

Uncertainty Management by Safety Measures in Probabilistic Product Design

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Motivation

Accuracy vs. Conservativeness

Accuracy vs. Efficiency

Wheel Loader (Caterpillar)

S-N Curve

structural stress $\mu_s(x)$

material strength $\mu_c(x)$


possible failure region

Efficiency

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Real advantages of probabilistic design

- Risk allocation for multiple failure modes
- Tradeoff for cheapest safety measures to satisfy reliability constraint
- Wing example

$$P_F = 0.5^{\frac{100 \Delta W}{W_0}} \cdot P_{F0}$$


	Wing		Tail		System	
	Weight	P_F	Weight	P_F	Weight	P_F
Initial	10,000lb	1%	1,000lb	1%	11,000lb	2%
Det. opt	10,200lb	0.25%	1,020lb	0.25%	11,220lb	0.5%
RBDO	10,120lb	0.435%	1,080lb	0.004%	11,200lb	0.439%

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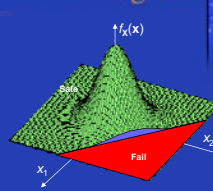
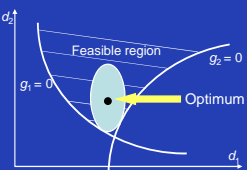
Well, RBDO seems good, but ...

- Conventional RBDO formulation

$$\text{Minimize } C(X, d)$$

$$\text{Subject to } P[g(X, d) \leq 0] \leq P_{\text{target}}$$

- can only incorporate **uncertainties that are given in the design stage**
- Too large uncertainty without considering safety measures
- Industry developed safety measures by trial-and-error

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Types of uncertainty

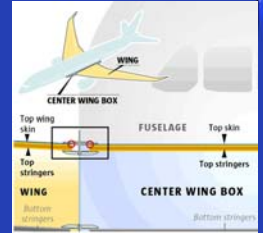
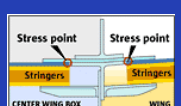
- Aleatory uncertainty**
 - Inherent variation of the physical system or the environment under consideration
 - Variability, type A, irreducible, or stochastic uncertainty
- Epistemic uncertainty**
 - Potential inaccuracy in any phase or activity of the modeling process due to a lack of knowledge
 - Subjective, type B, reducible, or cognitive uncertainty

Type of uncertainty	Definition	Causes	Reduction measures
Error	Departure of average from model	Simulation errors, construction errors	Testing and model refinement
Variability	Departure of individual sample from average	Variability in material properties, construction tolerances	Tighter tolerances, quality control

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Boeing 787 failed certification test

- Boeing expect I-stringers evenly distribute its load to the joint of the center wing box
- Excessive load is transferred through the upper part of stringers and wing surface and the joint is overloaded
- Error in internal load calculation!**

Proposed fix

To reduce stress, a U-shaped cutout is made in the end of the stringers. Fasteners are also added between skin and stringer.



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Design under Uncertainty

- How to design under uncertainty

- Live with it
 - Conservative design with a large safety margin
 - For aleatory uncertainty
- Improve knowledge
 - More accurate models
 - Tests
 - For epistemic uncertainty
 - Without shaping uncertainty, RBDO cannot satisfy reliability constraint in complex engineering systems

Designer needs to consider both uncertainties

Need a **tradeoff** between conservative design and more tests

The graph shows two normal distributions: a white one for stress and a red one for material strength. The mean of the stress distribution is μ_R and the mean of the material strength distribution is μ_C . The area where the stress distribution overlaps with the material strength distribution is shaded and labeled 'failure region'. The distance between the means is labeled 'Safety margin'.

Safety Measures

- How can the safety of a complex system be maintained?
- Activities of reducing uncertainty to make the system safer

The flowchart shows a process starting with 'Test (pre-design)', followed by 'Design', 'Manufacturing', 'Inspection, Health monitoring, Maintenance', and finally 'Accident investigation'. Arrows indicate feedback loops: from 'Design' back to 'Test (pre-design)', from 'Manufacturing' back to 'Design', and from 'Inspection, Health monitoring, Maintenance' back to 'Design'. A 'Truncated tail' is shown at the end of the process. A note states: 'Commercial airplanes are about $P_f = 10^{-6}$ due to various safety measures'.

