at this problem of
7 nomenclature and try to tie these changes in gene
8 expression to biologic changes in the tissues. It
9 is something that I know some of those
10 organizations will enjoy working on and will
11 probably do a very good job.

12 DR. BROOKS: I agree that FDA's
13 involvement in establishing guidelines now is a
14 good thing and that it is not going to hinder or
15 inhibit the development or the use of this data.
16 In fact, it may enhance it. Because of the fact
17 that there are so many different people, using so
18 many different technologies, doing so many
19 different things, without guidelines toward a
20 specific goal it is going to be much harder for
21 people to achieve that goal. I think even
22 independent programs, whether it is academia or
23 industry, are struggling with how they should be
24 doing things. So, some guidance from the right
25 perspective I think will be very helpful and I

1 think the FDA can be very constructive in that and,
2 as we learn more about the data and its ability to
3 be more informative for these applications, those
4 guidelines can become more rigid but right now they
5 can remain flexible.

6 With respect to the number of genes and
7 the data overload, there really are, you know, two
8 schools of thought and I think that some people
9 that started working immediately with specific
10 arrays are biological questions and if you make an
11 array where 99 percent of genes on that array
12 change as a function of your model, data analysis
13 becomes an even more difficult task. Biological is
14 broad; the arrays are broad and some of that
15 information that may not be used specifically for
16 biological inquiry is very important for
17 normalization and for understanding the systems
18 that you are interested in. So, I think data
19 analysis and the mathematical problems associated
20 with data analysis will continue to evolve.

21 But as Dr. Quackenbush stated, the fact of
22 the matter is you really do need to define your
23 question in order to be able to use this technology
24 effectively, and what the FDA has here with respect
25 to what they are interested in, toxicology, can be

1 a very well-defined question. If they can define
2 their question, they can use this technology
3 probably better in some instances and I think that
4 the question is here; it is just how well we can
5 define it.

6 With respect to building a database, I
7 think databases are good. We create them; lots of
8 people create them. I think that if the FDA wants
9 to start to look for its own development and for
10 its own information, not necessarily to hold that
11 information against sponsors but to use it to
12 continue to develop their question and their
13 guidelines, having that data at a raw level is
14 going to be important. So, as new mathematical
15 analytical models are established they can use them
16 to their benefit and not necessarily to the
17 detriment of their sponsors. Data analysis is the
18 one thing--you know, the technology has allowed us
19 to accelerate the development or the creation of
20 data tremendously. However, we really do in some
21 respects lag with respect to what we can do with
22 all of this data and being able to look at
23 thousands of genes at a time and how it relates
24 biologically. The guidelines I think should focus
25 on some of the technological variability which

1 allows us to focus on the biology. But from an
2 analytical standpoint for biology I think the FDA
3 needs to be involved in what analysis it feels
4 necessarily is important or what it will run or
5 expect to see, and that is probably the most
6 difficult question that I think faces some of the
7 guidelines that need to be created.

8 DR. WATERS: I would like to just pick up
9 a little bit on Dr. Hardisty's comments and try to
10 move them into the realm of toxicology. I think we
11 are really at an early stage in understanding how
12 to interpret molecular expression data in terms of
13 toxicology. I don't think we have put molecular
14 expression on toxicologic pathways yet. I think we
15 are just beginning to do that. I think we need to
16 understand those pathways in a molecular expression
17 context.

18 As we move towards that kind of an
19 endeavor and as we move towards building databases
20 we very definitely need to develop ontologies in
21 the toxicologic domain as well as the pathologic
22 domain. Those ontologies will be critical in
23 common understanding, common database query
24 capabilities in the future.

