1	NUCLEAR REGULATORY COMMISSION		
2	BEFORE THE ATOMIC SAFETY AND LICENSING APPEAL BOARD		
3	In the Matter of		
4	PACIFIC GAS AND ELECTRIC COMPANY) Docket Nos. 50-275 O.L.		
5	(Diablo Canyon Nuclear Project.)		
	Units 1 and 2)		
8			
0	DIRECT TESTIMONY OF GEORGE APOSTOLAKIS		
10	Q. Please state your name.		
10	A. George Apostolakis.		
11	Q. What is your business address?		
12	A. 5532 Boelter Hall, University of California, Los Angeles,		
13	California 90024.		
14	Q. What is the purpose of your testimony in this proceeding?		
15	A. I have been asked to render my professional opinion on the		
16	applicability of probability theory, decision theory, and		
17	statistics to the verification of the design of a nuclear		
18	power plant and to evaluate the adequacy of the Independent		
19	Design Verification Program (IDVP) to insure the adequacy		
20	of the design of Diablo Canyon Nuclear Power Plant, Units 1		
21	and 2. Specifically, my testimony pertains to contentions 1		
22	and 7.		
23	Ι.		
24	QUALIFICATIONS		
25	Q. What is your present position?		
26	A. I am a Professor in the School of Engineering and Applied		
27	Science at the University of California. Los Angeles, where I		
	l.		
I	P310200184 B31016		

PDR ADOCK 05000275

1		have taught since July 1974 I an a maken for
2		the Mechanical Accomputical and member of the faculty of
3		Department
4		Department.
5	0.	Please summarize your education.
	A.	I hold a Ph.D. in Engineering Science and Applied Mathematics
3		and an M.S. in Engineering Science, both from the California
7		Institute of Technology. I also hold a diploma in Electrical
8		Engineering from the National Technical University, Athens,
9	1000	Greece.
10	Q.	Are you a member of any professional organizations?
11	A.	am a member of the American Nuclear Society and the Society
12		of Risk Analysis. I am a past recipient of the well with
13		Award from the American Nuclear Conjute
14		Diesse summerican Nuclear Society.
15	2.	Please summarize your work experience in the fields of risk
16		assessment and nuclear engineering.
17	Α.	For the past ten years, I have been continuously engaged in
		research in risk assessment, including the conduct of
18		probabilistic risk analyses for nuclear power plants;
19		probability theory, decision theory, and statistics;
20		reliability analyses; and nuclear engineering.
21		Since 1977, I have served as a consultant to Pickard.
22		Lowe and Garrick, Inc., where I participated in probabilistic
23		risk analyses of the Oyster Creek. Zion, and Indian Doint
24		nuclear generating stations: I also served for Dickers
25		and Garrick on the technical review based for Pickard, Lowe
26		Probabilistic Safety Stale Teview board for the Seabrook
27		Probabilistic Safety Study. For the past three years, I have
		also served as a consultant to the Bechtel Power Corporation

on probabilistic risk assessment. In the past I have served as a member of the Peer Review Panel for the Load Combination Program of the Lawrence Livermore National Laboratory, as a consultant to the Seismic Safety Margins Research Program of Lawrence Livermore National Laboratory, as a consultant on risk methodology for geologic disposal of radioactive waste for the Sandia National Laboratories, and as a member of a research review group for the Probabilistic Analysis Staff of the U.S. Nuclear Regulatory Commission.

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My research work at UCLA has been both theoretical and applied. I have conducted research on the foundations and methods of probabilistic risk analysis, on data analysis, on fire risk analysis, and the general area of risk-benefit. I have developed and taught two courses on probabilistic risk analysis. I have also taught courses in nuclear engineering as well as basic engineering courses.

Do you regularly publish in the professional literature? 0. 18 Yes. I have edited one book and contributed to another on Α. 19 risk analysis. I have published numerous articles on 20 probabilistic risk assessment, nuclear engineering, and 21 related matters. I also serve as a reviewer for Nuclear 22 Safety, Nuclear Science and Engineering, Nuclear Technology, 23 IEEE Transactions on Reliability, AIChE Journal, Risk 24 Analysis, and Reliability Engineering. The list of my 25 publications has been submitted separately in my affidavit 26 of qualifications.

