

AT

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

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## P R O C E E D I N G S

**Call to Order**

DR. HAMMER: Good morning. Welcome, everybody, to what portends to be an interesting two-day discussion by Antiviral Drugs Advisory Committee on using RNA as a primary endpoint in HIV trials.

I would like to start the meeting by having the people at the table introduce themselves, and I will begin on the left with Dr. Feigal.

DR. FEIGAL: Good morning. I am David Feigal, FDA.

DR. FREEMAN: Donna Freeman, Acting Division Director, Antiviral Drugs.

DR. FLYER: Paul Flyer, FDA.

DR. ELASHOFF: Michael Elashoff, FDA.

DR. MURRAY: Jeff Murray, FDA.

DR. IACONO-CONNORS: Lauren Iacono-Connors, FDA.

DR. VALENTINE: Fred Valentine, NYU, Bellevue Hospital.

DR. DIAZ: Pamela Diaz, Chicago Department of Public Health.

DR. MATHEWS: Chris Mathews, U.C., San Diego.

DR. HAMMER: Scott Hammer from the Beth Israel Deaconess Medical Center and Harvard Medical School in

1 Boston.

2 MS. MCGOODWIN: Ermona McGoodwin, FDA.

3 DR. LIPSKY: Jim Lipsky, Mayo Clinic.

4 DR. EL-SADR: Wafaa El-Sadr, Harlem Hospital and  
5 Columbia University, New York.

6 DR. CHINCHILLI: Vernon Chinchilli, Penn State,  
7 Hershey Medical Center.

8 DR. VERTER: Joel Verter, George Washington  
9 University.

10 DR. MODLIN: John Modlin, Dartmouth Medical  
11 School.

12 MS. LEIN: Brenda Lein, Project Inform.

13 DR. HAMMER: Thank you. I would like to turn now  
14 to Ermona McGoodwin who will read the conflict of interest  
15 statement.

16 **Conflict of Interest Statement**

17 MS. MCGOODWIN: Thank you, Dr. Hammer. The  
18 following announcement addresses the issue of conflict of  
19 interest with regard to this meeting, and is made part of  
20 the record to preclude even the appearance of such at this  
21 meeting.

22 In accordance with 18 USC 208, general matters  
23 waivers have been granted to all Committee participants who  
24 have interests in companies or organizations which could be

1 affected by the Committee's discussion of plasma HIV-RNA  
2 measurement as an endpoint in clinical trials for drugs to  
3 treat HIV infection. A copy of these waiver statements may  
4 be obtained by sending a written request to the Agency's  
5 Freedom of Information Office, Room 12A-30, the Parklawn  
6 Building.

7 In the event that the discussions involve any  
8 other products of firms not already on the agenda for which  
9 an FDA participant has a financial interest, the  
10 participants are aware of the need to exclude themselves  
11 from such involvement and their exclusion will be noted for  
12 the record.

13 With respect to all other participants, we ask in  
14 the interest of fairness that they address any current or  
15 previous financial involvement with any firm whose products  
16 they may wish to comment on.

17 DR. HAMMER: Thank you. I would like to turn now  
18 to Dr. Feigal, who will introduce today's session.

19 **Introductory Comments, David Feigal, M.D. M.P.H.**

20 DR. FEIGAL: Good morning. In 1991 this Committee  
21 met to consider an application by Bristol-Myers Squibb to  
22 approve didanosine. At that time they had evidence which  
23 consisted of a control trial, showing an average of about a  
24 ten-cell increase in the CD4 count in patients with very low



1 CD4's.

2           It was wondered at that time whether or not that  
3 would be the basis to approve this drug. It had been about  
4 three and a half years since the approval of zidovudine. At  
5 the high doses that zidovudine was usually taken, the  
6 average duration that people could take zidovudine was less  
7 than a year. So there were many patients with HIV infection  
8 who really had no therapeutic options.

9           The question really was sort of what did ten cells  
10 mean, and was that an adequate basis to approve a drug  
11 product? The regulation for accelerated approval had not yet  
12 been written, although ideas of how to do such a regulation  
13 for conditional type of approval had been discussed by  
14 Commissioner David Kessler.

15           ddI went on to be approved based on those small  
16 changes but always with the understanding that the surrogate  
17 markers that the CD4 count and now later viral load with  
18 regards to the use of identifying promising drugs that would  
19 be followed up with clinical trials would show what the real  
20 benefit was.

21           One of the challenges for the Division right from  
22 the start was trying to decide what to do with the surrogate  
23 marker in the labels. What was it fair to tell people what  
24 really made sense, what really was the basis for clinical

1 information. Certainly, with the early drugs we didn't  
2 feel, and I don't think clinicians felt that individualized  
3 therapy based on small, transient changes in CD4 counts was  
4 a way of telling whether a drug was active or not.

5           As the therapies improved and as we moved into  
6 patient combination regimens, we could demonstrate that  
7 groups on average would have higher counts with new agents  
8 added to a new regimen. But it still really was not much  
9 basis for individualizing therapy, and although we described  
10 these CD4 results in the study section of the labeling and  
11 that information was available through the promotional  
12 literature and through the educational materials the  
13 companies had, it still was not clinically very satisfying.

14           Where we are now, however, is that we appear to  
15 have a measure of disease activity that is very sensitive in  
16 real time, is available commercially and has become a goal  
17 of therapy per se. Treatment panels have met to make  
18 recommendations about the optimal way to use the currently  
19 approved drugs, and have made recommendations about how to  
20 follow the load, and how to assess when someone has had good  
21 response and when a response is lost.

22           We began planning this meeting probably over a  
23 year ago as we began looking for trials to help us find a  
24 way to bring the data about viral load more systematically

1 into product labeling. The goal isn't simply to describe  
2 the studies that have led to an accelerated approval as to  
3 give the clinicians a sense of what the evidence is that the  
4 drug is active. The goal is really to describe the number  
5 of important features about the way that a drug performs and  
6 the way that individuals respond.

7           If you look over time in terms of how the  
8 therapies have been introduced for initial use when there  
9 was very little data available, we started with ddI with the  
10 rationale was, well, it appears to be an active drug and it  
11 should be used in patients with few alternatives, to a time  
12 when we realized that there was probably more promise with  
13 some of the new combinations than the old regimens where the  
14 trials had gone on long enough so that there was pretty  
15 uniform eventual failure, to now when there is a real need  
16 to be able to individualize therapy and assess response.

17           So part of the purpose of this meeting is to  
18 really take a look at how drugs affect viral load, and how  
19 should we describe those effects in the product labeling as  
20 a goal of therapy per se. This is not the same thing as a  
21 question of saying are we done with clinical endpoints  
22 because we still need to study these drugs in patients with  
23 clinically active disease. We need to understand the  
24 clinical toxicity and any adverse effects that offset the

1 clinical benefits. But we also feel we need to more  
2 systematically approach the way that we study the effects of  
3 the virus when the virus meets up with combinations of  
4 antiviral drugs.

5           When we began planning this meeting, we looked to  
6 commercial sponsors, we looked to the NIH groups of studies  
7 through the ACTG, CPCRA and other cooperative groups for  
8 studies that had information that could focus in on some of  
9 these questions; could tell us some information about how  
10 long should you wait to see response to a drug. How do you  
11 define an adequate suppression? How do you detect loss of  
12 response? And what is an appropriate evaluation or reasons  
13 for that loss of response?

14           This included studies that had both the measures  
15 of virology and immunology, and many of the studies also had  
16 the luxury of having measures of disease progression at the  
17 same time.

18           I think we have gotten past the simple question of  
19 is viral load validated that. I think we are looking at  
20 trying to define the metrics by which we think it will be  
21 useful to describe how these drugs work. We have looked  
22 forward to this meeting and, in particular, have appreciated  
23 the willingness of the study investigators to often break  
24 their studies apart and just show us one small focused part

1 of it to ask a question.

2 We met with the sponsors and we have asked them to  
3 try and follow a relatively uniform format in presenting the  
4 data so that it will be relatively easy to jump from study  
5 to study, but there will be times too when there are  
6 interesting other ways of analyzing these data.

7 The day will begin actually with some  
8 presentations by the FDA on some of our perspectives on  
9 these issues and at this point let me introduce the first  
10 speaker from the FDA, Dr. Lauren Iacono-Connors, who will  
11 talk a little bit about the properties of the viral load  
12 tests.

13 **Overview of HIV-RNA Measurements, L. Iacono-Connors, Ph.D.**

14 (Slide)

15 DR. IACONO-CONNORS: Good morning. I am Lauren  
16 Iacono-Connors, Division of Antiviral Drug Products. This  
17 morning we are going to hear three presentations on the  
18 subject of HIV-RNA quantitative assays. The purpose of this  
19 presentation is to generally review what these assays are in  
20 terms of their unique methodologies, what they actually  
21 measure and, most importantly, what we should keep in mind  
22 when reviewing data generated by these types of methods.

23 First I will present a very general overview of  
24 some of the methods currently used to estimate HIV-RNA in

1 clinical specimens. I will be followed by Dr. Don  
2 Brambilla, New England Research Institute, who will discuss  
3 data describing certain assay characteristics. In  
4 particular, he will focus on data which describes certain  
5 aspects of assay variability. Then Dr. Winston Cavert, from  
6 the University of Minnesota, will present data on tissue-  
7 specific HIV replication dynamics.

8           The goals of my presentation are to, one, review  
9 the general methodologies under development; two, define  
10 certain validation parameters which are essential when  
11 attempting to interpret HIV-RNA data sets; and, three,  
12 briefly discuss the target material for these types of  
13 assays, namely, HIV-RNA typically measured in plasma  
14 specimens.

15           Before I begin I would like to acknowledge and  
16 thank Dr. Indura Hewlett, from the Division of Transfusion  
17 Submitted Diseases from the Center of Biologics, for her  
18 contributions to this presentation.

19           (Slide)

20           This slide shows my outline. First I will discuss  
21 HIV nucleic acid quantification methods, then method  
22 validation parameters. I will follow this by a discussion  
23 briefly of HIV reservoirs, and then I will mention caveats  
24 to both the method and the target HIV specimen, and then I

1 will summarize.

2 (Slide)

3 Subsections of full-length HIV RNA, the virus  
4 genome, are the target of the five methods listed on this  
5 slide. All five methods require well-preserved HIV material  
6 in the clinical specimen. For most of the data which will  
7 be discussed today that specimen is plasma.

8 HIV particles in plasma have a wide range of  
9 potential concentration, upwards of seven logs. Several  
10 techniques have been developed which can systematically,  
11 directly through the RNA target or indirectly through the  
12 probe, amplify the HIV-RNA in a given specimen. As a  
13 result, the positive detection signal on the amplified  
14 specimen falls into a semi-quantitative range which allows  
15 for an estimate of the RNA copy number.

16 (Slide)

17 A well-documented direct nucleic acid  
18 amplification method is the polymerase chain reaction,  
19 commonly known as PCR. The majority of HIV-RNA clinical  
20 data which will be presented today and tomorrow were  
21 generated using this type of technology.

22 Due the inherent nature of this powerful molecular  
23 tool, the capacity for detection sensitivity can be most  
24 optimized. However, this technology also has the potential

1 for a broader range of values when attempting to quantify a  
2 single specimen. A greater degree of variability may be  
3 expected. Therefore, PCR-based assays may have may have  
4 greater sensitivity when attempting to detect small copy  
5 numbers but trade that feature for greater variability in  
6 all measurements.

7 Other methods which amplify the virus target are  
8 nucleic acid sequence-based amplification and strand-  
9 displacement amplification.

10 (Slide)

11 Techniques which are designed to indirectly  
12 amplify the HIV-RNA target material include branch DNA  
13 signal amplification, bDNA for short. Some of the data  
14 which will be presented here employed this technology.  
15 Since this method amplifies the probe which is directed at  
16 the target RNA, instead of amplifying the target RNA itself  
17 this technique appears to produce tighter results. Thus,  
18 the technique has lower variability characteristics,  
19 however, at the current state of development this type of  
20 technology appears to be less sensitive when measuring low  
21 copy count number material.

22 Another method which amplifies the probe for HIV-  
23 RNA is the ligase chain reaction. In order to interpret  
24 HIV-RNA copy numbers generated by a quantitative technique,



1 certain characteristics of the assay have to be defined.  
2 Essentially, the assay's capabilities and the limitations of  
3 those capabilities must be described for each specific  
4 assay. This is usually achieved through rigorous analysis  
5 of numerous analytical and clinical specimens. These assay  
6 characteristics are referred to as the validation parameters  
7 and, once determined for a given assay, they are unique and  
8 specific to that assay.

9 (Slide)

10 This slide lists the four basic areas of assay  
11 validation. Variability refers to how much wobble is  
12 detected when quantifying a single specimen. Two major  
13 contributors to assay variability are the assay itself due  
14 to its inherent design, its controls, standards, specimen  
15 handling and operator error, to name a few, and the  
16 biological variability associated with the specimen source.

17 Sensitivity refers to the lower limit of  
18 reasonable quantification of RNA copy number. Specificity  
19 simply requires that the assay detect the target material  
20 but not other biological components which may be present in  
21 plasma. Finally, the assay's linearity. This is described  
22 by graphing the expected and observed values in a set of  
23 control specimens. As a result, the upper and lower limits  
24 of the full range of linear detection are defined.

1 (Slide)

2 Now that we have generally identified methods for  
3 measuring HIV-RNA and some of their characteristics, the  
4 next question is where are we sampling for HIV. There are a  
5 number of well-documented tissue reservoirs for HIV. The  
6 virus can be found in both intracellular and cell-free  
7 compartments of the body. Cellular reservoirs include but  
8 are not limited to cells found in hematopoietic, central  
9 nervous system, skin and bowel tissues. Cellular  
10 subcomponents of these tissues, for instance lymph nodes,  
11 harbor actively replicating HIV, while other subcomponents  
12 harbor inactive or dormant HIV material.

13 Newly replicated and infectious HIV virion are  
14 shed from cells and either immediately infect an adjacent  
15 cell or move into a fluid component of the body, such as  
16 interstitial fluid, lymphatic fluid, plasma, cerebral spinal  
17 fluid and seminal fluid. Virus in the fluid component will,  
18 if unimpeded, cycle back into a competent cellular reservoir  
19 to infect new cells.

20 The HIV-RNA methods which I have already described  
21 are predominantly used to measure HIV-RNA in plasma.  
22 Therefore, the data to be presented here over the next  
23 couple of days describe estimated changes of detectable HIV-  
24 RNA by subcomopnent of one tissue reservoir of the body.

1 (Slide)

2 There are certain caveats to the methodology of  
3 quantitative HIV-RNA analyses to keep in mind when  
4 interpreting data. Each method is unique and every  
5 variation of a method should be considered unique.  
6 Therefore, data comparability between methods, unless  
7 rigorously studied, should be considered unclear.

8 Each method is different, meaning that those assay  
9 characteristics, referred to as validation parameters, are  
10 assay specific and vary between methods.

11 Finally, it is most important to remember that RNA  
12 copy number reported by an assay is a relative estimate of  
13 what may actually be present. In the absence of compelling  
14 comparability data, the estimate should also be considered  
15 assay specific.

16 (Slide)

17 There are also caveats associated with the  
18 selected HIV target source. When HIV is evaluated in plasma  
19 the copy number reported represents an indirect reflection  
20 of whole body HIV replication activity. It doesn't describe  
21 or allow predictions on any tissue compartment's  
22 contribution to the level of HIV-RNA present in plasma.

23 HIV nucleic acid material is the measurement  
24 target. However, this material is very labile. It's

1 potential random degradation may allow for inaccurate HIV-  
2 RNA estimates if the specimens are not properly managed.

3           Finally, all HIV particles present in plasma are  
4 assessed in these assays. Therefore, both infectious and  
5 defective HIV particles are being counted for RNA copy  
6 number.

7           (Slide)

8           My last slide just summarizes the brief  
9 presentation. There is a very wide range of HIV  
10 quantitative methodology being developed and used today to  
11 support analysis of viral burden in HIV-infected  
12 individuals. Because each method is unique, the  
13 characteristic validation parameters vary. In addition, due  
14 to differences in methods, the degree to which assay-  
15 specific data is comparable is not clear.

16           Although we have a vast amount of data on HIV-RNA  
17 copy numbers in plasma, it is important to keep in mind that  
18 we are assessing indirectly the replication activity of all  
19 competent tissues collectively. We have very little  
20 information on the degree of replication activity on  
21 individual tissues. The scientific and clinical community  
22 should continue to be suspicious of all competent tissue  
23 compartments and work towards the capability of HIV  
24 quantitative assessments of those tissues.

1           In closing, I want to point out that all these  
2 quantitative HIV methods are molecular-based state-of-the-  
3 art technology. We recognize the immeasurable amount of  
4 work and effort that has gone into the design and  
5 development of these methods. We also understand that these  
6 types of technologies are in a dynamic state of improvement  
7 and we look forward to each advancement.

8           This concludes my portion of the morning session,  
9 and now it is my pleasure to introduce Dr. Don Brambilla,  
10 from the New England Research Institute, who will present  
11 data which will further describe assay validation  
12 parameters. Thank you.

13                   **Assay Characteristics, Don Brambilla, Ph.D.**

14                   (Slide)

15           DR. BRAMBILLA: Good morning. Assay  
16 characteristics -- well, people write books about this in  
17 detail and have all kinds of characteristics in them. What  
18 I am going to talk about today are three basic issues that  
19 have to do with comparability of measurements between assay  
20 techniques and also reproducibility of measurements from  
21 different techniques.

22                   First, the relationship of variation in RNA  
23 measurements to RNA concentration, in other words, the viral  
24 load.

1           Second, differences in estimates of HIV-RNA copies  
2 per milliliter of RNA concentration among laboratories in  
3 which the same kit is used and also among kits. So we are  
4 going to look at the comparability of the kits.

5           Then, lastly, longitudinal variation within  
6 patients, and what I want to focus on there is overall  
7 variation in series of measurements in the same patient,  
8 then the contribution of assay variation and biological  
9 variation to the total or the relative contribution.

10           Then the effect of assay variation, different  
11 levels of assay variation, on confidence limits that we can  
12 place around a measurement. So I will take each of these in  
13 turn.

14           (Slide)

15           I am going to talk about three kits, three assays  
16 today: The Chiron ES bDNA is the second generation assay  
17 that was released last summer. Here I have linear ranges  
18 listed. I am more confident about the listing of the lower  
19 end of these ranges than I am about the upper end of these  
20 ranges. So if anybody sees something they disagree with up  
21 here, let me know later please. The Organon Teknika NASBA  
22 Assay and then the Roche Amplicor HIV Monitor Assay. Now, I  
23 have listed in the linear range two different values for the  
24 lower end of the Organon Teknika Assay. That is because the

1 assay can be run with two different sets of internal  
2 calculators, one set being simply a 1:10 dilution of the  
3 other, which drops the lower limit from 4000 copies to 1000  
4 copies.

5           There are two assays I am not going to talk about,  
6 the Roche Ultra-Sensitive Assay, which is still under  
7 development and characterization at Roche, and I am not  
8 going to talk about the NucliSens Assay from Organon  
9 Teknika, which was just released, and what that means is  
10 that part of my talk is already out of date.

11           (Slide)

12           A lot of the data that I am going to show you is  
13 derived from the Virology Quality Assurance Program, which  
14 is funded by the Division of AIDS, DAIDS in other words, to  
15 provide quality assurance for virologic assays in ACTG  
16 clinical trials and other NIH-funded studies, not just for  
17 RNA but for virologic assays in general.

18           The VQA laboratory is at Rush Presbyterian St.  
19 Luke's Medical Center in Chicago. The statistical center is  
20 in the New England Research Institute. That is me. There  
21 are several pieces of the program that are important to us  
22 today. First, the VQA laboratory provides external  
23 standards for RNA assays that are included in the assays  
24 that are run in ACTG clinical trials. Second, we conduct

1 routine proficiency testing of participating laboratories,  
2 and I will talk more about that. Third, we conduct studies  
3 to characterize and standardize virologic assays and, of  
4 course, that is what this talk is all about. The first two  
5 will play into what we are going to look at today.

6 (Slide)

7 The proficiency testing program, we need to  
8 describe that briefly since it is a basis for a lot of what  
9 I am going to say. The mandate in developing a proficiency  
10 testing program for RNA, the mandate that we were given by  
11 DAIDS was to develop a program which would show that a  
12 laboratory could maintain the precision needed to detect a  
13 five-fold difference between two measurements of RNA in the  
14 same assay batch.

15 We operationalize that by showing that the intra-  
16 assay standard deviation for  $\log_{10}$  RNA should not be  
17 significantly greater than 0.15. The reason for putting it  
18 that way is because 0.15 gives you 90 percent power to  
19 detect a 5-fold difference. In other words, given the  
20 design of our panels, this means that the standard deviation  
21 really has to be less than something on the order of 0.19 to  
22 0.21, depending on what panel we are looking at. The actual  
23 upper limit depends upon the number of specimens on the  
24 panel, number of dilutions.



1           We also have two other criteria. FDD errors must  
2 be less than the cutoff. Now, what is an FDD error? False  
3 difference detected. What this really means is that if you  
4 have a series of replicates at the same concentration on a  
5 coded panel that is sent to a laboratory, you compare the  
6 estimates that come back those specimens and if the  
7 estimates exceed a cutoff, which is the standard deviation  
8 of 0.15, then you have a false difference detected.

9           The other thing we do is TDU errors, a true  
10 difference between two specimens that differ by a factor of  
11 5 in concentration. If the estimates are too close  
12 together, again, based on this 0.15 standard deviation, then  
13 we declare that there is a true difference that went  
14 undetected. In order to be certified under this program, a  
15 laboratory must maintain a standard deviation and have  
16 counts of FDD errors and TDU errors on the panel below the  
17 levels that are, again, specified panel by panel depending  
18 on the number of specimens on the panel.

19           Lastly, we have to have a panel that is free of  
20 false positives. The last two rounds of testing have  
21 included 6-8 negative specimens. So a certified laboratory  
22 meets all of these criteria.

23           Now, I want to say one other thing. This 0.15 is  
24 not out of line with the levels of precision that are

1 claimed by the manufacturers for the assays that we are  
2 looking at.

3 (Slide)

4 The proficiency testing program -- let's talk  
5 briefly about panel composition because, again, a lot of  
6 what I am going to say is based on what is in the panel, the  
7 results from the panel, so I would like you to understand  
8 what is on them.

9 First, we have plasma that has been spiked with  
10 concentrated HIV, a series of five-fold dilutions. A  
11 typical one will have about four different five-fold  
12 dilutions on it. So one might have 100,000, 20,000, 4,000  
13 etc. There will typically be at least three replicates in  
14 each dilution. The two most recent rounds of tests have  
15 also included samples from HIV-infected patients. Again,  
16 what we have done, we have had is we have had one to three  
17 patients on the panel, so I think it is actually one or two  
18 on the panels I am going to talk about today, two to three  
19 dilutions of each. So we will take plasma from a patient  
20 and send it out neat and then send out a 1:5 dilution of it  
21 so we can compare it for TDU as well. Again, about three or  
22 more replicates of each one, depending on the panel. Then  
23 we are going to have panels that have three to eight samples  
24 of HIV-negative plasma, except the most recent panels have

1 been six to eight. So we have a mixture of prepared  
2 specimens and specimens from HIV-infected patients on the  
3 panels.

4 (Slide)

5 Here are two rounds of testing, round 6 and round  
6 7, which is going to be the basis of a lot of this. The  
7 concentrations on round 6 in the HIV-spiked specimens, the 3  
8 replicates of each, range from 2,200 copies/mL up to  
9 1,375,000 copies/mL.

10 Now, this is above the linear range of the Roche  
11 Monitor assay so this was not sent to labs that use the  
12 Roche Monitor Assay. The 2,200 copy specimens were. The  
13 2,200 is too low for the original Chiron assay, not for the  
14 ES bDNA assay but this panel was prepared before the ES bDNA  
15 assay was released. So the Chiron and Organon Teknika NASBA  
16 labs, or labs using those assays, received panels that  
17 ranged from 11,000 to 1,375,000 copies/mL.

18 Now, on panel 7 everybody got the same  
19 concentrations. You can see that they are slightly  
20 different from the previous panel but, again, we have a  
21 series of 5-fold dilutions.

22 (Slide)

23 The patient specimens -- we had 2 patients from  
24 panel 6. Sorry, this is actually incorrect. What we have

1 on panel 7 is 1 patient neat, 1:5 and 1:25. On panel 6 we  
2 had 1:5 and 1:25 dilutions of a specimen from that same  
3 patient, and then a 1:5 dilution of a specimen from patient  
4 2. So, sorry about the confusion. It is actually these 3  
5 here that are on panel 6 and then this one, this one and  
6 this one are on panel 7.

7 Here are the median concentrations estimated  
8 across all the laboratories participating in the testing,  
9 ranging from 14,000 copies/mL up to 320,000 copies/mL. It  
10 is the range over which we tested patient specimens in the  
11 proficiency program.

12 (Slide)

13 Recent results -- well, panel 6 was sent out on 2  
14 rounds. This is because the Chiron ES bDNA assay was  
15 actually released between round 6A and 6B. So the data that  
16 I present to you that is based on comparisons involving the  
17 ES bDNA assay is going to involve 6B and 7A. The data that  
18 I present that focuses only on the Roche assay, which will  
19 come later when we talk about variability, is going to  
20 involve 3.

21 The bottom line here is that what you see in the  
22 numerators is the number of laboratories that met the  
23 certification criteria and the denominator is the number of  
24 participants in each round. Roughly two-thirds to three-

1 quarters of labs on a given round meet the certification  
2 criteria.

3 (Slide)

4 Let's go on to our first topic, the relationship  
5 of variation to RNA concentration. Most of the statistical  
6 models that we use to analyze data assume that the variation  
7 in our measurements is homogeneous. For example, it does  
8 not depend upon the mean. In other words, variation is  
9 constant. It does not depend upon viral titer. Variation  
10 of measurements of titer does not depend on viral load  
11 itself.

12 In assays variation is often correlated with the  
13 mean. That is true whether you are looking at the assay  
14 variability or the overall variability. The same thing  
15 tends to hold. So this violates an assumption of the model.  
16 So what we typically do is transform the data prior to data  
17 analysis,  $\log_{10}$  transformation for example, to stabilize the  
18 variation.

19 (Slide)

20 Let's look at this. These are data from  
21 proficiency panel of the round 6 and 7 for the Roche Monitor  
22 assay. What you see here is each symbol on this plot is a  
23 standard deviation for 3 replicates. So this is assay  
24 variation. This is across all the laboratories that used

1 the Roche Monitor assay on round 6 and 7. The pluses are  
2 the spiked specimens. The circles, if you can see them, are  
3 the patient samples.

4           The only thing I want to point out here is that  
5 this is plotted on a log scale but actually the calculations  
6 of standard deviations are done without transforming the  
7 data. The thing I want to point out here is that the line  
8 connects the median standard deviations, simply just point  
9 out that the median standard deviation does go up with the  
10 mean.

11           (Slide)

12           If we transform to the log scale, you can see that  
13 what we get now is we transform the data to the log scale  
14 and calculate the standard deviations, and now we get a plot  
15 where the medians are pretty much constant across this range  
16 of concentrations.

17           (Slide)

18           If you do the same thing with the Organon Teknika  
19 NASBA assay you may actually get a standard deviation that  
20 rises somewhat at lower concentrations, but it is a lot  
21 flatter than you would see if you plotted standard  
22 deviations of untransformed data against the mean.

23           (Slide)

24           For the bDNA assay, it is pretty flat down to

1 about 10,000 copies. We have very limited observation in  
2 this range, below 10,000 copies so I am not going to say  
3 much about what goes on down there. This is under  
4 investigation.

5 Because of what I have shown you in the last three  
6 or four plots, we are going to talk about log transformed  
7 data when we do comparisons between kits, between  
8 laboratories and when we look at variability.

9 (Slide)

10 Next topic, differences in estimates of HIV-RNA  
11 concentration. What I am interested in here are two things  
12 that I want to show you: to what extent do estimates of HIV-  
13 RNA concentration vary among laboratories in which the same  
14 kit is used, and to what extent do estimates vary among  
15 kits. That is when you pool data across laboratories and  
16 prepare kits. So let's take these in turn.

17 (Slide)

18 A little bit about statistical methods and  
19 differences among laboratories. What I have done is fit  
20 linear regressions of log transformed estimates of RNA  
21 concentration to log transformed nominal concentrations.  
22 For the patient specimens we use those medians as the  
23 nominal concentrations. Then we compare the slopes and  
24 intercepts among laboratories within a kit. For the

1 differences among kits we use a random effects model, and  
2 what we are doing essentially is assuming that the slopes  
3 and intercepts are the regressions in the first part of the  
4 analysis are normally distributed around kit means, and we  
5 are simply comparing the kit means to get estimates of  
6 those.

7 (Slide)

8 Here is a comparison of laboratories that use the  
9 same kit. This is panel 6B and panel 7A. These are the  
10 fitted regressions. The panels that are labeled in yellow  
11 are panels in which either the slopes, the intercepts or  
12 both varied significantly among laboratories. For the ones  
13 labeled in white we didn't find any significant variation.  
14 This is on the spiked samples.

15 One of the interesting things I find about this is  
16 that there really is not a lot of difference in the visible  
17 spread of these regression lines between labs where you do  
18 get a significant difference among slopes and intercepts and  
19 where you do not. There are some fairly small differences  
20 in the error variance around regressions that seems to  
21 account for this. If you want to talk in terms of r-  
22 squares, the measure of error variance, as statisticians  
23 like to do, all of these models have r-square values of 0.95  
24 or greater. These two are down in the 0.95 range and these



1 are higher. So the small difference in percent of explained  
2 variation in the models seems to account for whether we can  
3 detect differences among laboratories or not. All in all  
4 though, the regression lines are reasonably close together.  
5 You can see that the spread is in all no more than about  
6 half a log in any one of these plots, with the exception of  
7 Roche where you have a large number of laboratories.

8 (Slide)

9 As I said earlier, one part of the VQA program is  
10 to send out a set of external standards that are included in  
11 assays that are run by the clinical trials group by the  
12 ACTG. They are also included on the proficiency panels.

13 What we can do is take these external standards  
14 and adjust the data across laboratories to a common set of  
15 standards and see if that reduces any of the variability  
16 that we see. We eliminate one significant difference.  
17 These lines are actually closer together. We still have  
18 some differences here. Although we detect a significant  
19 difference in intercepts here, the lines are actually closer  
20 together than they were on the previous plots. We still  
21 have a significant difference but we have shrunk the  
22 difference.

23 Here we have a bigger spread, and that is largely  
24 because of problems with this top regression line, one

1 laboratory that had problems with the standards themselves.  
2 But in general these lines are closer together. That is the  
3 effect of adjusting to a common set of external standards.

4 (Slide)

5 Now, there are no significant differences among  
6 regressions in the patient specimens, which I find  
7 interesting. These lines, here, the spread of regression  
8 lines in these groupings, is tighter than what we found for  
9 the spiked specimens, which is part of the reason we don't  
10 find a significant difference.

11 The other thing is that the r-square values around  
12 these regressions tend to be a little bit lower, still up in  
13 the 0.9 or greater range. So we may be losing a bit of  
14 power to detect differences. Again, for the most part, the  
15 spread among regression lines is no more than half a log.

16 (Slide)

17 If you correct to a common set of external  
18 standards, all of a sudden the effect of that correction  
19 seems to be to generate some significant differences,  
20 unfortunately. But here, you see, you still have very  
21 little difference among the regression lines. Here two have  
22 been separated from the others and that is the reason for  
23 the significance.

24 Now I am going to talk about how the kits compare.

1 What I am really going to do at this point is to talk about  
2 if you take these regression lines and compare them, you  
3 will sort of find an average regression here, using a random  
4 effects model, for an average regression through these to  
5 see how they compare.

6           If you do this, here are some p values. This is  
7 for the spiked specimens both for the kit-based estimates  
8 and then the estimates that are adjusted to the external  
9 standards. What you find is that we have significant  
10 differences on panel 6B and on panel 7A among the kids. The  
11 intercepts of the regressions differ in both cases. The  
12 slopes differ in 7A, and the slopes are close to different  
13 on 6B. With the VQA-adjusted estimates we have basically  
14 eliminated the significant differences on 6B by adjusting to  
15 the external standards. We still have significant  
16 differences here on 7A.

17           (Slide)

18           However, here is part of the data. Those of you  
19 that are familiar with coral snakes will like this color  
20 scheme I think. But Roche is in yellow, Chiron is in red,  
21 Organon Teknika is in black. It is hard to see, there is a  
22 black line here and there is one buried between the two  
23 yellow lines. The shorter segments are from 6B, the longer  
24 ones from 7A.

1           The main thing here is that there is a small  
2 difference between the sort of average regression for Roche  
3 Monitor and the Organon Teknika NASBA assays up here. But  
4 the primary reason for the significant differences for both  
5 6B and 7A between kits is because the bDNA assay is  
6 providing estimates that are lower than the estimates on the  
7 spiked samples and the estimates provided by other kits.