- How to allocate resources to provide public safety most efficiently?
 - Need to **quantify** the contribution of each safety measure to the system safety

Uncertainty Quantification in the Building-Block Process

The flowchart shows a process starting with 'Test (pre-design)', followed by 'Design', 'Manufacturing', 'Inspection, Health monitoring, Maintenance', and finally 'Accident investigation'. A box labeled 'Design' is highlighted with a red border, and a box labeled 'Test (in-design)' is shown below it, connected to the 'Design' box by a double-headed arrow.

Complex Engineering System Development

(Building-Block Test Process)

- Building-block test process: Systematic design and test evaluation at different stages
- Impossible to satisfy system reliability without shaping uncertainty at multi-level process
- Accumulation of uncertainties in the building-block process
- Calibration and redesign at each level based on calculation and test evaluation
- Want to find/correct design errors at a low-level (need to consider test variability)

The pyramid diagram shows levels of uncertainty quantification from bottom to top: COUPONS, ELEMENTS, DETAILS, SUB-COMPONENTS, and COMPONENTS. The pyramid is labeled 'STRUCTURAL FEATURES' on the left and 'DATA BASE' on the right.

Difficulties in the Current Building-Block Process

- Developed based on trial-and-error and adding safety margins
- Reducing epistemic uncertainty (error) at different levels through tests
- Current technology cannot quantify how much errors at each level are accumulated and contributed to the system
- Unable to identify how much tests at each stages can effectively reduce uncertainties in the system level
- Need a probability-based calculation methodology
- Possible to make the most efficient test evaluation plan within budget
- Possible to estimate the level of confidence at certificate

Basic Purpose of Coupon and Element Tests

Test stage	Objectives	Uncertainty sources
Coupon test	Estimate nominal value and variability of material strength	Variability in material strength and sampling error due to a finite number of coupons
Element design	Estimate multi-axial strength based on a failure theory	Incomplete knowledge of failure mechanism: error in failure theory
Element test	Reduce uncertainty in the multi-axial strength	Sampling error due to a finite number of elements

The diagram shows a 3D element and a 2D coupon. Arrows indicate the flow of information from the coupon to the element. Labels include 'ELEMENT', 'COUPON', 'Accumulated uncertainty', 'Incomplete failure theory', and 'Variability in failure strength'.

Coupon Tests – How to estimate material variability

- True material variability can only be found with an infinite No. of coupons $\hat{\mu}_{c,TRUE} = N(\mu_{c,TRUE}, \sigma_{c,TRUE})$ **Aleatory uncertainty**
- How to predict the true material variability from a test of n_c specimens?

Uncertainty in mean
 $\hat{\mu}_{c,TRUE} \sim N\left(\mu_{c,TRUE}, \frac{\sigma_{c,TRUE}^2}{n_c}\right)$

Uncertainty in STD
 $\hat{\sigma}_{c,TRUE} \sim \frac{\sigma_{c,TRUE}}{\sqrt{n_c-1}} \chi(n_c-1)$

Epistemic uncertainty

Estimated true distribution
 $\hat{\mu}_{c,TRUE} | (\hat{\mu}_{c,TRUE} = \mu_{c,TRUE}, \hat{\sigma}_{c,TRUE} = \sigma_{c,TRUE}) \sim N(\mu_{c,TRUE}, \sigma_{c,TRUE})$

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MCS for Conditional Probability

Estimated true distribution
 $\hat{\mu}_{c,TRUE} | (\hat{\mu}_{c,TRUE} = \mu_{c,TRUE}, \hat{\sigma}_{c,TRUE} = \sigma_{c,TRUE}) \sim N(\mu_{c,TRUE}, \sigma_{c,TRUE})$

