

Random Field Developments

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Introduction

Random Field:

- Random variation of a property (geometry, material, load, ...) over space.
- Spatial domain defined by the observed structure.
- A property takes random values at each point on the structure. Values at different locations may be correlated.





Case I: Given Random Field

Random field data are given

- from simulation of a production process (e.g. sheet metal forming)
- from measurements

and post-processed on the structure or re-imported to optiSLang



Data Reduction

Modal decomposition of covariance matrix:

$$\mathbf{\Psi}^T \mathbf{C}_{XX} \mathbf{\Psi} = \text{diag}\{\lambda_i\}$$

Transformation of data into sub-space:

$$\mathbf{Y} = \mathbf{\Psi}^T \mathbf{X} \quad \Leftrightarrow \quad \mathbf{X} = \mathbf{\Psi} \mathbf{Y}$$

... where statistical evaluations are performed.

Plan: use covariance matrix of data to identify important imperfection shapes.









Case III: Simulation of a Random Field

Stochastic properties given either by measured or computed data, or assumed.

Random field is modelled by mean vector and covariance matrix and simulated by series expansion: random amplitudes times mode shapes.

$$\mathbf{\Psi}^T \mathbf{C}_{XX} \mathbf{\Psi} = \text{diag}\{\lambda_i\}$$

$$\mathbf{Y} = \mathbf{\Psi}^T \mathbf{X} \quad \Leftrightarrow \quad \mathbf{X} = \mathbf{\Psi} \mathbf{Y}$$



Shrinkage Estimator of Covariance Matrix

The covariance matrix has to be estimated from given data

- maintaining positive definiteness
- obtaining high confidence.

One way to get this is the Shrinkage estimator:

$$\widetilde{\mathbf{C}} = \lambda \overline{\mathbf{C}} + (1 - \lambda) \mathbf{\Gamma}, \quad \lambda \in [0; 1]$$

 $l(\lambda) = \|\widetilde{\mathbf{C}} - \mathbf{C}\|^2 \to \min$





Example 2

Random geometric deviations at nodes selected by mesh coarsening.

Mapping of random field realisations by Shepard 3D-interpolation.

Performance criterion for lowest eigenfrequency $P[f_0 \le 400 \, Hz] = 10^{-4}$

After a pilot simulation by LHS, 7 relevant mode shapes were selected which influence most the lowest frequency.

Then, reliability is computed by ARSM using just 94 structural evaluations.

Random field modelling in <u>Slang</u>, Simulation, statistical evaluation in optiSLang.





Conclusions

SoS will develop from a random field analyzer to be a random field modeler and simulator as well.



Conclusions

Means of data reduction maintaining meaningful and interpretable results is a major topic in future developments:

- mesh coarsening,
- efficient modal decomposition,
- choice of relevant mode shapes,
- data interpolation,
- error measures.

For simulation, random field modelling is a key issue:

- estimation of correlations / covariance matrix,
- random field modelling between coarsely spread supports

Interfacing to optiSLang is planned in order to make use of parametrisation and post-processing.