25 So, I do believe there is an important

1 need for consensus building and for international
2 efforts in doing this sort of thing. The MGED
3 Society has made an important start. There was a
4 contrast between MIAME-Tox and the efforts that are
5 ongoing at CDER. The MIAME-Tox effort is just the
6 beginning of an attempt to put forth a potential
7 guideline in the toxicology domain. I think there
8 needs to be participation and there has not been
9 participation thus far in clarifying that
10 guideline. So, to me, there is a lot of room for
11 us to define the domain of toxicology, to separate
12 that domain to some degree from the domain of
13 pharmacology to really understand what we mean when
14 we talk about toxic effects in a molecular
15 expression context.

16 In order to do that, we do need a
17 database. The question is do we really know how we
18 want to build that database at the present time?
19 Do we really have enough standards? Do we really
20 have enough ontologies? These are things that I
21 think are important to consider. Thanks.

22 DR. KAROL: We have remarkable agreement
23 that we really should link molecular expression and
24 toxicology and pathology, and that we shouldn't be
25 too restrictive. But I would like to hear a little

1 bit more discussion about this database and what
2 you think should be involved in creating an
3 effective database. Frank, do you have comments?

4 DR. SISTARE: I was just going to say one
5 thing. I don't know if this is one of the things
6 that Tim was forgetting with respect to what Jay
7 had mentioned, but Jay mentioned something along
8 the lines of we ought to model data submissions to
9 the FDA along the lines of the way a paper would be
10 put together and submitted for publication. But I
11 think as John Quackenbush pointed out, those
12 journals are requiring the full gamut of gene
13 expression data derived from those experiments to
14 be submitted into a database. So, that is routine
15 now. Those journals are not publishing data
16 without people having documented that they have
17 submitted the full gamut of gene expression data
18 into a database.

19 So, it seems like that is becoming the
20 standard, the societal standard, if you will, for
21 supporting the conclusions of a well constructed
22 microarray gene expression experiment, that is,
23 full disclosure of the data that support the
24 conclusions of the paper for the inquisitive
25 scientists who look and evaluate on their own.

1 So, your question, Meryl, I think is spot
2 on and that was one of the first questions. You
3 know, format defines utility of everything, or the
4 shape of something is defined in utility of
5 something. If we ask for paper submission, it is
6 only going to be useful for that particular context
7 which the paper is being submitted to support.
8 That is all it is going to be useful for. If the
9 data is submitted electronically it now expands the
10 utility of that information.

11 So, I think that is the first fundamental
12 question we have to establish. FDA is moving
13 toward electronic data submission. It just happens
14 to coincide with the fact that now we are getting
15 10,000 data points on an experiment and the only
16 way you can really make sense of that is if it is
17 submitted electronically. You know, we are
18 establishing the first, fundamental question, which
19 should FDA ask for the data to be submitted
20 electronically? The first question is, is that a
21 reasonable request? Once we have established the
22 answer to that question, if the answer is no, okay,
23 we can go home but if the answer is yes--maybe we
24 should just ask that question first.

25 DR. ZACHAREWSKI: Just to follow-up, you

1 said that you are going towards electronic
2 submission. That means that minus the microarray
3 data you already have a database established to
4 capture all that information. Is that correct?

5 DR. LEIGHTON: We have to be careful here
6 in distinguishing between electronic submission of
7 paper data versus submission of electronic data. I
8 think the way we would be moving would be
9 submission of electronic data so that it is truly
10 searchable and can be searched across submissions.

11 DR. ZACHAREWSKI: But would you store that
12 within a database housed within FDA?

13 DR. LEIGHTON: I think ultimately, because
14 of the proprietary nature of the data, we would
15 have to do that. I doubt that it would be public.

16 DR. ZACHAREWSKI: So, that is the plan, to
17 develop a database to store that data only for FDA
18 use, period?

19 DR. SISTARE: Well, I think the initial
20 plan is to enable submission of electronic data in
21 a way that it is very easy for the reviewer to move
22 around that data and to pull things together and
23 pull it into programs to analyze the data
24 electronically. So, that is really the visible
25 rationale for doing it. By the way, once you do

1 that, now you can create a database and I think it
2 would be unwise not to. I am going to ask Randy to
3 address the question. I think you are asking sort
4 of the status of things right now. There is not a
5 lot of electronic data being submitted to my
6 knowledge.

