

# Aleatory Uncertainties in Computational Earthquake Engineering

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# Outline

Introduction

Uncertain Inelastic Dynamics

Formulation

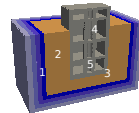
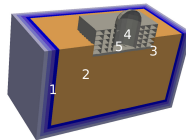
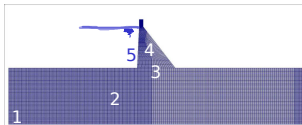
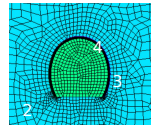
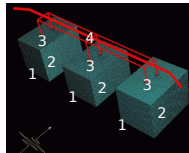
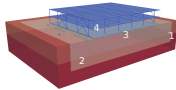
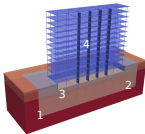
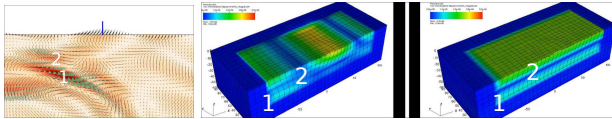
Example

Summary

# Motivation

- Safety and economy of infrastructure
- Design, build and maintain sustainable infrastructure
- Responsible Engineer, with Executive Powers
- Engineer with versatile, quality assured analysis tool to
  - Explore design concepts
  - Assess infrastructure performance
- Engineer needs to know!

# Civil Engineering Analysis Challenges



# Progress, Infrastructure



Nov1990



Jul2019



# Infrastructure Digital Twin

- Infrastructure exists in three dimensions 3D
- Material behavior is nonlinear, inelastic
- Soil and Structure work together
- Modeling, epistemic uncertainty, analysis sophistication
- Parametric, aleatory uncertainty
  - Uncertain material parameters
  - Uncertain loads

# Numerical Prediction under Uncertainty

- Modeling, Epistemic Uncertainty

  - Modeling Simplifications

  - Modeling sophistication for confidence in results

- Parametric, Aleatory Uncertainty

$$M\ddot{u}_i + C\dot{u}_i + K^{ep}u_i = F(t),$$

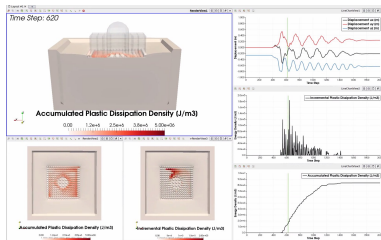
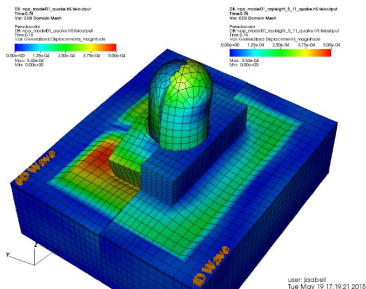
  - Uncertain: mass  $M$ , viscous damping  $C$  and stiffness  $K^{ep}$

  - Uncertain loads,  $F(t)$

  - Results are PDFs and CDFs for  $\sigma_{ij}$ ,  $\epsilon_{ij}$ ,  $u_i$ ,  $\dot{u}_i$ ,  $\ddot{u}_i$

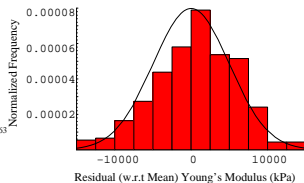
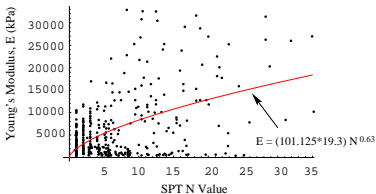
# Modeling, Epistemic Uncertainty

- Simplified modeling, 3D/2D/1D, 1C/2C/3C/6C. damping, viscous, elastic/el-pl, algorithmic
- Modeling simplifications are justifiable if one or two level higher sophistication model demonstrates that features being simplified out are not important (!?)

