

Probabilistic Analysis with LS-OPT

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Overview

- New Features in LS-OPT 2.1
- Probabilistic Structural Behavior
- LS-OPT Capabilities
- Example
- Summary
- Customer Feedback



New Features: LS-OPT 2.1

Probabilistic Modeling

Model and compute structural and response variation

Metamodeling

Kriging Metamodel added to RSM and Neural Nets

Search Methods

Sequential Random Search (SRS)

LS-DYNA interface

Binary interface (LS970)

Other:

- > Improved restart
- > GUI
- LS-OPT is free of charge on a LS-DYNA license



Probabilistic Analysis Objectives

1. Modeling of Variability

Repeatability of Response

2. Design Criteria

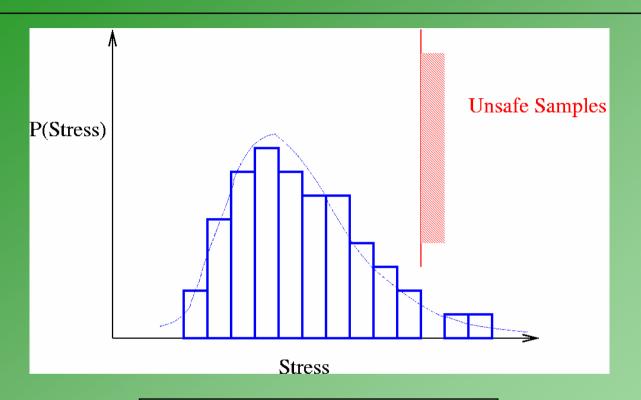
- Probability of failure
- Robustness (Variance)

3. Redesign

Source of variability



Response Variability



Response distribution

- > Mean
- > Standard deviation

Probability of Failure



Response Variation

Deterministic Variation

Due to change in a parameter value (controllable or uncontrollable)

Chaotic Variation

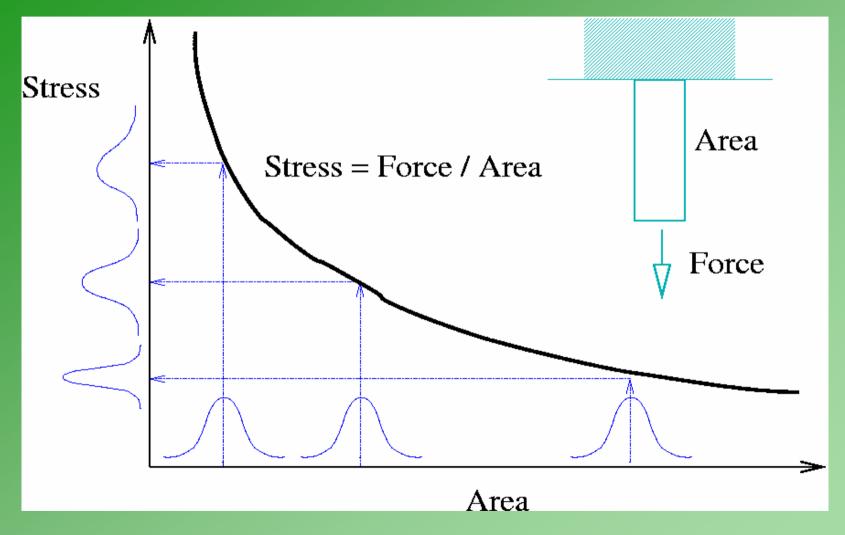
Bifurcation

Random Variation

Variability not explainable by the design model, e.g. mesh, roundoff, lack of convergence

