

Epistemic Uncertainty Reduction in the PSA of Nuclear Power Plant using Bayesian Approach and Information Entropy

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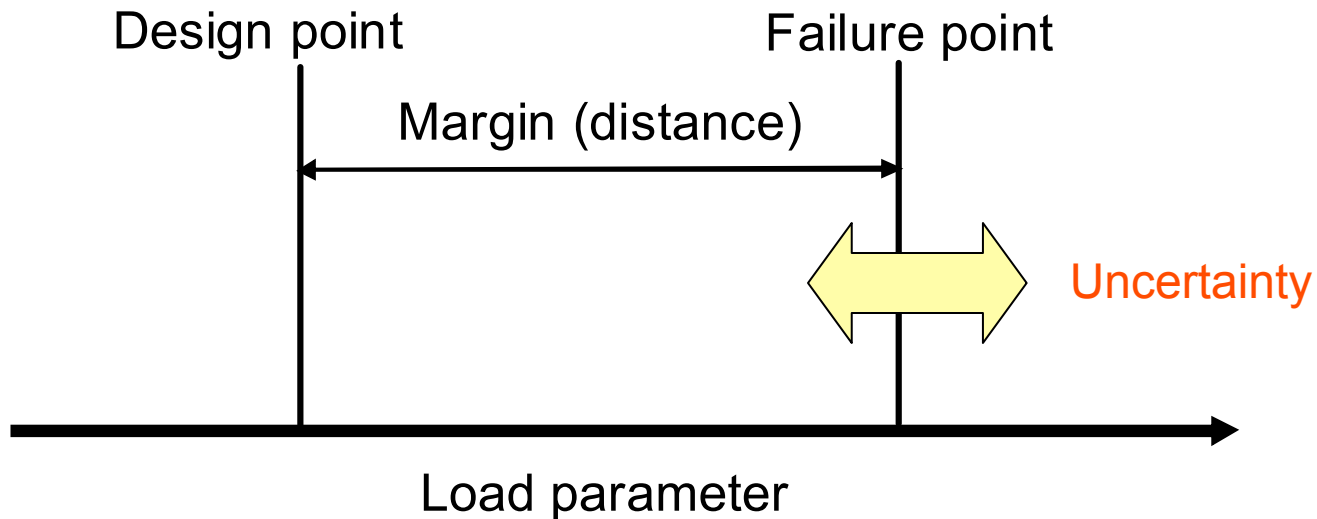
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Introduction

- Residual risk of nuclear power plants is mostly caused by rare events and lack of data is a major problem
- How to overcome the lack of data
 - Statistics, as a subject, is the study of frequency type information. That is, it is the science of handling data.
 - Probability, as a subject, we might say is the science of handling the lack of data. (Kaplan And Garrick, 1982)
- Some experienced earthquakes that exceed the design earthquake level
 - Tyuetsu-Oki earthquake in July 2007
 - Great East Japan earthquake in March 2011
- A methodology to utilize all the information is needed, that maximizes uncertainty reduction