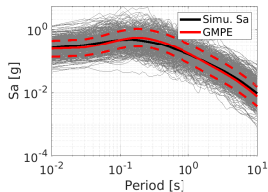
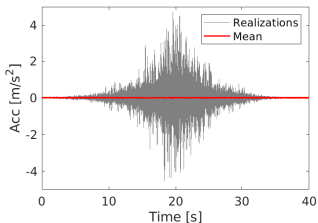




# Parametric, Aleatory Uncertainty



(cf. Phoon and Kulhawy (1999B))



(cf. Wang et al. (2019))

# Engineer Needs to Know!

- Forward propagation of uncertainty, full probabilistic, nonlinear/inelastic Earthquake-Soil-Structure-Interaction, ESSI response in time domain  
(Jeremic et al 2011, Wang et al 2019)
- Backward propagation, sensitivity analysis, quantifies the relative importance of input uncertain parameters on the variance of the probabilistic system response  
(Sobol 2001, Sudret 2008, Jeremic et al 2021)

# Forward Uncertain Inelasticity

- Incremental el-pl constitutive equation

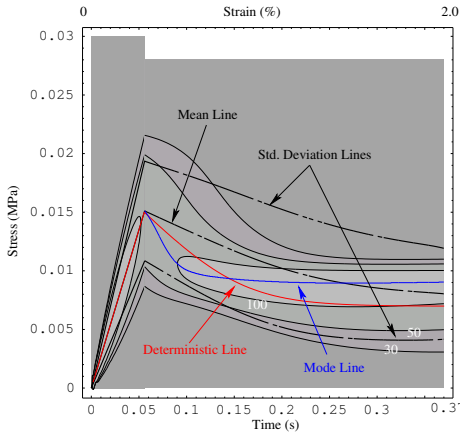
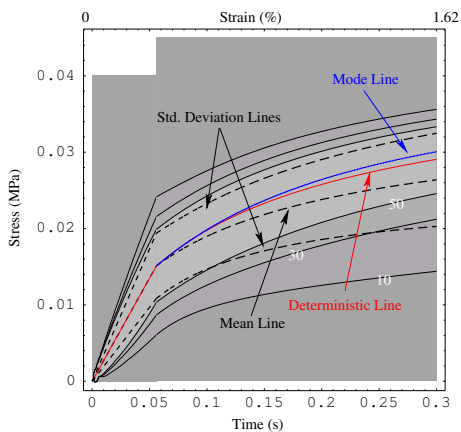
$$\Delta\sigma_{ij} = E_{ijkl}^{EP} \Delta\epsilon_{kl} = \left[ E_{ijkl}^{el} - \frac{E_{ijmnpq}^{el} m_{mn} n_{pq} E_{pqkl}^{el}}{n_{rs} E_{rstu}^{el} m_{tu} - \xi_* h_*} \right] \Delta\epsilon_{kl}$$

- Dynamic Finite Elements

$$M\ddot{u}_i + C\dot{u}_i + K^{ep} u_i = F(t)$$

- Material behavior (LHS) is uncertain
- Loads (RHS) are uncertain

## Formulation

Cam Clay with Random  $G$ ,  $M$  and  $p_0$ 

# Stochastic Elastic-Plastic FEM

$$\text{Dynamic Finite Elements } M\ddot{u}_i + C\dot{u}_i + K^{ep}u_i = F(t)$$

- Input random field/process (non-Gaussian, heterogeneous/non-stationary): Multi-dimensional Hermite Polynomial Chaos (PC) with known coefficients
- Output response process: Multi-dimensional Hermite PC with unknown coefficients
- Galerkin projection: minimize the error to compute unknown coefficients of response process
- SEPFEM eliminates Monte-Carlo inefficiency and inaccuracy

# Stochastic Elastic-Plastic Finite Element Method

- Material uncertainty expanded into stochastic shape funcs.
- Loading uncertainty expanded into stochastic shape funcs.
- Displacement expanded into stochastic shape funcs.
- Jeremić et al. 2011

