# Reliability and Sensitivity Analysis with OpenSees

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# Outline

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- Solution methods
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- □ Stochastic nonlinear dynamic analysis
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### Formulation of structural reliability



$$f_{\mathbf{X}}(\mathbf{x})$$
 Distribution of **X**

Failure domain  $\Omega_{\mathbf{x}}$ 

$$p_f = \int_{\Omega_x} f_{\mathbf{X}}(\mathbf{x}) d\mathbf{x}$$
 Failure probability



### Formulation of structural reliability

Component reliability problem

 $\Omega_{\mathbf{x}} \equiv \left\{ g(\mathbf{x}) \le 0 \right\}$ 

System reliability problem

$$\Omega_{\mathbf{x}} \equiv \left\{ \bigcup_{k} \bigcap_{i \in C_{k}} g_{i}(\mathbf{x}) \leq 0 \right\}$$



 $g(\mathbf{x}), g_i(\mathbf{x})$  limit-state functions (must be continuously differentiable) e.g.,  $g_i(\mathbf{x}) = \delta_{cr} - \delta_i(\mathbf{x})$  failure due to excessive *i*<sup>th</sup> story drift

#### Solution by First-Order Reliability Method (FORM)

$$\mathbf{u} = \mathbf{T}(\mathbf{x}), \quad g(\mathbf{x}) \to G(\mathbf{u})$$
$$\mathbf{u}^* = \arg\min_{\mathbf{u}} \left\{ \|\mathbf{u}\| \mid G(\mathbf{u}) \le \mathbf{0} \right\}$$
$$\beta = \mathbf{\alpha} \cdot \mathbf{u}^* \quad \text{reliability index}$$
$$\alpha = -\frac{\nabla G(\mathbf{u})}{\|\nabla G(\mathbf{u})\|}_{\mathbf{u}=\mathbf{u}^*}$$
$$p_f \cong \Phi(-\beta)$$

Der Kiureghian, A. (2005). First- and second-order reliability methods. Chapter 14 in *Engineering design reliability handbook*, E. Nikolaidis, D. M. Ghiocel and S. Singhal, Edts., CRC Press, Boca Raton, FL.



### Solution by Importance Sampling (IS)

$$p_f \cong \frac{1}{N} \sum_{i=1}^N I\{G(\mathbf{u}_i) \le 0\} \frac{\varphi(\mathbf{u}_i)}{h(\mathbf{u}_i)}$$

 $\mathbf{u}_i$  simulated according to  $h(\mathbf{u}_i)$ 



#### Uncertainty propagation

 $\delta = \delta(\mathbf{X})$  response quantity of interest

First-order approximations:

$$\mu_{\delta} \cong \delta(\mathbf{M}_{\mathbf{X}})$$
  
$$\sigma_{\delta}^{2} \cong \nabla_{\mathbf{X}} \delta \boldsymbol{\Sigma}_{\mathbf{X}\mathbf{X}} \nabla_{\mathbf{X}} \delta^{\mathrm{T}}$$

- $M_X$  mean vector
- $\Sigma_{\mathbf{X}\mathbf{X}}$  covariance matrix
- $\nabla_{\mathbf{X}} \delta$  gradient row-vector evaluated at the mean

### Response sensitivity analysis

For both FORM and FOSM, we need

Available methods in OpenSees:

- □ Finite difference
- Direct Differentiation Method (DDM)
   Differentiate equations of motion and solve for response derivative equations as adjoint to the equations of motion. Equations for the response derivative are linear, even for nonlinear response.

 $\partial \delta(\mathbf{x})$ 

DDM is more stable, accurate and efficient than finite difference.

- Zhang, Y., and A. Der Kiureghian (1993). Dynamic response sensitivity of inelastic structures. *Comp. Methods Appl. Mech. Engrg.*, 108(1), 23-36.
- Haukaas, T., and A. Der Kiureghian (2005). Parameter sensitivity and importance measures in nonlinear finite element reliability analysis. J. Engineering Mechanics, ASCE, 131(10): 1013-1026.

### Sensitivity/Importance measures

FOSM:

$$\frac{\partial \delta(\mathbf{x})}{\partial x_i} \sigma_i$$

FORM:  $\alpha$  = relative importance of **u** variables

 $\gamma$  = relative importance of **X** variables

$$\boldsymbol{\delta} = \left\{ \frac{\partial \beta}{\partial \mu_i} \boldsymbol{\sigma}_i \right\} = \text{reliabilit y importance of mean values}$$
$$\boldsymbol{\eta} = \left\{ \frac{\partial \beta}{\partial \boldsymbol{\sigma}_i} \boldsymbol{\sigma}_i \right\} = \text{reliabilit y importance of stdev values}$$

Haukaas, T., and A. Der Kiureghian (2005). Parameter sensitivity and importance measures in nonlinear finite element reliability analysis. *J. Engineering Mechanics*, ASCE, **131**(10): 1013-1026.