8           You can see that the spread seems to increase at  
9 lower concentrations. In other words, that is the  
10 difference in the slopes. For the most part, we are looking  
11 at a 0.3 to 0.5 log difference between these average  
12 regression lines, a 2- to 3-fold difference between the bDNA  
13 estimates and estimates from the other assays.

14           Now, if you do look at the data adjusted to the  
15 external standards, you can see that the regression lines  
16 are all on top of each other, which is interesting because  
17 it did still have a significant difference among regressions  
18 on 7A but it is very hard to see in this plot because the  
19 regression lines are so close together.

20           (Slide)

21           Here is the same thing for the patient specimens.  
22 Here we have differences in slopes on 6B, and that is about  
23 it. The only thing we see here is a difference in slopes.  
24 However, if you go to the VQA-adjusted estimates we pick up

1 a difference in intercepts on 6B as well. I will show you  
2 why.

3 (Slide)

4 There is something funny here. The longer  
5 segments are 7A again and you get the same pattern that the  
6 NASBA assay is producing estimates, a little bit higher than  
7 the Roche assay. Both of these are higher than the bDNA  
8 assay.

9 On 6B we have the slope of this segment, here,  
10 which is different from the slope of the Roche and NASBA  
11 assay. All the lines tend to collapse together again, even  
12 though we still do have a significant difference after  
13 adjustment to the VAQ standards, but the lines are much  
14 closer together. We have shrunk the difference between  
15 specimens.

16 (Slide)

17 Let's go on and talk about longitudinal variation  
18 within patients, the third topic, and then I will sum up and  
19 put everything together.

20 The three questions that I want to address are  
21 what is the overall level of variation of HIV-RNA within a  
22 patient? What I am thinking of here is if we track a  
23 patient over time during, say, clinical monitoring, quite  
24 apart from treatment effect, how much variability can we

1 expect to see in those numbers?

2           What are the relative contributions of assay  
3 variation and biological variation, overall variation? And  
4 how does assay variation affect the confidence limits on a  
5 measurement of RNA? These are the three topics that I named  
6 earlier.

7           (Slide)

8           There are several sources of variation that one  
9 can worry about when one talks about variability in RNA  
10 measurements. There is specimen handling prior to assay.  
11 So the blood is drawn. It has to get to the laboratory and  
12 then the assay takes place. But what happened in between?  
13 All sorts of things: storage, freeze-thaw, it sits on the  
14 shelf too long--lots of issues that can come up.

15           There is intra-assay variability. We have already  
16 talked a little bit about this. You would measure that by  
17 looking at differences among replicates in the same batch.

18           Inter-assay variability. If you took replicate  
19 subsamples from a sample of blood and put them in different  
20 assay batches what you would be measuring by inter-assay  
21 variability is really a combination. Measured in those  
22 terms, the difference between batches is really a  
23 combination of the variation within the batch and between  
24 the batches. So what I am talking about here is the added

1 component that comes up when you separate two aliquots into  
2 different batches.

3           Intra-kit, which is among kit lots, for example,  
4 which may be part of inter-assay variability, which is one  
5 way to view it. So if you use the same kit over time within  
6 a lab and keep changing kit lots, is that contributing to  
7 inter-assay variability?

8           Then inter-kit variability--we have already talked  
9 about this, differences between, say, the Monitor assay and  
10 the NASBA assay.

11           Then biological fluctuation, even if there is no  
12 systematic change in the patient and you had zero assay  
13 variability, you would not see constant numbers. Things  
14 would fluctuate. Those are all sources of variability.

15           What we are going to do now is to assume a patient  
16 is followed over time with the same kit in the same  
17 laboratory, with no change in treatment. So what we are  
18 going to do is focus on intra-assay variability, inter-assay  
19 variability and biological variability.

20           (Slide)

21           Measurements of variation is an important point  
22 because this tells us how we go about doing the estimates  
23 that we need to do to address these issues. Typically, when  
24 we talk about variability in an assay or measurements in a

1 patient, we talk about standard deviation of the  
2 measurements or some function of it, which might be a  
3 coefficient of variation which is the standard deviation  
4 divided by the mean. It might be the confidence limits you  
5 produce from a standard deviation. It might be the  
6 variance, simply the square of the standard deviation, or  
7 you could put the standard error, I suppose, up here as  
8 well. We typically discuss variation in terms of standard  
9 deviations or one of these functions but variances, not  
10 standard deviations, are additive.

11 (Slide)

12 Now, this is important. If you want to take apart  
13 the components of variation in an assay or in a series of  
14 measurements, say, you have to do that in a variance scale.  
15 What I am going to do is use this additive relationship here  
16 to take apart the overall variation, and then I am going to  
17 convert these values by taking the square root, two standard  
18 deviations, to present them to you since that is the scale  
19 of which we typically talk.

20 So our little model here is just overall variation  
21 in a series of measurements in a patient. It has three  
22 components that we are going to talk about. Biological  
23 variation may be inter-assay variation between batches and  
24 intra-assay variation.



1           Now, this model applies if you can get a series of  
2 clinical measurements in which each time the patient comes  
3 in and a sample is taken it is assayed. It doesn't apply if  
4 you are looking at samples that are assayed in a batch,  
5 typically done in a clinical trial where all the samples  
6 from a given patient are saved and assayed in a batch,  
7 because in that case the inter-assay component disappears  
8 and you end up with a somewhat simpler equation where the  
9 total is just the sum of the biological intra-assay  
10 components.

11           (Slide)

12           As I said, what we want to look at is the overall  
13 standard deviation, which is simply the square root of the  
14 sum of these three components, and we can define  $S_b$ ,  $S_e$  and  
15  $S_a$  to be the standard deviations that correspond to what you  
16 saw previously from the same components.

17           (Slide)

18           Ideally what we do is measure components of  
19 variation in a nested study so that we can measure all three  
20 components at the same time. What we would do is take a  
21 sample from a patient, divide it into four aliquots, assay  
22 two in one batch; wait a while and assay the other two in  
23 another batch. So we would have both within batch and  
24 between batch measurements. We would repeat the same

1 process with a sample from the same patient, and that way we  
2 would be able to look at within batch, between batch and  
3 between sample of variability all, as I said, in a nested  
4 study.

5 I don't have data from a nested study to show you.  
6 I actually asked the ACTG if it would be possible to do that  
7 and when they figured out what it would cost they said no.  
8 So we will take a somewhat different approach to doing this  
9 instead of a nested design.

10 (Slide)

11 What we are going to do is to estimate overall  
12 variability from serial measurements in several ACTG trials  
13 and in a natural history study. Then we will look at  
14 estimating the intra-assay and intra-assay variability using  
15 the standards and panels from the Virology Assurance  
16 Program. Then we will estimate biological variation by  
17 subtraction. Biological variation is simply the square root  
18 of the overall variation minus the intra-assay variation. I  
19 put it this way because the studies in which I am going to  
20 estimate overall variation all involve batched assays so we  
21 don't have the intra-assay component in that case.

22 (Slide)

23 Here we have estimates of the overall standard  
24 deviation. Four clinical trials, ACTG 175 is actually run

1 in three laboratories, and then the Women and Infant  
2 Transmission Study, which is a natural history study of  
3 transmission from mother to offspring during pregnancy.

4           What I have done here for the WITS, I have three  
5 or four samples per woman for 57 women. They were collected  
6 during pregnancy. For ACTG 076 there are 55 women. This is  
7 a subset. They were in the placebo arm and had measurements  
8 at study entry, labor and delivery and six months  
9 postpartum. They received no AZT. This is a study that  
10 demonstrated that AZT reduces the risk of transmission in  
11 pregnancy. Then clinical trials 175 and 229 obtained paired  
12 baseline measurements. So we can take the differences  
13 between the two baseline measurements, the interval between  
14 them varying somewhat, two weeks or so on average I think.  
15 From those paired baseline measurements we can estimate  
16 variability. These are sample sizes and numbers of patients  
17 involved. There are 663 total.

18           We can use linear models, either random effects  
19 models, if we have three or more measurements, or basically  
20 nested analysis of variance with the patient ID as the  
21 predictor if we have two measurements, to estimate overall  
22 standard deviations. For the most part, they are in the  
23 0.23 to 0.29 range. There is one exception here, which is  
24 quite small.

1           We can also get robust estimates. What I am  
2 getting at here is that the estimates from a model are  
3 sensitive to outliers. If you have a few points that are  
4 extreme values, they can blow up a standard deviation and  
5 give you what some would argue is an inflated estimate of a  
6 standard deviation that doesn't really represent what is  
7 going on in most of the data. If you take a robust estimate  
8 from paired measurements in all of these, in some cases you  
9 get pretty much the same thing because you don't have a lot  
10 of outliers. In other cases you get, as you do for Lab C in  
11 ACTG 175 and also 229, a pretty big drop because there are  
12 larger numbers of large differences between measurements in  
13 those studies.

14           But if you combine all the studies the overall  
15 estimate of 0.26 is what you get from the linear model; 0.18  
16 is what you get from the robust approach. I am going to  
17 work with 0.26 because, although I know it is inflated by  
18 the outliers, it represents what is going on in all of the  
19 data. As I said, the robust estimate represents what is  
20 going on in most of the data but I would rather work with  
21 what is in all of the data.

22           One caveat about this is that the overall estimate  
23 is 0.26. It turns out that there is a positive correlation  
24 between the time interval between measurements and the

1 variability that one observes in these studies. The  
2 correlation is extremely weak. The correlation coefficient  
3 is about 0.05 but it is significant because of the large  
4 number of samples that we have. So bear that in mind. I  
5 might comment on that further a little later.

6 All the laboratories involved in this study used  
7 the Roche Monitor Assay. So when we go to the VQA standards  
8 and the VQA proficiency panels to look at assay variability  
9 we are going to look at the standards that are in some of  
10 these laboratories or we are going to look at the  
11 proficiency panel data from labs that use the Roche assay.

12 (Slide)

13 The standards--ACTG 076, the data that I showed  
14 you were derived from 48 assay batches, and there were three  
15 positive VQA standards in each batch. For 175 there were 26  
16 and 37 batches in the three laboratories, and I think there  
17 were two standards included in each batch. So we estimate  
18 intra-assay standard deviations and inter-assay standard  
19 deviations from those standards. These are made up of  
20 plasma spiked with HIV.

21 The intra-assay standard deviations we have 0.09  
22 for three of them; we have 0.15 for the fourth one. the  
23 inter-assay standard deviations are 0.08, ranging from 0.05  
24 up to 0.10. If we put all the data together, all the

1 studies together, we end up with 0.12 and 0.08 as the  
2 estimates.

3           So the inter-assay component, the added component  
4 that one gets by running samples in the same patient in  
5 separate batches is a little bit smaller than the  
6 variability that one would see running samples repeatedly in  
7 the same batch. Of course, what you have to do if you want  
8 to get the assay standard deviation if you run samples in  
9 separate batches is square these, add them up and take the  
10 square root. I will show you a little later on what the  
11 effect of that is.

12           (Slide)

13           The Proficiency Testing Program, back to what we  
14 talked about earlier, let's look at our panels. Twenty-two  
15 laboratories participated in both round 6 and round 7 and  
16 used the Roche assay. So here we have estimates of intra-  
17 assay standard deviation from this. The median estimates  
18 from all 22 laboratories are on the order of 0.12 for the  
19 spiked samples and patient samples on round 7 to 0.16 for  
20 the same laboratories on panel 6. Here are the ranges for  
21 the intra-assay standard deviations. You can see that some  
22 of them are rather high. Those are laboratories that did  
23 not meet the certification criteria that we had set out  
24 which I talked about 20 minutes ago.

1 (Slide)

2 If you limit this to the laboratories--remember,  
3 there were 22 total--that meet the certification criterion  
4 just for having a standard deviation that is not  
5 significantly greater than 0.15, then you reduce this from  
6 22 to 20 or 17, 19, 18. The standard deviations drop a  
7 little bit. Instead of 0.15, they go down to the 0.11 or  
8 0.12 range for the median standard deviation, and you can  
9 see that the upper limit has dropped down.

10 So those are intra-assay standard deviations.  
11 There are consistent with what we saw from the standards  
12 that were included in the batches for clinical trials for  
13 ACTG 076 and 175. That is because the laboratories that  
14 participated in ACTG 076 and 175 met the certification  
15 criteria.

16 (Slide)

17 Biological variation. We have an estimate of  
18 intra-assay variability which we can take from the VQA data,  
19 and we have an estimate of overall variability. Here is our  
20 equation again estimating the biological standard deviation.  
21 As I said, if we take our larger value, 0.26, for the total  
22 then you can just square it to get the variance. If you  
23 take 0.12 ball park estimate of the intra-assay standard  
24 deviation, which is a good estimate of the median for the

1 laboratories that are meeting certification criteria, then  
2 what we end up with is that the biological standard  
3 deviation ends up being estimated at about 0.23.

4           The number itself doesn't mean much. What is  
5 important here is that if you take the biological variance  
6 and divide by the total variance, what you find is that 80  
7 percent of the variability that you would see in a sequence  
8 of measurements from a patient is biological rather than  
9 assay variability. This is true if you batch the specimens  
10 from a patient and assay them altogether. That is the first  
11 thing to keep in mind. The second is that it is a bit of a  
12 ball park estimate because there is some concern about what  
13 the actual assay variability is in a laboratory. You have  
14 to take that into account as well. I want to show you that  
15 and then I will wind up.

16           (Slide)

17           Let's talk about the effect of assay variability  
18 on the overall standard deviation. Here is 0.12 where we  
19 started. We said we had batched assay so there is no inter-  
20 assay component. We are assuming a standard deviation of  
21 0.23 and that gets us back to our 026.

22           If we simply allow the intra-assay component to go  
23 up to the upper limit of what is considered acceptable on  
24 the VQA program, then the overall standard deviation drops



1 from 0.26 to 0.30. If we let it go up a little bit higher,  
2 it jumps up to 0.34. If we start throwing in the inter-  
3 assay component and now we are looking at what happens if a  
4 patient is followed over time and samples are assayed as  
5 they are collected, then here we are back to 0.12-plus, the  
6 lower estimate of inter-assay variability that we had and we  
7 get a very small increase in going from here to here, 0.27.

8           So if the laboratory is performing well with  
9 regard to intra-assay standard deviation and has an inter-  
10 assay component that is at the lower of the two levels that  
11 we are comparing here, you get a very small increase in  
12 overall standard deviation. In other words, you don't lose  
13 much in this setting by going to real-time testing.

14           If the standard deviation creeps up to 0.15 for  
15 the inter-assay component, then we are up to 0.3 compared  
16 with our 0.26, and now let's just boost the intra-assay  
17 component to 0.2 and now we are up to 0.32 and 0.34.

18           Finally down here, at the bottom, we have a  
19 laboratory that is not meeting the performance criteria for  
20 intra-assay standard deviation set up by the VQA and also by  
21 the manufacturers, at least according to the package  
22 inserts. Then we have the two inter-assay standard  
23 deviations and we are up in the 0.35-0.37 range. So we have  
24 a substantial increase in overall standard deviation here.

1 (Slide)

2 What is the impact of this on our actual numbers?

3 Here is the same chart and I have added 95 percent  
4 confidence limits. Here, at 0.26 it is just 95 percent  
5 confidence limits plus/minus half a log. You can expect the  
6 confidence limits that you could put around a measurement on  
7 a patient in that setting would be plus/minus half a log.  
8 Then you can just count through here and see how the  
9 confidence limits would increase. Actually, the one I  
10 wanted to point out is right here, again, a small increase  
11 in standard deviation from going to real-time testing  
12 translates into a small increase in the confidence limits,  
13 from 0.51 to 0.53 to 0.59. However, if you got problems  
14 with the intra-assay standard deviation you are up to around  
15 plus/minus 7/10 of a log in terms of confidence limits.

16 We did all of this in terms of log scale because  
17 of the variability issues that I talked about, but we can  
18 now take plus/minus half a log to 7/10 of a log and  
19 translate back to real numbers, real RNA measurements, and  
20 say, okay, what is the confidence limit that we would place  
21 around a measurement of 10,000 copies/mL, just to pick a  
22 number that we can work with?

23 Here you have from 3,100 to 32,000. So if you had  
24 a measurement of 10,000 copies/mL in a laboratory that was

1 performing with this intra-assay standard deviation you  
2 might say that you got about a 10-fold range in which you  
3 would park that number. It lies somewhere between around  
4 3,000 and 30,000.

5           If you go to real-time testing the limits, again,  
6 don't increase very much. They are still in the range of  
7 3,000 to 34,000. But when you get down here where you are  
8 having intra-assay problems and you have a larger inter-  
9 assay standard deviation, now you are down to 1,900 to  
10 54,000 for the confidence limits on a single measurement.

11           This translates directly, although I don't have it  
12 on a slide here, into talking about the confidence limits on  
13 the difference in two measurements, in other words, in  
14 measurement of change within patients. As you increase the  
15 assay variability, the confidence limits for measurements of  
16 change also increase. I am sorry I don't have a slide to  
17 demonstrate that.

18           (Slide)

19           Summary. I covered three topics. First, I want  
20 to say that we use log transformation to stabilize variances  
21 when we analyze data. As I showed you on the last slide, we  
22 can always take that back to the untransformed numbers, the  
23 absolute RNA copies/mL when we want to look and see what the  
24 numbers actually mean.

1           What we found from this in kit comparisons is that  
2 estimates from the Chiron bDNA were lower than those from  
3 the Roche Monitor OT NASBA assays. Stay tuned when we start  
4 looking at the new version of the OT assay and we will do  
5 this all over again in about six months and see how things  
6 compare. The differences between the bDNA assay and the  
7 others actually vary with concentration. They are actually  
8 larger at lower concentrations.

9           Biological variation--I will put it this way, 80  
10 percent of total variation in a patient, excluding treatment  
11 effects and trends, assuming that the patient is not on  
12 average changing. I said that it is up to 80 percent  
13 because that is assuming some minimal level of assay  
14 variation. The percentage of variation represented by the  
15 patient will go down as assay variability goes up. The  
16 total variability will go up as assay variability goes up  
17 and the percentage represented by the patient will go down.

18           What the confidence limits are saying, what this  
19 really translates into is that unconfirmed changes of 0.5 to  
20 0.7 logs may have very little clinical meaning because those  
21 are just within the realm of possibility for fluctuations,  
22 including assay variability.

23           Lastly, I think it is important to monitor  
24 laboratory performance over the long term, perhaps using

1 external standards, however one chooses to do it, in order  
2 to ensure that that component is kept as small as possible  
3 so that what we are looking at is primarily the biological  
4 component. Thank you.

5 DR. HAMMER: Thank you, Don. Before we move on I  
6 might give the Committee a chance to ask questions of Dr.  
7 Iacono-Connors and Dr. Don Brambilla. This provides a lot  
8 of important data for us to discuss issues of application of  
9 these assays over the next couple of days. Are there any  
10 questions? Mark?

11 MR. HARRINGTON: It looks like there is a lot of  
12 variation even among the very best labs. My question is, we  
13 are talking about a standard of care that uses these  
14 measurements on a routine basis, how much greater variation  
15 do you think there is out there in the labs that were not  
16 certified but, yet, from which clinical data come from  
17 trials, and then how much even more variation is there in  
18 commercial labs that are actually giving people measurements  
19 that are used to make treatment decisions?

20 DR. BRAMBILLA: Well, I don't have any hard  
21 information to use to answer that question. You know, it is  
22 entirely possible that a clinical lab maintains its own very  
23 strict standards and, therefore, would meet these  
24 performance criteria. It is also possible that one doesn't

1 know unless one has some method of performance. I think  
2 that is the real problem. The answer to your question is  
3 that one needs to have a measure of performance and for a  
4 lot of laboratories I don't know what it would be.

5 MR. HARRINGTON: But what happens in an ACTG trial  
6 when the lab fails the certification but has been gathering  
7 data for the trial?

8 DR. BRAMBILLA: If I remember correctly, a  
9 laboratory has to be--typically in an ACTG trial the samples  
10 are saved in batches in the freezer and assayed at the end  
11 of the trial or at some point in the trial and the  
12 laboratory has to have maintained current certification in  
13 order to analyze the samples.

14 DR. HAMMER: Other questions?

15 DR. LIPSKY: When you look at variation are there  
16 particular trends, for instance, over replicates? Do the  
17 means go down? Do they stay the same? Do you get any hint  
18 of causes when you are doing an assay of why there is  
19 variation?

20 DR. BRAMBILLA: Oh, you mean from the proficiency  
21 testing program?

22 DR. LIPSKY: Yes.

23 DR. BRAMBILLA: Occasionally you do. A lot of  
24 times what you will see is that there are two different

1 kinds of variability that creep in here. Occasionally you  
2 will see a panel where everything looks great except one or  
3 two specimens in which something happened. Occasionally we  
4 can explain it easily. I have seen NASBA assays where it  
5 looked like the specimen never made it into the assay. The  
6 signal is basically zero. I have seen other assays where it  
7 looked like two specimens were switched. So occasionally  
8 you get a panel that looks fine except for a couple of  
9 outliers. Sometimes you do not. Sometimes you see panels,  
10 and I think this is the more common explanation, where at  
11 multiple nominal concentrations there is an increased spread  
12 over and above what you like to see. Why that happens I do  
13 not know. That is the point at which the VQA laboratory  
14 starts working with the personnel at the ACTG lab to try and  
15 figure out the problem.

16 DR. FEINBERG: Don, I think this is probably  
17 directed at the FDA and not you, but are there any  
18 comparable analogous data for biologic variance, etc. for  
19 the CD4 data that have been the basis for approval up to  
20 this time? I was not on the Committee when those things  
21 were laid out real carefully. I am just curious to know  
22 that.

23 DR. FEIGAL: We would probably have to get back to  
24 you with specifics, and you may have some comments too, but

1 I think it is of the same order of magnitude.

2 DR. BRAMBILLA: I have never looked at serial  
3 samples before from this point of view.

4 DR. EL-SADR: Does the biological variability  
5 differ by the actual RNA level in the patients? Did you  
6 look at people with higher values versus those with lower  
7 values?

8 DR. BRAMBILLA: Yes, I started to do that because  
9 I know that the transformation data I showed you was based  
10 on assay variability. No, I don't see a lot. If you go to  
11 the log scale, that tends to suppress a lot of the  
12 difference. If you don't go to the log scale, then, yes,  
13 absolutely the patients with a higher value show greater  
14 fluctuations. But on the log scale it is largely  
15 suppressed.

16 DR. VALENTINE: Just to respond to his question,  
17 the ACTG had a certification program for CD4s and the data  
18 was all analyzed by SDAC, and they did similar types of  
19 analyses.

20 DR. MATHEWS: Could you comment on what is the  
21 impact of these parameters if you were to pick clinical  
22 trial endpoints based on achieving, say, a certain threshold  
23 value, say, undetectable or whatever it is, a certain drop  
24 in baseline at a fixed point in time, whenever that might



1 occur? So time achieving a certain value versus trying to  
2 estimate a slope for an individual patient? In other words,  
3 it seems to me from what you have said that if there could  
4 be a 10-fold difference in a single measurement and 95  
5 percent confidence interval, it makes it much more difficult  
6 to pick a single threshold value as an endpoint versus  
7 trying to estimate changes by slopes over time, which might  
8 have less variability.

9 DR. BRAMBILLA: if you are talking about picking a  
10 single threshold value for making endpoint decisions for  
11 individual patients, I think the solution is to not believe  
12 anything until you confirm it with a second sample,  
13 something along those lines. One can define all sorts of  
14 regiments, whether it is two samples or three samples or a  
15 cutoff and then increase the confidence that the patient  
16 actually is below that cutoff by doing repeated  
17 measurements. I think that is the main thing you get here  
18 from those rather broad confidence limits.

19 The other problem that creeps in when you start  
20 talking about clinical trials, and it is one of the things I  
21 left out of this deliberately, is variation in treatment  
22 effect among patients or variance in treatment response. I  
23 leave that to my colleagues at SDAC to work with. That is  
24 more their job.

1 DR. HAMMER: Don, do you want to say something  
2 about the assay performance at the lower limits of  
3 quantification, where there is so much interest individually  
4 in patient management and certainly for clinical trial  
5 endpoints?

6 DR. BRAMBILLA: There is some tendency, and you  
7 saw this in the plot for the NASBA assay, for the standard  
8 deviations in log transformed data, at least on the NASBA  
9 assay, to increase at the low end. I don't know if this is  
10 true for the new NucliSens assay. I haven't seen any data.  
11 I am speculating in talking about the other assays because I  
12 really don't have as much data.

13 In fact, that is something that the VQA is working  
14 on, on characterizing. We have, for example, a set of  
15 panels in the field right now that are designed to  
16 characterize the Roche Monitor assay from 1,000 copies on  
17 down to 35,000 copies, looking at reproducibility and rates  
18 of values below detection limits. The data are not due in  
19 to us for analysis until next week.

20 DR. HAMMER: One quick last question, when you  
21 have your patient estimated numbers, those are taken from  
22 the mean of certified laboratories? What is the gold  
23 standard for identifying the patient numbers in your panels?

24 DR. BRAMBILLA: Oh, we just took the median across

1 all laboratories. That is a good point, but I think the  
2 thing to keep in mind is that the medians reproduce the 5-  
3 fold dilution series pretty well.

4 DR. HAMMER: Thank you.

5 DR. MATHEWS: I want to follow up on your comment  
6 to my question. You said that in individual patient  
7 management it is important to replicate. But in clinical  
8 trials you have to decide on an individual patient basis  
9 when they have achieved the endpoints. So you have the same  
10 problem as you do in individual patient management.

11 DR. BRAMBILLA: Absolutely, yes.

12 DR. MATHEWS: How many replicates do you need,  
13 based on data you have presented, to state that a person has  
14 confidently reached an endpoint, whatever that value may be?

15 DR. BRAMBILLA: Let's see, how do I answer that?  
16 I have to give you sort of a nebulous answer. It is hard to  
17 nail that down. The point is that if you have a value--for  
18 example, suppose you have a value that is so low that the  
19 upper 95 percent confidence limit on that value is less than  
20 the endpoint that you are talking about, then you might  
21 begin to think that one is enough. If you have two that are  
22 somewhat higher but the upper 95 percent confidence limit on  
23 the pair is less than that endpoint, then you might be happy  
24 with that. It all depends on how you want to define it.

1 That is the problem, the more replicates you have that are  
2 near or below that endpoint, the tighter the confidence  
3 limits that one can put around it.

4 DR. ELASHOFF: We are going to address that in  
5 more detail tomorrow so there will be a lot more data. I  
6 think maybe comments for Dr. Brambilla could may be put off.

7 DR. HAMMER: Thank you.

8 DR. VALENTINE: Just to pursue the point that Mark  
9 was making, clearly, we would presume that in clinical  
10 trials you would have quality assured laboratories, and so  
11 forth. Does the Agency know if there is some way of  
12 regulating or quality assuring commercial laboratories which  
13 are going to be predominantly used for patient management,  
14 which I suppose is what Mark was getting at I think? Is  
15 there a plan in place or how are those laboratories  
16 regulated? I know the College of Pathologists have sent out  
17 samples but I don't know with what stringency those are  
18 analyzed.

19 DR. FEIGAL: Laboratories are regulated by the  
20 accrediting bodies of those laboratories, which is not an  
21 FDA function. We would review the performance of the test  
22 and then it would be the College of American Pathologists  
23 review or sometimes the Joint Commission on Hospital  
24 Accreditation. Sometimes state licensing works. In fact,

1 it may vary state by state. Most commercial laboratories  
2 and hospital laboratories will have quality assurance  
3 programs and should be able to make it available how they do  
4 that.

5 DR. HAMMER: Thank you. Thanks, Don. I think we  
6 should move on. The next speaker is Dr. Winston Cavert,  
7 from the University of Minnesota, who will speak about  
8 comparative tissue compartment activities.

9 **Comparative Tissue Compartment Activity, W. Cavert, M.D.**

10 DR. CAVERT: I want to apologize to the Committee  
11 for using overheads. I found out on Thursday that all three  
12 of the locations within my neck of the woods that turn copy  
13 into slides were out of commission or on vacation.

14 (Slide)

15 We have clearly come a long way since Salk first  
16 coined the term viral load with regard to HIV in 1987, as  
17 this proceeding attests. But I think one of the implicit  
18 assumptions when we talk about plasma viral load is that the  
19 plasma reflects what is going on in the total body.

20 As I understand my charge from the Advisory  
21 Committee today, it is to see if that, in fact, the case?  
22 Does plasma viral load in some way reflect the viral burden  
23 of the entire body, in particular the lymphoid tissue,  
24 without therapy and during therapy?

1 (Slide)

2 None of the studies that I am going to cite today  
3 have the sort of statistical power that we just saw. All of  
4 the relevant studies really have probably less than 16  
5 patients in each. So I think we are still a long ways from  
6 having the kind of information that we would like about this  
7 question but there are some things that can be said.

8 I am going to review what was first called the  
9 dichotomy between the lymphoid tissue compartment and plasma  
10 or blood compartment, and a little bit of the viral kinetics  
11 models that has arisen out of that. I am going to talk  
12 about the correlation between plasma RNA and lymphoid tissue  
13 RNA in individuals that are untreated or what I am going to  
14 flippantly perhaps call lightly treated subjects, subjects  
15 with a single nucleoside perhaps; and then the comparative  
16 response of these two compartments during treatment; then  
17 some suspicion of what possible exceptions to the  
18 correlation between the loads between the two compartments  
19 might be; and then, finally, a brief comment about DNA load  
20 as well as RNA load, which is going to be the substance of  
21 this talk.

22 (Slide)

23 In 1991 Pantaleo and colleagues first identified a  
24 significantly higher burden of HIV in the lymph nodes or

1 adenoids as compared of the blood.

2 (Slide)

3 Since then the work from several laboratories and  
4 several sites I think has relatively incontrovertibly given  
5 us the impression that the lymphoid tissue contains the vast  
6 majority of the HIV viral burden within the body. Generally  
7 about three orders of magnitude perhaps more HIV is found in  
8 the lymphoid tissues than in the plasma.

9 In the lymphoid tissues we have identified major  
10 compartments or two major pools of virus. The first is  
11 follicular dendritic cells with germinal centers of lymph  
12 tissue. There, intact virions are complexed to antibody  
13 stored on the foot processes of follicular dendritic cells.  
14 Secondly, productive mononuclear cells within the lymphoid  
15 tissue. Most of these are CD4 cells but some as well are  
16 macrophages and other mononuclear cells perhaps.

17 (Slide)

18 When we are talking about lymphoid tissues we are  
19 primarily referring to secondary lymphoid tissues of lymph  
20 nodes, adenoids and tonsils, the spleen and the gut-  
21 associated lymphoid tissue within the intestines, and less  
22 so perhaps the thymus and the bone marrow.

23 (Slide)

24 To emblematically, if you will, put the two

1 different compartments within lymphoid tissue into some  
2 perspective, this is a computerized 3-D image of a series of  
3 sections from a single germinal center of a small lymphoid  
4 biopsy. What you see here, in the yellow is the follicular  
5 dendritic cell holding intact virions, and the red colored  
6 dots are individually productive mononuclear cells,  
7 producing virions with full-length HIV RNA.

8 (Slide)

9 There are at least as many, and actually probably  
10 a number of additional methodological issues that go along  
11 with looking at lymphoid tissue, even more so than there is  
12 in sampling blood.

13 For starters, there have been several different  
14 lymphoid tissue sites used for monitoring viral load: lymph  
15 nodes, tonsil and adenoids, otherwise sometimes referred to  
16 as mucosa-associated lymphoid tissue; gut-associated  
17 lymphoid tissue, most commonly from the colon; and spleen,  
18 taken from individuals who, for example, have  
19 thrombocytopenia and need, for clinical reasons, their  
20 spleens extracted.

21 There is also a number of different sampling  
22 techniques that have been used in these studies that we will  
23 be looking at. Autopsy specimens have been used. Lymph  
24 nodes have been biopsied with fine needle. Larger biopsies



1 have been done visually, either directly, for example of the  
2 tonsils, or direct larger needle biopsies of lymph nodes and  
3 also via endoscopy for the gut-associated lymphoid tissue,  
4 and finally tissue excision as in, for example, the spleen  
5 and whole lymph node excisions.

6 (Slide)

7 There are also different assays for dealing with  
8 this tissue, essentially variations on the same things that  
9 you have already heard about. There is a tissue bDNA assay,  
10 a number of versions of the RT PCR, including a tissue  
11 adaptation of the Amplicor assay, and at least two  
12 variations of the NASBA assay that I am aware of. There are  
13 also single-cell technology techniques, all of them  
14 essentially based on in situ hybridization, and some of  
15 those using more formal methods of quantitation including  
16 computerized image analysis or simply visual microscopic  
17 quantification.

18 (Slide)

19 All of this raises a number of potentially  
20 troublesome issues in analyzing HIV-RNA lymphoid tissue, and  
21 I have highlighted just some of the more prominent ones that  
22 I think are, at least some of them, unique to lymphoid  
23 tissue.