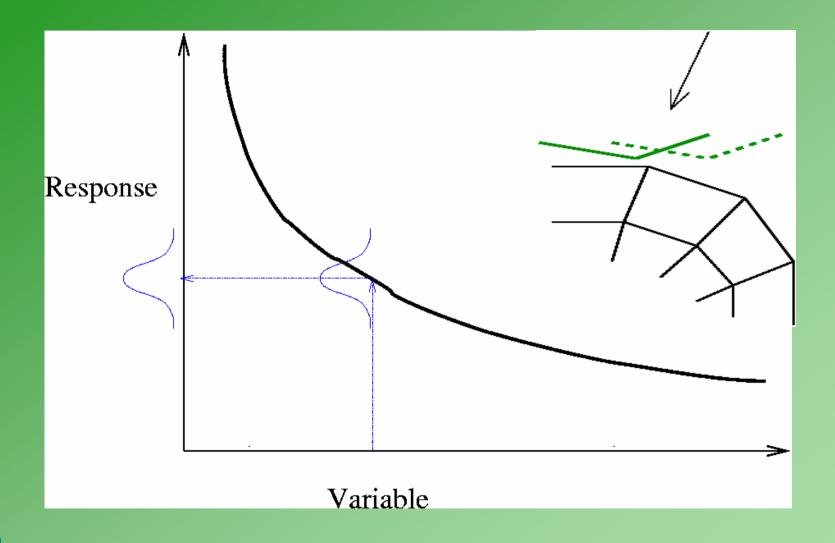


Deterministic Variation





Random Variation





Sources of Variation

Structural

Design parameters

Environment

> Load, material properties

Modeling

Mesh density

Analysis

> Algorithm convergence, contact

Roundoff



Parameter Variation

Control Variables

Values are controlled by designer

- Gauge
- > Shape

Noise Variables

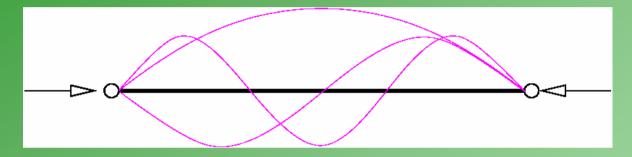
Values not controlled by designer but can vary

- > Load
- > Yield stress
- > Stiffness
- > Thermal

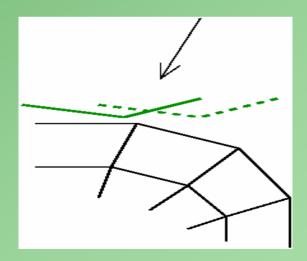


Sources of Variation Classification

- Analysis Variation
 - Physical buckling



> Algorithmic – contact





Sources of Variation Classification

Modeling Variation

- > FE mesh
- Postprocessing, time step size and filter selection, convergence
- Convergence: Iterative implicit solvers
- > Selection of node/element to monitor

Roundoff

- Machine precision
- > Different platforms give different results



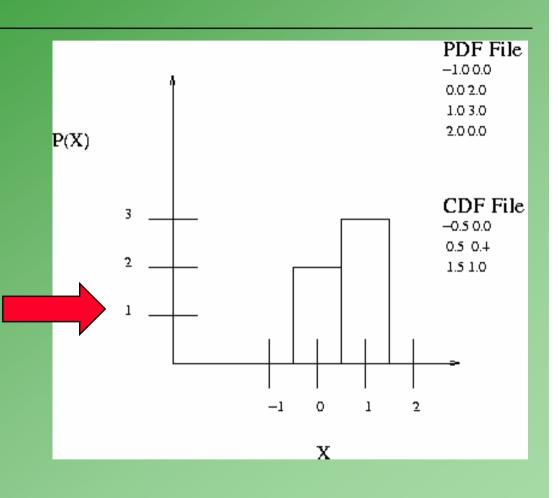
LS-OPT Probabilistic Capabilities

- Statistical Distributions
- Stochastic Sampling Techniques
 - > Monte Carlo
 - Monte Carlo using Metamodels: Polynomials, Neural Nets
- Design Criteria
 - Probability of Failure
 - Robustness (Variance)
- Distributed LS-DYNA job execution and data collection