Distance between DP and FP

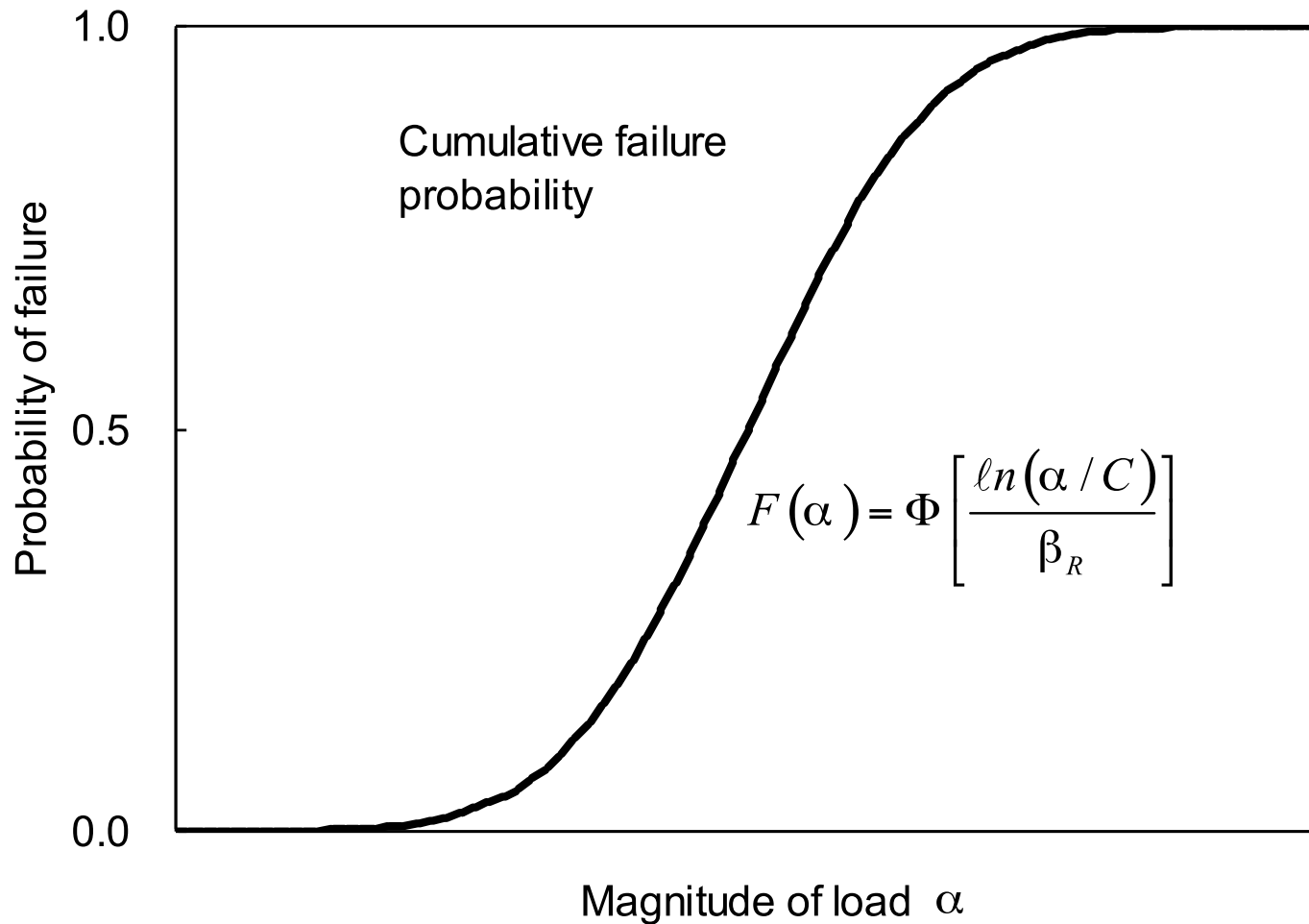
- Safety is assured by DBE as well as margins
- Uncertainty reduction
 - Failure limit and safety margin