24 The first is the question of whether HIV is

1 uniformly distributed within lymphoid tissue. This is both  
2 between different types of lymphoid tissue and also whether  
3 it is uniformly or relatively uniformly distributed within  
4 particular lymphoid tissues, for example within one lymph  
5 node. When one obtains a sample of lymphoid tissue there is  
6 an important question and that is does the sample actually  
7 have lymphoid tissue in it. In a number of the biopsies  
8 that have been looked at, for example from gut or other  
9 sites, other types of tissue are included in the biopsy. So  
10 it is important to know what percentage of the biopsy is  
11 actually lymphoid tissue and what percentage is stroma, fat,  
12 epithelium etc.

13           The question what the standard of reference by  
14 which we quantitate the viral load in the lymphoid tissue is  
15 I think still the major question. Several of the ways that  
16 it has been expressed so far is to express samples as per  
17 gram of lymphoid tissue or per the number of human cells,  
18 presumably T-lymphocyte but not necessarily always, cells  
19 that are counted, or perhaps the total amount of RNA that is  
20 extracted. Obviously, comparing between studies that use  
21 these different methods is going to be difficult.

22           There is also the question of total HIV-1 RNA in  
23 the lymphoid tissue versus looking at the amount of virus  
24 that is associated with two major cellular compartments,

1 that is, the FDCs versus the mononuclear cells.

2           Then there is the issue of assay sensitivity and  
3 range. The sensitivity calculations and considerations are  
4 analogous to those for plasma but quite unique or varied for  
5 lymphoid tissue. Finally, there is obviously the issue of  
6 quantitative reliability or validation.

7           (Slide)

8           At the risk of being, unfortunately, somewhat  
9 superficial due to constraints of time, I have lumped a  
10 number of the studies together in general categories as to  
11 essentially how read them.

12           I thought I would start off with a quick reference  
13 to six pioneer studies, very important studies but earlier  
14 studies that were cross-sectional studies, if you will, in  
15 individuals that were either untreated or treated perhaps  
16 with a single nucleoside or a single additional nucleoside.  
17 One of the earliest of these by Dr. Pantaleo in 1993--all of  
18 these studies that are on this page, I should mention, to my  
19 reading or to the reading of the authors do not show  
20 correlations, or at least do not show particularly strong  
21 correlations between the viral load in the lymphoid tissue  
22 and viral load in the blood. All of these studies are  
23 small, for the most part a dozen or fewer patients.

24           In the black, underneath each assay, I have put

1 some of the issues that I think make the data somewhat  
2 difficult to interpret. In the first study there are issues  
3 of assay sensitivity, and I think the assays are rather more  
4 sensitive at this point.

5           The second study, by Sei in 1994, used tissues  
6 that were from autopsy and some of them had been 24 hours or  
7 more before they were harvested and, therefore, I think it  
8 would be expected that the lymphoid tissue RNA would be  
9 rather degraded and so correlation wasn't seen in that  
10 study.

11           There are two papers published from DATRI 003, the  
12 first in a group of individuals, between 250 and 500, I  
13 believe it was, CD4 cells. There were 8 patients in this  
14 1995 Cohen study that were--actually, I should say that all  
15 of these patients were on AZT at the baseline point, and it  
16 is the baseline that I am focusing on from a cross-section  
17 of perspective. In general, as I look at this paper, I  
18 don't think that they found a correlation between the  
19 lymphoid tissue and the plasma. However, there are  
20 significant differences in how they expressed the  
21 denominators of those two compartments so I think it is a  
22 little difficult to read. I am not sure, there may be some  
23 correlation in these studies. I don't have access to the  
24 primary data from that but looking at their papers I am not

1 sure.

2 In a group of pediatric patients who were failing  
3 double nucleoside therapy in 1996, Mueller did not find a  
4 correlation between the two compartments.

5 Finally, Pesca Meylan published a paper last year  
6 using fine needle biopsy in 8 patients. Both the issue of  
7 fine needle biopsy, which is a sampling problem that needs  
8 to be looked at more carefully, as well as differences in  
9 how the denominators in the two compartments were expressed.  
10 They did not find a correlation either.

11 (Slide)

12 There are three papers published in the literature  
13 that, in my estimation do show a correlation between the  
14 relative amount of virus in the lymphoid compartment and the  
15 relative amount of virus in the plasma. All of these were  
16 in small numbers of patients. Mary Ann Harris and her  
17 colleagues in Vancouver in conjunction with Dr. Pete Daley  
18 and his colleagues working on tissue bDNA at Chiron provided  
19 an abstract at Vancouver last summer, a study comparing  
20 plasma bDNA, I think using the second generation assay, with  
21 the tissue bDNA assay. As you can see in these 14 patients  
22 or so, they found a reasonably decent correlation with a p  
23 value of 0.02.

24 So I guess what I am saying when I talk about a

1 correlation is when the plasma viral load higher, is the  
2 lymphoid viral load also higher? Do the two relatively  
3 reflect each other? And I think this study shows, at least  
4 for the small number of patients, that they do. The reasons  
5 why I think this is a better study than the others are two-  
6 fold. First, although this is a lymph node biopsy, they had  
7 histological touch confirmation from the fact that they were  
8 dealing with lymph node tissue, and almost entirely with  
9 lymph node tissue.

10           Secondly, I think the tissue bDNA assay, of all  
11 the sort of homogenation assays available, is probably the  
12 one that has gone the furthest so far in terms of validation  
13 in looking at the biologic variability and the intra-assay  
14 variability.

15           (Slide)

16           Two other studies for which I will not show the  
17 log-log plots in the interest of time but, essentially the  
18 plots look very similar. Faust published in Otolaryngology  
19 Surgery in 1996 a comparison of plasma DNA versus tonsil  
20 biopsy, again with histological confirmation that it was, in  
21 fact, lymphoid tissue and, again, using bDNA assay performed  
22 by Pete Daley at Chiron. Plasma bDNA assays in this case  
23 were performed by the ACTG lab at the University of  
24 Minnesota.

1           Then finally a study out of our lab, senior  
2 authored by Dr. Ashley Haase, published late last year in  
3 Science, correlating plasma bDNA second generation assay  
4 with tonsil using a new technique of quantitative computer  
5 analysis of in situ hibernization. Again, both of these  
6 studies show reasonable correlation with fairly good p  
7 values in the Faust study in 10 patients and the Haase study  
8 in 8 patients with 13 biopsies.

9           (Slide)

10           To look at the issue of what happens to the  
11 lymphoid tissue load during treatment, there are several  
12 studies of lightly treated patients where when we see a drop  
13 in the plasma viral load the lymphoid tissue may or may not  
14 drop. One of these is the previously cited Cohen paper from  
15 DATRI 003 of 6 individuals who had already been on AZT  
16 previously were treated in addition with ddI. As you can  
17 see, the log changes in the plasma RNA, in the closed black  
18 symbols, are relatively small but, nevertheless, in the open  
19 symbols for each of these 6 patients there was a somewhat  
20 correlating drop in 5 of them in the lymphoid tissue load.  
21 One of these patients was clearly an outlier and I think it  
22 is an important point to recognize that in statements that  
23 the two compartments seem to correlate in their loads we  
24 have seen, in addition to this study, a couple of other

1 circumstances, patients that are definitely outliers, where  
2 they have in particular low plasma virus load and quite high  
3 lymphoid burdens.

4           This is a study published last year by Dr.  
5 LeFalay, from Toulon in Marseilles, in southern France.  
6 There are 4 patients, previously naive, treated with AZT,  
7 ddI and 3TC in a pilot study of that combination. The  
8 squares are lymphoid biopsies and the circles are plasma.  
9 As you can see, each of these patients, although the times  
10 of biopsy varied as well as the times of plasma sampling  
11 varied in these patients, we see some corresponding drop in  
12 both compartments. The drop, at least in this early study  
13 on the lymphoid tissue looks like it was smaller than it was  
14 in the plasma, but I think that probably has to do with the  
15 sampling method.

16           (Slide)

17           That same group published a second study in ME  
18 just a couple of months ago in 10 patients treated with the  
19 same 3 nucleosides plus saquinavir, previously  
20 antiretroviral naive. In the top panel are the 10 patients'  
21 plasma RNAs. As you can see, they have roughly a mean 3-log  
22 drop between them. In the bottom panel are the plots of  
23 their lymphoid tissue viral load changes over 8 weeks, but a  
24 2.5-log drop in the lymphoid tissue.



1 (Slide)

2 Two final studies on the comparison between plasma  
3 and lymphoid burden in treated patients, this one from  
4 Switzerland, again Dr. Meylan's group. There were 9  
5 patients again with fine needle. This is interesting. This  
6 paper was just published, in February I believe it was.  
7 This is an interesting comparison that hasn't been done so  
8 far in any of the other studies that I am aware of, that is,  
9 the log change in the RNA level on the Y axis versus the log  
10 change with treatment in the plasma HIV-RNA on the X axis.  
11 As you can see, there is a correlation to some extent  
12 between these.

13 (Slide)

14 A study that we performed in conjunction with  
15 colleagues from University of Amsterdam, using triple  
16 therapy of AZT, 3TC and ritonavir, 34 adults were split  
17 between 2 arms, half of them received all the drugs from the  
18 start and the other half received just ritonavir from the  
19 start and added the nucleosides 3 weeks later. Onto the  
20 study a tonsil biopsy substudy was appended. Ten of these  
21 patients ultimately were included into the tonsil biopsy  
22 substudy.

23 (Slide)

24 This is from Dr. Don Notterman, one of the

1 principal investigators in Amsterdam. This is what happened  
2 with the plasma viral load with this triple therapy over the  
3 course of 24 weeks. As you can see, this curve of plasma  
4 viral load looks similar to that we have seen in other  
5 potent combination therapy trials. Over the 24 weeks they  
6 had about a 3-log drop in their plasma in both arms.

7 (Slide)

8 In the lymphoid tissue over those same 24 weeks,  
9 the analysis being done in our lab by in situ hibernization  
10 and computer image quantification, we saw a 3.4-log drop in  
11 the FDC pool and greater than a 2.3-log drop in the  
12 mononuclear cell, actively producing cell pool, in the  
13 frequency of those cells. If one were to homogenize the  
14 lymphoid tissue, the FDCs would be the predominating  
15 component in that. You can essentially say that the total  
16 lymphoid tissue viral burden drop was greater than 3.4 logs.

17 (Slide)

18 Unfortunately, I don't have direct comparison  
19 curves of this, but in the curves for individual patients  
20 there is an obvious similarity. Part of the difficulty in  
21 comparing curves too closely is because the plasma was  
22 sampled much more frequently than the lymphoid tissue. The  
23 lymphoid tissue was sampled at baseline, then at day 2, at  
24 day 22 and at 24 weeks.

1           Here is what the curve looks like from the first 3  
2 time points. The FDCs are the top line, in the red and the  
3 mononuclear cells, are the blue line underneath. As you can  
4 see, they both fell off rapidly within the first 2 days and  
5 then a slower phase, if you will, although it is a little  
6 hard to talk about a phase when you have only two time  
7 points and you draw a line between them but a slower phase  
8 of viral decline from the lymphoid tissues between day 2 and  
9 day 22.

10           One of the reasons I wanted to show this slide was  
11 because to us this expresses a model of what is going on in  
12 the lymphoid tissue and in relation to the plasma. It had  
13 been proposed in the past that because FDCs bind antigen  
14 antibody complexes, they would be a long-term storehouse.  
15 In fact, what we found is that the drop in the FDC lymphoid  
16 pool paralleled the drop in the productive CD4 cell pool.  
17 Therefore, we think that there is a quite fluid equilibrium  
18 between the FDC "storehouse" and the mononuclear cells in  
19 the lymphoid tissue. By comparison with the plasma curve,  
20 there also seems to be fluid equilibrium between both of  
21 these compartments in the plasma. So we think that what is  
22 driving the plasma, viremia and the FDC loading is the  
23 production of virus by the mononuclear cells, and when you  
24 shut that off with potent therapy, or nearly shut that off,

1 all three pools drop in a roughly parallel fashion.

2 (Slide)

3 Just a couple of brief words about the  
4 relationship between lymphoid tissue proviral DNA and plasma  
5 RNA. This is an important issue because if we can drop  
6 lymphoid tissue RNA down to levels that are getting close to  
7 detectable, then the issue starts to emerge what about cells  
8 that are latently infected that may have simply proviral DNA  
9 within them?

10 I think earlier studies suggested that there may  
11 be a correspondence between the amount of proviral DNA in  
12 lymphoid tissues and the plasma RNA viral load. This is  
13 part of a study published by Sei in 1994. This, again, was  
14 a study that had autopsy tissues. I did not show a relation  
15 probably because of that between the RNA components but the  
16 lymphoid tissue DNA, in the open triangles up here, and the  
17 plasma RNA, in the closed circles on the bottom panel, show  
18 a rough correlation or, actually, a pretty decent  
19 correlation. So that was one of the first studies  
20 suggesting that there may be a correlation between lymphoid  
21 DNA and plasma RNA.

22 (Slide)

23 Here is another one that seemed to confirm that.  
24 It was 8 patients, in a letter to Nature Medicine in the

1 middle of last year by Diazoni. Again, I think you can see  
2 the correlation pretty much by just looking at the number of  
3 DNA infected lymphoid cells in this column versus the RNA  
4 copy numbers.

5 (Slide)

6 However, a more recent paper, published just about  
7 a month and a half ago by a group at Johns Hopkins, Ocono  
8 was the main author, using really I think quite  
9 sophisticated PCR assay techniques, suggested that there  
10 may, in fact, not be a correlation between lymphoid tissue  
11 DNA and what is going on in the plasma RNA. So I think that  
12 is really an open question at this point.

13 (Slide)

14 One other point to mention is that when I am  
15 talking about lymphoid tissue we are ignoring a lot of other  
16 potential sites. Lymphoid tissue is obviously where most of  
17 the viral burden in the body is located but there may be  
18 small amounts of virus in other locations. One study showed  
19 16 percent of individuals with HIV infection have active  
20 replication going on in their bone marrow. CNS obviously is  
21 a potentially worrisome site of sequestration, as is genital  
22 tissue, and a number of others have been suggested as a  
23 possibility, including lung, myocardium etc.

24 One slide that somehow I lost out of the deck here

1 was to point out likely situations where the plasma may or  
2 may not reflect what is going on in the lymphoid tissues.  
3 Several of these possible situations include very early  
4 after seroconversion. What we thinks happens early on,  
5 after someone seroconverts is that mononuclear cells  
6 producing virus in the FDCs are slowly loaded over a period  
7 of days to weeks, possibly months. So in that early phase  
8 we would expect the lymphoid tissue burden to be relatively  
9 much lower compared to the plasma viral burden.

10           Second, that may also be the case in very late  
11 AIDS. It has been reported that lymph node architecture is  
12 quite destroyed in end-stage AIDS patients and, therefore,  
13 the follicular dendritic cell network that holds onto virus  
14 is going to be incapable of holding a significant viral  
15 burden theoretically. I think this is still open to  
16 question. End-stage AIDS may be another problem in trying  
17 to correlate these two.

18           There are also the issues of penetration of drug  
19 into lymphoid tissue and other tissues. So when I am  
20 showing these treatment trials, I think one cannot  
21 necessarily generalize across all potential antiretrovirals.  
22 It undoubtedly depends on drug penetration and also perhaps  
23 on individual metabolic features in different compartments.

24           Joe Wong and Doug Richman, at the January

1 retrovirus meeting, reported a couple of cases of  
2 individuals who had been on potent triple therapy for a  
3 period of time, then had a lapse in their therapy of several  
4 days and then had their lymphoid tissues and their plasma  
5 sampled two or three weeks after that. They found a  
6 disjunction between the plasma load and the lymphoid load,  
7 and the plasma load was uncharacteristically low compared to  
8 the lymphoid burden. They proposed at that point a short  
9 lapse in therapy perhaps resets the clock or, if you will,  
10 reloads the follicular dendritic cells.

11 Finally, there are reports in the literature of  
12 variable resistance patterns in body compartments, and I  
13 think that needs to be worried about as well.

14 (Slide)

15 In conclusion, as I said, the lymphoid tissue HIV-  
16 1 RNA is much greater than the plasma viral burden if you  
17 use comparable denominators.

18 The better cross-sectional series of untreated or  
19 lightly treated patients show in general a correlation  
20 between these two compartments. That is to say, when the  
21 lymphoid tissue burden is high the plasma burden is higher  
22 and vice versa.

23 There are occasional outliers and, as I mentioned,  
24 there are exceptional situations that are likely to exist.

1 The plasma RNA, as we have seen in the few studies that have  
2 been published to date in small numbers of individuals, in  
3 suboptimal treatment doesn't always seem to correlate, or at  
4 least in one of the studies there was some apparent  
5 correlation.

6 (Slide)

7 In individuals that are given potent combination  
8 therapy there seems to be parallel declines in the viral  
9 burden in the lymphoid tissue and the plasma. This is  
10 probably because there seems to be equilibrium between the  
11 lymphoid viral pool, both the FDCs and the mononuclear  
12 cells, and the plasma.

13 Finally, the lymphoid tissue HIV burden probably  
14 correlates with peripheral mononuclear cells in the blood,  
15 but the plasma RNA and the lymphoid tissue proviral DNA  
16 burden may not correlate. Then there is the question of  
17 other tissues besides lymphoid tissue. Thank you very much.

18 DR. HAMMER: Thank you. We are running a little  
19 behind but before we move on to the pediatric data, are  
20 there any pressing questions for Dr. Cavert from the  
21 Committee? If not, thank you very much.

22 I think Dr. Cvetkovich, from the FDA, is going to  
23 introduce the pediatric section of the program.

24 **Introductory Comments on Pediatric Data**



1 DR. CVETKOVICH: Except for the next three studies  
2 in this presentation at this meeting, we have been drawing  
3 exclusively from studies conducted in adults. However, the  
4 picture of the actual history and the response to treatment  
5 derived from viral load studies in HIV-infected infants and  
6 children is now beginning to emerge. While it may be  
7 premature to draw firm conclusions from these data, only now  
8 may we begin to evaluate what may be important similarities  
9 and differences between the adult experience and that found  
10 in the pediatric population.

11 We are very pleased to include in our meeting the  
12 next three speakers who will address the topic of viral load  
13 in HIV-infected infants and children. Dr. Lynne Mofenson,  
14 from the National Institute of Child Health and Development,  
15 will present current natural history data. Then Dr.  
16 Palumbo, from the University of Medicine and Dentistry of  
17 New Jersey, will present virology from ACTG 152. Finally,  
18 Dr. George Johnson, from Medical University of South  
19 Carolina, will present very recent virology results from  
20 ACTG 300. Dr. Mofenson?

21 **Natural History, Lynne M. Mofenson, M.D.**

22 DR. MOFENSON: I have been asked to summarize the  
23 results of the pediatric literature in 20 minutes. So what  
24 you are going to have is me speaking very rapidly.

1 (Slide)

2 Just to review what we see in infection in adults,  
3 with primary infection in adults there is an initial burst  
4 in viral burden with a peak of about a million, but after  
5 several weeks to months it declines to under 10,000 and most  
6 levels remain low or undetectable for many years, and RNA  
7 changes as well as CD4 changes together have been shown to  
8 be independently predictive of prognosis.

9 (Slide)

10 What is different about children that could affect  
11 the natural course of HIV viremia? First of all, HIV  
12 infection in most children perinatally infected is primary  
13 infection. Most children are negative in culture and  
14 virologic tests at birth and become positive only after  
15 about a week of age. So perinatal infection is really  
16 primary infection.

17 Uninfected newborns are immature both in cellular  
18 and humoral immunity. So we now have primary infection  
19 occurring in immunologically naive infants. Additionally,  
20 in normal newborns there is an increase in activation  
21 markers on CD4 and CD8 cells because they are rapidly  
22 expanding and differentiating, making them potentially more  
23 infectable by HIV. Infants also have a higher absolute  
24 lymphocyte number and higher CD4 levels than seen in adults

1 which, again, gives you more of a population to infect, and  
2 the CD4 cells are primarily naive.

3           So based on this, one might anticipate that there  
4 might be some differences in the natural history of RNA in  
5 perinatally infected infants.

6           (Slide)

7           I am going to review the published studies. The  
8 first study is by Paul, sitting over there, and this study  
9 was published in 1995 and looked at serial plasma samples  
10 during first year of life from 14 infected infants who were  
11 involved in a prospective natural history study. This was  
12 based on whole blood in heparin. RT PCR was used to  
13 measure. The mean RNA copy number in these infants was over  
14 500,000. The decrease over the first year of life was only  
15 2- to 10-fold less than 1 log, very much in contrast to what  
16 you see with primary infection in adults. Additionally,  
17 despite the high RNA levels, most of these infants had age  
18 appropriate CD4 counts and there was both rapid as well as  
19 slow progression.

20           (Slide)

21           This slide basically shows the data. As you see,  
22 levels remain high for a year and well near a million  
23 copies.

24           (Slide)

1           Moving along to the recently published data from  
2 the Women and Infants Transmission Study, this is a  
3 prospective natural history study of infected pregnant women  
4 and their infants from 6 different sites in the states that  
5 are listed on the slide. Blood is collected serially in  
6 infants, and the population that was reported in the recent  
7 New England Journal was perinatally infected singleton  
8 infants, born between 1990 and 1993 who had more than 12  
9 months of follow up.

10           (Slide)

11           There were 673 plasma specimens on 106 children,  
12 giving over 6 samples per child. Again, this was  
13 heparinized blood specimens with RT PCR, with RNA extraction  
14 via the Boom technique. The lower limit of detection for  
15 this assay was 400.

16           (Slide)

17           This just goes to show you the plasma HIV-1 RNA  
18 copy number by age. What you see is that there are very  
19 high copy numbers with the first few weeks of life.  
20 Although these copy numbers declined, they declined only  
21 very slowly, and most children have levels over 100,000  
22 during not only the first year of life but also only slowly  
23 declined to under 100,000 by three years of life. So early  
24 peak, slow decline.

1 (Slide)

2 This looks at rapid versus slow progression of  
3 disease in these children. Basically, what you can see is  
4 that children who had rapid progression, which meant that  
5 they developed AIDS or died before 12 months of age, had  
6 higher HIV-RNA copy number at almost data points with  
7 exception of the birth specimens.

8 (Slide)

9 This looks at a Kaplan-Meier curve based on  
10 whether or not the infant's median RNA level during the  
11 first 2 months of life was above or below the median. The  
12 median was nearly 300,000. You can see in adults RNA copy  
13 numbers significantly correlated with risk of disease of  
14 progression. These are kids who were above the median, a  
15 significant number progressed, versus those who were below  
16 the median. The difference one sees in adults is that the  
17 levels associated with progression are significantly  
18 different. We are not talking about levels of 10,000, we  
19 are talking about levels over 300,000.

20 (Slide)

21 This slide shows you the first 4 months of life in  
22 these children. Again, RNA copy numbers in progressors,  
23 which is here, were higher than in non-progressors.  
24 However, the important thing to look at is that there is

1 significant overlap for individual children between  
2 progression and non-rapid progression.

3 (Slide)

4 So basically in the WITS study RNA levels  
5 increased rapidly until about 1 to 2 months of age and then  
6 slowly declined through 2 years of age, and reached a median  
7 of about 35,000 copies at age 24 months, in distinct  
8 contrast to what we see in adults. This pattern was  
9 observed in both rapid progressors and non-rapid  
10 progressors, and was also observed in kids who had cultures  
11 positive at birth versus those who had cultures positive  
12 later.

13 (Slide)

14 Infants with rapid progression had higher peak RNA  
15 levels during the first 2 months of life than those with  
16 non-rapid progression. Infants with rapid progression also  
17 had higher mean geometric RNA levels during the first year  
18 of life. However, because of overlap between rapid and non-  
19 rapid progressors, there was no single threshold level that  
20 could be identified that would be very predictive of  
21 progression. However, it should be noted that infants who  
22 had RNA levels under 70,000 to 80,000 at 1, 2 and 4 months  
23 of age had rapid progression.

24 (Slide)

1                   Moving right along to the next study, Penn  
2 McIntosh and colleagues from Boston Children's Hospital  
3 recently published two articles. They looked at children  
4 with perinatal infection, followed in their site for  
5 clinical care between 1986 and 1993. The children were seen  
6 every 3 to 6 months, and they took stored sera from excess  
7 samples that were processed rapidly and stored at minus 70  
8 degrees, and again used RT PCR.

9                   (Slide)

10                  They looked at 48 children who had a mean age of  
11 29 months or a little over 2 years at first visit, and  
12 looked at them over the subsequent 2 years. There were  
13 about 6 samples per patient. Most of these children had  
14 received therapy at some point during the study.

15                  (Slide)

16                  This is very similar to what one saw in the  
17 Shearer paper. Basically, this is looking at a cross-  
18 sectional look. As you can see, RNA copy numbers were very  
19 high, over 100,000, in the first year of life and showed a  
20 consistent climb over time, persisting for long periods of  
21 time, many years, and this was seen regardless of treatment  
22 or non-treatment. The untreated kids are in the little  
23 triangles and the treated kids are in the little circles.

24                  (Slide)

1           This slide shows you basically the same thing  
2 except looking at slope. This is a negative slope, here, a  
3 positive slope here and zero slope. So you can see that  
4 there is negative slope for about the first 4 to 5 years of  
5 life and then a slow increase in RNA, more consistent with  
6 what one sees in adults.

7           (Slide)

8           So to summarize the McIntosh study, there was a  
9 slow consistent decrease in RNA from about 1 to 5 years of  
10 age that was about 1 log. RNA levels were highest in the  
11 youngest children. The slope was about -0.1 to -0.2 log per  
12 year. The set point was not reached until infants were  
13 nearly 5 years old, and this decline was seen regardless of  
14 degree of immunodeficiency or receipt of antiretroviral  
15 therapy.

16          (Slide)

17          This just shows you their next study, which was  
18 published just a month or two ago, that correlated RNA copy  
19 number with weight for age Z score and CD4 for age Z score,  
20 and this is just to show you that there was a significant  
21 correlation as RNA copy number increased, and these are  
22 different quartiles for the children, so did the weight for  
23 age Z score. So children had an increased level of weight  
24 loss as RNA copy number went up.



1 (Slide)

2 The same type of thing is seen for CD4, although  
3 it is not as well correlated with weight.

4 (Slide)

5 I would like to move on to the study I know the  
6 best, which is our recent publication from the Journal of  
7 Infectious Diseases, where we looked at stored specimens  
8 from the NICHD IVIG clinic file. This was a clinical trial  
9 of IVIG versus albumin placebo that occurred between 1988  
10 and 1991. We had 376 children enrolled in this trial, most  
11 of them perinatally infected. We obtained vital status  
12 updates on these children through September, 1996. So we  
13 had approximately 5-plus years of follow up on the children.

14 (Slide)

15 Blood was central storage at entry and every 3  
16 months during the study. It was stored at minus 20 to minus  
17 70 and then shipped to a central repository. Specimens from  
18 children who had more than one sample were retrieved and  
19 tested, and here HIV-RNA was measured by the NASBA assay by  
20 one ACTG certified lab.

21 (Slide)

22 This shows you what the patients were like. We  
23 had 254 patients that were in this study that were a mean  
24 age of 3.4 years, mean entry CD4 count a little over 1,000,

1 mean entry CD4 percent 20 percent. This is pretty much  
2 normal for children that are 3 years.

3 (Slide)

4 This slide shows the distribution at baseline. As  
5 you can see, there was a wide range of HIV-RNA copy numbers  
6 in these children. The median RNA copy number in these 3-  
7 year old kids was 105,000. The geometric mean was about the  
8 same. Only 8 percent of these children had undetectable  
9 levels; only 16 percent were under 10,000; a full 50 percent  
10 were over 100,000 and 2 percent were over 500,000, very  
11 different than one would see in a relatively asymptomatic  
12 adult population.

13 (Slide)

14 This shows you again the longitudinal geometric  
15 mean by age. You should be pretty used to this picture,  
16 very high levels early in life, near a million; slow decline  
17 that continues for a considerable period of time. The most  
18 rapid decline is within the first 2 years of life.

19 (Slide)

20 This shows you the same graph by age but in  
21 children who died versus those who did not die. As you can  
22 see, those children who died had significantly higher RNA  
23 levels at almost all ages than those children who did not  
24 die during follow up, although both of them showed some

1 decline.

2 (Slide)

3 This shows you the gradient of mortality with  
4 increasing RNA. Again, this is 5-year mortality. With  
5 children with RNA copy number under 10,000 the mortality  
6 rate at 5 years was 23 percent. For those between 10,000  
7 and 100,000 the mortality rate was similar, 24 percent.  
8 Between 100,000 and a million increased to 40 percent, and  
9 over a million it was over 70 percent.

10 (Slide)

11 This slide breaks it down into smaller categories.  
12 You can see that as RNA increases there is an increase in  
13 the risk of mortality. But really the sharp increase  
14 occurred at about 100,000. Just to note, for those children  
15 who were undetectable, 24 percent died by 5 years. Between  
16 4,000 and 50,000 the mortality rate was 28 percent, and  
17 50,000 to 100,000 was 15 percent. Then when you jump up to  
18 over 100,000 it goes up to 40 percent. So there didn't  
19 appear to be much differentiation between those children who  
20 had undetectable levels and those children who were about  
21 100,000.

22 (Slide)

23 This is a Kaplan-Meier plot broken into the same  
24 categories of under 10,000, 10,000 to 100,000, 100,000 to a

1 million and over a million. What you can see is that there  
2 is a difference in terms of the Kaplan-Meier plot. However,  
3 the under 10,000 children and the children between 10,000  
4 and 100,000 are not statistically significantly different  
5 from each other. This one, 100,000 to a million is  
6 different than these two and this one is, as you can see,  
7 clearly different.

8 (Slide)

9 This is an ROC curve. I just want to point out  
10 the difference in terms of sensitivity in false-positive  
11 rate depending on what threshold one picks to look at. For  
12 example, if one picks a threshold of 10,000 to define risk  
13 for mortality, you will be very sensitive and most children  
14 who are above 10,000 will have a death. However, you also  
15 have a significant false-positive rate, nearly 80 percent,  
16 and that means that 80 percent of children with levels over  
17 this much do not die. If you compare to the highest level,  
18 a million, the false-positive rate is much less but the  
19 sensitivity is also less. At about 100,000 sensitivity was  
20 about 80, false-positive rate was about 40.

21 (Slide)

22 This is to show you that again there was a  
23 difference by age. This breaks kids down to under 2 years  
24 and over 2 years. As you can see, in the under 2 children

1 one does not see a significant increase in 5-year risk of  
2 mortality until you hit a million copies, whereas in  
3 children who are over 2 you begin to see this rise at about  
4 100,000. So there is probably a point in between birth and  
5 6 years or so where the RNA predictive value changes from  
6 very high levels to lower levels. I think Paul will show  
7 this data as well.

8 (Slide)

9 This is just to show you CD4 was also significant.  
10 It correlated particularly in the lower CD4 range. For each  
11 5 percent drop you saw a rapid increase in mortality.

12 (Slide)

13 This shows you the correlation between HIV-1 RNA  
14 and CD4 percent. There was a borderline statistically  
15 significant relationship between the two. The r value was  
16 minus 0.4. This is very similar to the data published by  
17 Mellors.

18 (Slide)

19 This is a multivariate analysis looking at  
20 baseline RNA and baseline CD4 percent and their correlation  
21 with long-term risk of mortality. As you can see, they were  
22 both independently predictive with about a 2.5-fold  
23 increased risk of mortality for every log increase in RNA  
24 and about a 1- to 3-fold increase in risk of mortality for

1 every 5 percent decline in CD4 percent.

2 (Slide)

3 This just looks at change over time. Basically  
4 the data is the same for every log increase in RNA during  
5 the course of the study. Mortality risk increased about 2-  
6 to 8-fold and the CD4 percent increase was about the same as  
7 before.

8 (Slide)

9 This looks at the predictive value by baseline RNA  
10 quartile. This is RNA quartile 1, the lowest quartile, and  
11 this is the highest quartile. And now we have looked at  
12 mortality risk when you break it down by CD4 count. My main  
13 point here is that no matter what RNA level, CD4 count was  
14 still predictive of increased disease progression. It did  
15 increase mortality.

16 (Slide)

17 However, when you break it down and you look at it  
18 by CD4 percent, less than 15, 15 to 25 and over 25, in the  
19 children who were most immune suppressed the mortality risk  
20 appeared similar regardless of whether you had nearly  
21 undetectable RNA or very high copy number, and it was only  
22 in the children who had a better immune status where RNA  
23 levels showed the step-wise increase in terms of predicting  
24 disease progression.

1 (Slide)

2 This broke down HIV-RNA into different  
3 combinations. The most favorable combination was RNA under  
4 100,000, CD4 over 15 percent and the 5-year mortality risk  
5 in these children was 15 percent. As you can see, there is  
6 a step-wise increase, and when you get to bad parameters for  
7 both variables, over 100,000, less than 15 percent, the  
8 mortality rate at 5 years was a whopping 81 percent.

9 (Slide)

10 This shows you the Kaplan-Meier for those  
11 categories. The main thing I want to point out here is that  
12 when you use RNA and CD4 together you get better  
13 differentiation in terms of prognosis than when one uses RNA  
14 alone.