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	1	II.
2	8	PROBABILITIES AND STATISTICS
:	δ Q.	What do you mean by statistical inference?
4	A.	Statistical inference is the process by which evidence is
5	5	incorporated in our body of knowledge. This body of
6	-	knowledge is, in general, expressed by probabilistic
7	1	statements.
8	Q.	How is evidence incorporated in our body of knowledge?
9	A.	I view this question in the context of the Bayesian (or
10	1	Subjectivistic) Theory of Probability. According to this
11		theory, we always have some degree of knowledge of any
12		uncertain event of interest. Bayesian Theory asserts that
13		our degree of knowledge can be expressed in terms of
14		probabilities. As information becomes available, we modify
15		our state of knowledge; that is, we revise our probabilities
16		This modification is done in a consistent manner using
17		Bayes' Theorem.
18	Q.	What do you mean by "evidence"?
19	Α.	"Evidence" can be any kind of information
20		what is commonly referred to as "statistical ouidance"
21		well as such qualitative information as opinions of mende
22		scholarly literature, the results of experiments at
23	0.	What does the term "statistical ouidonce" ments, etc.
24	Α.	For present purposes. I use the torm "statistical
25		to refer to information concerning the formation to
26		given attribute is observed in a specific a
27		would include how many redboads we find in
		and in a given group of

1		people, the number of times a coin turns up heads in a
2		sequence of tosses, the proportion of heads in a
3		within a given income bracket and an american families
4	1	when in a given income bracket, and so on.
5	0.	What is the relationship between frequencies and
		probabilities?
	A.	Frequencies are observable quantities in a given sample or
1		population. Often we express a frequency as a proportion of
8		a sample or a population. Probabilities, on the other hand,
9		are not observable. They are numerical measures of degrees
10		of belief. In other words, frequencies are objective facts
11		and probabilities are subjective beliefs.
12	Q.	What is the distinction between probability theory and
13		statistics?
14	A.	Statistics is part of probability theory
15		is a set of rules that if obcurd guarantee ach
16		Statistics is that part of pucket/lite the
17		the schemest set of probability theory that deals with
18		the conerent use of evidence.
19	Q.	What do you mean by "coherent"?
20	Α.	Human beings dealing intuitively with uncertainty have been
20	lê kar	found to make inconsistent and unreliable use of the
21		information at their disposal. Probability theory, or, more
22		generally, decision theory, requires them to make their
23		reasoning process, their assumptions, and their use of
24		information consistent with certain principles of rational
25		behavior. This makes the decision process explicit and
26		visible.
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1 Q. What is the virtue of making the process explicit and 2 visible?

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- A. Probabilities are inherently subjective, as are decisions
 made under uncertainty, leading to differences of opinion
 among people. By making the process explicit and visible,
 we allow people holding different opinions, and third parties
 observing the differences, to approach resolution of the
 differences on a reasoned basis.
- 9 Q. What is the nature of the differences in opinion among people?
 10 A. People differ in their assessments of probabilities. They
 11 also differ in their assessments of the costs and benefits of
 12 different consequences of decisions.
- Q. What are the reasons for different probability assessments?
 A. Different decision makers may have different states of knowledge. In addition, there is evidence that human beings have great difficulty expressing their knowledge in terms of probabilities.

There is a substantial body of evidence indicating that people perform poorly in assessing probabilities, that is, in dealing coherently with a body of incomplete evidence. For example, Slovic, Fischhoff, and Lichtenstein, in their article "Facts and Fears: Understanding Perceived Risk" (published in <u>Societal Risk Assessment</u>, R.C. Schwing and W.A. Albers, Jr., Editors, Plenum Press, 1980), state, on the basis of their own experiments and research and those of others, that people tend to deny uncertainty, misjudge risks, and express unwarranted confidence in their judgments. The

same authors show that expert assessments are also susceptible to biases, particularly underestimation of risks.

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Kaplan, Garrick, Duphily, and I found similar evidence of expert underestimation of failure rates in a study we did of the performance of several components of a nuclear plant. We found, somewhat to our surprise, that the statistical evidence of failures at that plant indicated substantially higher failure rates that the experts had predicted. (Apostolakis, Kaplan, Garrick and Duphily, "Data Specialization for Plant Specific Risk Studies," <u>Nuclear</u> Engineering and Design, 56:321-329 (1980).)