7 DR. ZACHAREWSKI: Yes, there are two
8 questions, the status and will the system that you
9 have allow you to query across submissions?

10 DR. LEVIN: We are working on the tools
11 that will help us analyze that but we have found
12 that we are going to have to put that into a
13 database for those tools to work efficiently. So,
14 we are aiming toward a database that we put the
15 data into. If we develop a common terminology,
16 then we can potentially look across studies.

17 DR. ZACHAREWSKI: You mean like the
18 MIAME-Tox ones?

19 DR. LEVIN: Well, for example yes. The
20 thing that we are focusing on first is the
21 structure of the model, so not the terminology. We
22 need both to be able to look across studies.

23 DR. ZACHAREWSKI: The only other thing I
24 can say is that it sounds great but it won't happen
25 in my lifetime. So, when is this actually going to

1 be in place? That is the other thing. I think
2 that is going to be another major impediment
3 because these are not small undertakings and I am
4 sure you appreciate that.

5 DR. LEVIN: Well, we have gone pretty far
6 with the clinical data to define how we can
7 transport the information that we need for making
8 our regulatory decisions. We have a pilot project
9 for both the clinical and nonclinical data so we
10 are hoping that we start to receive some of this
11 data in from our pilot this year and to test the
12 model and see how good it is.

13 DR. ZACHAREWSKI: So, that means that you
14 could take this model and just add on to it a
15 subsystem for microarrays. Is that the plan?

16 DR. SISTARE: Yes, and I think what
17 Lilliam described is right now--we have a document
18 out there that says here is a way that you can
19 submit electronic data if you want to, right now.
20 I think the status is that we just haven't received
21 that many electronic data submissions but it has
22 been an option for sponsors to do at this point in
23 time. We are not making them, we are not requiring
24 them to but, again, allowing and enabling. So, now
25 within the context and the boundaries of what we

1 have established, if a sponsor chose to adhere to
2 the MIAME-Tox guidelines that are out there they
3 would be compatible. There are just a couple of
4 small things where we may have to wrinkle out some
5 things but otherwise they are compatible.

6 MIAME-Tox is more prescriptive, if you will.

7 DR. LEVIN: Actually, we have had some
8 success with carcinogenicity data and we have been
9 receiving that electronically for a long time.
10 More recently people have been following the
11 standard that was published in the 1999 guidance so
12 that has been pretty successful.

13 DR. GOODMAN: I think in terms of doing
14 things electronically it really is sort of a
15 no-brainer these days. We should move towards
16 doing more and more, if not everything,
17 electronically. When I said to submit like a
18 manuscript, obviously there would be appendices
19 that would include full data.

20 My concern, again, is that at the status
21 that I see toxicogenomics today I think to start
22 putting in place a proscribed database would be
23 less productive than over the next few years
24 letting the applicants submit their data in a file
25 form and then take and see what might be the best

1 of these, rather than start--once you start putting
2 something into guidelines--I hear you in terms of
3 that it can be flexible and it can be changed, but
4 it gets much more difficult. It gets difficult to
5 start changing once you have guidelines.

6 I just wonder out loud whether the notion
7 of comparing and sifting and sorting of these
8 database publicly is really something that is
9 realistic. It is my impression that you would be
10 dealing basically with proprietary data and that
11 this would not be that readily available. Maybe
12 there is a certain time span when it does become
13 available. But the point is that in order to
14 really move this field forward it is going to take,
15 I think, industry buying into it, and in order to
16 do that it has to be where you see that it is going
17 to be productive in terms of help, not only help
18 make better decisions but help in terms of working
19 with Food and Drug Administration. So, again, I
20 just think early on the less prescriptive and the
21 more working as partners, I think the more
22 productive everything will be.