$$\begin{bmatrix} \sum_{k=0}^{P_d} \langle \Phi_k \Psi_0 \Psi_0 \rangle K^{(k)} & \dots & \sum_{k=0}^{P_d} \langle \Phi_k \Psi_P \Psi_0 \rangle K^{(k)} \\ \sum_{k=0}^{P_d} \langle \Phi_k \Psi_0 \Psi_1 \rangle K^{(k)} & \dots & \sum_{k=0}^{P_d} \langle \Phi_k \Psi_P \Psi_1 \rangle K^{(k)} \\ \vdots & \vdots & \vdots \\ \sum_{k=0}^{P_d} \langle \Phi_k \Psi_0 \Psi_P \rangle K^{(k)} & \dots & \sum_{k=0}^M \langle \Phi_k \Psi_P \Psi_P \rangle K^{(k)} \end{bmatrix} \begin{bmatrix} \Delta u_{10} \\ \vdots \\ \Delta u_{N0} \\ \vdots \\ \Delta u_{1P_U} \\ \vdots \\ \Delta u_{NP_U} \end{bmatrix} = \begin{bmatrix} \sum_{i=0}^{P_f} f_i \langle \Psi_0 \zeta_i \rangle \\ \sum_{i=0}^{P_f} f_i \langle \Psi_1 \zeta_i \rangle \\ \sum_{i=0}^{P_f} f_i \langle \Psi_2 \zeta_i \rangle \\ \vdots \\ \sum_{i=0}^{P_f} f_i \langle \Psi_{P_U} \zeta_i \rangle \end{bmatrix}$$

# Sobol Sensitivity Analysis

- The ANalysis Of VAriance representation (Sobol 2001)
- Total variance of the probabilistic model response  $y = f(\mathbf{X})$

$$D = \text{Var}[f(\mathbf{X})] = \int_{I^n} f^2(\mathbf{x}) d\mathbf{x} - f_0^2$$

- Sobol' indices  $S_{i_1 \dots i_s}$ , fractional contributions from random inputs  $\{X_{i_1}, \dots, X_{i_s}\}$  to the total variance  $D$ :  $S_{i_1 \dots i_s} = D_{i_1 \dots i_s} / D$
- Total sensitivity indices, influence of input parameter  $X_i$

$$S_i^{\text{total}} = \sum_{\mathcal{S}_i} D_{i_1 \dots i_s}$$

## Sobol-Sudret Sensitivity Analysis

- PC expansion of response in ANOVA form (Sudret 2008)
- Multi-dimensional PC bases  $\{\Psi_j(\xi)\}$  decomposition

$$\Psi_j(\xi) = \prod_{i=1}^n \phi_{\alpha_j}(\xi_i)$$

- ANOVA representation  $\rightarrow$  PC-based Sobol' indices  $S_{i_1 \dots i_s}^{PC}$

$$S_{i_1 \dots i_s}^{PC} = \sum_{\alpha \in \mathcal{S}_{i_1, \dots, i_s}} y_{\alpha}^2 \mathbf{E} [\Psi_{\alpha}^2] / D^{PC}$$

- Total Sobol' indices  $S_{j_1 \dots j_t}^{PC, \text{total}}$

$$S_{j_1 \dots j_t}^{PC, \text{total}} = \sum_{(i_1, \dots, i_s) \in \mathcal{S}_{j_1, \dots, j_t}} S_{i_1 \dots i_s}^{PC}$$

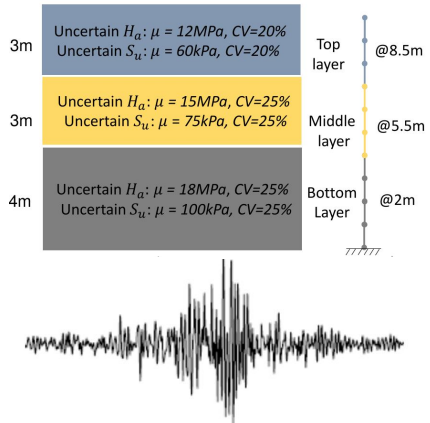
- Sobol-Sudret sensitivity indices within SEPFEM are analytic and inexpensive



## Example

## Example: Stochastic Site Response

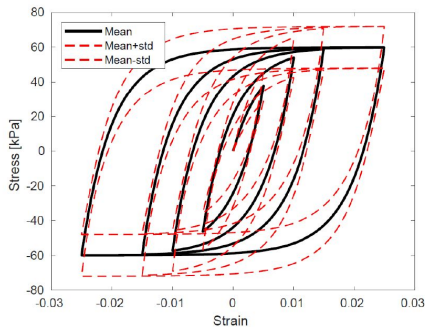
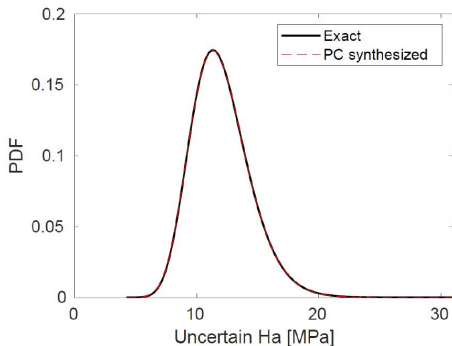
- Uncertain material:  
uncertain random field,  
marginally log-normal  
distribution,  
exponential correlation  
length 10m
- Uncertain seismic  
rock motions:  
seismic scenario  
 $M=7$ ,  $R=50\text{km}$