For rare events the difficulties people have assessing probabilities can lead to dramatically different opinions. Of course, this is one area where statistical evidence can be most useful. Bayes' Theorem tells us that when statistical evidence is strong, the prior beliefs (i.e., beliefs prior to obtaining the statistical evidence) become unimportant and the probability assessments are controlled by this evidence, that is, they are independent of the assessor. All this, of course, assumes that different assessors interpret the evidence in the same way, something that is not always true. III.

DESIGN ERRORS

Q. Has there been any formal research done on the frequency and significance of design errors in nuclear power plants?
 A. Yes. Three studies are particularly pertinent here:

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1 J. R. Taylor, "A Study of Failure Causes Based on U.S. (1) 2 Fower Reactor Abnormal Occurrence Reports," in 3 Reliability of Nuclear Power Plants (Proceedings of a 4 Symposium, Innsbruck, April 14-18, 1975), pp. 119-130, 5 Unipub, Inc., N.Y., 1975. Taylor studied Abnormal -Occurrence Reports (now known as Licensee Event Reports 7 (LERs)) submitted to the Atomic Energy Commission and 8 found that a large proportion of the failures in U.S. 9 plants involved design, installation, and operation 10 errors, with an unexpectedly large proportion of the 11 incidents involving multiple failures. Of 490 failures, 12 he classified 36 percent as being due to design errors. 13 The largest single cause of design errors was found to 14 be unforeseen conditions. 15 T. M. Hsieh and D. Okrent, "On Design Errors and System (2) 16 Degradation in Seismic Safety," in Transactions of the 17

4th International Conference on Structural Mechanics in Reactor Technology, San Francisco, Calif., August 15-19, 1977, T. A. Jaeger and B. A. Boley (Eds.), Vol. K, Paper K9/4, Commission of European Communities, Luxembourg, 1977. Hsieh and Okrent investigated the possible number and influence of seismic-related design errors by examining the historical record of such errors for a specific reactor. Their estimates of the core melt frequency were substantially higher than those of the Reactor Safety Study (WASH-1400), which had not taken into account the possibility of design errors.

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]	-	(3) P. Moieni, G. Apostolakis, and G. E. Cummings, "On
2		Random and Systematic Failures," Reliability
3	5	Engineering, 2:199-219 (1981). We analyzed the LERs for
4		two power reactors plus 100 design errors compiled by
5		Oak Ridge National Laboratory. We found that 18 percent
0	4	of all licensee events at one of the two reactors and 13
7		percent at the other were due to design errors. We
8		found that the most common design error was the failure
9	1	to foresee environmental conditions. That design error
10		alone accounted for nearly as many LERs as all
11		operational procedure errors.
12		It is important to keep in mind that these results are based
13		on each group of researchers' definitions of the term "design
14		error" and on their interpretation of the events reported.
15		Despite these reservations, there is a great deal of useful
16		information in these studies. For example, they show that
17		design errors are a more frequent cause of failures in
18		nuclear power plants than has been widely assumed.
19	۵.	What are the typical causes of design errors in nuclear power
20		plants?
21	Α.	The cited studies indicate that major causes appear to be
22		unforeseen environmental conditions, specification errors,
23		and wrong analyses.
24	Q.	Do these studies show that design errors are inevitable or
25		widespread in commercial reactors?
26	Α.	Not necessarily. Each of these studies has examined
27		previously identified operational failures and classified
		9.

them in various ways. There is no evidence from which one could conclude how representative the plants experiencing these events are of all commercial U.S. reactors. I know of no study of how frequent design errors are in general and of what their impact on the margin of safety is.

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So while these studies show that design errors are a more significant factor in plant failures than was previously thought, they do not tell us how frequent and how important to safety such errors are.

Q. Is there any basis for evaluating the safety significance of the design errors described in the literature?

12 Α. One must be very careful about the meaning of the term 13 "safety significance." If by that we mean actually causing 14 injuries to the public, then none of the errors were safety 15 significant. But if we are speaking about an error having 16 the potential for such harm under possible conditions that 17 were not actually experienced before the error was detected, 18 then it is more difficult to dismiss any error as not being 19 safety significant.