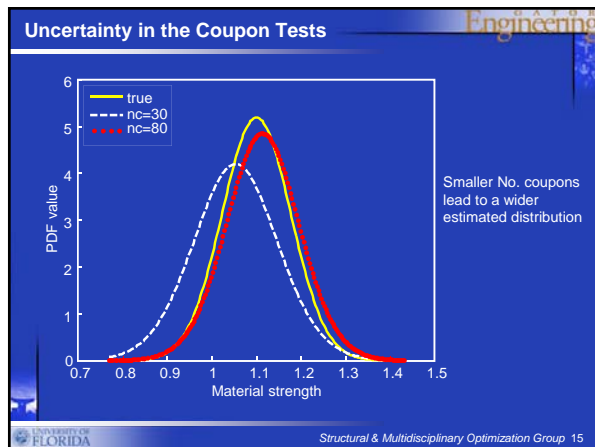
Outer loop

Inner loop

Generate M samples from each distribution
 $N(\mu_1, \sigma_1), N(\mu_2, \sigma_2), N(\mu_3, \sigma_3), \dots, N(\mu_N, \sigma_N)$
 $\sigma_{11}, \sigma_{12}, \sigma_{13}, \dots, \sigma_{1N}$
 $\sigma_{21}, \sigma_{22}, \sigma_{23}, \dots, \sigma_{2N}$
 $\sigma_{31}, \sigma_{32}, \sigma_{33}, \dots, \sigma_{3N}$
 $\sigma_{N1}, \sigma_{N2}, \sigma_{N3}, \dots, \sigma_{NN}$

$$f_c(\tau_c) = \int_0^\infty \int_0^\infty \phi(\tau_c | \mu_c, \sigma_c) f_{\mu_c}(\mu_c) f_{\sigma_c}(\sigma_c) d\mu_c d\sigma_c$$

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Uncertainty Accumulation in Building-Block Process

- Occurs at every step in building-block
 - Uncertainty in the lower-level + epistemic uncertainty in the upper-level
- Ex) Variability in coupons + error in failure theory

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MCS for Uncertainty Accumulation

Uncertainty accumulation
 $\hat{\mu}_{e,TRUE} = \hat{\mu}_{e,TRUE} \hat{\mu}_{c,TRUE}$

Estimated mean material strength (coupon test)

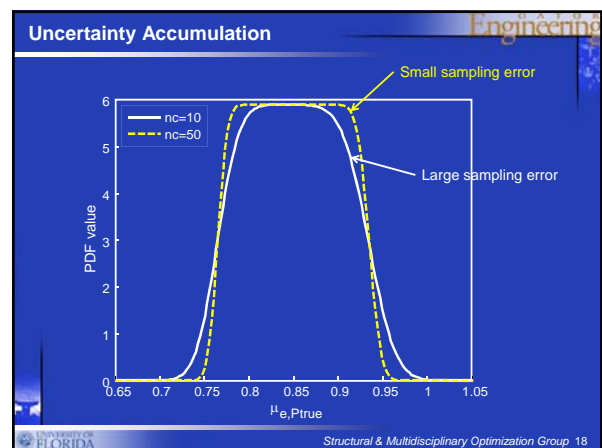
Estimated mean element strength (element design)

Multiply two random variables m_i, k_j for $i=1,2,\dots,N$ and $j=1,2,\dots,M$

$m_1, k_1, m_2, k_2, \dots, m_N, k_N$
 $m_2, k_2, m_3, k_3, \dots, m_3, k_3$
 $m_N, k_N, m_N, k_N, \dots, m_N, k_N$

$$f_{e,TRUE}(\mu_{e,TRUE}) = \frac{1}{\sqrt{2\pi} \sigma_{e,TRUE}} \exp\left(-\frac{(\mu_{e,TRUE} - \mu_{e,TRUE})^2}{2\sigma_{e,TRUE}^2}\right)$$