Statistical Distributions

- Normal
- Uniform
- Lognormal
- User defined
- Weibull



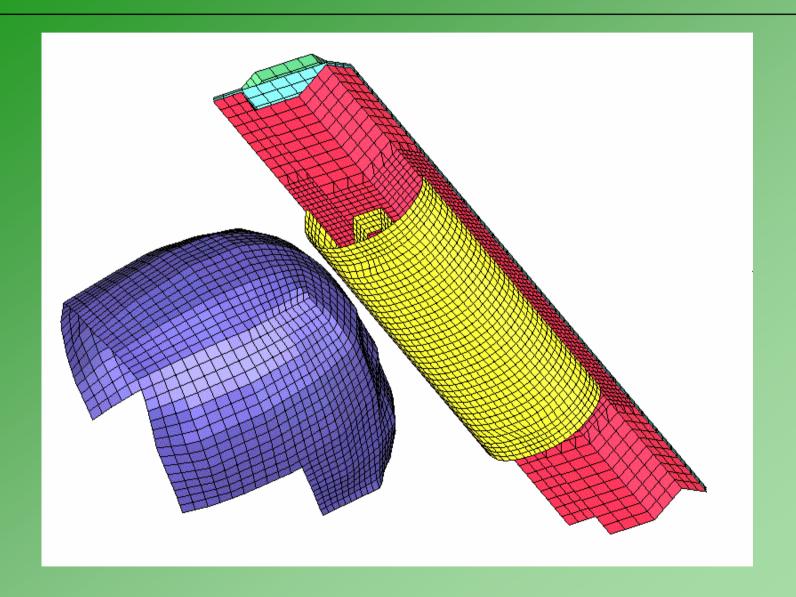


Sampling

- Random
- Latin Hypercube
 - > Structured Monte Carlo
- Space-filling
 - Maximizes minimum distance between any two points



Example – Head Impact Problem



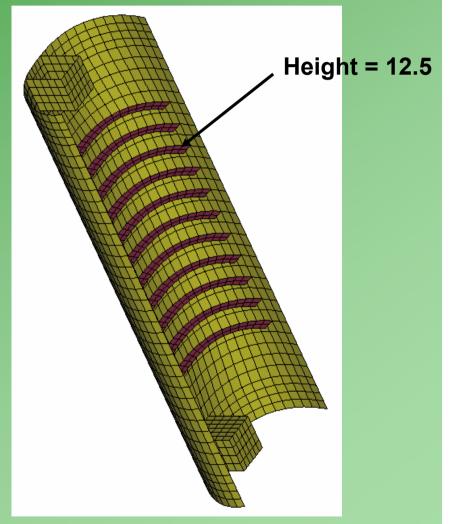


Head Impact Problem

Monitor: Head Injury Criterion:

HIC-d

- Variables:
 - ➤ Hor. Angle of impact
 - 15 degrees
 - 10% standard deviation
 - > Rib height
 - 12.5mm
 - 5% standard deviation

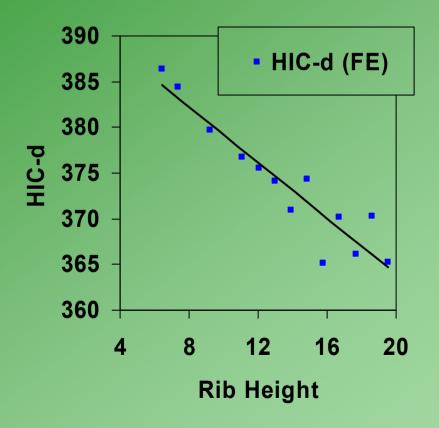




Variation

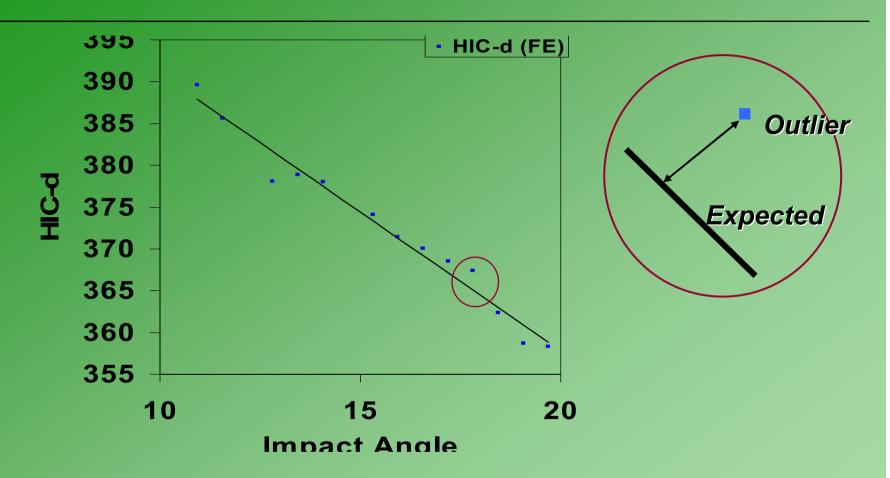
Vary one variable at a time to investigate curvature.

Linear response with some scatter (noise).