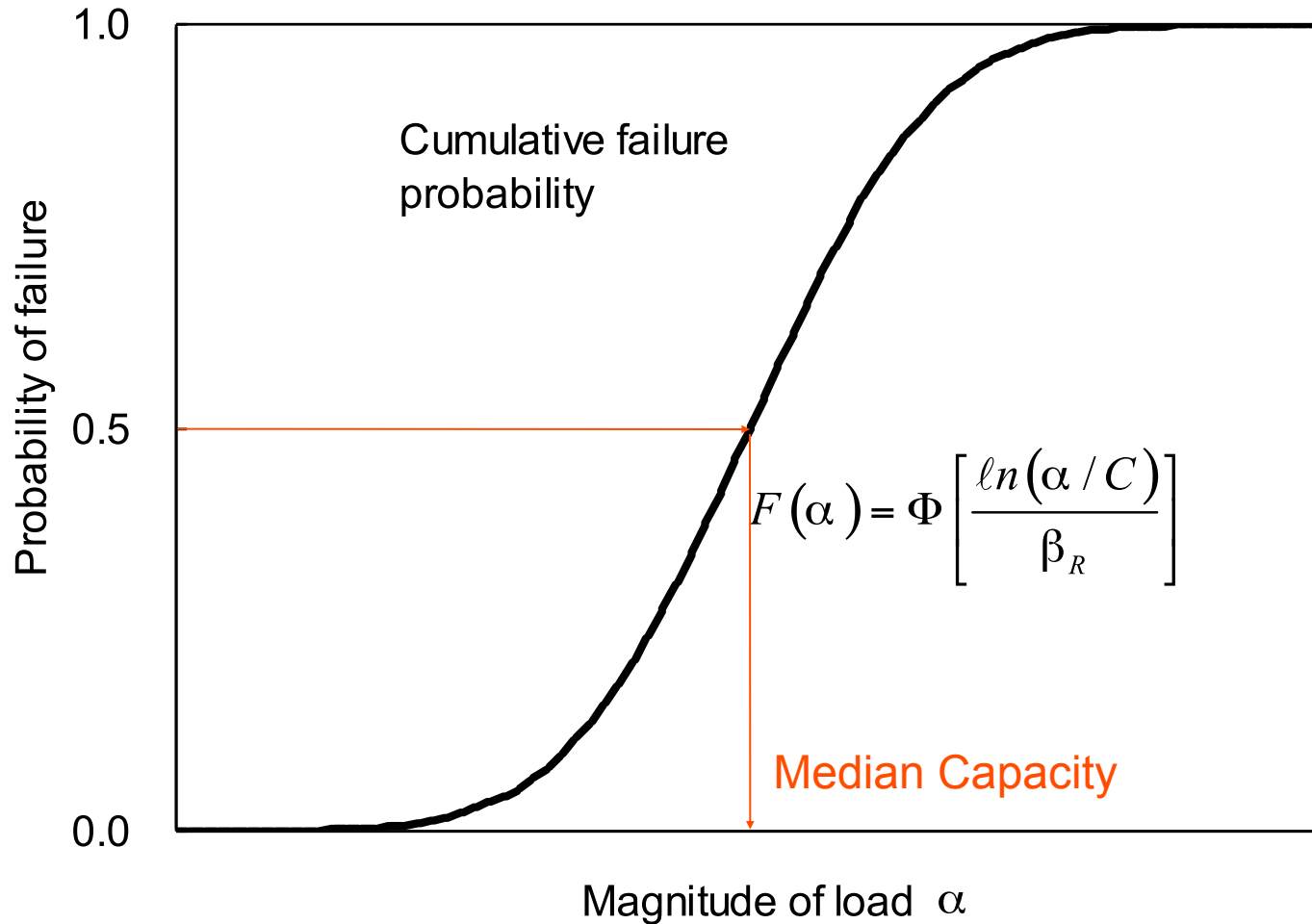
- Seismic capacity test (e.g. vibration table test)
 - Obtain additional information