I think that the most meaningful way to investigate these issues is based on the reduction in the presumed margin of safety. The only way I know to practically evaluate the safety significance of an error in these terms is to conduct a probabilistic risk assessment. This enables one to test the sensitivity of a given facility to designated system and component failures. In my experience, PRAs sometimes reveal failure paths not perceived by knowledgeable engineers

involved in the design of the plant. Furthermore, the potential of multiple failures of redundant components due to Jesign errors cannot be fully assessed without a PRA. In the probabilistic risk assessments with which you are Q. familiar, how have design errors been treated? 6 Α. Design errors have been treated only indirectly. By this I mean that, while something is usually done, the analysis is not as rigorous as other parts of PRAs are. For example, Appendix X to the Reactor Safety Study (WASH-1400, NUREG 75/014, October 1975) is entitled "Design Adequacy." The study team felt that they needed additional assurance that certain components would function as intended under severe conditions. Part of the reason for this was that the failure-rate distributions did not reflect experience with such environments. The design adequacy assessment was performed by the Franklin Institute Research Laboratories, which checked a sample of components, systems and structures. They found only minor problems, e.g., errors in assumptions used to calculate stresses and inadequate tests. The consequence of these errors was assessed to be a reduction in the safety margin.

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In more recent PRAs, like those for the Zion and Indian Point nuclear power plants, the issue of design errors was in the minds of the analysts when they quantified their judgment, so that very low values for failure rates were avoided. Design errors were part of the "other" category of failure causes, which means, causes not explicitly

1999	quantified. The notion of the "other" category has been
	proposed by Kaplan and Garrick (see Risk Analysis, vol. 1,
	p. 11, 1981), who were among the principal investigators
	performing these PRAs.
	IV.
1	VERIFICATION OF DESIGN
	USING PROBABILITY THEORY
Q.	Do you know of any case where the adequacy of a nuclear
	power plant's design was demonstrated using sampling?
Α.	No. There have been the studies of design errors I described
	above. But to the best of my knowledge, no nuclear power
	plant has ever been licensed using a sampling verification
	program as a substitute for a quality assurance program that
	was found to be inadequate.
٥.	What is the significance of the decision to verify the design
	by sampling?
Α.	Ordinarily, licensing decisions are framed in deterministic
	terms, i.e., does the plant design comply with the NRC
	criteria? A relatively straightforward answer to this
	question could be obtained by checking the entire design and
	fixing any errors found. If one decides to verify the design
	by sampling less than 100 percent of the design, then one
	transfers the problem into the realm of probabilities, i.e.,
	one is assessing the probability of an affirmative answer to
	the original question regarding compliance with the NRC
	criteria. In other words, one is no longer asking the
	deterministic question, "Does the design meet the licensing
	Q. A.

criteria?" Instead, one is asking, "What is the probability that the design meets the licensing criteria?" Or, more precisely, one is asking, "What is the probability that there are no deviations from the criteria in the existing design?"

The nature of the problem has now been considerably changed. One is now explicitly accepting the possibility of a deviation from the licensing criteria remaining undetected. Q. Can statistical techniques make a contribution to a program to verify the design of a nuclear power plant?

A. Yes, given my earlier discussion of statistics as part of
 probability theory. Once the decision has been made to
 characterize the problem in probabilistic terms, statistical
 techniques enable us to make full use of the information that
 we have available and furnishes the discipline and guidance
 that insures we are using the data properly.

16 Q. How do statistical techniques do so?

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A. These methods can provide guidance to the decision maker concerning both the qualitative aspects of the problem (e.g., what kinds of errors have been made, what can be done about them, etc.) and the quantitative aspects (e.g., how likely errors of a certain type are, how many errors remain undetected, etc.)

In this way, probability theory and statistics further the goal of making the analysis and evaluation explicit and visible.

26 Q. Is it possible to estimate the frequency of design errors in 27 a nuclear power plant using statistical techniques?

1 Yes. Again, one has to be very careful with one's Α. 2 terminology. Because there is no general definition of 3 "design errors," a definition would have to be established at 4 the outset of the study. The definition would have to 5 correspond to the purpose of the study and be precise enough 64 to permit consistent classification of observations. These 7 requirements are not substantially different from the 8 requirements for any engineering study, whether or not 9 statistics are used.

Assuming, however, that we are working with well-defined events, like selecting the wrong design pressure, we could, then, consider the universe of such selections and apply random sampling to estimate the frequency of such errors.

14 Q. What is a "random sample"?

A. A random sample of a population is one in which each element of the population has an equal chance of being drawn for the sample.