15 (Slide)

16 In conclusion, what can we say about the natural  
17 history? Well, high peak levels, over a million, are  
18 reached by 1 to 2 months of age in perinatally infected  
19 infants. In contrast to the several log fall over about 6  
20 months that one sees in adults, the RNA decline in  
21 perinatally infected infants is less than 1 log during the  
22 first year of life. RNA also continues to decline over the  
23 first several years of life regardless of therapy, and not  
24 appearing to reach a zero set-point slope until about 5

1 years of age, and when it does, it appears to be higher than  
2 that reported in adults.

3 (Slide)

4 As in adults, higher levels of RNA were associated  
5 with increased risk of mortality and also disease  
6 progression. You will see that in the next two  
7 presentations. However, similar to the CD4 threshold for  
8 PCP prophylaxis, RNA levels indicative of increased risk of  
9 mortality or disease progression appeared to be higher in  
10 infected children than has been reported in infected adults.

11 Just a comment, however, infected children  
12 progress more rapidly than adults. One of the reasons for  
13 their more rapid progression may be due to the higher RNA  
14 levels that they have.

15 Finally, RNA and CD4 when used together were  
16 better predictors of progression risk in terms of mortality  
17 than either used alone. Thank you.

18 **Response to Treatment: ACTG 152, Paul Palumbo, M.D.**

19 DR. PALUMBO: I want to thank the Committee for  
20 giving me the opportunity to present ACTG 152 and for  
21 recognizing that pediatrics actually is an important  
22 consideration at this meeting, with some very unique  
23 characteristics which Lynne very nicely outlined for us.

24 (Slide)



1           This is a relatively old study, between 1991 and  
2 1993. The entry criteria were quite broad and liberal, such  
3 that the majority of infants in our clinics were eligible  
4 for this particular protocol. The organizers of this  
5 protocol recognized that young infants were going to be  
6 considerably different than older children.

7           There were approximately 830 infants who were  
8 enrolled into this protocol over a 2-year period. They were  
9 stratified by age, realizing that young infants were going  
10 to have significantly different progression than older  
11 infants. The stratification was below 30 months and above  
12 30 months. This was a double-blind, randomized, placebo-  
13 controlled trial of monotherapy with 2 nucleosides, ZDV and  
14 ddI, compared with combination therapy.

15           (Slide)

16           It is important to consider the endpoints in this  
17 trial. They were entirely clinical and were composed of  
18 either time to death or time to first HIV disease  
19 progression. Far and away the majority of clinical  
20 endpoints were either weight-growth failure or CNS  
21 abnormalities, which is very typical of pediatrics. In some  
22 respects we are fortunate to have these endpoints that are  
23 very definitive and well defined in this population.

24           (Slide)

1           This is a Kaplan-Meier plot of the entire cohort  
2 for data through November, 1994. At this point a DSMB  
3 review recognized that the ZDV arm, which is in green, was  
4 performing less well than the other two arms, the ddI  
5 monotherapy and the combination arm, in yellow and black.  
6 The ZDV arm was stopped at that point but the other two arms  
7 were continued in a blinded fashion.

8           (Slide)

9           This is the data for the less than 30 months age  
10 group, which shows a more dramatic separation of the ZDV arm  
11 compared to the other two arms. There was no clinical  
12 difference between the ddI monotherapy arm and the  
13 combination arm.

14          (Slide)

15          That is a finding that was very similar to ACTG  
16 175 that Scott Hammer led for the adults, which is a very  
17 similar trial to this pediatric trial.

18          (Slide)

19          I mentioned that there were about 830 children  
20 enrolled in ACTG 152, and 579 are included in this virologic  
21 analysis. There were actually 566 who had both baseline RNA  
22 and CD4 results. So about 3/4 of the children in the study  
23 are represented in this virologic analysis that I will  
24 present today, and they were very representative of the

1 entire cohort. There were no significant differences  
2 between this subset that I will show you and the overall  
3 cohort. There were 311 under 30 months of age. We had  
4 about 1,500 specimens or 5 specimens per child in that  
5 group, and 268 in the children greater than 30 months of  
6 age.

7           The bottom two boxes here represent the baseline  
8 plasma RNA results. These are in untreated infants and  
9 children. So this was basically a natural history study,  
10 following up on what Lynne Mofenson just presented to us.  
11 In the top box I have eliminated the 19 individuals who had  
12 negative or undetectable RNA at baseline. Those individuals  
13 are represented in the lower box, if anyone wants to look at  
14 that particular one.

15           Again you see very high median baseline RNA  
16 results for infants under 30 months of age, 620,000, a mean  
17 of 5.7 log. For comparison purposes, in ACTG 175, the adult  
18 study, was similarly designed. The mean baseline RNA  
19 results were 4.2 log, so about 1.5 log lower than our  
20 infants who were less than 30 months of age. We see about a  
21 10-fold decrease in baseline RNA results for children over  
22 30 months of age, where the median was 61,000, the mean was  
23 4.8 log.

24           (Slide)

1           Graphically depicted through 18 years of pediatric  
2 untreated experience at baseline, you see the data presented  
3 here. Very high levels, over a million, in the early phases  
4 of the study, bottoming out or plateau-ing at around or a  
5 little below 100,000 in older children.

6           (Slide)

7           Depicted a little differently here by age groups,  
8 I do this for two purposes. One, to show the number of  
9 infants in each age group, and we had 164 infants who were  
10 less than or equal to 12 months of age in this study. That  
11 is a very high number to be studying. Again, you can see  
12 very high baseline RNA levels at entry in the youngest  
13 infants, greater  $10^6$ , and staying at or above  $10^5$  generally  
14 throughout the age groups that were entered into the study.

15           (Slide)

16           This graph depicts the mean changes from baseline  
17 in log RNA copies/mL at defined time points. In ACTG 152  
18 the first plasma specimen we had available to study after  
19 starting therapy was at 24 weeks, or about half a year. So  
20 we were missing a lot of potentially informative data before  
21 24 weeks. Nonetheless, it is quite interesting. This is  
22 the entire data set of all age groups combined and we do see  
23 differences between the 3 arms, significant differences, at  
24 all these time points between 24 weeks and 96 weeks.

1           The mean changes from baseline are very similar to  
2 adults, using similar antiretrovirals. Even though in  
3 pediatrics we are starting at a much higher level, we are  
4 seeing quantitatively similar decreases in pediatric  
5 patients.

6           (Slide)

7           This is a very similar plot for the subgroup who  
8 are less than 30 months of age. We see the most dramatic  
9 decreases in this population. At 24 weeks we are seeing 0.3  
10 log reduction in the ZDV monotherapy, about 0.6 logs in the  
11 ddI monotherapy arm, and 0.9 log in the combination arm.  
12 These are modest increases but they are associated with  
13 significant clinical benefit.

14          (Slide)

15          In the group over 30 months of age we are not  
16 seeing quite as much difference between the treatment arms.  
17 We do see significant differences at 48 and 72 weeks when  
18 the poorest performing ZDV arm is compared with the other  
19 two arms. Otherwise we see much more of a general trend  
20 between the three arms.

21          (Slide)

22          There was more extensive data for the ddI and  
23 combination arms. Those were continued in a blinded fashion  
24 for a longer period of time. We have data out to 144 weeks.

1 As you recall, we saw no clinical differences in the  
2 performance of these two arms but we did see significant  
3 differences between week 24 and week 72 in the reduction in  
4 RNA, with the combination arm performing better than the ddI  
5 monotherapy arm. This also was seen in ACTG 175. No  
6 clinical differences between these two arms in adults, but  
7 differences in RNA. So this is the group less than 30  
8 months of age.

9 (Slide)

10 Now, some Kaplan-Meier plots for tracking  
11 progression-free survival based on the baseline RNA. We  
12 divided the baseline RNA into 4 quartiles, 4 fairly even  
13 quartiles, undetectable to 150,000 being the lowest  
14 quartile, very different than what we see in any of the  
15 adult studies obviously; 150,000 to half million in the  
16 second, half million to 1.7 million in the third quartile,  
17 and greater than 1.7 million. And there is a very nice  
18 correlation between baseline RNA and ultimate course.

19 The other point here is that this is a very linear  
20 increase in risk with increase in RNA, with no evidence of  
21 any threshold above or below which there are dramatic  
22 differences in disease-free progression.

23 (Slide)

24 These are the children over 30 months of age. The

1 same can be said for relative linear effect. You can see  
2 that the quartiles are 10-fold less now than they were in  
3 the infants under 30 months of age. The quartiles are more  
4 like adult quartiles now as we get into the older age group  
5 of children. There were no endpoints seen in the 66  
6 children who were in the lowest quartile. So there may be  
7 some argument made for some threshold effect in the older  
8 children, but in general a linear effect of RNA with risk.

9 (Slide)

10 Another way of evaluated the issue of threshold in  
11 the infants under 30 months of age was to divide the RNAs  
12 into octiles if there was any information lost in the  
13 quartile information. But when one plotted baseline RNA  
14 with the hazards ratio we saw a very linear effect across  
15 the broad range of values that were present at baseline in  
16 infants under 30 months of age.

17 (Slide)

18 We also looked at hazards ratios or the Kaplan-  
19 Meier plot looking at the week 24 RNA, the absolute value  
20 obtained after 24 weeks of therapy. Again we divided it  
21 into quartiles. This is children over 30 months of age.  
22 What we start to see here is a threshold effect. That is,  
23 if you fell into the lowest quartiles, that is, less than or  
24 equal to 50,000, the risk was extremely small, 10 percent

1 less relatively speaking, for 24 month progression-free  
2 survival after that 24-week time point. If one got above  
3 50,000 the risk increased relatively dramatically for that  
4 particular population.

5 I don't have a similar plot for the infants under  
6 30 months of age but, again, the quartiles are higher and  
7 there is a threshold effect at about 100,000, with  
8 relatively low risk for progression or death in infants who  
9 fell under 100,000.

10 (Slide)

11 One of the questions that we wanted to address  
12 is, is pathogenesis or risk in older pediatric populations  
13 with lower viral load going to be different than younger  
14 children, that is infants, with higher viral load?

15 What I show you here are hazards ratios, and these  
16 were taken from the Kaplan-Meier plot. So this is the  
17 percent failure at 2 years plotted against RNA. The infants  
18 under 30 months of age are in the solid line and the  
19 quartiles are presented by the children. The children over  
20 30 months of age are in the dotted line and the quartiles  
21 are represented by the dots. What you can see is overlap  
22 here. Obviously there is 10-fold RNA in the younger infants  
23 than in the older children, but at a given RNA level the  
24 risk is essentially the same. So there is overlap.



1           The other point here is that in both cohorts we do  
2 see a linear effect of risk versus RNA, again without  
3 evidence of a threshold.

4           (Slide)

5           If one looks at the week 24 plasma levels and then  
6 follows these children for another 2 years after that week  
7 24 time point and calculates the percent failure at that 2-  
8 year time point, one again sees similar risk in the 2 age  
9 cohorts at similar RNA levels.

10           One also starts to see what I was alluding to  
11 before, that there may be threshold effects and that they  
12 are fairly similar for the 2 age groups. There may be  
13 somewhat higher thresholds for the younger infants, again  
14 around 100,000, and they may be lower for the older  
15 children, possibly around 50,000 or potentially a little bit  
16 lower.

17           (Slide)

18           As Lynne alluded to, RNA is not the only variable  
19 we should be considering and combining RNA with CD4 has  
20 potential for adding more power to our ability to calculate  
21 where a given individual is. All of the ACTG 152 data I am  
22 presenting today are, obviously, group data. I am not  
23 dealing with individuals. But a number of Cox proportional  
24 hazards models were generated with this data base for the

1 infants under 30 months of age, and this is what I am  
2 presenting here.

3           When RNA and age were included in a Cox  
4 proportional hazards model, RNA alone was independently  
5 predictive of outcome and of risk. The age fell out.  
6 Baseline in model 2, both CD4 and age were independently  
7 predictive of outcome. In model 3, when one compared both  
8 baseline RNA, CD4 and age. Both baseline RNA and CD4 were  
9 independently predictive of outcome in a fairly significant  
10 fashion.

11           The final model I will describe to you is model 6  
12 where all of the variables, baseline RNA, RNA attained at  
13 24, baseline CD4 and CD4 obtained at 24, are included in a  
14 complicated model. But, by and large, all 4 variables are  
15 independently predictive of where the child or where the  
16 group is going to go. We see risk reductions in the range  
17 of 40 to 50 percent for each log decrease in RNA at baseline  
18 or each log RNA decrease from baseline to week 24. These  
19 are fairly similar risk reductions to what has been  
20 described for adults in adult trials.

21           (Slide)

22           Just to show what some of our challenges are in  
23 the future, this is basically a graphic demonstration of  
24 where we are or post-ACTG 152. The top line here represents

1 baseline RNA data for infants less than 30 months of age.  
2 With nucleoside therapy as used in ACTG 152, I show a  
3 maximum of about a one log reduction in RNA, a modest  
4 reduction but clinically significant.

5           As you can see, we still have 3 to 4 logs to go to  
6 reach undetectable RNA. Whether this is going to be  
7 achievable in young pediatric populations is yet to be  
8 proven. I think it is very reasonable to expect in children  
9 over 2 or 3 years of age that they will probably perform  
10 similarly to adults on very aggressive therapy. However, it  
11 is still unsure how many of these younger infants we are  
12 going to get to undetectable RNA with aggressive therapy,  
13 and those trials are just beginning or ongoing.

14           (Slide)

15           I will just take one minute to acknowledge some of  
16 the people who played important roles in these analyses,  
17 Claire Raskino and Tim Ramacietti, at SDAC at Harvard School  
18 of Health, did most of the analysis and management of the  
19 data base. Jane England, Carol Baker and Steve Spector were  
20 the principals on protocol ACTG 152, and the four  
21 laboratories that performed the assay, and the assay that we  
22 used was the NASBA assay, are here. These are all ACTG  
23 certified laboratories. Thank you.

24           DR. HAMMER: Thank you very much. I think what we

1 will do before move to the third pediatric talk, is to take  
2 a 15-minute break here since we are running a few minutes  
3 behind. So we will return at 11:18 and start promptly.

4 [Brief recess]

5 DR. HAMMER: We are going to complete the  
6 pediatric data section. The next speaker is Dr. George  
7 Johnson of the Medical University of South Carolina, who is  
8 going to present the virologic results from ACTG 300.

9 **HIV-RNA Results: ACTG 300, George Johnson, M.D.**

10 DR. JOHNSON: I would like to thank the Committee  
11 for inviting me to speak. Actually, Dr. McKinney should be  
12 here as the protocol chair and driving force behind this  
13 protocol, but he is having a lot more fun refining his blues  
14 guitar skills in West Virginia right now.

15 (Slide)

16 This slide is to reflect the number of people that  
17 are involved in these protocols and give everyone credit.

18 (Slide)

19 ACTG 300 was substantially similar in design to  
20 ACTG 152 but we are at a much more preliminary stage of  
21 analysis of the data so our virology is going to be  
22 extremely limited and I am just going to present the data  
23 that we have, with a summary of the trial.

24 The objective was to compare AZT-3TC combination

1 therapy with the better of ddI or ddI plus AZT therapy with  
2 respect to progression, with a decision being made as to  
3 which of the two ddI-containing arms should be determined  
4 based on the clinical results of 152. Actually, the AZT-ddI  
5 dual therapy arm was discontinued in May of 1996 on the  
6 basis of the clinical comparison of the 152 results. That  
7 was prior to our virology data being completely available.

8           The patients were very similar. They were  
9 symptomatic HIV-infected children, and they had to have less  
10 than 8 weeks of prior antiretroviral therapy. This was also  
11 slightly different from 152 initially in that this allowed  
12 for perinatal AZT therapy, and they were between 7 weeks and  
13 15 years of age.

14           (Slide)

15           The endpoints were very similar but there were  
16 some minor differences, primarily related to neurologic  
17 endpoints. But the principal endpoints were development of  
18 a new category C disease or death, inadequate weight growth  
19 velocity of deterioration neurologically.

20           (Slide)

21           As of April 4, 1997 when the data set was frozen  
22 for DSMB analysis, there were 615 children who had been  
23 enrolled in the study; 596 were able to be evaluated, and  
24 the data was on the comparison of ZDV-3TC or ddI with 471

1    evaluatable children.  The data was current within 2 months  
2    for 98 percent of the patients on treatment, and the median  
3    follow-up at that point was 9.4 months.

4                   (Slide)

5                   This is just to show the similarity of the  
6    population in the 2 treatment arms that were analyzed for  
7    the DSMB and to give you baseline data relative to CD4  
8    counts, which were very comparable, and median RNA log  
9    titers, which were also comparable.

10                   (Slide)

11                   The primary analysis was on the time to clinical  
12    progression or death, and there was a statistically  
13    significant difference between the arms, the top solid arm  
14    being the ZDV-3TC, the bottom being ddI.

15                   (Slide)

16                   This just gives you a breakdown of the number of  
17    endpoints.  There was a total of 53 endpoints which  
18    contributed to that analysis, 38 in the ddI arm, 15 in the  
19    ZDV-3TC arm.  As you can see, there was also a large  
20    difference in death as a first endpoint, but a large number  
21    of the endpoints, similar to 152, were CNS deterioration and  
22    weight-growth failure.

23                   (Slide)

24                   This is the breakdown and stratum looking at

1 children under age 3 relative to time to clinical  
2 progression or death. Again, this is more impressive than  
3 the total set and, actually, 44 of the clinical progression  
4 or death endpoints were in this subgroup or stratum of the  
5 population. That was 83 percent of the endpoints.

6 (Slide)

7 This is looking at the same data relative to those  
8 in the over 3 age stratum, and there was not a significant  
9 difference. Looking just at survival for the whole  
10 population, you can see there was a significant difference  
11 in survival in favor of the ZDV-3TC arm.

12 (Slide)

13 I am going to through this very fast because this  
14 is just weight growth presented as Z scores for these, with  
15 zero being normal growth. You can see that even with the  
16 ddI arm, which is the lower arm, there is some benefit in  
17 that it increases above zero, and the ZDV-3TC arm did better  
18 as far as weight growth than the ddI monotherapy arm.

19 (Slide)

20 Similarly for height Z scores, although there was  
21 much more variability in measuring height, particularly in  
22 young children--and the data presented is only for the under  
23 3 stratum--there is a similar trend in the over 3 stratum  
24 but it is not nearly as impressive.

1 (Slide)

2 Probably data that people want to look at more is  
3 on CD4 counts from baseline. This is a change in absolute  
4 CD4 count from baseline, with the bottom line being ddI  
5 monotherapy, zero being no change from baseline; and the top  
6 line, solid, and being ZDV plus 3TC, and there was a  
7 substantial increase, particularly early on, but it was  
8 sustained out through 48 weeks of therapy. This is under 3  
9 years of age.

10 (Slide)

11 This is the same data presented as percentage,  
12 which is more stable in the young infants. There is a  
13 gradual decrease over the first several years of life in CD4  
14 counts, as has been presented. So this is CD4 percent,  
15 younger stratum, and similar in the older stratum, both  
16 favoring the ZDV-3TC arm.

17 (Slide)

18 This is the mean change in viral RNA on a log 10  
19 scale comparing the ZDV and 3TC. There was a sustained  
20 difference out to 36 weeks, again favoring the ZDV-3TC arm.

21 (Slide)

22 This is including all of the patients that were  
23 simultaneously enrolled to the first 3 arms prior to  
24 discontinuation of the ZDV-ddI arm, and comparing them for



1 time to clinical progression. You can see that the ZDV-3TC  
2 and ZDV-ddI arms, at the top, overlap and were  
3 indistinguishable, both being substantially better than the  
4 ddI monotherapy arm.

5 (Slide)

6 Survival was similar in the 2 dual therapy arms  
7 compared to the ddI monotherapy arm, which is different than  
8 what had been found in 152 and we don't have an answer at  
9 this point yet as to what the difference was.

10 (Slide)

11 This is looking at the viral load data including  
12 the ZDV-ddI dual therapy arm, which resulted in levels  
13 between the ZDV-3TC and ddI monotherapy effect. This is  
14 again limited to the patients who were simultaneously  
15 enrolled in all 3 arms.

16 (Slide)

17 So our conclusions are that ZDV plus 3TC was  
18 superior with respect to HIV disease progression, survival,  
19 weight and height growth rates, and CD4 and RNA changes.

20 These effects were sustained through 48 weeks of  
21 therapy. There was a mean increase in CD4 cells of 125 to  
22 190 cells, and a mean increase between 0.7 to 0.9 logs of  
23 viral RNA.

24 There were no significant differences in toxicity

1 between the arms. There was a slightly higher rate of  
2 hepatotoxicity in the ddI monotherapy arm.

3 (Slide)

4 When patients were randomized to all 3, the 2 dual  
5 therapy arms were superior to the monotherapy arm with  
6 respect to HIV disease progression and survival.

7 The differences between this study and 152 may be  
8 due to slight differences in progression criteria, and that  
9 will be looked at; some other factors which we really don't  
10 have any handle on; and chance.

11 The combination therapy should be preferred to  
12 monotherapy for antiretroviral therapy of infected children,  
13 particularly under 3.

14 And I have lost the last slide. The other thing I  
15 think I have already mentioned so we can just end it and see  
16 if there are any questions.

17 DR. HAMMER: Thank you very much. I would just  
18 ask the panel if there are any immediate, pressing questions  
19 for Drs. Mofenson, Palumbo or Johnson on the pediatric data.  
20 Dr. Feinberg?

21 DR. FEINBERG: I just have one clarification  
22 question. For the ACTG 152 group, what was the eligibility  
23 criterion around prior antiretroviral therapy?

24 DR. PALUMBO: The ACTG 152 pediatric population

1 essentially had to ...[not at microphone; inaudible.]

2 DR. HAMMER: Any other questions? Chris?

3 DR. MATHEWS: Also for Dr. Palumbo, on the 152  
4 data, where you presented those Cox models with main effects  
5 for CD4 change and RNA change at, I think, 24 weeks, did you  
6 look in the analysis at discordant responders, or what was  
7 the prevalence of discordant responses between those 2  
8 markers? In other words, CD4 going up, viral load going up  
9 or the opposite?

10 DR. PALUMBO: Those analyses are ongoing. Other  
11 than the standard analyses presented, the other analyses  
12 looking at trends and counter-trends, are ongoing. So we  
13 don't have a look at whether RNA going one direction and CD4  
14 in another are issues that we see commonly or uncommonly in  
15 this pediatric population.

16 DR. EL-SADR: It seems to me that looking at the  
17 contrasting pediatric experience and the adult experience is  
18 that the clinical events, primarily CNS and growth events,  
19 happened pretty rapidly. You can see differentiation in  
20 these trials pretty early. So it is different from the  
21 adult clinical trials where you have to continue the trial  
22 for a very long time until you get the clinical events. How  
23 much of an advantage is it to look at the 24-week viral  
24 endpoint event versus 36 weeks clinically then? I mean, did

1 you actually try to look at it this way?

2 DR. JOHNSON: I can say for 300 that we haven't  
3 yet looked at that. This is about what we have. The  
4 virology that we have is about 70 percent of what needs to  
5 be done.

6 DR. PALUMBO: Yes, I would concur with George.  
7 ACTG 300 ended rather abruptly and quickly from start to  
8 finish with clinical endpoints only. The problem I think we  
9 are going to see though, as we progress in pediatrics and as  
10 we have seen with our adult colleagues, is that as we are  
11 able to lower viral load to lower quartiles, shall we say,  
12 the risk is going to become much lower for disease  
13 progression and I think in pediatrics, hopefully, we will  
14 lose the ability to use clinical endpoints routinely in very  
15 aggressive protocols. So I think we are really going to  
16 need to use CD4, RNA and any other laboratory markers we can  
17 find to better define how clinical regimens are working or  
18 not working in future clinical trials.

19 DR. JOHNSON: That is ongoing in the pediatric  
20 ACTG, that transition to using virologic and other surrogate  
21 markers.

22 DR. HAMMER: I just have one quick question for  
23 Dr. Johnson. One thing that is striking in the ddI arm  
24 therapy is the 0.2 to 0.3 reduction at week 12, a lower

1 reduction than one might expect as seen in other studies  
2 with ddI. Is there any drug adherence data? One wouldn't  
3 expect that much in the way of resistance.

4 DR. JOHNSON: That was collected but we don't have  
5 that analyzed yet. That was collected by report so there  
6 was no monitoring of levels on a compliance type of basis.

7 DR. HAMMER: Thanks very much. I think we should  
8 move on now to the open public hearing. We have a number of  
9 speakers signed up.

10 There are a couple of issues first. Please limit  
11 your comments to five minutes or less. Also please make any  
12 financial disclosures. If there are no financial  
13 disclosures to make, please state so.

14 The first speaker in the open public hearing  
15 session is Dr. Victor DeGretolla, from the Harvard School of  
16 Public Health.

17 **Comments, Victor DeGretolla, Ph.D.**

18 DR. DEGRETOLLA: Dr. Victor DeGretolla, from the  
19 Harvard School of Public Health. In terms of the financial  
20 disclosures, I am one of the principal investigators of a  
21 project funded by Glaxo to look at the role of markers  
22 across a wide range of studies.

23 As everyone knows, a number of drugs have been  
24 approved in the past few years that have lowered viral

1 burden and extended AIDS-free survival, and the effects are  
2 seen in national surveillance, as well as clinical trials  
3 and every doctor's office. So it is natural to want to  
4 change standards for regulatory purposes.

5 I have a few concerns about this. Drugs, of  
6 course, have a number of important effects. In addition to  
7 reducing viral burden, they can also induce resistance both  
8 to the drug the patient has taken and to other related  
9 drugs.

10 So in addition to knowing RNA, you have to know  
11 something about whether a drug maintains or reduces future  
12 treatment options, and whether it works in patients who may  
13 have reduced options because of multi-drug resistance. The  
14 problem, of course, is that new drugs will be used in  
15 combination and patients take sequences of treatments. So  
16 how can we evaluate the contribution of an individual drug?

17 I think one thing we can do is try to do long-term  
18 follow-up regimens that include the drug of interest to find  
19 out the duration of suppression and other longer-term  
20 consequences, like what treatment options remain after a  
21 patient has had an adverse virologic response.

22 In addition, I think studies of the best  
23 strategies for how to use drugs may aid regulators in  
24 determining what features a drug should have. For example,

1 if a strategy study showed that the best way to initiate  
2 therapy is to use a protease-containing regimen and if they  
3 showed that patients should switch therapy as soon as there  
4 was any detectable virus it would be clear that what a drug  
5 should do is either work well in a protease-containing  
6 regimen to maintain virus below detection as long as  
7 possible, or work well in a group of patients who have had  
8 and already failed a protease-containing regimen.

9           On the other hand, if we were to see that the best  
10 way to initiate therapy was not to use a protease right away  
11 and the best time to switch would be after a patient had a  
12 viral load count of 5,000, then there would be different  
13 standards that a drug might be held to.

14           Now, there has been a lot of discussion about  
15 strategy trials. I saw Ellen Cooper in the audience, who  
16 has spoken about the usefulness of such trials. We haven't  
17 yet seen one. Perhaps if regulators and the Advisory  
18 Committee were to consider whether such evidence would be  
19 useful for regulation, it might help encourage the  
20 development of studies.

21           My only question might be do we need clinical  
22 endpoint studies any more or does RNA provide enough  
23 information? I think it might depend on patient class. If  
24 you have a class of patients for whom you can get nearly

1 complete and durable suppression, as well as CD4 count  
2 increases and no other serious consequences, that  
3 information might well be enough. But if you are talking  
4 about patient populations that are harder to treat, with  
5 drug-resistant virus where you can't attain full antiviral  
6 suppression, then partial suppression of RNA may not be  
7 enough and clinical endpoints might be useful.

8           So in conclusion, the recommendations that I would  
9 like the Committee to consider are the usefulness of long-  
10 term studies of suppression; evaluation of the effect of  
11 drugs on future options of treatment for the patient, as  
12 well as the initial RNA response; and evaluation of whether  
13 a drug might work well in a hard to treat population; and,  
14 finally, some recommendations regarding strategy studies  
15 that might demonstrate how best to use RNA or resistance  
16 information.

17           DR. HAMMER: Thank you very much, Victor. The  
18 next speaker is David Scondras, from Boston.

19                           **Comments, David Scondras**

20           MR. SCONDRAS: Thank you. I represent a non-  
21 profit organization called Search for Cure that is funded by  
22 a great many sources, community groups, Roche Diagnostics,  
23 IRC, Agrom Pharmaceuticals, state departments for public  
24 health.



1 I have passed to each of you a more complete  
2 rendition of the testimony that I wanted to introduce today.  
3 Given the burden of testimony that you are under, I will try  
4 to make this brief and limit my comments to a reflection on  
5 the nine points that I put forth in the document which was  
6 circulated to each of you.

7 The first three really would be no surprise to you  
8 inasmuch as the community reflects the general community  
9 feeling, which I think is fairly widespread and which I  
10 infer from conversations with dozens, if not hundreds, of  
11 people over the last six months in the HIV-positive  
12 community and advocates that, essentially, there seems to  
13 exist sufficient data to warrant the use of viral load for  
14 the approval of therapeutics.

15 Having said that, I would like to turn my  
16 attention to those things that might not come on your table,  
17 and to bring to the table a handful of concerns regarding  
18 that issue that might otherwise not be discussed and that we  
19 would feel a need for some reassurance.

20 The first is the assumption we make, and we would  
21 hope this Committee would communicate that assumption, that  
22 the FDA will continue to ensure that companies take  
23 responsibility beyond full approval for seeing, in a long-  
24 term sense, what the effects of the approved drugs are, both

1 in combination, side effects and so forth. That is a  
2 presumption we make. How do you ensure that there are Phase  
3 IV trials that are, in fact, adhered to?

4 Another assumption we are making is that the FDA--  
5 and this is a bit of an irony, we are sitting here,  
6 discussing the use of these tests for FDA to determine which  
7 treatments are effective while they are not permitted to be  
8 used by clinicians to make treatment decisions; they are  
9 approved for prognosis at this point. In fact, the best of  
10 them, second generation tests that Chiron has and Roche,  
11 have not been approved at all and, yet, we know there is a  
12 need for them. I think a mentioning of that to the FDA  
13 would be greatly appreciated.

14 A third concern of ours is that we really hope and  
15 we assume that if, indeed, you find that viral load is an  
16 applicable surrogate marker to approve with therapies, are  
17 we referring to all therapies, not just antivirals?  
18 Inasmuch as there is a wealth of data showing independence  
19 of the viral load numbers, whether you are talking about  
20 reduction in or the actual absolute value of, and clinical  
21 progression etc., it would be extremely disturbing if it  
22 turns out that there was a therapy that had implications for  
23 viral load that was not approved or supported because it was  
24 not a drug. That would create a significant concern.

1           We also assume that duration as well as viral  
2 reduction is viewed as important by the FDA. And we applaud  
3 them for the work they have already done and the concern  
4 they have expressed about the whole issue of duration.

5           However, the most important concern we have is one  
6 that I know you will find hard to focus on and I question  
7 the extent to which you can address it. But it goes like  
8 this, the FDA decision to use viral load could very well  
9 have the unintended consequence of reducing research under  
10 development of immune-based therapies, including those to  
11 restore immune function.

12           It is the old story of folks looking for their  
13 lost wallet while the light is good. The problem that we  
14 have with this is that there are a lot of folks who at this  
15 point do not see any reason to believe will necessarily  
16 benefit a great deal from the new therapies, people for whom  
17 drugs have failed, people in whom the immune system damage  
18 is such that it will not be reconstituted with  
19 antiretrovirals. We can show you studies, if you would like  
20 to see them, on the extent to which FDA decisions in effect  
21 directly impact the amount of dollars that go into research  
22 in certain areas.

23           We would, therefore, ask you, given the potential  
24 negative consequences of this decision, to tell the FDA to

1 take a step to ameliorate that. And I think it has an  
2 ethical responsibility to do that, in particular, to convene  
3 an assemblage of immunological experts to identify and cause  
4 data to be assembled on immunologically relevant surrogate  
5 markers so that the message that comes out from here is not  
6 that this is the gold standard at the end of the rainbow to  
7 which every dollar in the world flows but, rather, as one of  
8 many, to be followed by others.

9 I know it is unusual to request that the FDA  
10 become proactive instead of reactive, but I know that David  
11 Feigal has done an extraordinary job in that area at CDER  
12 and will do so at CBER. And I am hoping that you will take  
13 the time and be willing to mention to the FDA the extent to  
14 which many people are concerned that the decision will  
15 actually end up not narrowing the tunnel of researches to  
16 those chemotherapies which reduce viral load rather than  
17 broadening them to look at restoration and look at other  
18 concerns that we all have.

19 Thank you for taking the time to listen to me, and  
20 I really appreciate the work that you are doing.

21 DR. HAMMER: Thank you very much. The next  
22 speaker is Jules Levin, from the National AIDS Treatment  
23 Advocacy Project.

24 MR. LEVIN: I pass today. I would like to speak

1 tomorrow.

2 DR. HAMMER: Okay. Make sure you are listed for  
3 tomorrow's public session. The next speaker then is Alan  
4 Norburn, from the AIDS Treatment Project in London.