Variation



- Quadratic Surface should fit accurately
- Range of response surface is 2σ



Response Variation

■ Baseline Design: HIC-d = 374.4

Monte Carlo Analysis: 150 FE analyses

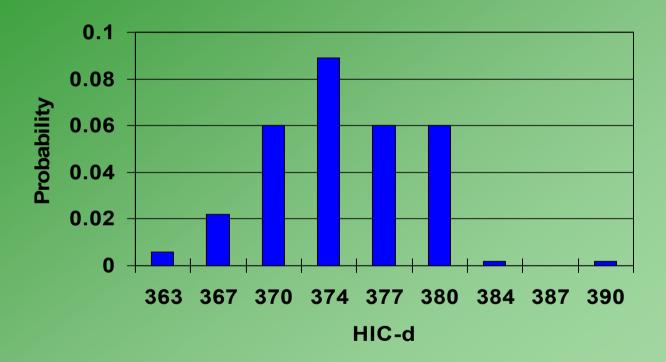
Quadratic Response Surface: 60 FE analyses.
 Residuals have standard deviation of 2.35.

	Monte Carlo	Metamodel
	(150 simulations)	(60 simulations)
Mean	373.9	373.9
Standard deviation		
Deterministic		4.21 (87%)
Deterministic + Outliers	4.85	4.83



Probability of Value

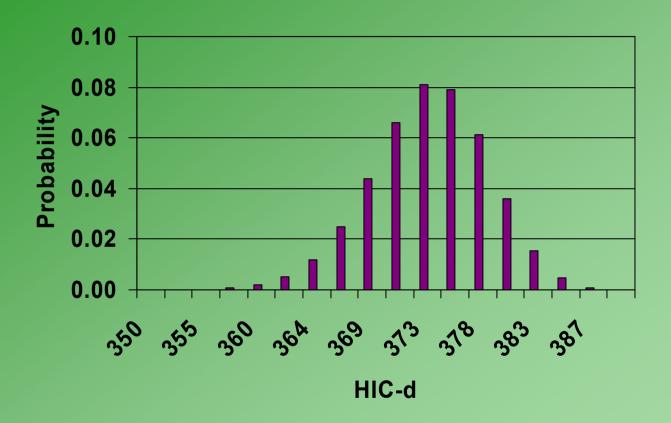
Monte Carlo Analysis Values





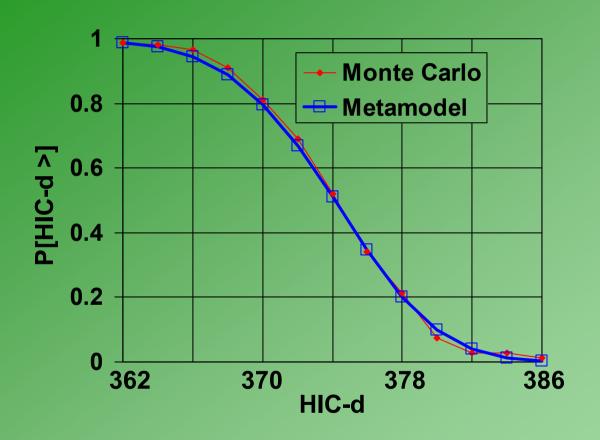
Probability of Value

Metamodel values





Probability of Exceeding Bound





Derivatives

	Angle of Impact	Rib Height
P[HIC-d >378]	-0.15	-0.17

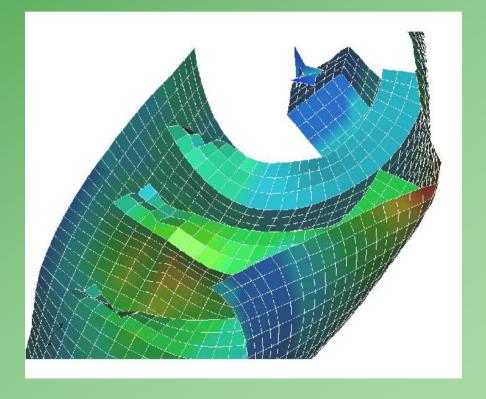
	Std Dev(Angle)	Std Dev(Height)
Std Dev(HIC-d)	2.1	1.0



Displacement Variation

Some displacements may be:

- > Unrelated to a design variable change
- > Not repeatable



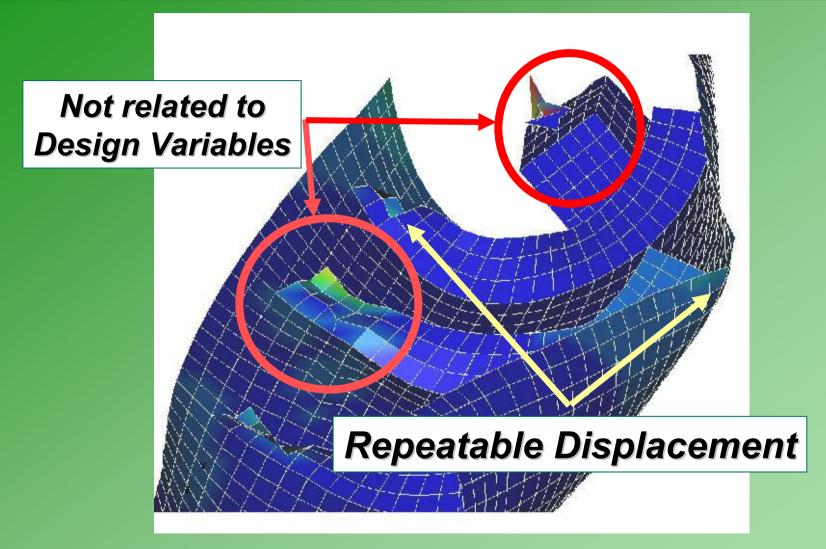


Displacement Variation

- Create metamodel for each nodal displacement.
- Collect outliers (noise) not predicted by metamodel.
- Plot on model
- Investigate
 - Modeling
 - > Bifurcation