23 DR. ZACHAREWSKI: No, for that session.
24 The problem is that if you don't set up some
25 guidelines, when you do finally set up guidelines

1 you will lose that information because it will be
2 very difficult, if everybody submitted their data
3 in a different format, to then reformat, you know,
4 what you have collected for the last five years and
5 put it into the proper format to put into the
6 database. If you only have two formats being
7 submitted it is not so much of an issue. If you
8 have 15 or 20 or more, whose responsibility is it
9 to reformat that so that it is acceptable into the
10 database?

11 DR. GOODMAN: Do you have a crystal ball
12 at this time to start setting up these databases?
13 Why not just see how the information flows for a
14 while and then try to revisit this issue?

15 BROOKS: Maybe the definition of
16 guidelines is where we are getting hung up with
17 respect to the kind of data to be submitted. Maybe
18 if we start with more simple things as formatted
19 data, as someone said CEL files or raw data versus
20 processed data. Raw data gives you the ability, as
21 new analytical tools for what you want to do across
22 databases come out, the flexibility to do that
23 without restricting you to guidelines with respect
24 to other ancillary information that goes along with
25 it so you use maybe MIAME-Tox as a standard and say

1 we are going to take raw data. After you start
2 taking that data and working with it, then you can
3 refine or establish specific guidelines about
4 information that is more pertinent to what you are
5 trying to do. But I think the form of data is
6 probably the most critical right now.

7 DR. SISTARE: Yes, I would add one of the
8 things that Randy pointed out to me and I should
9 have mentioned earlier too, and that is what is
10 important here I think is to specify the transport
11 file, as you point out, the format that you want
12 the data to come in. Then, you can modify that and
13 change that any way to fit a database.

14 The one place where it does get a little
15 dicey is when you start specifying ontology, words
16 and vocabulary and things like that. If you do
17 that up front, that may be difficult and you may
18 lose some aspect of the flexibility of the use of
19 that information if you don't do that up front but
20 I hear what you are saying, if you are a little too
21 prescriptive and the Society of Toxicological
22 Pathology hasn't quite developed a consensus on the
23 best definitions of the terms.

24 FDA can maybe proceed judiciously and
25 carefully along that line but are we getting the

1 general gist that this is a wise endeavor for us to
2 go down; this is a path we should be going down in
3 terms of setting up and preparing ourselves in a
4 way to receive the data, that it could be useful
5 and populate a database without being prescriptive?

6 DR. GOODMAN: I think the answer is yes.

7 DR. KAROL: Randy, did you want to say
8 something?

9 DR. LEVIN: Yes, I think Frank was saying
10 that from our experience and with the clinical
11 data--many things that you were just bringing
12 up--we can define the transport, just the
13 information how to communicate with each other.
14 Our database may change over time but we are hoping
15 that the transport information would stay the same
16 so you would have that stable.

17 Another piece that might be interesting is
18 the annotated ECG waveform data. We were talking
19 about receiving that in an electronic format. At
20 first we might not have the full database but we
21 would have the standard of how to receive the data.
22 Then eventually, once we got everything worked out,
23 we could have it put into a database. We could
24 take that data we received in the past and put it
25 into a database because it is all standardized.

1 Then, the other thing is that once we have
2 the database there is a possibility to look at some
3 of that data for research issues beyond just a
4 review of that particular application. So, looking
5 at it and saying is there some way we can monitor
6 drugs for cardiac toxicity because we look at this
7 ECG data. So, it does offer something beyond the
8 initial use, and something you would consider for
9 your work here too.

10 DR. HARDISTY: I agree. I think it is a
11 good time to probably start a database and it
12 should have some minimal standards. I think that
13 is what you have recommended. If someone wants to
14 go beyond that, so be it. So, it is not really
15 restrictive or prescriptive but there is some
16 minimum that you want everybody to conform to.