5 **Comments, Alan Norburn**

6 MR. NORBURN: I am speaking today on behalf of  
7 Rafid Babikianian of the AIDS Treatment Project, Executive  
8 Director of the AIDS Treatment Project. ATP is funded by a  
9 wide range of drug companies, corporate donations and  
10 organizations such as the Elton John AIDS Foundation.

11 There are several unique difficulties in the  
12 currently facing regulatory authorities evaluating anti-HIV  
13 drugs. The European AIDS Treatment Group believes that  
14 there is a way forward navigating between these obstacles to  
15 rational and compassionate system of HIV drug approval.

16 Twenty-four week surrogate marker trials can show  
17 the efficacy of the new drug in combination. Viral load and  
18 CD4 are important tools in determining prognosis and  
19 treatment efficacy. But 24 weeks is not long enough to  
20 determine the duration of the surrogate marker changes and  
21 24-week analyses should be part of longer surrogate marker  
22 trials, one year or more, which would continue to look at  
23 duration of effect, including longer-term safety and when to  
24 switch drugs as an endpoint.

1           I can illustrate the point with my own personal  
2 circumstances. I am now in week 26 of combination therapy  
3 which is failing to realize the goal of complete suppression  
4 of viral replication. Having already been through the  
5 trauma of deciding which therapy to use as a first-line  
6 treatment, I am now in something of a hit and miss situation  
7 in deciding how to continue.

8           The reality of the situation is that surrogate  
9 marker trials are highly unlikely to fully predict all of  
10 the clinical ramifications of a particular combination  
11 therapy, and it is probably impossible to conduct  
12 traditional clinical endpoint studies with most drugs that  
13 will be seeking licensing approval in the next few years.

14           Ideally, the day a drug is approved I would like  
15 to know everything about it, not only how safe and efficient  
16 it is, but also guidelines on how and when to use it. This  
17 reality urgently necessitates exploration of non-traditional  
18 methods of assessing clinical efficacy of drugs, including  
19 treatment strategy trials and observational data bases.  
20 Such methods may not be as statistically sound or  
21 scientifically accurate as traditional blinded clinical  
22 endpoint studies, but at the moment they are the best hope  
23 we have for obtaining some clinically useful information.  
24 They could not say how efficient a drug is, but they would

1 complement the information we already have, helping us  
2 attain maximum methods once efficiency has been shown by  
3 surrogate markers. They could provide valuable information  
4 with direct clinical relevance, particularly if there were  
5 to be uniform data collection from trial to trial, allowing  
6 for easier meta-analysis.

7           In addition to focused safety and interaction  
8 studies, wide extended access programs should be encouraged.  
9 Targets or guidelines for the number of people who should be  
10 exposed to a drug prior to filing for licensing approval and  
11 the lengths of exposure time will help to ensure that  
12 additional safety and interaction problems are identified  
13 early on.

14           Patent extensions could be offered to drug  
15 companies as an incentive to conduct uniform studies and  
16 access programs. This would mean a degree of cooperation  
17 between the drug companies and frequently convened antiviral  
18 expert groups. Such groups could be responsible for  
19 advising on evaluation on post-approval commitment.

20           New drugs are desperately needed by HIV-positive  
21 people for whom current drugs are no longer working or whose  
22 regimes have become too unrealistic to comply with. Each  
23 month that passes by without new drugs more people become  
24 ill and die. We hold the fate of these people in our hands.

1 Thank you for listening.

2 DR. HAMMER: Thank you very much. The next  
3 speaker is Francois Houyez, from the European AIDS Treatment  
4 Group in Paris.

5 **Comments by Francois Houyez**

6 MR. HOUYEZ: Good morning.

7 DR. HAMMER: I don't think your mike is on. For  
8 your overheads, just stand aside when you show them so the  
9 entire audience can see them. We had some trouble with the  
10 earlier overheads.

11 MR. HOUYEZ: My name is Francois Houyez. I am  
12 with the European AIDS Treatment Group. This group is based  
13 in Europe. This group is funded by a wide range of grants  
14 from the European Union.

15 DR. HAMMER: I think your microphone is still not  
16 on.

17 (Slide)

18 MR. HOUYEZ: ACTG presented a month ago a document  
19 to the European Medicinal Evaluation Agency which is  
20 reviewing the criteria to evaluate new drugs. These new  
21 guidelines should be published in September or October. So  
22 in a few words, we came to a decision on such indications  
23 like these results.

24 (Slide)



1 Dale Kemp and others presented in St. Petersburg a  
2 few days ago data which clearly shows that maximum  
3 suppression of viral load is one of the main predictors of  
4 the duration of the effect of an antiretroviral treatment.

5 (Slide)

6 Based on this information, we looked at the kind  
7 of information that we have and the kind of information  
8 which people need to make their decision and to evaluate the  
9 treatment.

10 (Slide)

11 We came to the point that it is not enough just to  
12 know the percentage of people who achieve results of a viral  
13 load below such-and-such a threshold. The main information  
14 we have to know is what maximum suppression of viral load we  
15 can achieve and how long this will last, and how we decide  
16 that we have to change our treatments.

17 So on these curves I show you some indication of  
18 first treatment which comes to the nadir, the maximum  
19 suppression of viral load, and the different situations  
20 afterwards. The question is how long a time clinical trials  
21 have to be run to show such maximum suppression. Ideally,  
22 if the treatment really works well, you won't be allowed to  
23 see such results for a long period of time. So we came to a  
24 decision that a 24-week trial should be the best average

1 time, not too short but not too long. We came to the point  
2 that the slope of decay could be an indicator of treatment  
3 potency from a single drug if you compare to another regimen  
4 which doesn't contain that drug.

5 (Slide)

6 So the principles are to grant conditional  
7 approval on the basis of 24-week surrogate marker with more  
8 preliminary safety data. The 24-week surrogate marker can  
9 show initial efficacy of the new drug combination. HIV-RNA  
10 and CD4 evolution have to be evaluated or else you wouldn't  
11 be able to really know what the treatment benefit is. As I  
12 said, there is a strong correlation between HIV-RNA nadir  
13 and duration of maximum response, and there is not such a  
14 correlation between RNA at baseline and duration of  
15 treatment.

16 (Slide)

17 But 24 weeks is not long enough. This analysis  
18 should be part of a much longer trial which will document  
19 the time to switch from one treatment to another one. So if  
20 approval should be based on a 24-week trial, then we should  
21 go on with such trials to document more data about the best  
22 way and when to switch treatment.

23 (Slide)

24 In addition to this, we addressed this point with

1 the European agency, that trials need to look at a wider set  
2 of drugs when considering drug interactions, not especially  
3 recreational drugs but also a broader range of drugs.

4 (Slide)

5 And to compile the theories and profile in vivo as  
6 a part of the dossier. We consider resistance to be a  
7 safety issue as well because of individual consequences when  
8 you fail because of resistance. It also has public health  
9 issues from the point of view that resistance is a safety  
10 issue. This is why we would like to address the resistance  
11 data as soon as any other data in the development of new  
12 drugs, which is not the case right now.

13 So these are the main points that should be raised  
14 to the CPMP and I will stop here and thank you for your  
15 attention.

16 DR. HAMMER: Thank you very much. The next  
17 speaker is Linda Grinberg, from the Foundation for AIDS and  
18 Immune Research Project Reform. If she is not here, the  
19 next speaker is Bill Bahlman, from ACTUP, New York.

20 **Comments, Bill Bahlman**

21 MR. BAHLMAN: Thank you. Bill Bahlman, from Act-  
22 Up, New York. I am also an officer of the Community  
23 Advisory Board at NYU Bellevue. I wear a number of hats.  
24 ACT-UP, New York accepts no pharmaceutical grants, mostly

1 their own fund-raising efforts.

2 I am going to be speaking more at length tomorrow  
3 but I wanted to raise one point, and that was in reference  
4 to what Don Brambilla said today bout the variability of  
5 response of viral load tests and I just wanted to touch on  
6 that before giving my full remarks tomorrow.

7 I think it is very important that what he said  
8 relates very closely and very importantly to one's own  
9 personal care, to not make quick judgments on a drug regimen  
10 that somebody may be taking with their private physician or  
11 in a clinic, and to have confirmation viral-load tests done,  
12 as well as those who enroll in clinical trials should not  
13 jump out of a clinical trial based on one viral-load result.

14 But when you relate it to what the FDA is  
15 proposing in terms of accepting viral load and percentages  
16 of patients that go to undetectable levels, the 6-month  
17 period and then 48-week period, it doesn't represent the  
18 same kind of problem because you are going to be doing a  
19 viral-load tests every 2 weeks initially, then every 4 weeks  
20 and then maybe less so often but still doing them regularly  
21 for the second 24 weeks of a study, and there outlining test  
22 results should not present a problem to analyze percentages  
23 of patients that go undetectable. Thank you.

24 DR. HAMMER: Thank you. The next speaker is Dr.

1 Iris Long.

2 **Comments, Iris Long, Ph.D.**

3 DR. LONG: My name is Iris Long. I reside in  
4 Jackson Heights, Queens, New York. I am a pharmaceutical  
5 chemist by background. Since 1987 I have been an advocate  
6 of people living with AIDS HIV, focusing on medication  
7 development, especially during the experimental phase.

8 I am a member of AIDS Coalition to Unleash Power,  
9 ACT-UP, New York, a volunteer organization, AIDS activist  
10 organization, and a member of the Community Advisory Board  
11 for the AIDS Clinical Trial Unit at Mt. Sinai Medical  
12 Center, both in New York City. I have no financial  
13 association with any drug company or device company.

14 Women AIDS treatment activists would give their  
15 full support to new viral-load trial designs and associated  
16 drug approval process if women living with AIDS HIV are  
17 included in the process of developing and implementing these  
18 trials.

19 Before such trials are designed, issues concerning  
20 women's access to these new trials need to be addressed.  
21 Women with life-threatening AIDS HIV disease are still being  
22 explicitly and implicitly excluded from many drug trials  
23 sponsored by the federal government and pharmaceutical  
24 manufacturers. Monitoring women's access to these

1 treatments has not been effectively done by either the FDA  
2 or the government-funded ACTG program.

3           In 1996 Centers for Disease Control reported that  
4 the nation's female adult, adolescent population represented  
5 20 percent, with the race ethnicity profile being 21 percent  
6 white, 59 percent black and 19 percent Hispanic. According  
7 to the ACTG's inflated data, women's participating in AIDS  
8 ACTG trials is 15 percent. However, they include in their  
9 15 percent female participation figure a significant number  
10 of pregnant women participant enrollees, 1,400 or the 6,000,  
11 of the pediatric ACTG program. Pediatric trials are not  
12 testing treatments for women. Subtracting pregnant women  
13 gives 12 percent women participants, not 15 percent.

14           What needs to be done is that an ethical national  
15 research policy should be developed. Women living with AIDS  
16 HIV around the country should have ample opportunities to  
17 discuss with the FDA government-funded clinical trial  
18 programs and drug companies, patient options and issues  
19 involved in new antiretroviral trial design. The FDA must  
20 require the inclusion of women in new trial designs so that  
21 meaningful analysis of women can be done.

22           The National Institute of Allergy and Infectious  
23 Diseases and the FDA should closely monitor women's access  
24 to government and non-government new viral-load based trials

1 in all phases of development, including local site  
2 monitoring. Effective community advisory boards should be a  
3 part of this process. Thank you.

4 DR. HAMMER: Thank you. That concludes the open  
5 public hearing section. There is one final thing, Dr.  
6 Johnson has asked just to show his last slide to conclude  
7 our morning.

8 (Slide)

9 DR. JOHNSON: The first thing that I have already  
10 talked about was the very suboptimal therapy with ddI.  
11 Monotherapy did have some beneficial effect on disease, and  
12 that is useful information for pediatricians where  
13 compliance may be even a greater problem than in adults, and  
14 some families will only be able to comply with a monotherapy  
15 approach, at least initially.

16 The speculative second one there about more than a  
17 0.3 log drop at 12 and 24 weeks may be associated with  
18 differences in rates of survival and clinical progression.

19 But the last, relative to this portion of the day,  
20 is the imperative to study combinations, particularly those  
21 containing non-nucleoside reverse transcriptase inhibitors  
22 and protease inhibitors rapidly in children, to allow access  
23 of those therapies.

24 In addition to this, just as a pediatric advocate,

1 I would like to push for both development and testing of  
2 formulations which are able to be administered to  
3 particularly young children, which is lagging behind adult  
4 drug development.

5 DR. HAMMER: Thank you very much. I have just  
6 been apprised that there is one additional request to speak  
7 at the open public session, Ron Baker, from San Francisco.

8 **Comments, Ron Baker**

9 MR. BAKER: Good morning and thank you for the  
10 opportunity to speak here this morning. I am Ron Baker,  
11 Editor in Chief of the AIDS Treatment publication from the  
12 San Francisco AIDS Foundation. In terms of financial  
13 disclosure, in the current fiscal year the San Francisco  
14 AIDS Foundation has been the recipient of educational grants  
15 from Glaxo Wellcome, Inc. and from Hoffman-La Roche.

16 I am here today to present the views of the  
17 Foundation on three issues related to the use of HIV-RNA  
18 testing in clinical studies. First, we feel that the time  
19 has come to eliminate suboptimal treatment arms in  
20 comparative clinical studies. Second, a number of studies  
21 reviewed here, this morning, have shown that viral-load  
22 measurement demonstrates the usefulness of experimental  
23 drugs more quickly and just reliably as clinical endpoints,  
24 in our view. Thirdly, we feel that drug labeling should



1 reflect the way in which drugs will actually be used  
2 clinically.

3           The recently issued guidelines for the use of  
4 antiretroviral agents are clear regarding decisions to  
5 initiate or change therapy. Such decisions, according to  
6 these recommendations, should be guided by measurement of  
7 HIV-RNA and CD4 T-cell counts, as well as by the clinical  
8 condition of the patient.

9           At the San Francisco AIDS Foundation we feel that  
10 it is no longer justifiable to extend clinical studies until  
11 clinical endpoints, such as the deaths of study participants  
12 have been reached. In our view, accelerated approval can be  
13 granted, as it is now, based on viral-load suppression in a  
14 significant number of study participants with perhaps as few  
15 as 18 to 24 weeks of data. Full approval could be granted  
16 to a new drug based on about a year's data, provided that  
17 the magnitude of viral suppression is great enough and  
18 remains durable. Defining "great enough" and "durable  
19 enough" remains problematic.

20           At the same time, it is very important for  
21 clinicians to know who does not respond well to drug so that  
22 individuals can switch to a potentially more effective  
23 regimen.

24           Concerning drug labeling, the Foundation supports

1 the use of language in the FDA indication that a drug should  
2 be used "to achieve maximal viral suppression" rather than  
3 treatment of HIV disease. This more precise and descriptive  
4 language in labeling will provide greater clarity to  
5 physicians and to patients concerning the real objective of  
6 antiretroviral therapy, which is to reduce the viral load to  
7 as low a level as possible for as long as possible.

8           In summary, the Foundation asks the Committee to  
9 recommend, first, the elimination of suboptimal study arms.  
10 Second, we strongly support the use of viral-load  
11 measurement as an endpoint in AIDS drug studies. Finally,  
12 we advocate the use of labeling that accurately reflects how  
13 a drug affects HIV viral load.

14           These are urgent concerns and we feel they should  
15 be implemented by FDA as quickly as possible. Thank you.

16           DR. HAMMER: Thank you very much. If there are no  
17 other requests to speak at the public hearing, I am going to  
18 adjourn the morning's session. We can reconvene in one  
19 hour, at 1:20.

20           [Whereupon, at 12:20 p.m., the proceedings were  
21 recessed, to be resumed at 1:20 p.m. this same day.]

1 AFTERNOON SESSION

2 DR. HAMMER: We are waiting for the sound  
3 engineer, but if you will please take your seats, we will  
4 start in a moment. We will start with this one microphone  
5 that appears to be working. This afternoon's session is  
6 devoted to a series of presentations, first, a review of  
7 antiretroviral guidelines, and then clinical confirmation of  
8 HIV-RNA changes. I would like to begin with Sherilyn  
9 Stanley, from NIAID, who will review the antiretroviral  
10 guidelines.

11 **Review of Antiretroviral Guidelines, S. Stanley, M.D.**

12 DR. STANLEY: Thank you very much. I would like  
13 to take this opportunity to thank the organizers for  
14 inviting us to review these for you all.

15 (Slide)

16 I am sure I don't have to remind this group of the  
17 rapid evolution that we have experienced in antiretroviral  
18 therapy over the past several years. This slide just sort  
19 of summarizes that schematically. In '87, with monotherapy  
20 we could get nice suppression of viremia but it did not  
21 persist. By '94 we could give combination 2-drug therapy,  
22 and again we could get nice suppression of viremia but  
23 generally it did not persist. Now, in 1997 we have  
24 available to us the combination therapies, including the

1 protease inhibitors, and we are seeing again in a lot of our  
2 patients marked suppression of viremia with some long-term  
3 suppression in some patients.

4 (Slide)

5 The evolution of therapy and the advances, along  
6 with the rapid increase in number of drugs that have been  
7 approved for HIV therapy, have caused a lot of confusion in  
8 the field of the clinical practitioner. This was recognized  
9 by the Secretary and she requested that Eric Goosby, head of  
10 the Office of HIV AIDS policy, in conjunction with Henry J.  
11 Kaiser Foundation, convene a panel that would have a three-  
12 year life span at a minimum that would address clinical  
13 issues of therapy for HIV-infected patients, not just  
14 antiretroviral therapy but also other issues, for instance  
15 pediatrics perhaps, perhaps managed care issues, whatever  
16 clinical aspects of HIV management this panel wished to  
17 address. The first issue that with some urgency the panel  
18 felt needed to be addressed was antiretroviral therapy.

19 (Slide)

20 This is the panel that was put together, again,  
21 co-convened by the Henry J. Kaiser Foundation and OHAP. The  
22 co-chairs were picked to be Dr. Fauci, as the federal  
23 representative, Dr. John Barton, from Johns Hopkins, as the  
24 private community representative, thus, again solidifying

1 that this was a unique public-private partnership.

2 The panel members were drawn from a large variety  
3 of community activists, active commissions, researchers and,  
4 as well, multiple HHS agencies were represented as members  
5 of the panel.

6 Again, this panel has been constituted for three  
7 years and they have just released, in the Federal Register,  
8 the draft guidelines for antiretroviral therapy. Those  
9 guidelines are available through a 1-800 number. We are  
10 still in a period of public comment. At the end of that  
11 period the comments will be considered by the panel and a  
12 final document will be released.

13 Let me just summarize for you the process that the  
14 panel went through. The panel met three different times and  
15 discussed various aspects of antiretroviral therapy and the  
16 data that was available. They relied on the recently  
17 formulated principles of antiretroviral therapy, which is a  
18 document that is traveling in companion form with the  
19 guideline document.

20 The principles document was derived by an NIH-  
21 convened panel to develop principles of antiretroviral  
22 therapy. These are basically the eleven pathogenetic  
23 principles that provide the rationale for why to treat and  
24 how to treat. So the clinical practices panel used that

1 information as well as the experience of experts and their  
2 own clinical experience to derive the draft guidelines.

3           Critical in their considerations was one of the  
4 principles of HIV disease, which is the fact that HIV  
5 replication is constant and active during all stages of HIV  
6 disease. This would lead one to think that one should  
7 perhaps treat early and treat aggressively.

8           (Slide)

9           However, the panel also considered very seriously  
10 the data by Mellors et al., showing that depending on the  
11 level of RNA copies, time to progression to AIDS is very  
12 much faster with the higher viral burdens as opposed to  
13 those with less than 9,000 copies who have a prolonged  
14 progression to AIDS and a fairly prolonged symptom-free  
15 survival.

16           (Slide)

17           This is also shown in this graph where, as virus  
18 levels increase going in this direction and as CD4 levels  
19 decrease going in this direction, you get increasing  
20 progression with likelihood of development of AIDS within  
21 three years.

22           (Slide)

23           Considering this data, the panel felt that there  
24 were also other considerations that would weigh either for

1 or against early therapy in the asymptomatic individual.  
2 These were summarized in a table in the document. They were  
3 also discussed at length in the text of the document.  
4 Again, this is for making the decision for the practicing  
5 physician when do you treat the asymptomatic HIV-infected  
6 individual. The aggressive clinician, on the basis of  
7 active viral replication, might say treat early. The more  
8 conservative clinician or the patient that has other  
9 considerations might say let me look at my chance of  
10 progression to symptomatic AIDS, and let me weigh these  
11 other factors in here and decide whether I wish to delay  
12 therapy for some time or aggressively treat at this point.  
13 The panel tried to reach this sort of center view that  
14 allows the aggressive clinician to treat early, but also  
15 allows for and recognizes the legitimacy of delaying therapy  
16 in the asymptomatic individual, such that the  
17 recommendations for therapy are really summarized in Table 5  
18 in the document.

19           It was absolutely universal that any patient with  
20 symptomatic HIV infection or AIDS, no matter what the CD4 or  
21 HIV-RNA levels, should be treated. However, in the  
22 asymptomatic individual with CD4 T-cells less than 500 or  
23 HIV-RNA greater than 10,000 by bDNA or 20,000 by RT PCR, in  
24 this category of patients treatment should be offered based

1 on the scientific principles that we understand with HIV  
2 disease. But the strength of the recommendation for  
3 treatment should be balanced on the prognosis for disease-  
4 free survival and the willingness of the patient to accept  
5 therapy and, some of those other considerations that I  
6 showed you in the last slide. So this is really the group  
7 where the approach of the clinician, whether aggressive or  
8 conservative, and the considerations of the patient will  
9 make the most difference.

10 Then in the asymptomatic patients where the CD4  
11 counts are greater than 500 and the HIV-RNA is very low to  
12 undetectable, some experts would delay therapy and observe  
13 but still there are some experts who would treat these  
14 patients based on the principle that there is always ongoing  
15 viral replication and this is detrimental to the immune  
16 system. Therefore, this option was allowed by the panel.

17 (Slide)

18 In order to help make this document as user  
19 friendly as possible and to help the clinician make these  
20 decisions, there is a table in the document that gives the  
21 Mellors data: plasma viral load, CD4 count and percent  
22 developing AIDS in three years, six years and nine years.  
23 This is the number of patients that fell into each category.  
24 These are the less than 350; these are 351 to 500. Again,



1 allowing clinicians to use either bDNA or RT PCR, with an  
2 explanation of how these convert from the MAX data which, of  
3 course, was generated on frozen, stored specimens. Then for  
4 the patients with greater than 500 CD4s this data is given.  
5 So the document really provides the clinician with the  
6 survival best data that we have to date, as well as the  
7 other rationale, for or against, for the early treatment of  
8 the asymptomatic patient.

9 (Slide)

10 Once the decision to treat has been made, the  
11 panel was unanimous in stating that three drugs are the  
12 first choice of therapy including a protease inhibitor, so  
13 that two nucleosides or reverse transcriptase for instance  
14 and a protease inhibitor, or perhaps a non-nucleoside,  
15 reverse transcriptase inhibitor. But it was unanimous that  
16 once you decide to treat, you need to be aggressive so that  
17 you achieve maximal viral suppression with all the benefits  
18 that that obviously gives, which is decreased development of  
19 resistance and other positive factors.

20 (Slide)

21 That decision was really based on the multitude of  
22 clinical trials we have showing benefits of triple-drug  
23 therapy, either virologic or clinical benefits.

24 (Slide)

1           Table 6 in the document shows what the panel  
2 considered to be the optimal choices and the alternative  
3 choices. You will notice that in this table the preferred  
4 choice is a highly active protease inhibitor with two NRTIs,  
5 and the potential combinations are shown.

6           An alternative regimen, again, for initiating  
7 therapy in the asymptomatic patient was considered to be  
8 nevirapin and two NRTIs, as above, saquinavir in its current  
9 hard-gel capsule formulation.

10           (Slide)

11           Less desirable and, in fact, not recommended  
12 unless there was some sort of special clinical situation, is  
13 the use of only two NRTIs without a protease inhibitor or  
14 other third drug. Again, absolutely not recommended and  
15 probably contraindicated are all monotherapies and these  
16 particular combinations of two NRTIs.

17           (Slide)

18           The panel addressed the issues of when do you  
19 change therapy. In relationship to this meeting here, based  
20 on the data that was reviewed by the clinical guidance panel  
21 as well as the principles panel, the scientific principles  
22 enumerate in the principles document that a ten-fold  
23 reduction in viremia at four weeks is a sign of successful  
24 therapy, and that virus that is not under ten-fold by four

1 to six months after therapy probably reflects suboptimal  
2 treatment.

3           This is a table from the document. The first  
4 bullet is discussed at more length in the text of the  
5 document. I will recommend that that be read. The  
6 guidelines for changing therapy would include suboptimal  
7 reduction of plasma viremia after initiation which, as I  
8 said, the panel took to be ten-fold reduction at four weeks;  
9 reappearance of viremia after suppression or detectable  
10 virus at four to six months after therapy; significant  
11 increases in viremia from the nadir of suppression; decline  
12 in CD4 numbers and clinical deterioration. Again, there  
13 were several caveats of things to consider when a clinician  
14 is considering changing therapy. I won't go through all of  
15 those. They are included in the document.

16           (Slide)

17           The panel went so far as to try, with the minimal  
18 data is available, to help the practicing clinician by  
19 making some recommendations for what a potential change  
20 would consist of. Again, this table, which is in the  
21 document, outlines what the prior regimen might be; what you  
22 might consider switching to, given again that this is mostly  
23 expert opinion because of the paucity of data that is  
24 available.

1           The panel did consider acute retroviral syndrome.  
2 It was felt it was important to reeducate the clinician on  
3 what that is so that we can perhaps achieve better  
4 recognition of this. So in the document is included a list  
5 of the symptoms. The panel basically said that most experts  
6 would probably treat a recognized case of acute retroviral  
7 syndrome but was unable to come to a firm conclusion about  
8 the length of such treatment, and ended up stating that  
9 probably treatment should go on indefinitely once it has  
10 been initiated, again because of a lack of data.

11                   (Slide)

12           Again, relative to this meeting, the panel  
13 obviously put great emphasis on HIV-RNA testing and felt  
14 that although it is still very important to get CD4 counts  
15 to understand the immunologic condition of the patient,  
16 viral burden data is what people are really using out there  
17 in the academic settings and in the clinical settings to  
18 guide their decisions about antiretroviral therapy.

19           So there is a table that instructs the clinician  
20 on what the clinical indications might be for RNA testing;  
21 what information you would hope to get; and how you would  
22 use that in your decision making.

23                   (Slide)

24           This is simply the second half of that table.

1 (Slide)

2 Just to give a flavor of some of the tables that  
3 are in the document, I am not going to go through them but  
4 there is a table describing characteristics of the NRTIs,  
5 the NNRTIs--

6 (Slide)

7 --the protease inhibitors--

8 (Slide)

9 --and this one goes on for some time.

10 (Slide)

11 Drugs that should not be used, so that this will  
12 be available, hopefully, for the clinician's reference. We  
13 ran these past David Feigal and his staff at the FDA, and we  
14 hope that they are accurate and we will continue to update  
15 them as more information becomes available. We appreciate  
16 David's staff time on this.

17 (Slide)

18 Again, drugs which should not be used in  
19 combination by category.

20 (Slide)

21 Various clinical drug interactions that are of  
22 significance.

23 (Slide)

24 This continues.

1 (Slide)

2 Also interactions between the PIs and the NNRTIs  
3 listed to date.

4 (Slide)

5 And this is the second half.

6 (Slide)

7 Finally, the panel also addressed the use of  
8 antiretroviral therapy in pregnancy, and the recommendation  
9 is basically that the woman's clinical status should be the  
10 primary determinant factor in treating or continuing therapy  
11 with, of course, special considerations that one must think  
12 of the unborn fetus and possible teratogenicity and  
13 carcinogenicity effects. There is a table in the document  
14 that gives what information we have on that.

15 So where are we at? We have a week left in the  
16 public comment period. I urge you, if you haven't obtained  
17 the document to get it and to please give us feedback.  
18 After the comments are then put together, the panel will  
19 review these, will make appropriate changes in the document  
20 and the document will then be published, hopefully, within  
21 the next several months.

22 There is a mechanism being put in place for this  
23 panel not only to go on and address other issues, but to  
24 continually update this document so that as we get more data

1 on various agents or new agents appear they can be added to  
2 the document.

3 I will be glad to answer any specific questions.

4 DR. HAMMER: Are there any questions? I think no.  
5 Thank you very much. We will now move on to the session on  
6 clinical confirmation of HIV-RNA changes. The first speaker  
7 will be Dr. Jeffrey Murray from the Division of Antiviral  
8 Drug Products, FDA.

9 **Introduction, Jeffrey Murray, M.D.**

10 (Slide)

11 DR. MURRAY: I am Jeff Murray, one of the FDA  
12 reviewers who helped in planning this Advisory Committee  
13 session.

14 I would like to spend just a moment to acknowledge  
15 the other individuals and groups who helped put this meeting  
16 together because it was quite a bit of an effort. We have  
17 members from FDA Antiviral Drugs, but also we had a lot of  
18 help from the Surrogate Marker Collaborative Group, other  
19 IND and NDA holders, and also this meeting was put together  
20 with some feedback from community groups at a meeting that  
21 we hosted a month or so ago.

22 (Slide)

23 We are now beginning the part of the meeting at  
24 which analyses of the clinical trial data will be presented

1 and explored. This afternoon there will be presentations on  
2 the clinical correlation of treatment-induced HIV-RNA  
3 changes. Tomorrow we will examine behavior or pattern of  
4 HIV-RNA in response to various antiretroviral treatments.

5 (Slide)

6 In preparation of the following presentations and  
7 preparation for the Committee's discussion tomorrow, I want  
8 to comment briefly on our rationale and our objectives for  
9 this meeting and for the presentations that will immediately  
10 follow.

11 There are several reasons this is an opportune  
12 time to reevaluate clinical studies supporting traditional  
13 approval. First, there are new therapeutic goals and  
14 guidelines, as you have just heard from Dr. Sherilyn  
15 Stanley. Specifically, we are in the midst of a shift in  
16 the way HIV-infected patients are managed clinically. In  
17 contrast to the setting in which the past clinical endpoint  
18 trials were conducted, we are now working in the setting of  
19 HIV monitoring, potent drug combinations and new goals of  
20 therapy, that is, to maximally suppress virus.

21 Second, we realize that it was an opportune time  
22 because there was a sizeable accumulation of clinical trial  
23 data describing relationships between virologic changes and  
24 clinical disease progression. There is also a large amount



1 of clinical trial data describing the behavior of HIV-RNA  
2 response to some potent drug therapies. We felt that the  
3 state of this is sufficiently rich to start making some  
4 recommendations for improving trial design and we thought it  
5 would be a missed opportunity not to explore this wealth of  
6 information.

7 (Slide)

8 So, in short, our overall goals for this meeting  
9 were to explore using HIV-RNA as an endpoint, and also as  
10 another option for additional approval, and also as a label  
11 indication. We anticipated that using HIV-RNA as a primary  
12 endpoint could have potential advantages in clinical trial  
13 design for participants, for investigators, for sponsors and  
14 also for FDA.

15 (Slide)

16 The reasons for considering the use of RNA as a  
17 clinical endpoint for efficacy trials are really some  
18 potential advantages. That is, it is a less complex  
19 endpoint than the current clinical endpoint which includes  
20 approximately twenty different infectious diseases,  
21 malignancies and other conditions.

22 Second, using RNA as an endpoint coincides with  
23 medical practice, and you can't ignore this. Physicians use  
24 this test to make recommendations.

1 Third, we felt that an endpoint that we have  
2 actually proposed, an endpoint such as time to loss of  
3 virologic response could be appealing for trial participants  
4 and also for those analyzing the data. This type of  
5 endpoint could easily allow treatment switches before  
6 clinical failure and, at the same time, treatment switches  
7 would not necessarily disturb the study analysis because an  
8 endpoint would have been achieved before a treatment switch  
9 was actually required.

10 Last, with the advent of more potent regimens,  
11 powering studies for the relatively infrequent clinical  
12 endpoints was and is becoming more difficult. So powering  
13 studies with respect to detecting differences in RNA  
14 endpoint could offer us economy with respect to sample  
15 sizes.

16 (Slide)

17 Although we want to consider RNA endpoint studies  
18 as another option for traditional approval for  
19 antiretroviral drugs, we were confident that certain  
20 conditions have to be preserved. First, clinical endpoint  
21 studies need to remain an option and still should be  
22 encouraged for answering certain clinical questions as  
23 necessary. That is, RNA studies can support an indication  
24 for lowering RNA. Clinical studies can support an

1 indication for delaying disease progression, for AIDS  
2 dementia and for other type of clinical questions.

3           Second, we are fully committed to keeping all  
4 early access mechanisms intact, including accelerated  
5 approval based on earlier, such as 16 to 24-week changes in  
6 both CD4 and RNA. For the RNA endpoint short-term changes  
7 would just simply be confirmed by longer-term RNA studies  
8 demonstrating durability.