18 Q. What is "judgmental sampling"?

19 This is not a term I had encountered before my involvement in A. 20 this case. I gather from the IDVP materials I have read that 21 the IDVP uses this term to refer to the process of selecting 22 elements from the population by using engineering judgment. 23 Are both kinds of sampling used in statistical analysis? Q. 24 There are places for the use of informed judgment, including Α. 25 engineering expertise, in a statistical study. For example, 26 judgment is used to formulate hypotheses. However, once a 27

population is identified for study, samples are drawn from the population randomly.

Why? Q.

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In statistical terms, any sample that is not drawn randomly Α. is suspect of biases. Once one departs from random 6 selection, the danger exists that the selection mechanism contains a bias, presumably unintended, that will lead to an unrepresentative sample and results that cannot validly be generalized to the population from which the sample was drawn. 0. Can you state a pertinent example?

11 A. There are many well known examples of biased samples 12 rendering invalid results. One of the best known is the 13 Presidential preference poll taken by the Literary Digest 14 before the 1930 election. Over two million respondents to 15 the poll showed a preference for Landon over Roosevelt by a 16 57% to 43% margin. In the election, President Roosevelt got 17 62% of the vote.

Any time one departs from random sampling one hazards similar errors. For example, it has been stated that the IDVP sampled the Diablo Canyon design work emphasizing complex designs on the assumption that those were the designs where errors were most likely to be found. However, it is entirely possible that the managers who oversaw the design work recognized the complex problems and assigned them to the most competent engineers and designers. If so, sampling in this way could underrepresent the work of those people most likely to make errors.

1	Q.	Are you saying that what the IDVP calls judgmental sampling
2		has no place in a design verification program?
3	A.	No. If one has information leading one to suspect the
4		location or type of errors, that information should be
5		exploited. But I do not believe that a sample drawn
-		non-randomly can validly be used to generalize about the
7		frequency of errors in the unsampled portion of the
8		population.
9		v.
10		EVALUATION OF THE IDVP
11	Q.	What have you reviewed concerning the Diablo Canyon
12		Independent Design Verification Program?
13	Α.	Parts of the Phase II Program Management Plan, the IDVP Final
14		Report, NUREG-0675 (Safety Evaluation Report, Supplement 18),
15		the IDVP Program Management Plan for Phase II, Interim
16		Technical Reports 1, 8, 34, and 35, and certain depositions
17		and interrogatory answers.
18	Q.	What is your understanding of how the IDVP sought to verify
19		the adequacy of the non-seismic design?
50	Α.	Three systems were selected (the auxiliary feedwater system,
21		the control room ventilation and pressurization system and
25		the safety-related portions of the 4160-V electrical
23		distribution system). I am told that the IDVP verified
24		completely the design of these systems in Unit 1. The IDVP
25		examined the design of these systems and identified errors.
26		It grouped these errors into classes according to whether or
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1		not the errors caused criteria or operating limits to be
2		exceeded.
3		The IDVP then sought to group some of these errors into
4		"generic concerns." Five generic concerns were raised and
5		all systems where these could apply were verified. No other
q	•	samples were taken.
7		On the basis of this examination, the IDVP drew
8		conclusions about the adequacy of the overall design of
9		Unit 1, including the systems not sampled.
10	Q.	In your opinion, did the IDVP proceed in an appropriate
11		way?
12	A.	It is not clear to me why they chose to sample and use
13		probabilistic arguments rather than a full deterministic
14		review. Given, however, that they decided to sample, the
15		available statistical methods, particularly random sampling,
16		that would justify extrapolation of their findings to parts
17		of the plant not sampled, have not been used.
18	۵.	In your opinion, was the IDVP's judgment concerning the five
19		generic concerns sound?
20	Α.	I do not have enough information to judge. I do recognize
21		that issues like this involve extensive use of judgment.
22		Therefore, different analysts may classify errors in many
23		different ways. Nevertheless, I find the presentation of the
24		IDVP's classification unconvincing.
25		For example, the selection of system design pressure,
26		temperature, and differential pressure across valves is
27		identified as a generic concern. I can see a more general

concern being the selection of system design parameters, which would also include other variables, such as stress, enthalpy, humidity, etc. Since the literature I cited above suggests that incorrect selection of design parameters in general is a common source of errors, I find no adequate justification for limiting this generic concern to incorrect selection of pressures, temperatures, and differential pressures across valves.