17 The other thing about restrictive
18 nomenclature I think is probably a good thing and
19 not a bad thing, particularly in histopathology or
20 any of the toxicology endpoints. We have been
21 doing toxicology studies for years and we are
22 trying to take the information we get from
23 toxicology studies today and correlate it with gene
24 expression. So, the things that we are seeing in
25 the tissues aren't going to change. We are trying

1 to correlate those changes with gene expression.
2 So, we should be able to go ahead and restrict the
3 terminology based on what we already know. What we
4 are trying to do is eliminate synonyms in our
5 database so that you can search it without having
6 to worry whether the study was done in England or
7 whether it was done here, in the United States.
8 So, I think that we already have the information
9 there. It is just a matter of setting it down and
10 deciding what you want in your database and how you
11 want to handle it.

12 DR. BROOKS: One thing that was mentioned
13 in the first talk with respect to the goals--one is
14 to, obviously, find more sensitive or different
15 ways of assessing toxicological assessment. The
16 other is being able to make predictions based on
17 the efficacy of drugs and their toxic events on
18 specific individuals. So, I just wanted to note
19 that without collecting data from individuals or
20 studies that are specific in having that full
21 dataset it is going to be virtually impossible to
22 achieve that second goal. So, having a database is
23 going to help you make greater strides with
24 individual sponsors or academic labs that are
25 trying to achieve that information. It is a much,

1 much larger endeavor that needs to be at the level
2 of the federal government I think.

3 DR. WATERS: I would just like to comment
4 that I think the FDA can play a very important role
5 in consensus building with regard to some of the
6 data standards. I am not sure that you have been
7 involved extensively up to this point. I think it
8 would be very good if you were engaged in that
9 activity. The international standard setting
10 effort for databases is very important and, as
11 well, the ontology building efforts that a number
12 of the societies are becoming engaged in. So, I
13 think to become engaged actively in those processes
14 and work towards the evolution of also publicly
15 available data so that there could be a consensus
16 in understanding the way in which one would
17 interpret those datasets would be to your advantage
18 because everybody really needs to get on the same
19 page. Everybody really needs to have a common
20 understanding of molecular expression datasets, not
21 only the regulated community and the regulators but
22 also the other academic members of the scientific
23 community, as well as other governmental agencies.

24 So, I think as well inter-agency efforts
25 would be laudable at this point and there should be

1 an effort to extend to other parts of the federal
2 government. So, for example, the National Cancer
3 Institute is also developing large databases and it
4 is also interested in the clinical domain. I think
5 there would be natural synergy to work with them in
6 their database efforts. Similarly, NIEHS is very
7 interested in animal toxicology and is engaged
8 directly in developing a public database in that
9 domain.

10 The other aspect that I think is important
11 is an international one. I think we don't live in
12 isolation anymore in the U.S. We are definitely a
13 part of the international community and we also
14 have to engage in the international sector with
15 regard to development of standards.

16 DR. HARDISTY: One of the questions was
17 what major obstacles would you expect down the
18 road. Most of the work that has been done with
19 gene expression and genomics has been done in
20 universities or non-GLP type settings. Not that
21 they are not good studies, but it is a different
22 type of environment than in the regulatory GLP
23 laboratory and validation of the systems that you
24 are using and all those types of things are
25 something that the manufacturers and some of the

1 people who are doing this work need to start
2 thinking about now, rather than later. If these do
3 become regulatory requirements, then they are going
4 to have to work in the GLP environment. Right now,
5 toxicology may be outpacing the science in that
6 area so it is hard to keep--you don't want to not
7 continue the technologic development but imposing
8 GLP requirements on those people at this point.
9 But if these are going to be used in a regulatory
10 setting, then you are going to have to try and
11 limit those areas.