9           Third, we are also committed to an accurate  
10 evaluation of safety and tolerability. Although RNA studies  
11 would be powered for an RNA endpoint, clinical disease  
12 progression data would still be collected. Information on  
13 safety, tolerability, CD4 data, we all realize this is still  
14 important to be looking at.

15           (Slide)

16           Before RNA could be used as a primary endpoint for  
17 traditional approval, we believe the following questions  
18 need to be addressed: First, we need to be confident that  
19 there is evidence that RNA reduction in itself is associated  
20 with decrease in clinical progression rate.

21           Second, we need to know best how to measure and  
22 analyze these changes, what was most clinically relevant.  
23 Third, we want to explore if there are any considerations  
24 for using this endpoint in special subpopulations, and that

1 is why we devoted a considerable amount of time this morning  
2 to looking at the pediatric data that is available. The  
3 clinical correlates presentations that follow will address  
4 numbers one and three.

5 (Slide)

6 For this set of presentations that will  
7 immediately follow we developed a number of analyses  
8 questions pertinent to the relationship between RNA  
9 reduction and clinical benefit. These are, to evaluate  
10 whether reduction of plasma RNA confers benefit; to describe  
11 the relationship in magnitude and duration of HIV-RNA  
12 reduction and clinical disease progression; to describe the  
13 relationship between virologic nadir and clinical disease  
14 progression.

15 (Slide)

16 To explore whether the prognostic significance of  
17 HIV-RNA is dependent on baseline factors; to describe  
18 appearance HIV-RNA changes around the time of the clinical  
19 event, before and after; and also to explore the proportion  
20 of antiretroviral treatment effect that is mediated by  
21 changes in both virologic and immunologic endpoints.

22 (Slide)

23 So we are asking the Committee to evaluate a  
24 collection of clinical studies. This involves more than

1 just pooling of data to produce a single mathematical  
2 estimate of treatment effect explained. In fact, this has  
3 not been done and you will not hear that.

4           Instead, what will follow are five separate  
5 presentations. Some use data from one trial and others have  
6 combined data when appropriate. There is a heterogeneity in  
7 the studies. Not all studies use the same assays; not all  
8 studies measure the same time points or study the same  
9 populations.

10           Although this heterogeneity can sometimes prevent  
11 pooling of data, it is also very informative. For example,  
12 you will see that all of these analyses show that reduction  
13 of plasma HIV-RNA is associated with a decreased risk of  
14 disease progression, and they all show that the more HIV-RNA  
15 is lowered the greater the reduction for the risks for  
16 disease progression. It is the consistency of results  
17 across different studies and different patient populations  
18 that constitutes very strong evidence.

19           However, making a decision about the relevance of  
20 RNA requires beyond the presentations today. There are  
21 other pieces of this puzzle. For example, we must apply  
22 what we know biologically, particularly with respect to the  
23 development of resistance. We can't ignore the fact that  
24 concerns about drug resistance and the continued usefulness

1 of the regimens that we are approving is directly linked to  
2 the maximal suppression of virus.

3 We will now begin these very interesting  
4 presentations. The first speaker will be Dr. Ian Marchner,  
5 from the Harvard School of Public Health, and he will be  
6 presenting some data from various ACTG trials.

7 **Presentation by Ian Marchner, Ph.D.**

8 DR. HAMMER: Thank you very much.

9 DR. MARCHNER: My name is Ian Marchner. I am a  
10 statistician with the AIDS Clinical Trials Groups,  
11 statistical data analysis center at Harvard.

12 (Slide)

13 I am going to be talking about a pooled cross-  
14 protocol analysis of a number of ACTG studies that we  
15 conducted to assist the utility of treatment-mediated  
16 changes in plasma HIV-RNA for predicting clinical  
17 progression rates.

18 (Slide)

19 Just to give you some background with regard to  
20 the studies we are analyzing, this was a cross-protocol  
21 analysis of seven ACTG studies involving a variety of  
22 different treatment regimens. Most of the treatment  
23 regimens we are dealing with involved nucleoside reverse  
24 transcriptase inhibitors, particularly ZDV, ddI and ddC.

1 There was one study that included nevirapin and one study  
2 that included saquinavir.

3 We made a decision in this analysis to include  
4 even monotherapy arms in the analysis, the reason being that  
5 we are interested in looking at the association between RNA  
6 responses with clinical progression, rather than the effect  
7 of therapies on response.

8 The sample that we are dealing with is going to be  
9 all individuals in these studies that had baseline RNA  
10 measurements and CD4 measurements. Some of the analyses  
11 that were performed included all such individuals, and they  
12 numbered 1,330 altogether. Most of the analyses, in fact  
13 probably all the analyses that I am going to present today  
14 are concerned with changes in RNA up to week 24, and their  
15 association with clinical progression. So we could only  
16 include individuals who were at risk of clinical progression  
17 at week 24 and who also who had CD4 and RNA data at week 24.  
18 These individuals numbered 1,000.

19 Everything that I am going to be talking about  
20 today, with the exception of one brief comment about the  
21 Chiron assay, is concerned with the Roche PCR assay, the  
22 original assay.

23 (Slide)

24 This is just a very small summary of the data that

1 I have included in the packet that the Committee got by way  
2 of background. The crucial elements in this table are,  
3 firstly, the sample size which is 1,000, primarily for the  
4 individuals who had the data that I talked about on the  
5 original slide. There were 120 clinical progressions,  
6 defined as a new AIDS-defining clinical event or death among  
7 those 1,000 individuals.

8 The CD4 at baseline was approximately 200, the  
9 median RNA was 50,000 and, in terms of the follow up, the  
10 median follow up overall was approximately 1 year, with  
11 follow up out to about 3 years in some individuals.

12 Here also is a summary of the different therapy  
13 arms that we have on the studies. You can see that there  
14 are monotherapy arms together with triple therapy arms, and  
15 triple therapy here as well.

16 (Slide)

17 So what I am going to do is basically just present  
18 you with three or four key questions, for the purpose of  
19 brevity, and associate with each question just a single key  
20 graphic to give a picture of what I think the answer to the  
21 question is.

22 The first question I am going to deal with is how  
23 is the magnitude of response related to the reduction in  
24 clinical disease progression rate? What I am going to use



1 as response is the change in HIV-RNA over the first 24 weeks  
2 of each of these trials.

3 (Slide)

4 This is really what I would see as the key  
5 graphical information relating the magnitude of response to  
6 the reduction in clinical risk. What I have on the X axis  
7 is the change from baseline to week 24 in HIV-1 RNA. what  
8 we have done is to split the change up into 8 groups,  
9 octiles, and then to estimate essentially the hazards ratio  
10 for clinical progression in each of these octiles. Then the  
11 hazards ratio is plotted against the median change in each  
12 of the 8 octiles.

13 The first thing that obviously stands out is that  
14 individuals with no reduction or even a slight increase are  
15 at the greatest risk of progression, whereas individuals who  
16 had greater reductions are at less risk for progression.  
17 You will notice that for this data set we are dealing with  
18 reductions at about 1.5 logs from baseline.

19 But probably the more striking feature is the  
20 fairly strong linearity between the adjusted hazards ratio--  
21 I should say that this is adjusted for the baseline level.  
22 So this is the effect of changes adjusting for the baseline  
23 level. There is a very strong linear relationship,  
24 indicating that the decrease in clinical progression risk is

1 proportional to the reduction in HIV-RNA out to 24 weeks.  
2 Implicitly what that tells us that larger reductions are  
3 more beneficial than smaller reductions, but still smaller  
4 reductions have some clinical benefit.

5           This plot that I just showed you in some sense  
6 gives us information about the next question, although I  
7 prefer another plot for answering it in more detail. The  
8 question is what descriptors of the magnitude of response  
9 are most clinically relevant. I have sort of taken this to  
10 mean are we interested in absolute response or absolute  
11 level of RNA achieved, or is there some threshold beyond  
12 which there is no real change in clinical progression risk?

13           (Slide)

14           The analysis that we have done is analogous to the  
15 one on the previous plot, but now on the X axis what we are  
16 dealing with is the absolute week 24 RNA level as opposed to  
17 the response or reduction over the first 24 weeks. We can  
18 see a very striking linear relationship, indicating a  
19 proportional relationship between the risk of clinical  
20 progression and the absolute value achieved after 24 weeks  
21 of therapy on these studies.

22           What we see here is no evidence of a threshold  
23 relationship in the sense that the lower your week 24 RNA  
24 is, the better your clinical progression risk is, and this

1 doesn't seem to level out as we get to the lower week-24 RNA  
2 values.

3           The next thing I want to talk about is the effect  
4 of baseline RNA level on the interpretation of treatment-  
5 mediated reductions and, in particular, this question really  
6 has two aspects to it. The more transparent aspect is, is  
7 baseline RNA level an independent predictor of clinical  
8 progression risk over and above the reduction in RNA? And  
9 we would perhaps expect that it would be. But a more subtle  
10 question is whether or not the baseline level modifies the  
11 interpretation of the HIV reduction over the first 24 weeks,  
12 and that is what statisticians like to refer to as  
13 interaction between baseline and treatment-mediated  
14 response.

15           (Slide)

16           In terms of the independent predictive ability of  
17 baseline and response, this is just a crude summary giving a  
18 feel for the fact that baseline level and reduction are both  
19 independent predictors if you do more sophisticated  
20 analyses, and that certainly comes out.

21           What we have here is individuals divided into 4  
22 categories depending on whether or not they had any decrease  
23 in HIV-RNA out to week 24. So there are two groups, one who  
24 had a decrease; one who didn't have a decrease. Then

1 whether or not they were above or below the median of 55,000  
2 at baseline. What we see is, for example, the top 2 curves  
3 refer to groups below the median at baseline. We have  
4 separation here, statistically significant separation  
5 dependent on whether or not they had a reduction in RNA  
6 after week 24 and then, likewise, for the individuals who  
7 were below baseline we again have separation dependent on  
8 their response, suggesting to us that baseline and response  
9 are independently predictive.

10 (Slide)

11 Now getting at the more subtle question of whether  
12 or not there is interaction between the two or whether the  
13 baseline level modifies the interpretation of an HIV  
14 response, what we have here are hazards ratios again. On  
15 the X axis what we have is individuals divided into whether  
16 or not they had 1 of 3 week-24 responses: no decrease, or an  
17 increase, a moderate response, from 0 to 0.5 log reduction  
18 and a better response of greater than 0.5 log reduction. In  
19 the 3 lines are just the baseline level categorized  
20 approximately into 1/3 percentiles, so less than 20,000,  
21 20,000 to 100,000 and above 100,000.

22 In each of the 3 baseline categories we see that  
23 the risk increases as the week-24 response gets worse. What  
24 we see between each of the 3 curves is that the higher the

1 baseline level, the higher the risk of progression.

2           The important point though is that the increase in  
3 risk or the decrease in risk, however you want to look at  
4 it, is very similar for the 3 different baseline categories,  
5 suggesting that the interpretation of a given week-24  
6 response is similar for the different baseline categories.  
7 So, for example, a 0.5 log reduction decreases your risk of  
8 clinical event by about the same ratio regardless of what  
9 your initial baseline level was.

10           (Slide)

11           We had a fairly substantial talk today about  
12 issues of variability in interpreting these responses. I  
13 won't spend very much time at all on this. But I just want  
14 to say that this is basically an intersection of some of the  
15 data that Don Brambilla collected. This is looking at peak  
16 baseline measurements of HIV-RNA. In terms of the Roche  
17 assay, we found basically about 90 to 95 percent of  
18 successive measurements, in other words, repeated baseline  
19 measurements in the absence of a treatment effect, were  
20 within a range of about 0.5 log peps up to about 6 logs.

21           So the Chiron assay, and this is the only comment  
22 I will make about the Chiron assay, I don't have any  
23 clinical progression data related to the Chiron assay, but  
24 what we did find for another study, ACTG 306 in fact, was

1 that we had a similar level of variability for individuals  
2 who started out at greater than 10,000 copies but the  
3 variability with the Chiron assay was somewhat greater than  
4 with the Roche assay for individuals less than 10,000  
5 copies. This is a result that we have picked up for  
6 descriptive analyses and it requires further investigation.  
7 It was hinted at in some of the data, I think, that Don put  
8 up today.

9 (Slide)

10 What I will do before summarizing the results is  
11 just make some comments about the relationship between CD4  
12 response and RNA response. What we have here is a plot of  
13 the week-24 RNA response against the week-24 CD4 response.  
14 As you would probably expect, there is a negative  
15 relationship in the sense that individuals with reductions  
16 in RNA tend to have increases in CD4.

17 Importantly, this response is not particularly  
18 strong. It is statistically significant but the correlation  
19 coefficient is only 0.3. In fact, if you look at any given  
20 range of RNA responses, say, zero to 1 log reductions, the  
21 range of CD4 responses is very wide-ranging, from almost 200  
22 cell decreases to 200 cell increases.

23 (Slide)

24 So what this suggests is that given the potential

1 for discordance between the RNA response and the CD4  
2 response, the two markers could potentially both be useful  
3 in assessing treatment response and predicting clinical  
4 progression.

5           What I have here is just one summary of the  
6 various joint analyses that we did for CD4 and RNA. What we  
7 have again on the X axis is the RNA response divided into 3  
8 groups, no response, moderate response and better response,  
9 and 3 curves corresponding to the CD4 response, no response,  
10 moderate response, greater response.

11           In all cases we see, as we go from the better RNA  
12 response to the moderate response, we get an increase in  
13 clinical progression risk. This is, in fact, statistically  
14 significant. But then as we move from the moderate response  
15 to no response, the increase in clinical progression risk is  
16 really very dependent on the CD4. If you combine a poor RNA  
17 response with a poor CD4 response, that is clearly much  
18 worse than if you combine a poor RNA response with perhaps a  
19 moderate CD4 response.

20           So that is really suggesting to us that both CD4  
21 and RNA contain prognostic information and perhaps both  
22 should be used in assisting prognosis.

23           (Slide)

24           By way of summary of the key points, the reduction

1 in clinical progression is proportional to the reduction in  
2 HIV-RNA out to 24 weeks. That basically implies the second  
3 point here, the larger the reduction, the better the  
4 decrease in clinical progression risk.

5 I should point out the caveat that we didn't have  
6 a high percentage of individuals reaching below the limits  
7 of delectability by the Roche assay in this data set, and we  
8 were dealing with responses out to about 1.5 logs so whether  
9 this proportional relationship persists for larger  
10 reductions is not clear from these data.

11 The implication then from this proportionality  
12 relationship is that any response beyond what could be  
13 deemed to be just assay variability or biologic variability  
14 could be seen as clinically beneficial. So one figure that  
15 has been banded around is about a 0.5 log drop as being  
16 indicative of biologic variability. Don had a slightly  
17 larger figure this morning, maybe 0.5 to 0.7 .

18 (Slide)

19 The last two points are to summarize for you that  
20 the baseline level of HIV-RNA doesn't seem to modify the  
21 clinical interpretation of the week-24 RNA response in the  
22 sense that a given response can be interpreted singly for  
23 individuals of different baseline levels.

24 Although the baseline level is an independent



1 predictor of clinical progression, it just doesn't modify  
2 the interpretation of the RNA response.

3           The last point that I would just reiterate again  
4 is the importance of CD4 as an additional indicator of  
5 prognosis, over and above RNA. Thank you.

6           DR. HAMMER: Thank you. I would like to ask the  
7 Committee members to hold their questions until the end of  
8 this section and we will have a combined and answer period  
9 later.

10           DR. LIPSKY: I would just like to clarify one  
11 thing, please.

12           DR. HAMMER: One clarification, sure.

13           DR. LIPSKY: Could you put up the only Kaplan-  
14 Meier plot that you have, please?

15           (Slide)

16           Does that indicate that a group where the viral  
17 counts actually went up greater than baseline did better  
18 than some groups where they actually went down?

19           DR. MARCHNER: Yes, in a sense because you might  
20 have a group that started out lower and had perhaps no  
21 decrease or slight increase, and then comparing that with  
22 individuals that started out very high, the former group may  
23 have a better prognosis because they started, say, from  
24 10,000 and went to 15,000 or 20,000 compared with starting

1 at 100,000 and maybe going down to 80,000 or going up to  
2 120,000.

3 DR. LIPSKY: Clearly, something like that is  
4 happening?

5 DR. MARCHNER: Yes, that is what you would expect  
6 because the former group had a lower level to begin with.

7 DR. LIPSKY: Even though they are virologically  
8 getting worse?

9 DR. MARCHNER: Right. You have to distinguish  
10 between improvement in progression rate and absolute  
11 progression rate. The first group that you talked about  
12 didn't have any improvement but they were starting from a  
13 better beginning point than someone who had a very high  
14 absolute progression rate and had a slight improvement to  
15 improve that high progression rate but still didn't get  
16 down.

17 DR. LIPSKY: So your data would say that an  
18 increase in viral RNA is not necessarily a bad thing?

19 DR. MARCHNER: Well, it is a bad thing in the  
20 sense that you haven't got the improvement of the therapy.  
21 You haven't gotten any benefit from the therapy. But if you  
22 have an increase in RNA, it doesn't necessarily mean that  
23 your absolute risk is going to be worse than someone who had  
24 a moderate decrease. That person with the moderate decrease

1 started out from an extremely high level.

2 DR. HAMMER: If you went up from 2,000 to 10,000,  
3 you went up but your risk is going to be lower than if you  
4 went from 500,000 to 100,000.

5 DR. LIPSKY: It seems though, you know, we are  
6 looking at a cutoff at 55,000 on baseline--that is up there.  
7 I mean, there is something that seems a little bit  
8 disturbing about that Kaplan-Meier plot because we are not  
9 talking about 10,000 and 2,000. You are up at 55,000.

10 DR. MARCHNER: Yes, but we are dealing with very  
11 large categories here. If you break things down into  
12 smaller categories or deal with things on continuous levels  
13 the same sorts of results are going to apply. I don't think  
14 it is particularly surprising that a person's absolute level  
15 should be primarily indicative of where their absolute risk  
16 is at any given point of time. Therefore, someone with a  
17 lower value could well be better off even if they didn't  
18 have a good response.

19 DR. LIPSKY: Well, I think that certainly has  
20 strong implications for changing therapy etc., because you  
21 clearly have a group that appears to have done better than  
22 another group --

23 DR. MARCHNER: The point is you have to take into  
24 account two things, not just the response but where the

1 person is.

2 DR. LIPSKY: You showed some graphs that were  
3 called linear. Could they not be sigmoidal, particularly  
4 the first one, actually log linear and intercepts below  
5 zero. Does that not have implications about threshold?

6 DR. MARCHNER: The evidence for a sigmoidal  
7 relationship in any of those plots wouldn't have been very  
8 strong. I certainly want to draw the caveat that we don't  
9 have a lot of individuals going down to extremely low  
10 levels, and whether or not that trend persists is not 100  
11 percent clear from these data. But I wouldn't be prepared  
12 to argue that those curves were supporting any sort of  
13 threshold relationship.

14 DR. LIPSKY: But it is certainly log linear if it  
15 is a linear relationship.

16 DR. MARCHNER: Yes. That is the sort of natural  
17 mathematical scale to present risks on.

18 DR. HAMMER: Thank you. There will be time for  
19 more questions later. We are going to start now having a  
20 group of four presentations by pharmaceutical manufacturers.  
21 The first presentation will be by Ralph DeMasi and Lynn  
22 Smiley, from Glaxo Wellcome.

23 **Presentation by Lynn Smiley, M.D.**

24 DR. SMILEY: On behalf of Glaxo Wellcome, we

1 appreciate the opportunity to present our data today. I  
2 would also like to echo the acknowledgements given by Dr.  
3 Jeff Murray at the reopening of the session earlier this  
4 afternoon. The work presented today and tomorrow really  
5 represents a culmination of collaboration over the past year  
6 and a half among many groups.

7 (Slide)

8 Following some introductory overviews of the  
9 trials, we have analyzed, and that will be presented today,  
10 I am going to turn this over to Dr. Ralph DeMasi, a  
11 statistician within our group, who was project leader  
12 internally within Glaxo Wellcome's initiative.

13 (Slide)

14 What you are going to see today is a cross-study  
15 analysis done retrospectively of six controlled trials,  
16 sponsored by Glaxo Wellcome, that were completed within the  
17 past two, two and a half years. About two-thirds of the  
18 data are on patients who received zidovudine plus 3TC. The  
19 remaining one-third were randomized to the control  
20 treatments.

21 The cross-study analysis included the CAESAR  
22 trial. The CAESAR study which was our adult clinical  
23 endpoint study for 3TC, 85 percent of those patients were  
24 treatment experienced.

1           The NUCA and NUCB3001 studies, which were  
2 surrogate marker trials in less advanced patients, and these  
3 patients were naive. The NUCA and NUCB3002 studies were  
4 conducted similarly but included treatment a experienced  
5 population.

6           The AVANTI01 trial was a trial of ZDV-3TC versus  
7 ZDV-3TC plus and investigational nucleoside reverse  
8 transcriptase inhibitor.

9           Our data are from 1,581 patients who had RNA done  
10 at baseline and at least one follow-up visit. Our endpoints  
11 are from 197 patients and included 268 clinical disease  
12 progression events, either new AIDS-defining events or  
13 death.

14           (Slide)

15           All clinical trails were randomized, double-  
16 blinded, controlled studies, with a mean duration follow-up  
17 of 1 year. All AIDS events in the CAESAR trial were  
18 independently reviewed by an endpoint committee. The plasma  
19 samples were tested using the Roche Amplicor assay in all  
20 studies. As I mentioned, RNA was measured at baseline and  
21 approximately every 4-8 weeks on study.

22           (Slide)

23           The metrics of RNA response included looking at  
24 nadir or the lowest level achieved, as well as the peak

1 response, which was the maximum change from baseline. Ralph  
2 will show some data that includes the 8-28 week mean change  
3 from baseline, as well as an 8-52 week mean change from  
4 baseline. Cox multiple regression model was used and the  
5 intent-to-treat population was analyzed.

6 (Slide)

7 The demographics and baseline characteristics  
8 showed that approximately half of the population in this  
9 cross-study analysis was treatment naive, as defined by  
10 having received less than 6 months of nucleoside therapy.  
11 Predominantly male population. The disease stage is shown,  
12 partition across CD4 A, B and C. The mean CD4 count was  
13 around 200 and the mean RNA level at baseline was about 4.8  
14 logs, or about 60,000 copies/mL.

15 At this point I am going to turn it over to Dr.  
16 Ralph DeMasi to present the results and conclusions.

17 **Presentation by Ralph DeMasi, Ph.D.**

18 (Slide)

19 DR. DEMASI: I would like to start off this part  
20 of my presentation with some descriptive analyses, looking  
21 at the correlation between the magnitude and duration of RNA  
22 reduction and the reduced incidence of risk of clinical  
23 progression. I realize that some of this may be hard for  
24 you to read so I am going to walk the axes, the Y axis and

1 the X axis and then the particular points that we are  
2 looking at on the plot.

3           This particular plot looks at the progression  
4 incidence by duration of RNA below 5,000 copies/mL. What we  
5 have here on the Y axis is progression incidence, defined as  
6 the number of events per patient year of exposure. The X  
7 axis represents the duration of RNA reduction of 5,000  
8 copies/mL. The particular points along the X axis, the  
9 first point here is zero weeks. This means the patients who  
10 have actually shown no reduction in RNA to below 5,000. The  
11 next point is 0-12 weeks or up to 12 weeks reduction; 12-24  
12 week reduction and then greater than 24-week reduction to  
13 the far right. Again, the Y axis is the number of events  
14 per year of exposure, and the higher it is, the higher the  
15 risk for incidence of clinical progression.

16           What this first analysis indicates is the  
17 relationship here between the more durable RNA response and  
18 the decreased risk of clinical progression, and that the  
19 effect of longer duration of suppression is roughly  
20 proportional to the decreased risk in clinical incidence.  
21 The numbers here indicate the actual number of events, or  
22 204 events and the 722 at the lowest point, and then at the  
23 highest 7/529 patient years.

24           (Slide)



1           Now I would like to turn to looking at different  
2 metrics of the magnitude of response. This first metric is  
3 the maximum change in RNA. It is a peak response over the  
4 treatment period, and now I have stratified this analysis by  
5 baseline RNA values.

6           The Y axis here is actually the progression  
7 incidence over 100 patient years of exposure. The X axis is  
8 the magnitude of the peak response categorized into certain  
9 distinct categories. A 2 log reduction, 1.5-2 log  
10 reduction, 1-1.5, 0.5-1, 0-0.5 and an increase in RNA over  
11 the treatment period.

12           We have two lines here, the top or pink line is  
13 for patients who started out with RNA values above the  
14 median. So these were patients who had higher baseline  
15 RNAs. The green line here is for the patients who had lower  
16 baseline RNAs.

17           What we can see here again is proportionality of  
18 effect between a better RNA response or a 2 log reduction on  
19 the far left, translated to a very rare incidence of  
20 clinical progression, and on up to patients with a very  
21 modest or no reduction having the highest risk of clinical  
22 progression. Then in between you can see roughly the  
23 proportionality of the effect.

24           Another point to recognize is that the patients

1 who start out with higher baseline values have a higher risk  
2 of clinical progression independent of the actual RNA  
3 response. I would just like to note that these bars here  
4 are the 95 percent confidence intervals for the estimates.

5 (Slide)

6 This is an analogous presentation showing again on  
7 the Y axis the progression incidence over 100 patient years  
8 of exposure, and on the X axis the magnitude of the  
9 reductions with the same categories as I just discussed.

10 The metric we are looking at now is an 8-28 week  
11 change from baseline and, once again, that is stratified by  
12 baseline RNA. So the pink line is for patients having a  
13 higher baseline; the green line is for patients having a  
14 lower baseline.

15 We can also see using this metric the  
16 proportionality of effect. Patients having a better RNA  
17 response measured by this metric have a lower incidence of  
18 progression. Patients having a worse response have a higher  
19 incidence of progression.

20 I would just note that the uncertainty here, the  
21 stratum having a 2 log reduction, is reflected in the width  
22 of this confidence band.

23 (Slide)

24 The next metric we are going to be looking at is

1 the actual nadirs. So this is the lowest value achieved on  
2 treatment. Once again, that is stratified by baseline RNA.  
3 The Y axis again is number of events per 100 patient years.  
4 On the X axis I have now used the categories of actual  
5 response. To the far left you see less than 400 copies/mL.  
6 The next group is 400 to 5,000 and 5,000 to 20,000 and  
7 greater than 20,000. Once again, we have two lines here,  
8 the pink for patients with higher baseline values and the  
9 green for patients with lower baseline RNA values.

10           Once again here we can see the proportionality of  
11 the effect between the actual level achieved and the  
12 incidence of clinical progression. I would also like to say  
13 that in this plot it appears that the baseline value is not  
14 significant looking at it this way, but when we do this type  
15 of analysis in a multiple regression setting, using the  
16 actual baseline values and the actual nadirs, the baseline  
17 value is still significant.

18           (Slide)

19           This is the same plot as the previous plot but we  
20 pooled data over the patients who have lower and higher  
21 baseline RNAs. There are two differences here. The Y axis  
22 now is the clinical progression rate and the X axis is 10  
23 categories of nadir achieved instead of the 4 that we had in  
24 the previous plot.

1           What you can see here from this presentation is  
2 this proportionality of effect, linear effect relationship  
3 between the actual nadir achieved and the risk of clinical  
4 progression, with patients achieving the lowest values at  
5 lowest risk and patients achieving the highest values at the  
6 highest risk, and in between this proportionality of effect.

7           Before I move on I would like to say that these  
8 analyses have been done looking at Kaplan-Meier estimate of  
9 progression rate and also Cox model, showing that these  
10 results are fairly consistent.

11           (Slide)

12           Now I would like to turn to looking at the joint  
13 effect of CD4 and RNA, looking at the relationship between  
14 these two variables and the risk for incidence of clinical  
15 progression. This plot shows the correlation between 8-52  
16 week CD4 count and RNA and clinical disease progression.

17           The X axis is RNA response in terms of the log  
18 scale and the Y axis is the CD4. The reference lines on  
19 this plot represent a value of about 3.7 log-10, which is  
20 about 5,000 copies, and 200 CD4 counts on the Y axis. The  
21 blue dots are patients who did not progress on the study and  
22 the red squares are patients who progressed on the study.

23           What we can see looking at it this way are a  
24 couple of points. We know that the highest clinical

1 progression incidence occurs when patients have RNA values  
2 greater than 5,000 and CD4 counts less than 200, in  
3 particular, the rate is about 28 percent. When the CD4  
4 count is above 200 and the RNA value is less than 5,000  
5 clinical progression is relatively greater. We can also see  
6 a slight correlation here between CD4 and RNA but,  
7 nevertheless, a wide variability. For patients with a given  
8 RNA value, they have a range of CD4 values.

9 (Slide)

10 This is a similar presentation to the previous  
11 one. What I have looked at here is the 8-28 week mean  
12 changes from baseline and now I am looking at subsequent  
13 clinical disease progression. So this is clinical disease  
14 progression after 28 weeks and we are trying to look at the  
15 temporal relationship between the changes in CD4 and RNA and  
16 the subsequent risk of clinical progression.

17 The X axis ranges from minus 3 to 1.5 and the  
18 reference line is at minus 1. So this is patients who have  
19 had a 1 log reduction. The CD4 axis runs from minus 200 to  
20 300, with the reference line drawn at 50 CD4 cell rise.

21 What we can see here is that patients who did not  
22 achieve 1 log reduction or 50 cell CD4 rise in the 8-28 week  
23 period were at the highest risk of clinical progression and  
24 that patients who achieved such a CD4 and RNA response were

1 at the lowest risk of clinical progression. Once again here  
2 we see a slight correlation, about minus 0.5, between CD4  
3 response and RNA response.

4 (Slide)

5 Now I would like to look at the effects of  
6 baseline CD4 and baseline RNA and CD4 response and RNA  
7 response. What we are doing here is using a Cox multiple  
8 regression model to predict subsequent clinical disease  
9 progression based on an 8-28 week metric. So this is a mean  
10 change from baseline at 8-28 weeks. I have fitted this Cox  
11 model, calculated the hazards ratios and then from those  
12 obtained the percent reduction in risk of clinical  
13 progression. These estimates are for 50 cell CD4 increase  
14 or 1 log reduction. In other words, the hazards ratio for  
15 baseline CD4 count was about 0.5 so that corresponds to  
16 about a 50 percent reduction in risk. The confidence  
17 intervals are noted here. You can see that the fact that  
18 they don't overlap indicates that all these variables are  
19 highly statistically significant, particularly the effects  
20 of baseline CD4 and RNA of about 50 and 60 percent reduction  
21 in risk, or 50 cell increase in CD4 or 1 log reduction in  
22 RNA. Then for the CD4 and RNA response we have a 60 to 70  
23 percent reduction in risk of subsequent clinical disease  
24 progression. It is important to note that we looked at the

1 interaction between CD4 and RNA and this was not  
2 significant.

3 I would now like to spend a couple of minutes  
4 looking at RNA and CD4 as surrogate markers for clinical  
5 disease progression to new a new AIDS event or death in the  
6 CAESAR trail, the clinical endpoint trial. What I am going  
7 to be showing you is the concordancy between the treatment  
8 effects on the CD4 response and the RNA response and the  
9 clinical response as measured by progression to new AIDS  
10 event or death. Then what I would like to show you is what  
11 happens to the treatment effect in the clinical progression  
12 if you remove the treatment effects on the CD4 and RNA  
13 responses.

14 (Slide)

15 The objectives of such a surrogacy analysis, the  
16 main objective of this type of analysis is to answer the  
17 question of whether or not the effect of antiretrovirals on  
18 delayed clinical disease progression is actually mediated by  
19 the antiretroviral therapy on immunologic and virologic  
20 endpoints as measured by CD4 and RNA.

21 I just want to note that there are two other  
22 methods in looking at surrogacy for CD4 and RNA. One of  
23 these is looking to see whether or not treatment regimens  
24 which confer immunologic and virologic benefit compared to

1 control regimens also confer clinical benefit compared to  
2 these regimens. Then, whether or not the converse is true.  
3 In other words, do treatment regimens which do not confer  
4 immunologic and virologic benefit compared to control  
5 regimens also do not confer clinical benefit compared to  
6 these control regimens?

7 I just want to note that we are conducting some  
8 collaborative work with other sponsors and ACTG to look at  
9 this and that work is still in progress.

10 (Slide)

11 I am going to show you now the RNA responses for  
12 the two treatment arms in the CAESAR trial, the placebo arm  
13 here in the pink, and the 3TC arm in the green. What this  
14 plot shows is a median change from baseline in log RNA for  
15 each treatment arm, and this is the time on study.

16 So we can see that the current therapy plus  
17 placebo arm remains essentially unchanged, flat throughout  
18 the treatment period for about one year, as opposed to the  
19 3TC arm which shows a sharp reduction, a 1.5 log drop, and  
20 then a gradual return to baseline but, nevertheless, a  
21 sustained 0.5 log reduction out to about a year of study.

22 The treatment comparisons of the 3TC arm with the  
23 placebo arm here, they are all highly significant. We  
24 looked at different metrics of response, shorter term as



1 well as more median-term responses, using simple metrics  
2 such as the mean.