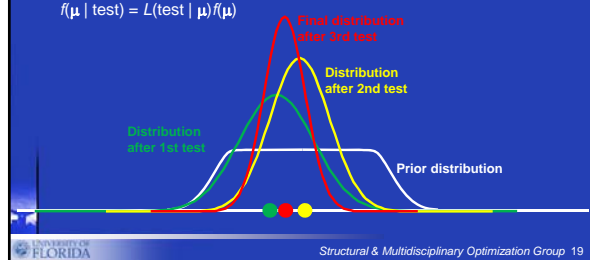
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Uncertainty Reduction by Tests

- Accumulated uncertainty between levels as a prior
- Likelihood based on experimental error
- Posterior distribution based on **Bayesian inference** after tests

$$f(\mu | \text{test}) = L(\text{test} | \mu) / f(\mu)$$

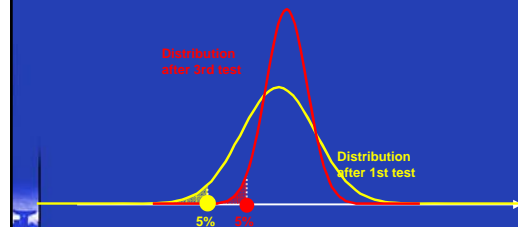


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Effect of the Number of Tests

- Distribution depends on the number of tests
- Possible to estimate the effect of system weight based on 5% conservative allowables



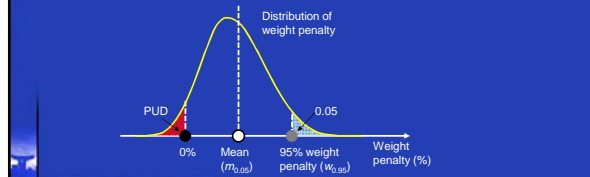
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Weight Penalty

- A way of accessing the merits of the number of coupon and element tests
- Weight penalty is calculated by conservative 5th percentile against the true distribution

$$w_i = \left(\frac{A_{i,0.05}}{A_i} - 1 \right) \times 100 = \left(\frac{F / A_{i,0.05}}{F / A_{i,true}} - 1 \right) \times 100 = \left(\frac{A_{i,true}}{A_{i,0.05}} - 1 \right) \times 100 (\%)$$

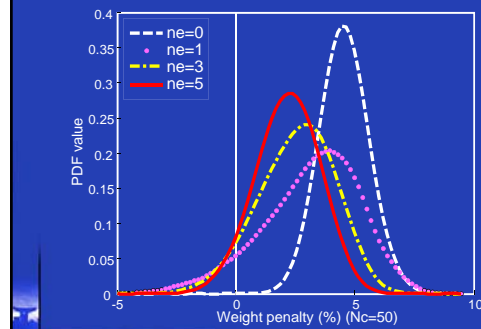


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Ex) Unconservative Failure Theory

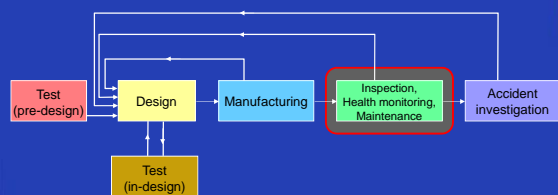
$$n_c = 50, \mu_{e,true} = 0.95, b_e = 10\%$$



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Prognostics and Health Management



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Damage Tolerance Design and Preventive Maintenance

- Damage-tolerance design
 - Flaws can exist and propagate with usage as long as they can be detected and repaired through preventive maintenance
- Preventive maintenance (PM)
 - Inspection (NDI) for all panels & stiffeners
 - Repair/replace for detected cracks
 - Type C : Cost ~ \$2M
 - Downtime loss
 - Removal of installed equipment and wires



Aloha Airlines, 1988

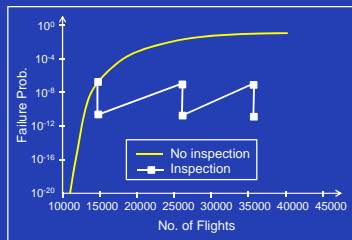


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What's the Effect of Inspection on Safety?