12 DR. BROOKS: I think one of the other
13 hurdles you might need to be prepared to overcome,
14 with respect to any time you put guidelines in
15 place, is that you are going to get questions about
16 those guidelines and ask for recommendations with
17 respect to how people are going to do things. So,
18 there was a lot of talk about biological
19 replicates, and experimental design and study
20 design. Everybody does things a little bit
21 differently. I think it has gotten a whole lot
22 better over the years with this, but I think that
23 you need to be able to be prepared, given the model
24 and once your question is defined, to be able to
25 answer questions with respect to suggestions. If

1 we want to generate this data or we want to submit
2 it, you know, what is going to be better, more
3 replicates, less replicates, with respect to our
4 design as these experiments and studies are being
5 built. If you have the guidelines and can't
6 provide some suggestions or information I think
7 that people will be less reluctant to provide that
8 kind of data, fearing that they might miss the
9 mark.

10 DR. JACOBSON-KRAM: I think it is kind of
11 interesting that the dichotomy that is developing
12 here is the way that we are going to deal with this
13 kind of data versus traditional. For example,
14 somebody submits the results of a carcinogenicity
15 study; you don't ask for the slides. You pretty
16 much believe what the report says and if you are
17 very unhappy with it you can go back and audit it.
18 Here what you are asking for is essentially the
19 equivalent of the slides so that you can reexamine
20 it and perhaps re-interpret it. That is really a
21 change in paradigm for how we have done toxicology
22 in the past.

23 I think that could also be part of some of
24 the needs in the pharmaceutical industry because
25 basically you say here is carte blanche; go ahead.

1 Here is how we interpret it; what do you think?

2 DR. SISTARE: I think part of what appears
3 to be a dichotomy there--I think Kurt Jarnigan
4 expressed it well when he talked about the
5 youthfulness of the technology, the youthfulness of
6 using RNA transcript measures as endpoints to link
7 definitively to outcome, as opposed to the maturity
8 of the two-year bioassay and not asking for slides.

9 We are striking a compromise and what
10 William proposed is we want a suggestion, a
11 consideration for discussion and for some input in
12 terms of what our thinking here is, not to actually
13 ask for the 40 MB TIF image files. That would, I
14 think, be asking for the histopath slides. So, we
15 are asking for something in between, not just the
16 process report but, you know, the data--the data.
17 I think, again, we are asking for the raw output
18 data. Even that is not completely raw because some
19 algorithm has to be applied to get a signal out of
20 background and, you know, you are allowing the
21 experimenter to do that and not questioning that in
22 a sense when you go to the CEL file, intermediate
23 file. So, you are actually asking for number
24 output.

25 It is a fair question and it is something

1 that we have wrestled with and had dialogue on,
2 that is, how far back do you go, and I would like
3 to get some feedback and some dialogue here from
4 the experts who have wrestled with these datasets
5 and know the state of the technology. Should we
6 ask for a polished, final expression ratio report,
7 or should you ask for something like a CEL file?

8 DR. HARDISTY: I don't see it as a whole
9 lot different than what you get on a
10 carcinogenicity study. You don't get the glass
11 slides but you get the individual data and every
12 data point in that dataset. If you get it in a CEL
13 file and you evaluate and your interpretation is
14 different than the sponsor's, they are going to get
15 a letter from you--

16 [Laughter]

17 --so I would see it the same way. You are
18 not asking for the microarray, it is the data that
19 they are submitting so you are not going to repeat
20 the generation of the data, which is what you would
21 do if you had the glass slides. You are repeating
22 the analysis of the data, or could repeat the
23 analysis of the data, which you can do with routine
24 toxicology data today.

25 DR. BROOKS: I think a lot of it stems

1 from the interpretation of these datasets and I
2 don't think that the problem is going to be with
3 any given sponsor, that you are going to
4 necessarily disagree with their interpretation but
5 when you look at compounds or things within the
6 same class across sponsors how do you interpret
7 each of their individual interpretations if they
8 are all using different platforms, or even if they
9 are using the same platform, even though they have
10 given you their MIAME-Tox standards tell you that
11 their labeling samples quite differently?