# Methods implemented in OpenSees

Propagation of uncertainty: Estimate second moments of response	First-Order Second- Moment (FOSM) Monte Carlo sampling (MCS)	<ul> <li>Mean</li> <li>Standard deviation</li> <li>Correlation</li> <li>Parameter importance at mean point</li> </ul>
Reliability analysis: Estimate probability of events defined in terms of limit-state functions	First-Order Reliability Method (FORM) Importance Sampling (IS) Second-Order Reliability Method (SORM)	<ul> <li>"Design point"</li> <li>Probability of failure</li> <li>Parameter importance/ sensitivity measures</li> </ul>
Response sensitivity analysis: Determine derivative of response with respect to input or structural properties	The Direct Differentiation Method (DDM) Finite Difference scheme (FD)	• $\frac{\partial \text{ response}}{\partial \text{ parameter}}$ (Used in FOSM and FORM analysis)

# The I-880 Testbed Bridge



### **First-Order Second-Moment Analysis**



# FORM Analysis, $g_1$ , $\lambda_o=0.20$





1	-0.603	Element	141	$\sigma_{y}$
2	-0.538	Element	142	$\sigma_{\rm v}$
3	-0.280	Element	151	$\sigma_{\rm v}$
4	0.240	Element	142	f'c
5	0.232	Element	142	ε <sub>cu</sub>
6	-0.188	Element	152	$\sigma_{y}$
7	-0.177	Element	1502	E
8	0.135	Element	142	f'c
9	-0.122	Element	1602	Е
10	-0.100	Element	161	$\sigma_{y}$
11	0.091	Element	141	f'c
12	0.083	Element	152	f'c
13	-0.073	Element	141	b
14	-0.058	Element	142	ε <sub>c</sub>
15	-0.056	Element	162	$\sigma_{y}$
16	-0.048	Element	142	b
17	0.046	Element	142	ε <sub>c</sub>
18	0.040	Element	152	ε <sub>cu</sub>
19	-0.040	Element	1502	Е
20	0.040	Element	152	f'c
21	-0.032	Element	141	Е
22	0.031	Element	162	f'c
23	0.029	Element	151	f'c
24	-0.027	Node	14002	y-crd.
25	0.026	Element	141	ε <sub>c</sub>
26	0.026	Node	14005	y-crd.
27	-0.023	Element	1602	E
28	-0.022	Element	142	E
29	-0.022	Element	151	b
30	0.021	Element	162	ε <sub>c</sub>

### **Parameter Importance**

# FORM Analysis, g<sub>2</sub>, u<sub>o</sub>=0.30



# Continuity of response derivative









# FORM Analysis, g<sub>3</sub> and g<sub>4</sub>



0.17由

0.2

0.4

 $CDF = P[g_4 < 0]$ 

0.6

0.8

1

0.17

0

10

20

Corresponding PDF

30

40

50

Methods and object-oriented software for FE reliability and sensitivity analysis with application to a bridge structure. *Journal of Computing in Civil Engineering*, ASCE, **21**(3):151-163.

### Stochastic nonlinear dynamic analysis

Representation of stochastic ground motion

$$A(t, \mathbf{u}) = q(t) \sum_{i=1}^{n} s_i(t) u_i = q(t) \mathbf{s}(t) \mathbf{u}$$
  
Temporal nonstationarity  
$$\begin{array}{c} \text{Spectral nonstationarity} \\ \textbf{Spectral nonstationarity} \\ \textbf$$

### Stochastic nonlinear dynamic analysis

Representation of stochastic ground motion

$$A(t,\mathbf{u}) = q(t) \sum_{i=1}^{n} s_i(t) u_i = q(t) \mathbf{s}(t) \mathbf{u}$$



Rezaeian, S. and A. Der Kiureghian (2009). Simulation of synthetic ground motions for specified earthquake and site characteristics. *Earthquake Engineering & Structural Dynamics*, 39:1155-1180.

### Stochastic nonlinear dynamic analysis

 $\Pr[x \le X(t, \mathbf{u})]$  tail probability for threshold *x* at time *t* 

Solution of the above problem by FORM leads to identification of a Tail-Equivalent Linear System (TELS).

TELS is solved by linear random vibration methods to obtain response statistics of interest, e.g., distribution of extreme peak response, fragility curve.



Fujimura, K., and A. Der Kiureghian (2007). Tail-equivalent linearization method for nonlinear random vibration. *Probabilistic Engineering Mechanics*, **22**:63-76.

### Application to MDOF hysteretic system



### Fragility curves for story drifts



Der Kiureghian, A., and K. Fujimura (2009). Nonlinear stochastic dynamic analysis for performancebased earthquake engineering. *Earthquake Engineering and Structural Dynamics*, **38**:719-738.

### Summary and conclusions

- OpenSees is a general-purpose structural analysis platform with unique capabilities for sensitivity and reliability analysis.
- Reliability analysis requires specification of uncertain quantities, their distributions, and definition of performance via limit-state function(s). It provides probability of exceeding specified performance limit(s).
- Uncertainty propagation provides first-order approximation of mean, variance and correlations of response quantities.
- Sensitivity and importance measures provide insight into the relative importance of variables and parameters.
- Stochastic dynamic analysis is performed via tail-equivalent linearization. TELM can be used to generate fragility functions (e.g., in lieu of performing IDAs).