- PM removes dangerous tail part and recovers reliability (safety)
- Without inspection, the current panel thickness must be increased by 30%

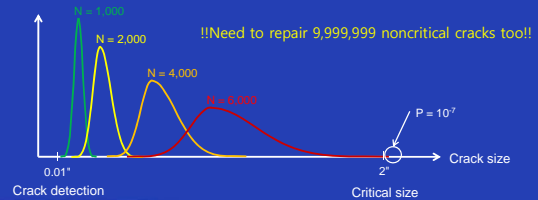


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Preventive Maintenance and Uncertainty

- Periodic NDI -> repair & replace
 - FAA regulation: every 6,000 flights, repair cracks larger than **0.01"**
 - Not causing fracture before next inspection
- Uncertainty in crack growth
 - Uncertainty in material properties, applied loadings, etc
 - Probability to grow to the critical size before next inspection: 10^{-7}

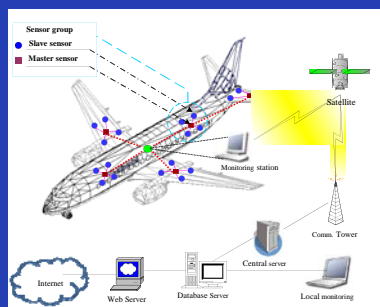


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Structural Health Monitoring (SHM) System

- Damage detection using sensors installed on panels

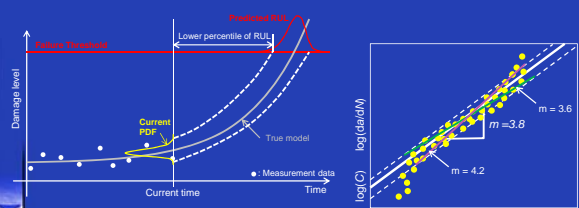


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What Is Prognostics?

- To predict the behavior of damage (crack size, performance degradation) by combining measurement data with physical model
- To Identify model parameters using noisy data: $\frac{d\sigma}{dN} = C(\Delta K)^m$
- To predict RUL (remaining useful life, remaining time before maintenance)

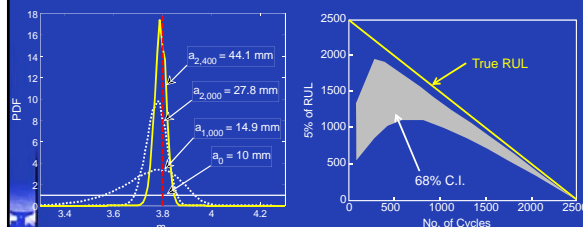


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Parameter Identification (with noise and bias)

- Initial distribution: $m \sim U(3.3, 4.3)$
- Updated m converges to $m_{true} = 3.8$
- Fast convergence as crack grows fast
- RUL: remaining cycles until critical size a_c
- 95th percentile converges to the accurate RUL from a conservative side



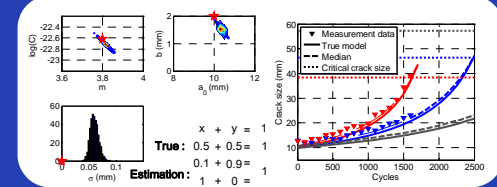
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Multiple Parameters – Challenge in Correlation

- In general multiple parameters need to be identified
- Some of them are correlated
 - Strong correlation between m and C
 - Strong correlation between a_0 and $bias$
- Difficult to identify exact values of parameters, but **good enough to estimate the RUL**

D. An, et al. Identification of correlated damage parameters under noise and bias using Bayesian inference, Structural Health Monitoring 11(3), 292-302, 2012



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Summary

- Conventional RBDO needs to include safety measures, which shape uncertainty before/after design
- Bayesian inference is used to quantify the effect of uncertainty reduction by tests during building-block process
- Bayesian inference is also used to reduce uncertainty in Inspection & maintenance process to accurately predict the remaining useful life

Contributors and Supporters

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Supporters

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ONR

Thank you for your attention