12 So, I think by having intermediate with
13 the absolute raw data to some unprocessed data
14 allows you then the flexibility to potentially
15 compare across platforms and, more importantly,
16 compare applications as to whether or not there is
17 a consistency for those compounds or those
18 submissions. I think in the case of Affymetrix,
19 the CEL file is a good compromise because it leaves
20 you open for different kind analyses you can do to
21 explore the interpretation, I mean within the
22 context of what they are trying to say. If you had
23 some kind of a measure, as William said, that would
24 tell you if there was a defect with respect to
25 image file, and the same can be true for

1 slide-based arrays where there is a standard
2 background subtraction, and I think most people
3 won't necessarily argue with respect to array
4 performance and then, instead of getting ratios,
5 getting the signal data along with those would be
6 equivalent to a CEL file.

7 DR. LEIGHTON: I had a question that goes
8 to the question that is on the board here. For the
9 FDA to specify a transparent data processing
10 protocol and the single statistical analysis
11 method, would this be viewed as moving the field
12 forward or being too prescriptive? Or, should this
13 really be deferred until the issues of standard
14 development are more evolved?

15 DR. GOODMAN: I think it is too
16 prescriptive. Frankly, I think we have problems in
17 terms of making sometimes too many mistakes in
18 toxicology and we don't want to bring on a new
19 technology and make more mistakes quicker. It is
20 not ready to jump in now in terms of prescribing an
21 approach.

22 DR. ZACHAREWSKI: I would like to agree
23 with my esteemed colleague--

24 [Laughter]

25 --if that is worth anything. But I think

1 this is one of the issues in terms of what data do
2 you get. So, I would say that if you were to try
3 and prescribe a specific data analysis, which one
4 are you going to choose? And, if you asked
5 everybody in this room, they would probably give
6 you at least two opinions. So, there is no
7 prescribed method at this point in time. However,
8 let's say five years from now when there is, you
9 are going to have to go back to each one of those
10 pharmaceutical companies on bent knee potentially
11 and ask them for their raw data files to be able to
12 reanalyze all that information and repopulate your
13 database using a standard normalization or
14 quantitation type protocol.

15 DR. BROOKS: That is if you don't collect
16 the raw data now. That is what you are saying.

17 DR. ZACHAREWSKI: Right. But if you do
18 that now you could go back and do that yourselves
19 with respect to the interpretation, not to go back
20 and, like I said before, penalize what has happened
21 in the past but move in a better direction for the
22 future. So, I agree that right now is absolutely
23 not the right choice. Actually, if you guys have a
24 transparent statistical analysis method, I would
25 like actually to take that back with me on the

1 plane but I don't think that exists at this point
2 in time.

3 DR. SISTARE: We could name one but you
4 might not like it. I mean, the rationale behind
5 this question is this whole concern about FDA
6 taking a dataset, analogous to what Jerry brought
7 up--and say here is how we are going to analyze the
8 data when we get it; this is what we are going to
9 do with it; these are the rules we follow. I think
10 there is a lot of anxiety if data is submitted to
11 us by sponsors. They may feel that this is the
12 best way to analyze the data. If we don't agree
13 with their approach and we analyze it another way,
14 you know, will the conclusions be markedly
15 different? Probably not, but it is an attempt for
16 FDA to try to be somewhat transparent and to say,
17 you know, at this point in time this is how we are
18 going to look at the data when you give it to us so
19 you might want to look at it that way first too.
20 You can use whatever other way you want, what you
21 think is best, but you might want to do this
22 because this is what we might do. But if you are
23 what you are suggesting is there is just no way we
24 could do that--with Affymetrix we could say, you
25 know, use 5.0 and we are going to use this

1 approach.

2 DR. ZACHAREWSKI: What I would do then is
3 I would encourage Dr. Rosario, when she is working
4 with Schering Plough, for them to analyze their
5 data two different ways at least.