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9 As a second example, it is stated on page 6.3.4-2 of the 10 IDVP Final Report that three EOIs (8001, 963 and 1069) 11 involve the misapplication of computer programs. Because 12 there was no commonality between the programs involved in EOI 13 8001 and the other pair, and because the types of errors were 14 different, a generic concern was not identified. It may be 15 reasonable, however, to identify "misapplication of computer 16 codes" as a generic concern.

Q. What is the significance of the fact that the IDVP found what it called "random errors," that is, errors that were not covered by the five generic concerns?

A. If the three sampled systems were really representative of
the unsampled systems, this implies that there are similar
errors remaining to be found in the unsampled parts of the
plant. On the other hand, if the three systems are
unrepresentative, we have almost no information about the
unsampled elements of the design and no basis for confidence
in the adequacy of the design.

Q. Is the safety significance of the errors uncovered relevant?

 A. It depends on what the issue is. If the issue is whether the plant's design meets licensing requirements, safety significance of the design errors is not relevant.

If the issue is the safety of the plant, then safety significance of errors is obviously relevant, but, as I stated earlier, the only way I know to perform such an evaluation is in the context of ϵ PRA.

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In your opinion, does the IDVP's work provide a basis for 0. 9 estimating the number of as yet undetected design errors? 10 The failure to use random sampling techniques makes a Α. No. 11 reliable extrapolation impossible and creates the suspicion 12 that there may be errors whose types are not known yet. 13 Furthermore, the same lack of random sampling does not allow 14 the estimation of error frequencies or absolute numbers. The 15 design of the IDVP was not amenable to providing a basis for 16 estimating frequencies.

Q. Does the IDVP provide a basis for concluding that the rate of undetected errors is acceptable?

To decide that a given rate of errors is acceptable, one A . No. 20 must know two things: what the rate of errors remaining in 21 the plant is and what rate is acceptable. For the reasons I 22 have just given, one cannot get from the IDVP's work an 23 estimate of the rate of remaining errors at Diablo Canyon. 24 And nowhere have I seen anyone attempt to set and justify an 25 acceptable rate. The decision that I identified earlier, 26 namely, to recast the problem in probabilistic terms has 27 created the need to have a criterion for acceptability. The

1		issue of an acceptable rate of design errors has not been
2		studied and resolved.
3	Q.	Could one not attempt to set a rate that provides reasonable
4		assurance of safety?
5	A.	The term "reasonable assurance" is not defined. This term is
¢		usually used in NRC regulatory matters to refer to the level
7		of assurance sought in setting the design criteria. Thus, we
8		say that the criteria, if met, will provide a reasonable
9		assurance of safety. It would be a significant departure to
10		talk about a reasonable assurance that the criteria are even
11		met. Then one is talking about a reasonable assurance of
12		meeting license criteria that, if met, would provide a
13		reasonable assurance that the plant is safe. This is a powel
14		notion, the implications of which are not obvious
15	0.	What can be said about the adequacy of Diablo Capyon Unit 2
16		from the verification program for Unit 12
17	A.	I have already said that the findings of the IDVD in Unit 1
18		cannot be generalized to the portions of Unit 1 not evamined
19		That is obviously true of Unit 2, for which the IDVP does not
20		have a sample at all.
21	0.	Do we know whether the rates and distribution of errors in
22		the two units are the same?
23	Α.	No. We know of certain similarities and contain differences
24		between the two units. To be able to say anything about the
25		error rates in the two units random samples would be needed
26		from both units
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1	Q.	What can now be done to achieve confidence in the design of
2		Diablo Canyon?
3	Α.	As a first step, the decision to cast the problem in
4		probabilistic terms should be fully understood. Given the
5		decision to verify by sampling, the objectives of the study
æ		and the decision criteria should be explicitly stated, and
7		the populations should be defined. Random samples should be
8		drawn to determine the nature and frequency of the errors
9		This would permit one to draw valid conclusions about the
10		design as a whole.
11		VI
12		CONCLUSION
13	0	How would you summarize your outlustion of the Touble would
14	*•	The seneral it appears that a such deal of the IDVP's work?
15	A.	In general, it appears that a great deal of good engineering
16		work has been done. In my opinion, the greatest weakness of
17		the IDVP effort has been its failure to recognize the
18		implications of the decision to cast the verification program
19		in probabilistic terms and its failure to use the principles
20		and methods appropriate to a probabilistic analysis. These
21		shortcomings are particularly manifested in the lack of
20		explicit and visible decision rules and the failure to use
03		random samples.
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