Limited number of the test..... Best test conditions

Mathematical model for seismic margin

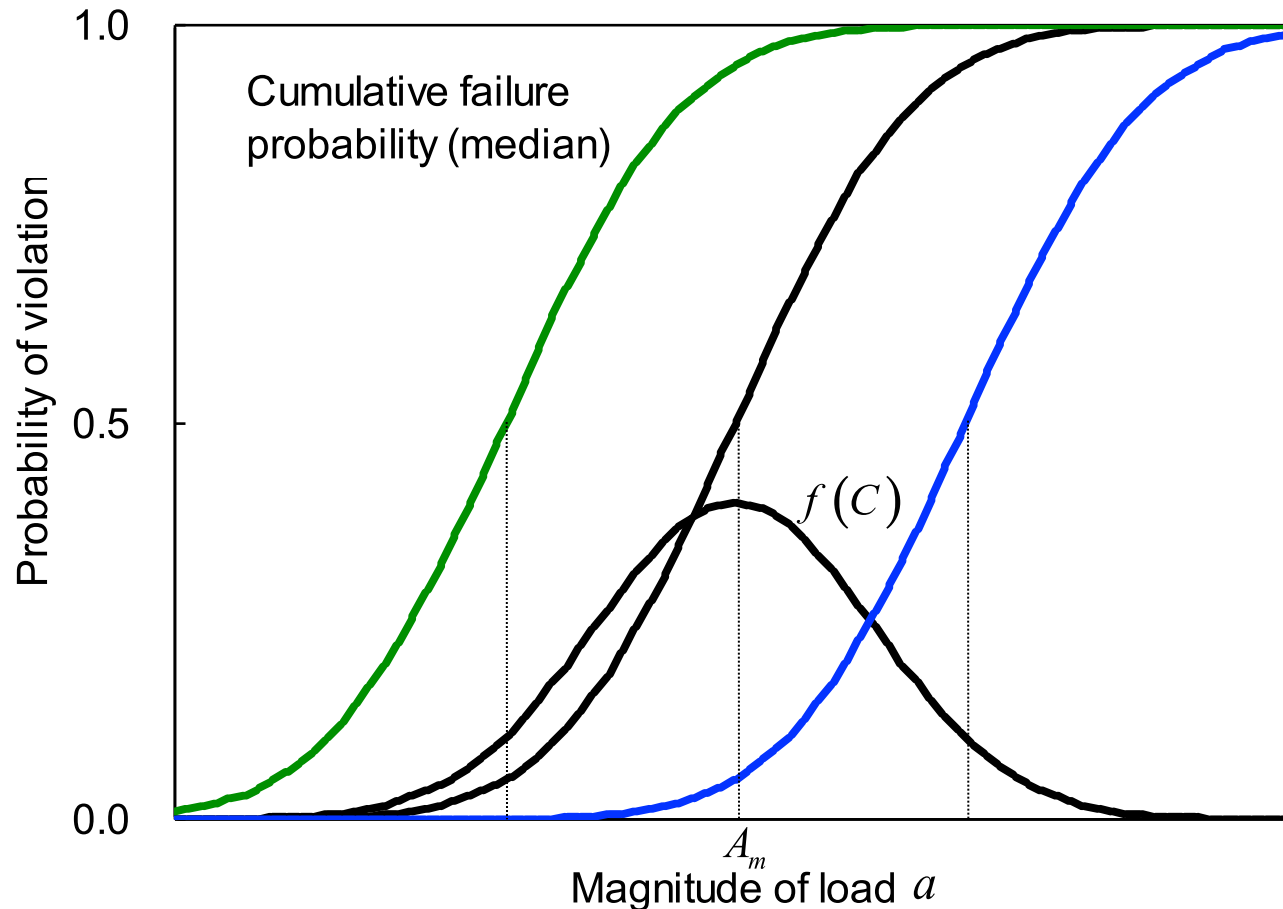


Mathematical model for seismic margin



Mathematical model for seismic margin and uncertainty

$$f(C) = \frac{1}{\sqrt{2\pi} \beta_U C} \exp \left[-\frac{1}{2} \left\{ \frac{\ln(C / A_m)}{\beta_U} \right\}^2 \right] \quad F(\alpha) = \Phi \left[\frac{\ln(\alpha / C)}{\beta_R} \right]$$



Response-Strength Model

$$\xi = R / S$$

$$R = LN(R_m, \beta_{Res})$$

応答の確率分布

$$S = LN(S_m, \beta_{Str})$$

強度の確率分布



損傷パラメータ

$\xi \geq 1$: *failure criterion*

$$f(\xi) = \Phi \left[\frac{\ln(\xi / \xi_m)}{\beta_\xi} \right]$$

$$\xi_m = R_m / S_m$$

$$\beta_\xi = \sqrt{\beta_{Res}^2 + \beta_{Str}^2}$$

Conditional failure probability fragility and R-S model

Fragility model

$$F(\alpha) = \Phi \left[\frac{\ln(\alpha / A_m) - \beta_U \cdot \Phi^{-1}(p)}{\beta_R} \right]$$

Response-strength model

$$F(\alpha) = \Phi \left[\frac{\ln(\alpha / A_m) - \ln(C / A_m)}{\beta_\xi} \right]$$

$$\beta_R = \beta_\xi = \sqrt{\beta_{Res}^2 + \beta_{Str}^2}$$

Variability of response to strength (random)