3 (Slide)

4 The next thing to look at then is the treatment  
5 effect on the CD4 count, and here we see the concordancy  
6 between the treatment effect on RNA and the treatment effect  
7 on the CD4 count. What we see here in the placebo arm, and  
8 this is time on study. This is CD4 count median change from  
9 baseline. There is an actual reduction in the placebo arm,  
10 current therapy plus placebo, versus about a 35 cell rise in  
11 the 3TC arm. Then this is followed by a return to baseline  
12 but, nevertheless, we see about a 35 cell difference  
13 throughout the treatment period.

14 (Slide)

15 This is the Kaplan-Meier of estimates of AIDS-free  
16 survival. What we have here again is the 3TC arm in the  
17 green. I know it is probably difficult to see. In the red  
18 line is the placebo arm. We can see here a very highly  
19 statistically significant relationship in that adding 3TC  
20 was beneficial to placebo with respect to clinical disease  
21 progression.

22 I would like to note that the time scale for the X  
23 axis is from 0-32 weeks because on the next overhead we are  
24 going to be looking at a metric of 12-20 weeks for RAN

1 response and seeing how well that explains this observed  
2 treatment difference during the 20-32 week time frame.

3 (Slide)

4 When we do that, what we do is we fit a Cox  
5 proportional hazards model, and we cover the baseline  
6 survival function and then use that to obtain the treatment  
7 specific survival curves. These are like Kaplan-Meier  
8 curves. What we want to demonstrate is that after we  
9 account for the fact that the treatment arms are different  
10 with respect to CD4 and RNA. If, in fact, they were the  
11 same, what would the difference in the clinical progression  
12 rates be?

13 When we do that, we see that the treatment effect  
14 is clearly non-significant and that the adjusted Kaplan-  
15 Meier curves are essentially superimposing.

16 (Slide)

17 In conclusion, I would like to reiterate that  
18 treatment-induced reductions in RNA reduce the risk of  
19 clinical disease progression. That CD4 adds independent  
20 prognostic information to RNA on the risk of clinical  
21 disease progression.

22 Furthermore, the prognostic significance of RNA  
23 does not depend on baseline CD4 or CD4 response, and that  
24 disease progression is rare for patients with very low RNA

1 and is most common for patients with very high RNA.

2           Finally, delayed progression to new AIDS or death  
3 caused by antiretroviral therapy in the CAESAR trial is  
4 actually mediated by the antiretroviral effect on CD4 and  
5 RNA.

6           DR. HAMMER: Thank you very much. Unless there  
7 are pressing clarification questions, we will hold them  
8 until the general questions. The next speaker is Christy  
9 Chuang-Stein, from Pharmacia and Upjohn.

10                   **Presentation, Christy Chuang-Stein, Ph.D.**

11           DR. CHUANG-STEIN: Well, I hope everybody is still  
12 awake. Good afternoon, everyone.

13                   (Slide)

14           During the next twenty minutes I will share with  
15 you results from Pharmacia's and Upjohn's effort with the  
16 Food and Drug Administration to understand the role of  
17 viral- load reduction in evaluating AIDS treatment.

18                   (Slide)

19           The data that formed the basis of my presentation  
20 came from two large trials conducted by Pharmacia and  
21 Upjohn. Over 2,300 patients were enrolled into these two  
22 trials, and the data of the two studies were combined for  
23 this exercise.

24           The baseline CD4 count for the combined analysis

1 population was around 230 and the geometric mean of the  
2 baseline HIV-1 RNA was around 75,000 copies. The median  
3 follow-up duration was over 1 year, and the longest follow-  
4 up duration exceeded 2 years.

5           Even though the treatment regimens used in these  
6 two trials have become obsolete by today's treatment  
7 standards, the data collected in these two trials,  
8 nevertheless, has offered a wealth of information on the  
9 relationship, or the lack of it, between the viral-load  
10 reduction and the risk of clinical progression.

11           This morning we heard two presentations on the RNA  
12 PCR assay as well as the assay's characteristics. Because  
13 of the information shared with us by Dr. Iacona-Connors and  
14 Dr. Brambilla, I will only mention very briefly here the  
15 results from Pharmacia's and Upjohn's own efforts to examine  
16 the combined HIV-RNA assay variability and the short-term  
17 biologic variation in HIV-1 RNA.

18           For the two trials here, Pharmacia and Upjohn  
19 Company used and RNA PCR assay developed by its own clinical  
20 research laboratories. This RNA PCR has been thoroughly  
21 validated against the Roche Amplicor assay, and was found to  
22 produce results highly correlated with those produced by the  
23 Roche assay, with a correlation coefficient around 0.93.  
24 All the RNA PCR values included in my presentation have been

1 converted to their Roche equivalent.

2           These two trials offered a great opportunity to  
3 examine the combined HIV-RNA assay variability and the  
4 short-term biologic variation in HIV-RNA because two  
5 pretreatment RNA measurements were taken 14 days of each  
6 other after each patient was adequately washed out of their  
7 current antiretroviral therapy.

8           (Slide)

9           Applying the tolerance limits technique to the  
10 difference between the two pretreatment RNA measurements, we  
11 concluded that a viral-load reduction of 0.5 logs --

12           (Slide)

13           --or more was beyond the combined HIV-RNA PCR  
14 assay, as well as the short-term biologic variation in HIV-  
15 RNA.

16           (Slide)

17           We will next concentrate on the prognostic value  
18 of baseline CD4 count, baseline HIV-RNA, as well as the  
19 change in RNA for the subsequent risk of clinical  
20 progression. For convenience I will use the term clinical  
21 progression to include death.

22           There are several metrics that one can use to  
23 characterize the change in RNA or the RNA response. The  
24 metric we chose was the peak reduction in RNA during the

1 first 12 weeks of treatment, with the peak reduction  
2 expressed on a log 10 scale. To evaluate the merit of using  
3 this peak reduction during the first 12 weeks of treatment,  
4 we restricted the analysis to those individuals who were in  
5 the studies for at least 12 weeks and did not experience any  
6 clinical progression during these 12 weeks.

7           Two measures were used to quantify the risk of  
8 clinical progression. The first was simply the proportion  
9 of individuals who experienced clinical progression. The  
10 second one was based on the incidence rate of clinical  
11 progression, defined as the number of individuals  
12 experiencing clinical progression with 10,000 days of follow  
13 up.

14           (Slide)

15           We used the FDA's suggestion to classify the  
16 baseline HIV-1 RNA into 5 categories ranging from less than  
17 20,000 copies to greater than half million copies. As for  
18 the peak reduction, we first classified the peak reduction  
19 into 3 categories, greater than 1 log, between 0.5 and 1 log  
20 and less than 0.5 log. This classification was based on an  
21 earlier observation that a 0.5 log or greater reduction  
22 represented a real RNA response in our trials. Our second  
23 classification was based on the quintile of the peak  
24 response of the analysis population. The second

1 classification was used to decide if the risk of clinical  
2 progression was a monotonal function of the RNA reduction in  
3 the first 12 weeks.

4           These two graphs show the risk of clinical  
5 progression for the 3 reduction categories. The X axis here  
6 represents the baseline HIV-1 RNA on the original scale.  
7 This first graph, here, shows a greater risk of clinical  
8 progression for smaller reduction in the HIV-RNA during the  
9 first 12 weeks. This relationship was especially clear for  
10 higher baseline HIV-1 RNA values.

11           (Slide)

12           An almost identical pattern was observed when we  
13 expressed the risk in terms of the incidence rate of  
14 clinical progression for 10,000 days of follow up.

15           (Slide)

16           The next two transparencies show the risk for the  
17 5 reduction categories. Except for a few instances where  
18 things get a little bit switched, the inverse relationship  
19 between the RNA reduction and the risk of clinical  
20 progression was apparent in this graph. In addition, the  
21 risk of clinical progression continued to decline with  
22 higher baseline HIV-1 RNA values. Furthermore, these  
23 findings were independent of the measures used to quantify  
24 the risk of clinical progression.

1 (Slide)

2 We have also looked at the role of baseline CD4  
3 count and the risk of clinical progression. In order to do  
4 that, we fit a proportional hazards model using the 5  
5 reduction categories as well as the 2 baseline marker values  
6 as predictors. Recall that the 5 reduction categories were  
7 determined using quintiles. Therefore, there were about the  
8 same number of patients in each of these 5 categories here.  
9 In addition, the baseline CD4 count, which was treated as a  
10 continuous variable, was expressed in a unit of 25 cells and  
11 the baseline HIV-1 RNA was expressed as a unit of 0.5 log.

12 Our analysis showed that both baseline CD4 as well  
13 as baseline HIV-1 RNA were highly correlated with clinical  
14 progression. Under the fitted model, the model suggested  
15 that a higher baseline CD4 count in the amount of 25 cells  
16 was associated with a 15 percent less risk of clinical  
17 progression. On the other hand, a higher baseline HIV-1 RNA  
18 value in the amount of 0.5 log was associated with a 65  
19 percent increase in the risk of clinical progression. The  
20 estimated hazards ratio of the various reduction categories  
21 relative to the first one decreased monotonally as the  
22 amount of reduction increase.

23 I would like to point out here that in including  
24 these 5 categories in the model we treat this one, the



1 category with the least amount of reduction, as the  
2 reference group. Therefore, the risk was measured against  
3 the reference group. That is where the hazards ratio came  
4 from. Therefore, the hazards ratio for the reference group  
5 was set equal to 1.

6 (Slide)

7 So far we have shown you results from classifying  
8 the baseline HIV-1 RNA according to the suggestion of the  
9 FDA. We have also conducted an analysis using the baseline  
10 RNA classification suggested by the Surrogate Marker  
11 Collaborative Group.

12 The baseline RNA categories suggested by the SMCG  
13 ranged from less than 5,000 to greater than 200,000 copies.  
14 Here I have shown you the calculated risk of clinical  
15 progression for each of the 5 baseline HIV-1 RNA categories.  
16 You will notice that the distribution of patients into these  
17 5 categories is not as smooth or not as even as that among  
18 the 5 categories recommended by the FDA.

19 The striking finding here is that none of the 123  
20 patients in our analysis population who started the trials  
21 with a baseline HIV-1 RNA less than 5,000 experienced any  
22 clinical progression during the trial's period.

23 (Slide)

24 In addition to this observation, our earlier

1 estimate of the hazards ratio also suggested that an 0.5 log  
2 reduction in the HIV-1 RNA during the first 12 weeks was  
3 associated with a 38 percent less risk in the risk of  
4 clinical progression. The corresponding figure of 1 log  
5 reduction during the first 12 weeks was 57 percent. Also,  
6 the clinical benefit of a viral-load reduction beyond 1.3  
7 logs appears to flatten out in our trials.

8 (Slide)

9 In addition to looking at the peak reductions  
10 during the first 12 weeks, we also looked at the duration of  
11 virologic response to see whether this duration has any  
12 prognostic value for subsequent risk of clinical  
13 progression.

14 For this analysis we included only those patients  
15 who were in the studies for at least 24 weeks and did not  
16 experience any clinical progression during the first 24  
17 weeks. Also, in order to conduct this analysis we needed a  
18 definition for virologic response. Based on our earlier  
19 observation of what constituted a true RNA response, we  
20 defined for our analysis for the data from our trials a  
21 virologic response by at least 0.5 log reduction in HIV-1  
22 RNA from the baseline value. For those individuals who did  
23 achieve a response during the first 24 weeks, we defined  
24 virologic failure for our analysis a viral-load rebound to

1 be within 0.5 log from the corresponding baseline.

2 For this analysis, when we looked at the duration  
3 we looked at the 24-week period, and the response duration  
4 during this 24-week period is then calculated as to the  
5 period between the time when the virologic response was  
6 first observed and the time when the virologic failure was  
7 first declared.

8 For those individuals who did not fail by week 24,  
9 we truncated their response at week 24 and calculated the  
10 response duration accordingly. This convention, indeed, did  
11 not differentiate between virologic failure from the  
12 continued response at week 24. However, since more than 95  
13 percent of the responders responded by week 12, a short  
14 response duration, such as less than a 8 weeks, implied a  
15 true virologic failure at week 24. Therefore, the results  
16 from this analysis can be best interpreted who never  
17 responded and those who had very short response duration and  
18 those who had a long response duration.

19 (Slide)

20 We divided the response duration into 5  
21 categories, again by the quintile of the response duration  
22 distribution. Because the plots based on the true risk for  
23 the two measures for risk of clinical progression are  
24 extremely similar, I will only show you here the one based

1 on the incidence rate of clinical progression.

2           The curve for the non-responders is above all the  
3 remaining curves. This might be a little hard to see. In  
4 order to better differentiate among the four middle response  
5 duration categories, I added green to the four corresponding  
6 curves. In addition, I added blue to the curve pertaining  
7 to the longest response duration group.

8           As can be seen from this graph, there is a trend  
9 for a decrease in the risk of clinical progression or an  
10 increase in the responders' duration.

11           (Slide)

12           The estimated hazards ratio for the various  
13 response categories relative to the non-responders using the  
14 proportionate hazards model are given on this transparency.  
15 The pattern among the estimated hazards ratio confirms the  
16 trend observed earlier, and there is a trend of decreasing  
17 risk with a longer response duration, consistent with our  
18 earlier observation on the relationship between the baseline  
19 CD<sub>4</sub> and the baseline HIV-1 RNA with clinical progression.  
20 This analysis only confirms that highly significant  
21 association between the baseline marker values with the  
22 clinical progression.

23           (Slide)

24           What are the implications of our findings for

1 study designs of trials to evaluate the HIV treatment?  
2 Obviously, it is important to consider stratifying  
3 randomization by the baseline CD4 count and baseline HIV-1  
4 RNA values because of their strong association with clinical  
5 progression.

6 Our second point here is really a question for the  
7 members of the Advisory Committee as well as the experts in  
8 the AIDS arena. Remember, in our trials we did not observe  
9 any clinical progression in those individuals in our  
10 analysis population who started the trial with an HIV-RNA  
11 less than 5,000 copies.

12 In view of that observation, how can we conduct  
13 the benefit against risk of the highly active antiretroviral  
14 therapies in patients with very low viral load while the  
15 clinical benefit of HAART might not be realized for a long  
16 time? The risk of the HAART can be felt acutely through  
17 treatment side effects and drug toxicity. We don't have an  
18 answer to this question. We would simply like to bring the  
19 issue up for the community to consider.

20 Finally, considering the relationship between  
21 response duration and the risk of clinical progression  
22 identified in our analysis, we feel it is important that the  
23 trials be long enough to capture information on the response  
24 duration. In our opinion, trials to evaluate a regimen's

1 ability to suppress viral load should be at least 24 weeks  
2 long.

3 This concludes the Pharmacia and Upjohn  
4 presentation this afternoon. We would like to thank the  
5 Agency for the opportunity to participate in this important  
6 project and the chance to share the important scientific  
7 knowledge learned from Pharmacia's and Upjohn's effort to  
8 combat HIV infection. Thank you.

9 DR. HAMMER: Thank you very much. It is not  
10 scheduled but I am going to take the prerogative to have a  
11 10- or 15-minute break now. Then we will return promptly  
12 and finish up the afternoon's presentations.

13 [Brief recess]

14 DR. HAMMER: Please take your seats. We are going  
15 to continue the pharmaceutical sponsor presentations. The  
16 next speaker is Lesley Struthers, Hoffman La Roche.

17 **Presentation, Lesley Struthers**

18 (Slide)

19 MS. STRUTHERS: My name is Lesley Struthers. On  
20 behalf of Hoffman La Roche, I will present the data analysis  
21 we performed exploring the relationship between RNA regimens  
22 and the time to first AIDS-defining event or death. I will  
23 present data on study design and baseline information.

24 We looked at various definitions in defining a

1 virologic responder and how long this definition identifies  
2 those patients who later go on to have AIDS-defining events  
3 or die. We explored the impact of duration of RNA effect,  
4 and I will also show you the importance of CD4 as a  
5 surrogate marker.

6 (Slide)

7 All analyses presented here are taken from one  
8 large study, ND-12456, U.S. study. This was a double-blind,  
9 randomized study involving 940 patients in the intent-to-  
10 treat analysis. The patients who had experienced AZT but  
11 were naive to protease inhibitors, ddC and ddI. There were  
12 three treatment arms, ddC, saquinavir and the combination of  
13 these two treatments.

14 The primary endpoint was time to first AIDS-  
15 defining event or death. There were 223 endpoints, not 243  
16 as on this slide. These were patients who were followed for  
17 a median time of 17 months. Both RNA, using the Roche  
18 Amplicor kit, and CD4 were measured every 4 weeks up to 16  
19 weeks, and then every 8 weeks up to 80 weeks. So you can  
20 see this is a large study with a considerable amount of  
21 information.

22 (Slide)

23 The distribution of baseline CD4 and RNA are  
24 represented here in the pie graphs. A large proportion of

1 patients, 70 percent, had a baseline CD4 between 100-300.  
2 The median baseline was approximately 170 across all 3  
3 treatment arms.

4 For RNA, the patients were split to 26 percent  
5 having RNA values below 50,000, 36 percent of patients with  
6 a value between 50,000 and 200,000 and 39 percent of  
7 patients with a value greater than 200,000. The median  
8 baseline across the 3 treatment arms was approximately 5.1  
9 logs.

10 (Slide)

11 This graph shows the absolute levels of CD4 and  
12 RNA together, looking at the area under the curve over the  
13 48 weeks, with RNA along the X axis and CD4 along the Y  
14 axis. The red circles show all patients who were alive and  
15 did not have an AIDS-defining event. The squares indicate  
16 the patients who either died or had an adverse AIDS-defining  
17 event.

18 It is clear from this graph that the majority of  
19 patients with an AIDS-defining event who had died, the  
20 yellow squares, had CD4 values of less than 200 and RNA  
21 values greater than 10,000.

22 Using these cutoff values, we can also see that a  
23 large number of red circles are in this lower right-hand  
24 quadrant, meaning we are also falsely identifying these



1 patients.

2 (Slide)

3 One of our first objectives was to see whether  
4 there is a virologic response cutoff level identifying  
5 patients who would later have an AIDS-defining event or die.  
6 Virologic response is defined as an absolute level of RNA  
7 which the patient had to reach in the first 24 weeks before  
8 they could be classified as a responder.

9 In this graph we looked at two cutoff values,  
10 10,000 and 100,000. The bottom line is the time to an AIDS-  
11 defining event or death for those patients whose absolute  
12 RNA level remains above 100,000 during the first 24 weeks.  
13 The patients in the top line have at least one value of less  
14 than 10,000 in the first 24 weeks, and the middle line  
15 containing the remaining patients.

16 This clearly shows that an absolute RNA cutoff  
17 level in the first 24 weeks has an effect on time to first  
18 AIDS-defining event or death.

19 (Slide)

20 In addition to looking at the 10,000 and 100,000,  
21 we also looked at different RNA cutoff levels. In fact, we  
22 looked at 15,000, 20,000, 30,000, 40,000 and 50,000. In the  
23 graph we tried to summarize our information.

24 (Slide)

1           These cutoff levels are displayed along the X  
2 axis. We looked at the effect of having 1 response value in  
3 the first 24 weeks. This is the blue line. We looked at 2  
4 consecutive response values, the pink line, and then 3  
5 consecutive values, the yellow line. The Cox model was used  
6 here to calculate the relative risk and includes baseline  
7 CD4 and RNA and treatment, where the RNA cutoff value  
8 defined the patient as a responder or a non-responder.

9           The relative risk, displayed along the Y axis, is  
10 the ratio of the hazards of progression to AIDS-defining  
11 event or death for responders and patients who did not  
12 respond. The greater the difference from a relative risk of  
13 1, the greater the difference between responders and non-  
14 responders. All levels of RNA on this graph are important  
15 with the relative risk between 0.5 and 0.3, and all of them  
16 clearly very different from 1. A cutoff value of 10,000 or  
17 above is equally as effective in predicting AIDS-defining  
18 event or death as the higher cutoff values are.

19           (Slide)

20           We also looked at the effect of change from  
21 baseline over the first 24 weeks. Here, the patients are  
22 split into 3 groups according to whether they had greater  
23 than 1 log decrease in RNA. This is the top line. Between  
24 a 0.5 and 1 log decrease, the middle line, and less than a

1 0.5 log decrease, the bottom line.

2           The clear difference in lines indicates the  
3 smaller the change from baseline, the greater the  
4 probability of suffering an AIDS-defining event or death.  
5 Interestingly, this metric did not split the patients as  
6 widely as when we looked at the virologic response  
7 definition as absolute difference in RNA of 100,000.

8           (Slide)

9           Next we looked at the additional effect of  
10 baseline CD4 and RNA on time to first AIDS-defining event or  
11 death after the RNA responses were taken into account.

12           (Slide)

13           Here, when looking at patients who had a value of  
14 RNA less than 100,000 during the first 24 weeks, and then  
15 switching these patients by their CD4 median baseline value,  
16 there is a clear effect of CD4 baseline on the first AIDS-  
17 defining event or death.

18           (Slide)

19           Now we look at the effect of baseline RNA on the  
20 percentage of patients who progressed to the first AIDS-  
21 defining event or death. The baseline is split in quartiles  
22 so each of these 4 arms has the same number of patients in  
23 it. The change from baseline in the first 24 weeks is also  
24 split into quartiles.

1           This 3D plot demonstrates that the smaller the  
2 change in RNA over 24 weeks, the higher the block. So more  
3 patients have progressed. This trend is demonstrated over  
4 the different baseline levels. It is also clear that the  
5 baseline RNA has an effect and the higher the baseline, the  
6 higher the block, and so the greater the likelihood of  
7 progression.

8           (Slide)

9           One of the objectives was also to examine the  
10 duration. The area under the curve, AUC, takes into account  
11 the level of RNA as well as the duration. The patient who  
12 has a low RNA and maintains this will have a low AUC,  
13 whereas, a patient whose RNA drops initially and then  
14 increases will have a higher AUC. This graph demonstrates  
15 that the metric AUC for the first 24 weeks, when split by  
16 quartiles, is clearly a nice, strong prognostic indicator.

17           In the next graph we are going to show you the  
18 importance of following RNA throughout the study and using a  
19 virologic failure definition.

20           (Slide)

21           We used virologic failure cutoff levels of 5,000,  
22 10,000, 15,000, 20,000, 30,000, 40,000, 50,000 and 100,000.  
23 These are displayed along the X axis in this graph. As soon  
24 as the patient has 1 RNA level above our cutoff used in this

1 virologic failure definition, they are classified as a  
2 failure, the blue line, at the time point that the failure  
3 occurred. We have also examined the effect if patients have  
4 2 consecutive RNA values as failures, the pink line, and 3  
5 consecutive failures, the yellow line.

6           The Cox model used here, in this analysis,  
7 includes baseline RNA, CD4 and treatment, with the RNA value  
8 defined as failure or not a failure being used in the model  
9 as a time-dependent variate. So this means that we don't  
10 just look at the RNA values during the first 24 weeks; we  
11 look at them through the life of the study.

12           The relative risk on the Y axis is the ratio of  
13 Cox hazards regression based upon an AIDS-defining event or  
14 death. A failure is compared to patients who do not fail.  
15 Again, the further the relative risk from 1, the greater the  
16 difference between failures and non-failures.

17           Here, the relative risk demonstrates that with  
18 cutoff levels of 10,000 or above the patients defined as  
19 failures are more likely to progress than patients who have  
20 not failed. The risk also increases as we raise the cutoff  
21 level along the X axis.

22           If we just look at 2 examples on this graph, and  
23 look at the pink line, so those patients who have 2  
24 consecutive measures that count as a virologic failure,

1 using a cutoff of 10,000 the relative risk is 1.5, which  
2 means patients classified as failures have a 50 percent  
3 increase in having an AIDS-defining event or dying compared  
4 to those whose RNA stays below 10,000. With a cutoff level  
5 of 100,000, the relative risk is 2. So patients classified  
6 as failures have a 100 percent increase in having an AIDS-  
7 defining event or dying compared to those who remained below  
8 100,000.

9 (Slide)

10 This slide looks at the effect of CD4 during the  
11 study, as well as the effect of RNA. Here the data is split  
12 into 3 groups based on likely change of response in CD4 and  
13 RNA, using the area under the curve minus the baseline over  
14 the 24 weeks compared to their respective means. The plus  
15 indicates patients who were above the mean. This means they  
16 have a high CD4 or high RNA. The minus indicates patients  
17 below the mean.

18 The top line shows that patients who have a high  
19 CD4 and a low RNA do considerably better compared to  
20 patients who have a low CD4 and a high RNA.

21 We have seen from the data so far that both CD4  
22 and RNA are important prognostic factors, and that baseline  
23 CD4 and RNA provide additional prognostic information.

24 I will now go through surrogate marker analysis

1 that we used to investigate how effective our surrogate  
2 markers are using the Prentice criteria.

3 (Slide)

4 For this analysis 2 treatment arms are used, ddC  
5 and saquinavir plus ddC.

6 (Slide)

7 The Prentice criteria states that first we need to  
8 have a clear treatment effect in our time to AIDS-defining  
9 event or death, and we clearly demonstrate this between the  
10 2 arms, with a significant difference, a p value of less  
11 than 0.0001.

12 (Slide)

13 The Prentice criteria also states that we need to  
14 have a clear treatment effect in our surrogate markers.  
15 Here, looking at RNA, we have a clear difference between the  
16 2 treatment arms, looking at the area under the curve minus  
17 the baseline over 24 weeks.

18 (Slide)

19 With CD4 we also see a clear difference over the  
20 first 24 weeks in the area under the curve minus the  
21 baseline.

22 (Slide)

23 Given that we have a clinical difference and  
24 surrogate marker differences, we used a Cox model initiated

1 by the baseline CD4 and RNA in the model, which are both  
2 highly significant, with p values of less than 0.01. We can  
3 see that this is significant.

4 (Slide)

5 Next we add both RNA and CD4, area under the curve  
6 minus the baseline. We can see that both baselines and the  
7 area under the curve minus the baseline relative to CD4 and  
8 RNA are significant. We can also see that the treatment  
9 effect is no longer significant. This means that the  
10 surrogate markers are explaining some of the treatment  
11 effect.

12 (Slide)

13 Pooling data together from several models, the  
14 AUCMB RNA by itself explains 35 percent of the treatment  
15 effect; 49 percent of the effect is explained by CD4 AUCMB  
16 in the 24 weeks. Together, as in this model, they both  
17 explain 61 percent of the treatment effect.

18 (Slide)

19 Here I show this graphically. Here we have the  
20 model which is baseline CD4 and RNA values and we can see  
21 that there is a treatment effect there.

22 (Slide)

23 Once we put AUCMB RNA and CD4 in the model, we can  
24 see that the tap is closed, the yellow lines compared to the



1 blue lines.

2 (Slide)

3 In summary, our conclusion from this analysis is  
4 that RNA is a very strong prognostic indicator. It appears  
5 that the level of RNA may be more important than change from  
6 baseline. The duration of RNA when we look at the area  
7 under the curve indicates that duration of effect is  
8 important. From this data, based on experienced patients,  
9 we also clearly show that CD4 is equally as important as  
10 RNA. Thank you.

11 DR. HAMMER: Thank you very much. The next  
12 speaker is Margo Heath-Chizzoi, from Abbott Laboratories.

13 **Presentation, Margo Heath-Chizzoi, M.D.**

14 (Slide)

15 DR. HEATH-CHIZZOI: I am Margo Heath-Chizzoi, and  
16 on behalf of Abbott Laboratories I would like to thank the  
17 FDA for inviting us to present the correlations of HIV  
18 changes with clinical benefit demonstrated in the Abbott  
19 study M94-247.

20 (Slide)

21 The M94-247 study evaluated clinical benefit in  
22 patients with advanced HIV illness. At baseline the  
23 patients had to have CD4 cell count less than 100 cells/mL<sup>3</sup>,  
24 at least 9 months of prior approved antiretroviral therapy.

1 The patients were allowed to continue up to 2 but not more  
2 than 2 concurrent, approved agents during the time the study  
3 was conducted. It included only nucleoside agents. They  
4 had to have a Karnofsky score of greater than 70, and they  
5 needed to not have active opportunistic infections requiring  
6 induction of therapy. Maintenance therapy and secondary  
7 prophylaxis agents were included in the study.

8 (Slide)

9 The study enrolled 1,090 patients who were  
10 randomized to either ritonavir 600 mg b.i.d. to their  
11 chronic regimen, and that arm included 543 patients, or  
12 placebo added to their current regimen in 547 patients.

13 The primary objective of the study was to evaluate  
14 clinical endpoints, which were defined as new AIDS-defining  
15 illness, with the exception that recurrent PCP, Candida  
16 esophagitis or prolonged mucocutaneous herpes were allowed  
17 as recurrent events, and death was included as a primary  
18 endpoint if patients didn't have another new disease  
19 diagnosed prior to death.

20 A surrogate marker sub-study was nested in this  
21 study for the first 16 weeks of evaluation. During these 16  
22 weeks patients were asked to continue the current  
23 antiretroviral regimen that they studied the study on, with  
24 the exception that they could stop an agent if they had

1 toxicity during the first 16 weeks. There were 80 ritonavir  
2 patients and 79 placebo patients, and those were the first  
3 patients who were involved in the study with a baseline RNA  
4 better than 15,000 copies/mL.

5           During the time that it took us to identify these  
6 virology subset patients, patients who enrolled in the study  
7 were also analyzed for CD4 and CD8 changes. The enrollment  
8 for the entire study occurred during May and June of 1995,  
9 and the 191 events that were assumed to be required to have  
10 an 8 percent power to detect a one-third reduction in events  
11 between the 2 arms were approved and analyzed by the middle  
12 of December of 1995, which showed a highly statistical  
13 significant benefit both in disease progression and survival  
14 between the two arms. That allowed us to offer open-label  
15 ritonavir to all patients in early January, 1996.

16           Since these were the first patients enrolled in  
17 the study, they overall had about 7 months of evaluation  
18 during the placebo-controlled period. The remainder of this  
19 presentation will focus only on the 80 ritonavir patients in  
20 the virology subset, looking at changes at their HIV-RNA and  
21 how it correlated with clinical benefit over the 7-month  
22 observation.

23           (Slide)

24           The baseline demographics of the virology subset

1 are similar to the overall patient group. They were  
2 primarily men, 94 percent men; 89 percent Caucasian, with an  
3 average age around 39 years. They were more than 5 years  
4 from their diagnosis of HIV infection, the majority of the  
5 patients and, unfortunately, 88 and 15 belong over here  
6 where the majority of patients acquired HIV by sexual  
7 transmission.

8 (Slide)

9 These, indeed, were advanced patients. Here are  
10 their mean and median baseline RNA numbers. The baseline  
11 RNA median was 5.4 logs, with a median CD4 cell count of  
12 20.8 cells and a median CD8 of 411 cells. The concurrent  
13 antiretroviral agents taken by this group reflected sort of  
14 the standard care in the mid-1995's range. There was a  
15 large group of patients who actually had no concurrent  
16 regimens; a fair number of patients who were taking either  
17 long-term therapy AZT or D14; a small number on ddC. There  
18 was one patient on monotherapy, ddI, and the other patients  
19 were on ddI combinations, and a fair mixture of AZT dual  
20 combination regimens, and different regimens were also  
21 included.

22 (Slide)

23 Nineteen patients had a clinical endpoint at 7  
24 months of evaluation, and similar to the larger study

1 population, the diagnoses spanned the spectrum of AIDS-  
2 defining illnesses, including 4 patients who died prior to  
3 the diagnosis or a new event.

4 (Slide)

5 The first correlation we looked at looking at RNA  
6 changes with the clinical endpoints, we looked at early  
7 changes in the magnitude of HIV-RNA increase at 12 weeks or  
8 16 weeks, stratified by change into either 0.5 log or a log  
9 at those time points.

10 As you can see, the 0.5 log cuts at both 12 and 16  
11 weeks had a more balanced distribution of patients than the  
12 greater or less than of 1 log cuts at both time points, and  
13 using the Fisher's exact test and proportional hazards model  
14 we were unable to identify any statistically significant  
15 difference in the events rates in these 2 groups.

16 (Slide)

17 Here the Kaplan-Meier shows the trend towards a  
18 difference when you took changes of less than a log, in  
19 yellow, compared to greater than a log at the 16-week time  
20 point. Again, there is a fairly small number of patients in  
21 this greater than a log decrease which may be damping our  
22 ability to detect a statistically significant difference  
23 between the groups.

24 (Slide)

1           The next analysis looked at maximal suppression at  
2 nadir, and here you see more splaying of the groups when you  
3 take a maximal suppression of less than a log at any time  
4 point, here in green, compared to 1-2 logs, in yellow, or  
5 further than a 2-log decline nadir, here in blue. You can  
6 see that you do begin to get a separation of the curves even  
7 in this early small sample size.

8           (Slide)

9           To explore that further, we looked at thresholds  
10 for absolute RNA values at the nadir, looking at the level  
11 of detection of the assay for the Roche Amplicor, using 200  
12 copies as our level of detection, a threshold of above and  
13 below 1,000 copies or a threshold of above or below 5,000  
14 copies.