## Example

## Stochastic Material Parameters

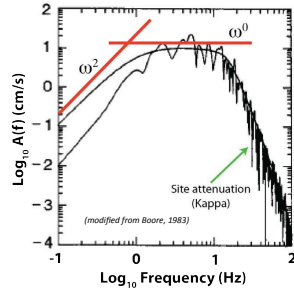
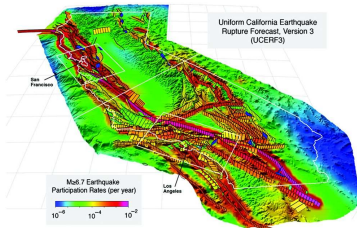
Log-normal distributed random field with PC Dim. 3 Order 2



## Example

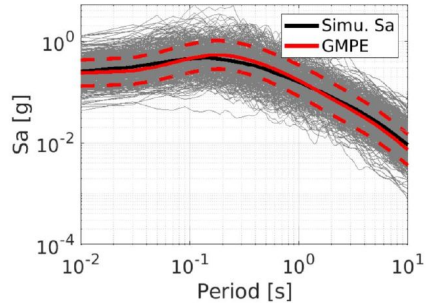
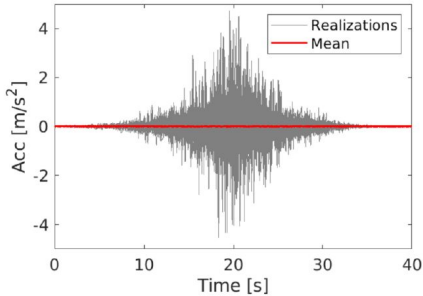
# Stochastic Seismic Motion Development

- UCERF3 (Field et al. 2014)
- Stochastic motions (Boore 2003)
- Polynomial Chaos Karhunen-Loève expansion
- Probabilistic DRM (Bielak et al. 2003, Wang et al. 2021)



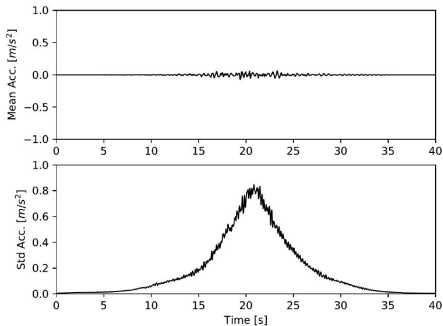
## Example

## Stochastic Seismic Motions

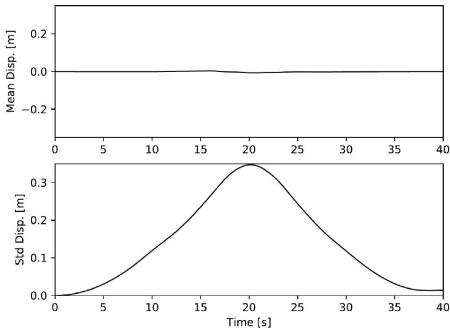


## Example

## Stochastic Site Response, Mean and St.Dev.



Acceleration



Displacement

# Sensitivity Analysis

Total variance in PGA, in this case (!), dominated by uncertain ground motions

49% from uncertain rock motions at depth

2% from uncertain soil

49% from interaction of uncertain rock motions and uncertain soil

## Summary

- Numerical analysis to predict and inform
- Engineer needs to know!
- Real-ESSI Simulator System: <http://real-essi.info>
- Collaborators at UC Davis: Wang, Yang, Lacour, Staszewska
- Collaboration with and financial support from the US-NRC, US-NSF, US-DOE, UN-IAEA, is much appreciated
- "Le doute n'est pas une condition agréable, mais la certitude est absurde" (Voltaire)