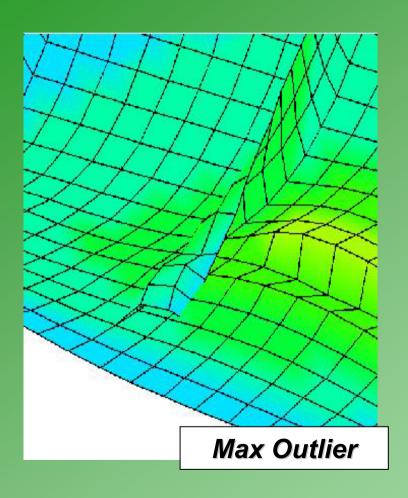
Standard Deviation of Outliers

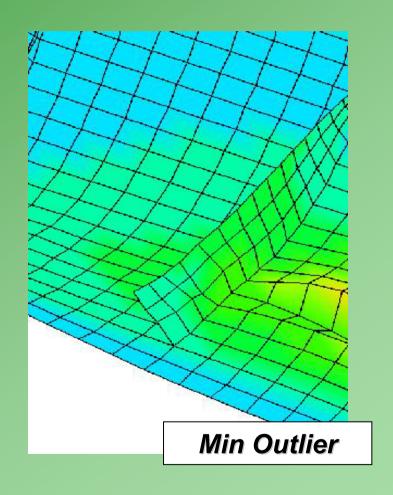




Investigate Outliers

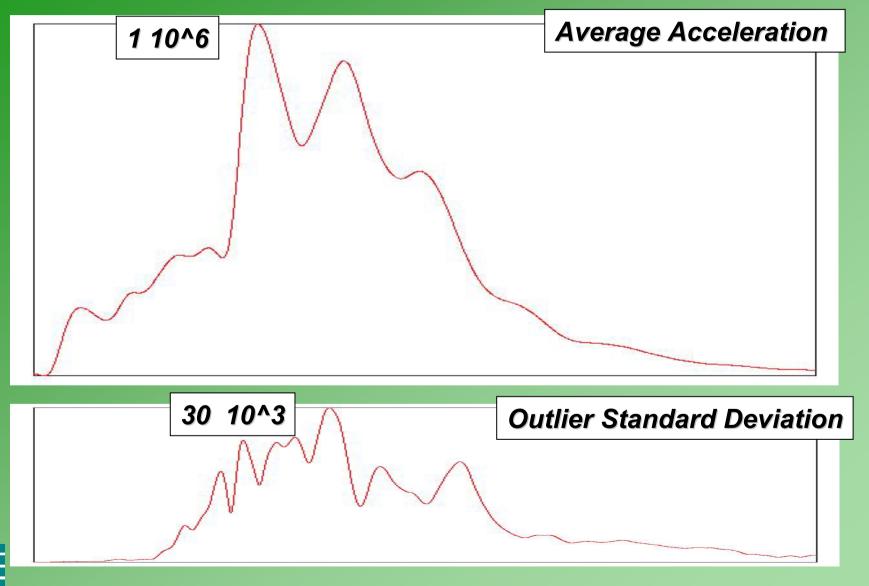
Different buckling modes



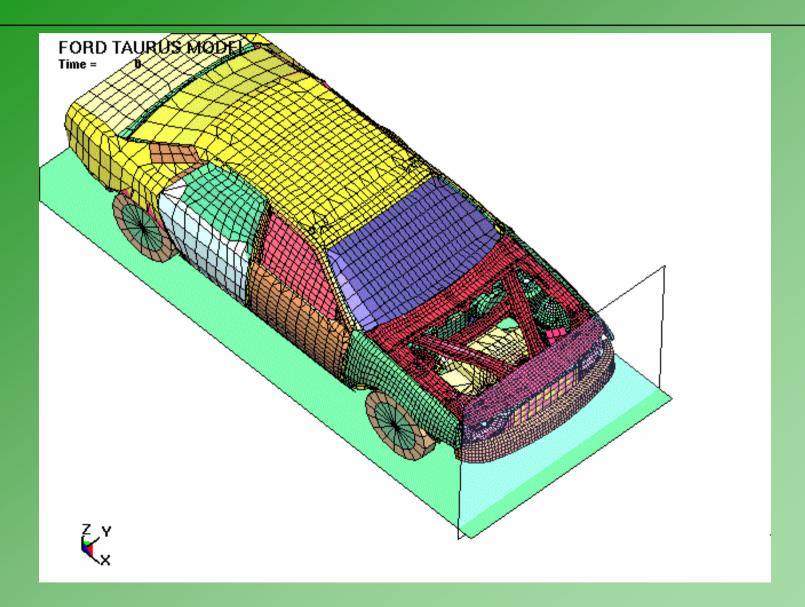




Acceleration History Variation

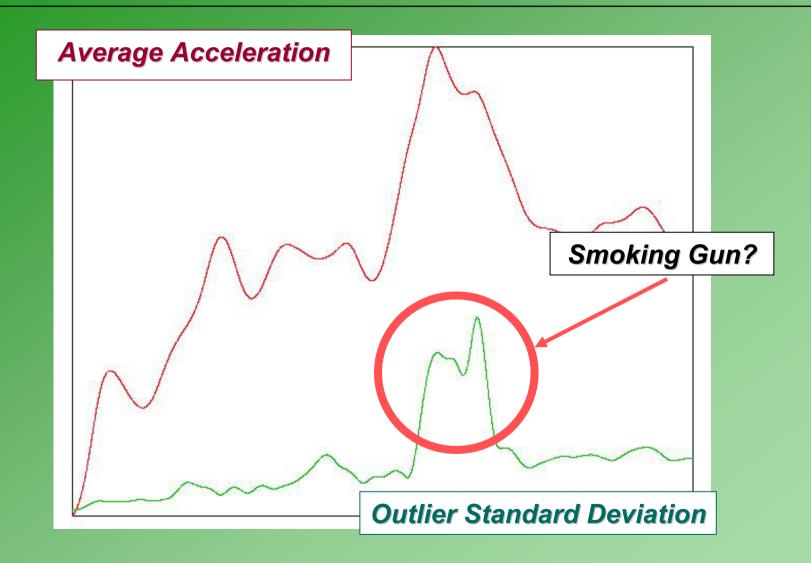


Vehicle Crash





History Variation

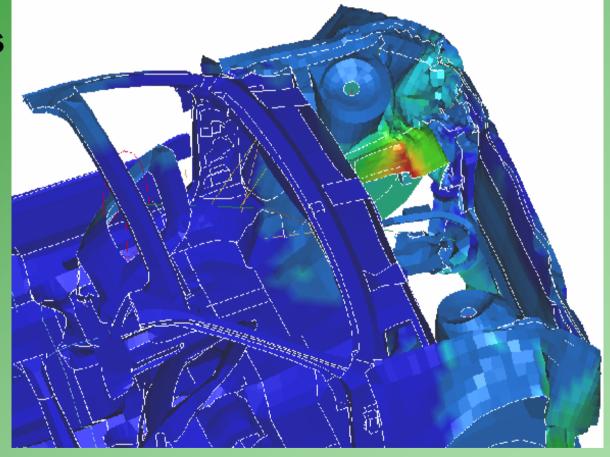




Displacement Variation

Vary angle of impact

25 FE Runs





Summary

- Objective : Repeatable performance of design
- Monte Carlo
 - Mean value
 - Standard deviation
 - Reliability
 - Indication of extreme values
- Metamodels
 - > Allow cost savings
 - > Separation of random components
 - Effect of design variable changes
- Outliers can be informative