15           Here, you can see that using the level of the  
16 5,000 copy threshold we got a more even distribution of  
17 patients than at either of the 2 lower thresholds. This  
18 above and below 5,000 copies did lead to statistically  
19 significant changes in the comparisons between the 2 groups.  
20 There was a trend toward clinical benefit in the less than a  
21 1,000 compared to greater than 1,000 analysis, but the  
22 statistical significance between the groups is sort of  
23 dampened by the unequal distribution of patients. When you  
24 got down to using a threshold of the level of detection of

1 the assay, we got even more imbalance in the patients and we  
2 lost our ability to detect a difference between the groups.

3 (Slide)

4 To assess the influence of this a little bit more  
5 closely we plotted a Kaplan-Meier analysis. As you can see,  
6 the group that had a nadir of greater than 5,000 appeared to  
7 develop clinical events early compared to the group whose  
8 nadir was less than 5,000. That appeared to be sustained  
9 through the 7 months of evaluation.

10 (Slide)

11 To assess the influence of baseline  
12 characteristics on that observation, we looked at grouping  
13 the patients by a baseline RNA of less than 300,000 copies  
14 compared to greater than 300,000 copies, this being a round  
15 number fairly close to a median number for the overall  
16 patient group. In the patients that had less than 300,000  
17 copies there are actually more patients with fewer overall  
18 clinical events and we couldn't detect statistical  
19 significance in that comparison.

20 However, when you go to patients with a baseline  
21 greater than 300,000 copies, despite having a slightly  
22 smaller overall number of patients, you do have more  
23 clinical events that separated nicely or had the  
24 characterization of a nadir of less than 500,000 compared to

1 a nadir of greater than 500,000, shown here in green. This  
2 really isn't surprising from the clinical standpoint since  
3 these patients who had very high RNA levels at baseline and  
4 didn't really see an appreciable change in their RNA would  
5 be the ones you would expect to have the highest rate of  
6 disease progression in the short observation period.

7 (Slide)

8 In a similar analysis we looked at the impact of  
9 baseline CD4 cells above and below the median of 20 on a  
10 similar split between nadirs of less than 5,000 and greater  
11 than 5,000. Again, it was the group with the most advanced  
12 disease, baseline CD4s less than 20, that had the most  
13 apparent difference in the nadir of 5,000 compared to a  
14 nadir of greater than 5,000. And patients with greater than  
15 20 CD4 cells counts at baseline really didn't show that much  
16 difference between this nadir cut of above and below 5,000.

17 (Slide)

18 An additional analysis was conducted to look at  
19 the impact of duration of suppression on clinical benefit.  
20 Here patients were stratified by time to rebound from their  
21 nadir of less than 85 days compared to greater than 85 days,  
22 which was very close to the median in rebound from nadir for  
23 the overall patient group.

24 This was using a fairly stringent definition of



1 nadir and rebound where we required the patients to have at  
2 least 0.6 logs decline in order to be called a nadir, and  
3 they have to have 0.6 logs rebound. As you can see, using  
4 those stringent definitions, we have a very small sample  
5 size and couldn't detect a difference in the groups.

6 (Slide)

7 So within the limits of this data set on the 80  
8 patients with 19 clinical endpoints observed for 7 months,  
9 we feel comfortable making the following conclusions: That  
10 a nadir decrease of greater than 2 logs during 16 weeks is  
11 associated with a greater clinical benefit than having a  
12 nadir decrease of less than 1 log, and that having an  
13 absolute nadir value of less than 5,000 copies is associated  
14 with clinical benefit compared to having a value greater  
15 than 5,000 copies.

16 This ability to replicate it in the subset of  
17 patients only with greater than 300,000 copies may well be a  
18 function of the advanced patient population of the study  
19 rather than something of virologic significance. We think  
20 the patients who have baselines less than 100,000 should  
21 also have clinical benefit with longer observation and a  
22 minimal cut of less than 5,000.

23 (Slide)

24 Basically, we don't feel that duration of

1 observation in this limited sample is enough to be able to  
2 make any statements about long-term benefit in the group.  
3 Taken altogether, we feel that these data do support the  
4 general conclusion that HIV-RNA should be decreased as much  
5 as possible for as long as possible to maximize clinical  
6 benefit. Thank you.

7 DR. HAMMER: Thank you very much. The final  
8 speaker is Michael Elashoff, who will provide some summary  
9 comments and then we will open this up for questions.

10 **Summary, Michael Elashoff, Ph.D.**

11 (Slide)

12 DR. ELASHOFF: I will be summarizing the company  
13 presentations and presenting some of our conclusions. In  
14 planning this meeting we tried to achieve some consistency  
15 in the format of presentations so that it would be easier to  
16 draw conclusions. At the same time, the trials had varying  
17 populations, regimens and sizes so no one analysis could be  
18 dictated for all presentations. The result is that each  
19 company examined their data in a somewhat different way.  
20 This allows for us to examine the relationship in RNA  
21 changes and clinical events in several different ways and  
22 judge how consistent this relationship may be.

23 (Slide)

24 The basic question the presentations addressed was

1 how an initial RNA response predicted the subsequent  
2 clinical events. The RNA response was measured in several  
3 different ways, change from baseline, nadir and durability  
4 of the effect. The area under the RNA curve was also looked  
5 at for durability and effect. Clinical events were also  
6 examined in several different ways, Kaplan-Meier curves and  
7 proportional hazards models, to address the association  
8 between RNA and clinical events.

9 (Slide)

10 Overall, all five presentations found evidence for  
11 an association between RNA changes and subsequent clinical  
12 events. This association was seen in all the  
13 characterizations of RNA changes and clinical event analyses  
14 shown in the previous slide. Together, the data provide  
15 compelling evidence for this relationship.

16 (Slide)

17 The presentations covered many clinical trials and  
18 drug regimens. Here I have summarized the studies and  
19 regimens analyzed.

20 (Slide)

21 As I have mentioned, the study populations varied  
22 across the spectrum of HIV disease. Subjects in the Abbott  
23 study RNA subset provide by far the most advanced, as  
24 evidenced by the median RNA and CD4. This allowed for the

1 small number of subjects to yield valuable information. The  
2 other larger trials primarily represented less advanced  
3 subjects, with a mixture of naive and experienced patients.  
4 Overall, the number of subjects analyzed was almost 5,500.

5 (Slide)

6 I will now go over some of the specific analyses  
7 from the presentations. One of the standard ways in which  
8 RNA responses are quantified is by the change from baseline,  
9 usually measured in log units.

10 (Slide)

11 The ACTG analysis found a strong relationship  
12 between change from baseline over the first 24 weeks. Shown  
13 here on the X axis are units of 0.5 log drop, and the  
14 clinical event relative risk is on the Y axis. The primary  
15 story of this graph is in the slope and shape of these  
16 lines, rather than the particular lines here which just  
17 represent different ways of analyzing the data. This  
18 relationship was seen after adjustments for baseline RNA,  
19 CD4, treatment and study.

20 (Slide)

21 The Glaxo analysis also identified a strong  
22 relationship. Here are two curves, one for high baseline  
23 RNA and one for low baseline RNA. Change from baseline RNA,  
24 again in units of 0.5 log decrease, is shown on the X axis

1 and clinical incidence rate is on the Y axis. The curves  
2 seem to flatten with large decreases, especially for those  
3 who started lower to begin with. This effect is in part due  
4 to the assay lower limit of detection. More follow up will  
5 be necessary to detect differences down in this range.

6 (Slide)

7 Pharmacia's results also showed association  
8 between change from baseline and clinical progression rate.  
9 Here we see that subjects who started with a low RNA  
10 experienced few events regardless of their change. But the  
11 relationship is more dramatic for higher baseline RNA  
12 values.

13 (Slide)

14 They also presented these results in a  
15 proportional hazards model, which indicated a dose-response  
16 type relationship between change from baseline and clinical  
17 event rates after adjustment for baseline CD4 and RNA  
18 values.

19 Notice here that above about 0.5 logs, no  
20 reduction was seen in the hazard rate. In the region above  
21 0.5 logs, there was a significant reduction. The two  
22 categories here above 0.86 logs showed the most decrease in  
23 clinical event rate.

24 [Slide.]

1 Roche presented Kaplan-Meier curves stratified by  
2 the magnitude of change from baseline. The curves showed  
3 separation between the three categories indicating that  
4 greater decreases are associated with a lower risk of  
5 disease progression.

6 [Slide.]

7 Overall, the results indicated that even the  
8 smallest decrease studied across the trials, about 0.5 log,  
9 was associated with clinical benefit. Further, greater  
10 decreases resulted in lower clinical-event rates. The  
11 relationship for large decreases remains to be clarified  
12 where the limit of detection and small event rates make  
13 characterizations difficult.

14 [Slide.]

15 Another way of measuring RNA response is via the  
16 RNA nadir, or lowest value achieved. Since this is an  
17 absolute number, cross-study comparisons could be made more  
18 easily.

19 [Slide.]

20 Glaxo found, as might be expected, that lower RNA  
21 nadirs were associated with lower clinical progression  
22 rates. This was true for both high and low baseline RNA  
23 levels. In this graph, you can also see a flattening of the  
24 curve below about 5,000 copies. However, the small number

1 of events indicate that longer follow up will be necessary  
2 to clarify this range.

3 [Slide.]

4 Glaxo also showed the relationship between RNA  
5 nadir and clinical progression rate. They found an  
6 approximate linear relationship. The ACTG also found a  
7 similar result.

8 [Slide.]

9 Here are Kaplan-Meier curves stratified by the RNA  
10 nadir level achieved while on treatment and eventual  
11 clinical progression in Kaplan-Meier curve format. Again,  
12 lower RNA nadirs were associated with longer times to  
13 clinical events.

14 [Slide.]

15 Roche also presented their data in terms of hazard  
16 rates. This graph found lower hazards for smaller RNA  
17 values.

18 [Slide.]

19 Overall, the data showed a clear association  
20 between the lowest RNA value achieved and subsequent  
21 clinical-event rates. The RNA values achieved to date do  
22 not appear to have reached the so-called threshold effect  
23 beyond which further reduction would convey no further  
24 advantage. Since these studies were started some years ago,

1 the regimens were not, in general, optimal by today's  
2 standards.

3 This meant proportionately few subjects had  
4 responses down to the limit of these assays so that  
5 characterizing the response curve in that range remains  
6 imprecise. Longer follow up and better treatments will be  
7 needed to address this area.

8 [Slide.]

9 The final characterization of RNA response was in  
10 the durability of that response.

11 [Slide.]

12 Pharmacia showed that durability was associated  
13 with clinical-event rates. In this graph, individual lines  
14 represent differing durabilities of effect. For this  
15 analysis, effect was defined as a half-log reduction. For  
16 the higher baseline RNA values, the lines are seen to  
17 separate indicating that less durable responses were  
18 associated with higher clinical-event rates.

19 [Slide.]

20 This effect is easier to see in the proportional-  
21 hazards model. In this model, responses that persisted past  
22 16 weeks or about 114 days resulted in the fewest events  
23 while less durable responses meant a higher rate of clinical  
24 events.



1 [Slide.]

2 Roche addressed the issue of durability using a  
3 area-under-the-curve, or AUC. The AUC is analagous to the  
4 product of duration times effect. When they broke down  
5 subjects into four groups by their AUC, a clear difference  
6 was seen in the Kaplan-Meier curves. But since AUC is a  
7 function of duration and amount of change, this analysis is  
8 an indirect measure of the association between durability  
9 and clinical-event rates.

10 [Slide.]

11 Durability of response and subsequent clinical-  
12 event rates were found to be associated in the Pharmacia  
13 analysis. This result was supported by Roche's analysis of  
14 the area-under-the-curve. Although we had hoped that the  
15 trials considered would have dealt more completely with  
16 durability, we should not be surprised at the limited  
17 information available. This is due to the age of the  
18 trials, the treatment regimens and the population studied.

19 Long-term durability--that is, longer than 24  
20 weeks--was infrequent and remains to be addressed.

21 [Slide.]

22 Both Roche and Glaxo also provided surrogate-  
23 marker analyses. These approached the question in a  
24 somewhat different way from the other analyses by

1 incorporating treatment comparisons into the analysis.

2 Initially, Roche found a significant treatment effect.

3 [Slide.]

4 After incorporating RNA and CD4 responses while on  
5 study, presumably induced by the treatment, they found that  
6 the treatment effect was no longer significant. In other  
7 words, the RNA and CD4 responses seemed to mediate the  
8 treatment's effect on clinical-event rates.

9 [Slide.]

10 Glaxo showed this visually for their data as the  
11 treatment difference in AIDS-free survival between treatment  
12 and control seen here--

13 [Slide.]

14 --was virtually gone when RNA and CD4 responses in  
15 the first part of the study were incorporated into the  
16 analysis.

17 [Slide.]

18 While these analyses were suggestive that much of  
19 the treatment effect is mediated by changes in RNA and CD4,  
20 it is important to point out that the goal of these talks  
21 was not to formally validate RNA as a surrogate. When drugs  
22 are approved for lowering RNA, the claim of clinical benefit  
23 will not follow automatically.

24 To claim clinical benefit, sponsors will have to

1 show exactly that. Clinical practice and treatment  
2 strategies are largely based on RNA. As Dr. Feigal stated  
3 in the introduction, we may be read to move past the  
4 question of surrogacy. Tomorrow, we will discuss in more  
5 detail, the design of clinical trials making use of long-  
6 term RNA changes. The role of clinical endpoints and CD4 in  
7 such trials will be considered further.

8 [Slide.]

9 In summary, both change from baseline and RNA  
10 nadir were found to be associated with clinical-event rates.  
11 The analyses presented, on concert with several other  
12 considerations, suggest that the RNA nadir may be preferable  
13 to change from baseline. This avoids the problem of where  
14 you start pointed out by Dr. Marchner.

15 It makes the inevitable comparisons across studies more  
16 interpretable and may avoid some of the problems associated  
17 with the quantitative analysis of mean changes.

18 Further, the assays seem more suited to  
19 qualitative interpretations such as RNA decreasing versus  
20 RNA increasing or RNA below the limit versus above the  
21 limit. The limit of detection is simply the lowest nadir  
22 possible. Overall, clinical progression seems to be more  
23 related to the absolute RNA level rather than the change  
24 from baseline.

1 [Slide.]

2 The primary focus of the talks was on the  
3 relationship between RNA changes and subsequent clinical  
4 events. While these trials are not the final story,  
5 particularly for long-term responses at very low viral  
6 levels, they provide a compelling evidence that the lower  
7 the level of RNA achieved during the study, the lower the  
8 risk of clinical progression.

9 More limited data suggested that the more durable  
10 response, the better the outcome. We expect that further  
11 information will emerge as more data become available.  
12 Finally, these conclusions were based on multiple studies,  
13 treatments and methods of analyses and covered over 5,000  
14 subjects.

15 DR. HAMMER: Thank you very much.

16 **Questions**

17 We now have approximately a half an hour for  
18 questions from the committee. I would suggest that  
19 questions be directed to any of the afternoon speakers. The  
20 warmth in the room has quelled some of the enthusiasm.

21 DR. MATHEWS: I could address this to Dr.  
22 Elashoff, but some of the data that raises this question in  
23 my mind were from the Roche presentation dealing with nadir  
24 RNA values as an endpoint. The question is while RNA nadirs

1 seem to perform reasonably well prognostically, is there  
2 come confounding with baseline value and interpretation of  
3 the nadir?

4           In other words, the interpretation of an endpoint  
5 as a nadir really mixes the baseline value with the  
6 treatment effect since, if a drug is known to produce, say,  
7 on average, a 0.75 log drop from wherever you start at a  
8 certain point in time, that is going to determine the nadir  
9 value that is achievable in a given patient population.

10           DR. ELASHOFF: I don't think that baseline affects  
11 the analysis of the data more than change from baseline. I  
12 think just the opposite that baseline--I think that this  
13 confusion was also noted somewhat after the ACTG talk where,  
14 when you look at change from baseline, really you can't  
15 think about change from baseline numbers without knowing  
16 where you started whereas the RNA nadirs are a more absolute  
17 comparison.

18           Certainly, baseline is important. Nadir and  
19 baseline would be important. But if you say that a drug is  
20 a 0.7 log drug, that is only true for the population in  
21 which it was originally studied. That doesn't necessarily  
22 imply that a different population with a different baseline  
23 is going to get that same reduction.

24           DR. MATHEWS: I agree that that is true but it is

1 problem true of all the metrics that have been presented.  
2 Some of the analyses that were presented gave adjusted  
3 values in the use of the nadir metric, adjusting for  
4 baseline. But many of them did not. They were just  
5 unadjusted or crude effects, whether it was using hazard  
6 ratios or incidence rates.

7 I am just not convinced. In my own mind, I guess  
8 haven't processed all this data yet that the nadir value is  
9 as attractive as your conclusion suggests compared to all  
10 metrics. I guess I would argue, just as we have looked at  
11 in most of the other drugs I have seen come before this  
12 committee, for consistency of effect across whatever metric  
13 is used, and not relying on a single metric.

14 DR. ELASHOFF: I guess I would agree with that  
15 although when you have a more effective regimen where, say,  
16 you have two treatment arms both of which are getting you to  
17 very low levels, differences in change from baseline will  
18 only reflect differences from where they started and won't  
19 reflect a true treatment difference.

20 I would agree with you that if they are telling  
21 you different things, then that would warrant further  
22 exploration.

23 DR. VALENTINE: I think many of us have been  
24 functioning under the paradigm that if we can really

1 diminish viral replication then we will prevent the  
2 selection of resistance. Yet all of us have seen patients  
3 who became below "levels of detection" for a period of time  
4 of a number of months only to rebound back again.

5 All the data we have seen today, obviously, is  
6 using a lower limit of detection of the 400, 500 or 1000  
7 range, even. I hope that tomorrow we see some data going  
8 down lower than 500. I would urge all the sponsors to go  
9 back to their frozen samples and repeat these assays. Data  
10 is emerging that there is still replication going on less  
11 than 400 and it is conceivable that the nadir measurement  
12 would be much more meaningful if it were down around 20 so  
13 that there would be a long durability of effect and less  
14 chance of selection that may be occurring between 20 and  
15 400.

16 DR. ELASHOFF: Actually, two of the presentations  
17 tomorrow are going to address exactly that comparing nadirs  
18 in the 400 to 500 range with values down as low as 20 and,  
19 actually, a dramatic difference was seen.

20 DR. VERTER: I must confess that I am a bit  
21 apprehensive about making the following comments.

22 DR. HAMMER: That's okay.

23 DR. VERTER: I feel a little bit like the little  
24 kid with his finger in the dike. But I am going to raise

1 some cautionary here. First, I would like to thank all five  
2 presenters for the analysis. I must confess that it was  
3 almost overwhelming. I was writing notes on the first one  
4 when the third one, I think, was starting to do their  
5 presentation. So I am sure I missed some subtle points.

6           Also, I confess to no expertise in this area of  
7 retroviral analysis or AIDs trials, just some years of  
8 background in other clinical trials. However, I must say  
9 that I found that a number of the analyses led me to raise  
10 the following cautions.

11           One, I think that if we had days to go over each  
12 analysis, we could probably answer some of these. But my  
13 concerns start with the potential for selection biases. I  
14 believe if I interpreted all the analyses presented  
15 correctly that none of them were actually the randomized  
16 cohorts but they represented selected cohorts based on  
17 either survival to eight weeks or 16 weeks or 24 weeks or  
18 whatever period of time the investigators could get a  
19 reduction measure.

20           So, immediately, we have some concern, I would  
21 think, that not everyone is in the study. In fact, you had  
22 to survive to that point in time, or at least you had to  
23 have a measure to that point in time before you could get  
24 into the analysis. This, I think, should raise a cautionary



1 note, indeed, as to what implication that has for any drug  
2 use.

3           The worst case would be, of course, that the drug  
4 was actually harmful in those who didn't get to that point.  
5 But if you made it to that point, you were okay. In fact,  
6 there was some suggestion in other trials, not in this area,  
7 that responders might do well. For example, in  
8 antiarrhythmic disease, a number of antiarrhythmic agents  
9 have been licensed by the FDA because antiarrhythmia is  
10 known to increase the risk of sudden death.

11           Antiarrhythmic agents reduce arrhythmias;  
12 therefore, the giving of antiarrhythmics should reduce  
13 death. In fact, until about six, seven years ago when a  
14 landmark trial was done, this was thought to be true. But  
15 then two drugs were tested and found to have a threefold  
16 increase in the risk of death even though they were shown,  
17 prior to the beginning of the trial, to reduce arrhythmias  
18 in that cohort that was eventually randomized.

19           So I think that is one example. There are others.  
20 In addition to that, and I don't want to take too long a  
21 time, but there were a few other things that I would like to  
22 note. One, a couple of the presenters pooled studies. I  
23 can see where, with the short time frame available to the  
24 committee to hear all the presentations, that may have been

1 a necessary evil.

2 But I hope, at some point in time, that the FDA or  
3 the companies present the individual studies. Pooling  
4 studies can actually mask or enhance an effect and I think  
5 the individual studies, given that there are different  
6 treatments, different risk groups, the individual results  
7 should be presented.

8 There were also multiple subgroup analyses that  
9 were done. Here, again, the cautionary note that there may  
10 have been a differential effect amongst the subgroups, but  
11 it may not have been a qualitative difference. It may have  
12 been quantitative. In other words, there was one analysis  
13 put up of under and over 300,000 copies per ml.

14 If you noticed quickly, as it went by, there was  
15 an effect in both groups. It was just better in one than  
16 the other. So one shouldn't preclude the use of the drug  
17 just because one is better than the other one.

18 Maybe I will stop here for now and let someone  
19 else get a chance.

20 DR. HAMMER: I was just wondering whether any of  
21 the speakers want to comment on the points raised.

22 DR. FEIGAL: I think many of the comparisons  
23 actually were across randomized groups. Your point about  
24 you are going to using a laboratory marker is that you have

1 to return to clinic to have the laboratory to be in some of  
2 these analyses is well taken.

3 Most of these trials are studies which have been  
4 previously reviewed because they have shown clinical benefit  
5 across randomized groups. I think that there are some  
6 aspects of the technology of the virology and the way the  
7 specimens were collected that sometimes left us with subsets  
8 where specimens were only collected in a sample of patients.

9 Part of the rationale and why there were so many  
10 similar presentations is I guess we were looking to see what  
11 kind of case could be made for the consistency of the  
12 evidence.

13 The issue of how to examine a relationship and how  
14 to describe it when you have often a categorical outcome  
15 such as progression or survival and you are looking for a  
16 dose response across some other variable is the reason, I  
17 think, that you saw that the data was often split looking  
18 for the kind of dose response.

19 I recognize that there is the problem of  
20 multiplicity but I think the kinds of evidence we are  
21 looking for was an ordered response across the direction and  
22 some consistency in that from study to study.

23 DR. HAMMER: Thank you.

24 DR. LIPSKY: I have a question for the FDA who saw

1 all the data; we realize historically we are looking at  
2 therapy which would, by 1997 standards, be considered  
3 outmoded. Although we are trying to glean some good  
4 information from it, it is possible we may be misled. One  
5 of the things, nadirs, et cetera, could be misleading.

6 I am just curious for today, for what we were  
7 presented today, how many patients were able to go down to  
8 levels that were undetectable in viral copies?

9 DR. ELASHOFF: I think it is a small percentage.  
10 I am not sure what the overall one would be; maybe 10  
11 percent or less. Actually, it would depend on the  
12 particular study. The studies with more advance populations  
13 had actually much lower rates. Some of the NUCA studies I  
14 think were in the 60, 70 percent range, maybe a little  
15 higher initially and then they bounced back up again.

16 So it covered a spectrum, but there really weren't  
17 long-term response which, as Mike had pointed out earlier,  
18 by 24 weeks, it was a relatively small proportion across the  
19 board.

20 DR. LIPSKY: So one potential is I think we would  
21 have to be very careful to make conclusions, at least at  
22 that level on the data today. Maybe tomorrow we will see  
23 some things more clearly.

24 DR. EL-SADR: I think most of the data were in

1 patients with median CD4 about 200 or so. How comfortable  
2 should we be with extrapolating to people with higher CD4,  
3 especially regarding the new guidelines. It seems none of  
4 the data presented today were relevant to those with higher  
5 CD4.

6 DR. ELASHOFF: I think the median represents just  
7 that, the median, and not the entire range. So the actual  
8 CD4 counts across these thousands of subjects range from low  
9 to high.

10 DR. FEIGAL: Typical cutoffs were 350, sometimes  
11 500, but typically 350. There was, I think, adequate amount  
12 of data on patients with very low counts. I think one of  
13 the things that the data brings into question, though, and  
14 particularly looking at some of the quartile plots, is the  
15 adequacy of assessing risk from CD4 count alone and whether  
16 or not you need to also look at baseline viral load at the  
17 same time.

18 DR. MATHEWS: It seemed from the background  
19 documentation and the presentations that the paradigm we are  
20 being asked to examine as a primary endpoint is is viral  
21 load sufficient versus the preceding paradigm which relied  
22 on clinical endpoints. But I think the discussion has  
23 implied that we are lumping clinical endpoints with CD4  
24 responses.

1 DR. FEIGAL: None of these progression endpoints  
2 were the mixed CD4 count progression endpoints that were  
3 seen in some trials. These were all data with clinical  
4 progression. Part of it, I think, Chris, it is a little bit  
5 more subtle shift than just clinical endpoints or not.

6 One of the things that was emphasized in the  
7 previous studies where we tried to get six months of data on  
8 a new regimen compared to an old regimen was to try and get  
9 complete data for that whole time period because, as people  
10 dropped out, we got biased estimates of what the average  
11 response was because, quite predictably, the people with the  
12 crummy counts dropped out.

13 What we were asking participants to do was to  
14 remain in a trial even though we had evidence that they  
15 probably were no longer responsive to the drugs that they  
16 were in. So part of the emphasis is trying to break the  
17 process down a little bit.

18 An initial phase of studying the response of a  
19 drug by asking how deep is the response and what proportion  
20 of subjects get an adequate response and then looking, once  
21 you have that data on response, from there on from time to  
22 failure. So if you can define a loss of a viral-load  
23 response or some other type, you don't ask the participants  
24 to simply stay in the study on fixed regimens as though

1 there is no information, no prognostic information, when the  
2 viral counts rise or the CD4 counts begin to fall.

3           So it is, in a sense, trying to change the  
4 structure of the way that we study the viral load per se.  
5 The issue of whether there are clinical endpoints or not is  
6 not an either/or kind of a decision. We are saying that the  
7 way in the past that we have asked for information about the  
8 viral load with fixed regimens for periods of time without  
9 therapeutic monitoring, whether the patient had a response  
10 or not, is just not clinically appropriate any longer.

11           DR. MATHEWS: I agree with that but the point of  
12 my inquiry is that the bulk of the data that has been  
13 presented, and it is fairly consistent from every one of the  
14 trials examined, is that there is independent information  
15 from CD4 count and RNA response, that the responses are not  
16 always concordant.

17           For example, in the MAC study that was reported,  
18 there was a 20 percent discordance rate. Since both  
19 measures are laboratory measured in real time, if you are  
20 going to specify a laboratory marker as a primary endpoint,  
21 it seems to me it should include both viral load and CD4  
22 response and not just viral load alone.

23           DR. HAMMER: That probably will come up again  
24 tomorrow afternoon, I suspect. I have a question for Drs.

1 DeMasi and Struthers. It follows up on Chris' point about  
2 discordance. Both presentations included a quadrant plot.  
3 Obviously, the bulk of patients who clinically progressed  
4 were in the low CD4, high RNA, quadrant. But has either  
5 investigation looked at the patients in the other quadrants?

6 I know those are small numbers but what you can  
7 tell us about those discordant patients, particularly those--  
8 --and I know it is very small numbers--but those in the  
9 quadrant of low RNA, higher CD4. One of the questions that  
10 we will have to debate is this issue of a marker endpoint, a  
11 single biologic test, for a traditional approval with that  
12 indication.

13 I think there is some concern about the small  
14 proportion of patients who don't exhibit a "classic"  
15 response in both CD4 and RNA, at least with the drug classes  
16 we are talking about. Is there any comment from either Dr.  
17 DeMasi or Dr. Struthers since both showed a quadrant plot?

18 If there is no comment, there is no comment.

19 DR. DeMASI: If I understood the question  
20 correctly, it was the discordancy in the quadrants, the ones  
21 on the lower left and the upper right?

22 DR. HAMMER: There is about 9 percent of patients  
23 I think in your plot that don't fit into the neat right  
24 lower quadrant. Of course, we understand these are biologic



1 tests and the patients don't always fit into these neat  
2 categories but it does raise an interesting question about  
3 the discordancy potentially. I just was wondering if you  
4 have looked at that or have any comments about it.

5 DR. DeMASI: Yes. Actually, we did look at those  
6 events that occurred with, say, a high CD4 count and a low  
7 viral load it was relatively rare, as you saw. But what we  
8 did see is that some of these events were presumptive  
9 diagnoses. Some of them were non-AIDS-related deaths such  
10 as a heroine overdose and one accident.

11 When we restrict the analyses, what we have done  
12 further more is look at specific types of AIDS events in the  
13 particular bottom right. When we do that, when we limit to  
14 certain events such as CMV, disseminated or retinitis, we  
15 can even see this more of a concordancy in the CD4 and RNA  
16 responses.

17 DR. HAMMER: Thank you.

18 DR. STRUTHERS: On our data, we had four patients  
19 that had an AIDS-defining event in the high CD4 but low RNA.  
20 But we haven't looked at those patients individually to see  
21 the reason, what their event was.

22 DR. HAMMER: One of the interesting things about  
23 DDC has been, at least in some either as monotherapy or with  
24 ZDB in the 175 trial, you have seen virologic responses

1 without much in the way of CD4 responses. So that drug,  
2 alone or in combination, in some combinations, have shown  
3 some of this discordance and, for example, in the 175,  
4 experienced subjects, we know, in the virology substudy,  
5 there was a virologic response but no clinical benefit that  
6 could be determined in that study.

7 So it is just a note of caution, I think, that the  
8 group has to deal with. So thank you.

9 DR. VALENTINE: Just while we are discussing  
10 discordances, another type of discordance that is seen now  
11 in several studies is that with a large suppression of RNA  
12 copy number and the concomitant rise in CD4 cells, when the  
13 RNA copy numbers come back up toward baseline, at least, the  
14 CD4 cells do not come down but tend to hang in there at  
15 least for 20-plus weeks. That also remains to be explained.

16 DR. VERTER: I was wondering of any of the persons  
17 who presented did a kind of sensitivity analysis; that is, I  
18 thought I saw 25 percent of the cohorts that weren't  
19 included because they didn't have changes from baseline to  
20 16, 24, whatever weeks, to put in a worse change and rerun  
21 the analysis. I wonder if anyone has done that.

22 DR. HAMMER: That is a general question to all of  
23 the presenters this afternoon. No one is leaping to the  
24 microphone. Ian has a comment.

1 DR. MARCHNER: First of all, I think the figure of  
2 25 percent might be inflated a little by the fact that not  
3 only are we deleting people if they don't have the marker  
4 data, we are also deleting them if they had a clinical  
5 progression in the time frame where we are defining  
6 response, which we have to do.

7 That was one of your earlier points. If you are  
8 looking at the prognostic indication of a response over a  
9 given period, you have to look at the effect of that  
10 conditional on not having the clinical event during that  
11 period. For example, if you were to include people with  
12 clinical events during that period, you run into problems of  
13 whether or not having the clinical event will affect the  
14 marker value which is probably fairly likely.

15 So what I am saying is if someone has AIDS, that  
16 could bump their RNA up, for example.

17 DR. VERTER: That is my point, is giving the  
18 person the worse possible outcome and including the clinical  
19 thing. If you are trying to evaluate a treatment, it seems  
20 to me that by excluding the clinical events, you  
21 potentially--and I use the word very carefully, could wind  
22 up with something that looks beneficial whereas, in reality,  
23 you have excluded those in which it was harmful.

24 DR. MARCHNER: You have to look at the two groups

1 separately. You have got individuals who have short-term  
2 failure and you need to look at that group separately from  
3 the individuals who had their response over the period in  
4 which you are defining the response.

5 I think it makes sense to look at both groups. I  
6 don't think it makes sense to combine them in a single  
7 analysis.

8 DR. HAMMER: We can return tomorrow.

9 MS. LEIN: I saved it until last because it is  
10 quick. I just wanted Dr. Feigal, for the record, if you  
11 could just let folks know that this discussion is really  
12 relevant more so to antiviral therapies and, in the context  
13 of immune-based therapies, unless the mechanism is supposed  
14 to be HIV-specific, that HIV RNA is really looked at as a  
15 safety issue more than evaluating efficacy.

16 DR. FEIGAL: I take your point. I think this is  
17 specifically for antiviral drugs.

18 DR. HAMMER: On that note, I would like to thank  
19 the speakers today for providing just an enormous amount of  
20 data for us to digest. This day is adjourned. We will  
21 reconvene at 8 a.m. tomorrow. Thank you.

22 [Whereupon, at 4:30 p.m., the proceedings were  
23 recessed to be resumed at 8:00 a.m., Tuesday, July 15,  
24 1997.]