$$\beta_U = \frac{1}{\Phi^{-1}(p)} \ln \frac{C}{A_m}$$

Uncertainty of equipment capacity

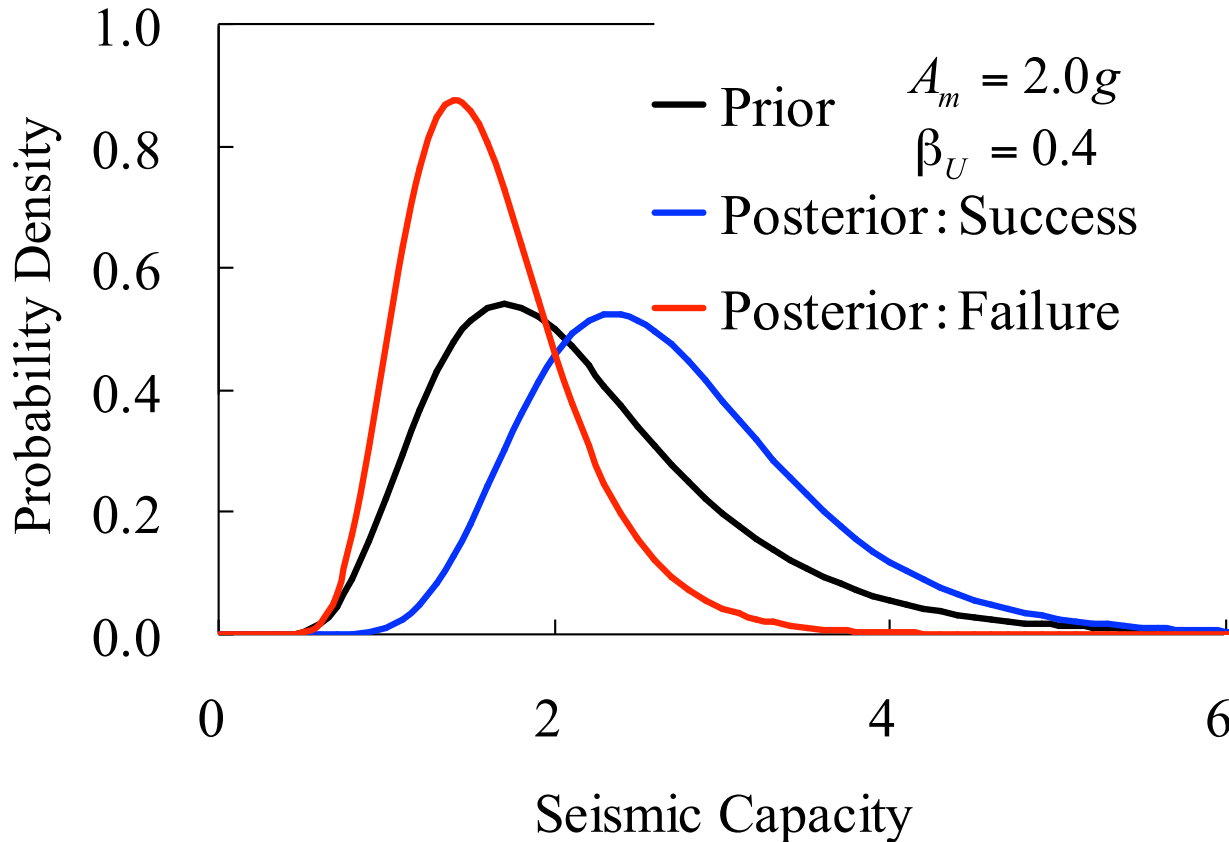
Bayesian Update of Capacity

Bayes Equation

$$f(C|E) = \frac{f(C)L(E|C)}{\int_0^\infty f(C)L(E|C)dC}$$

Shaking table test (Bernoulli trial)

$$L_k(\alpha, N|C) = \binom{N}{k} F(\alpha|C)^k \{1 - F(\alpha|C)\}^{N-k}$$



N=1 (single test)

$$L_0(\alpha|C) = 1 - F(\alpha|C)$$

$$L_1(\alpha|C) = F(\alpha|C)$$

α : test acceleration
 What the test level
 should be?

Conclusions

- Epistemic uncertainty reduction is necessary in seismic PSA
- Extension of our knowledge concerning uncertain failure phenomena reduces the epistemic uncertainty
- Bayesian method updates seismic fragility based on additional information (seismic test)
- Effective seismic test condition can be determined by information entropy
- Reduction of fragility uncertainty can be predicted by generalized uncertainty Index