6 The other thing that I would really do is
7 I would encourage for you to approach other
8 pharmaceutical companies and see whether they would
9 do it, and see how they would do it differently. I
10 don't know whether they would go and talk to
11 Schering Plough or not and just copy what they are
12 doing, but I would think that the idea of getting
13 different perspectives from different
14 pharmaceutical companies--you know, you could then
15 merge and pick what you like and ask them to
16 resubmit what you didn't like.

17 DR. WATERS: I think actually Tim brought
18 out a major point, and if you look at the LCF I
19 think it bears is, that is that in the effort that
20 was undertaken involving 30 different
21 pharmaceutical companies so much was learned by
22 looking at divergent opinions. I think at this
23 point in time we would all be well advised to look
24 at divergent opinions. We just don't know enough
25 and I think that we have an opportunity here to do

1 it right and, if we do it right then this
2 technology will become established and we will be
3 able to use it and we will have all we want out of
4 the effort. But I think if we push it too far too
5 fast, then it really may backfire on us.

6 DR. BROOKS: I think that your sponsors
7 now that would risk--risk is a bad word but that
8 would go ahead and submit data of this nature are
9 sort of at an advantage because I think that you
10 are going to gauge some of your interpretation in
11 the analysis based on these submissions and how
12 effective they are and how well they work, whereas
13 if they wait until guidelines are established they
14 might be changing things in a big way. So, I think
15 that by submitting data it has to be clear that you
16 are not going to necessarily change now the
17 interpretation of the data based on your learning
18 curve or based on how it might be used to establish
19 other kinds of tools. You know, the earlier you
20 get in and can justify your interpretation and your
21 model with your data, it might actually become a
22 better established guideline.

23 DR. ZACHAREWSKI: Actually, I have another
24 suggestion. Why don't you ask the PhRMA companies
25 how they want to submit the data?

1 DR. SISTARE: We actually have. We have
2 had at least one sponsor come to us and say we have
3 some data we want to submit; how do you want it? I
4 put the mirror up and I said challenge us. You
5 submit the data to us in a format that you think is
6 the best, the most advisable, productive format,
7 but I did share one word with them, an adverb
8 actually. I said electronically. I did say that
9 but I said in whatever format you choose and, you
10 know, tell us how you would like to submit the data
11 and maybe we can get some dialogue on that and give
12 you some feedback. But we haven't seen it yet.

13 DR. ZACHAREWSKI: But this might be
14 something that ILSI-HESI might want to pick up. I
15 mean, the organization and the structure is there
16 for them to do that since they meet regularly
17 anyway.

18 DR. KAROL: Frank, I think we have
19 addressed all of the questions.

20 DR. SISTARE: I think the feedback we have
21 gotten has been really excellent. I really want to
22 thank all of the speakers and all of the committee
23 participants today. This has really helped us and
24 this is a landmark meeting for all of us. As Helen
25 pointed out, this is the first time we have

1 assembled this subcommittee. I want to thank Meryl
2 for chairing this beautifully, for getting us back
3 on time and for allowing for full discussion of the
4 issues. Again, I think we got all the issues out
5 there that needed to be. We missed Roger; there
6 was a void there. There was one gap there in some
7 of the practical applications of some real live
8 scenarios that we were hoping to get. But,
9 otherwise, I think we got everything on the table.
10 We have achieved our goal of being as transparent
11 as we can. Now the ball is in our court, and we
12 will try to get back to the committee members
13 something in writing within the next six to eight
14 months that captures some of the feedback we have
15 gotten today and allows FDA to move forward.

16 DR. KAROL: I also want to thank the
17 committee for a very wonderful discussion and just
18 a very exciting topic. I am really looking forward
19 to seeing just how this new technology can be used
20 in an effective regulatory role. So, I thank
21 everybody for their participation, the agency and
22 Kimberly as well. The meeting is officially
23 adjourned.

24 [Whereupon, at 3:25 p.m., the proceedings
25 were adjourned.]