

PBEE Assessment Methods



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Outline



1. Conditional Probabilistic Approach

- Introduction

- SAC/FEMA

- PEER PBEE (**very brief**)

2. Unconditional Probabilistic Approach

Conditional Probabilistic Approach: Introduction



- Aimed to be **practice-oriented**
 - Currently employed mostly in the **academic community**
 - Expected to gain increasing acceptance in **professional practice** in the **very near future**
- Common standpoint of the methods: Use of **intensity measure (IM)** as an **interface** between seismology and structural engineering
 - IM is commonly represented with a **hazard curve**
 - *Structural engineers* need to have **basic information** about the **process of obtaining a hazard curve**, otherwise high potential to end up with **incorrect seismic hazard representation**

Excellent Review Article: Why Do Modern Probabilistic Seismic-Hazard Analyses Often Lead to Increased Hazard Estimates? By J.J. Bommer and N.A. Abrahamson [*Bulletin of the Seismological Society of America*, **96**(6):1967–1977, Dec. 2006]

Conditional Probabilistic Approach: SAC⁽¹⁾/FEMA⁽²⁾



- ❑ During 1994 Northridge earthquake, some steel-moment-resisting-frame (SMRF) buildings underperformed by **showing fractures** in many beam-column joints which were **supposed to remain elastic**
- ❑ Originally developed for investigation of this **unexpected behavior** and **assessing the seismic performance** of these SMRF buildings
- ❑ Applicable to **all building types** with some adjustments

(1) SAC is a joint venture of the **Structural Engineers Association of California (SEAOC)**, the **Applied Technology Council (ATC)**, and **California Universities for Research in Earthquake Engineering (CUREe)**, formed to address both immediate and long-term needs related to solving the problem of the WSMF connection.

(2) US **Federal Emergency Management Agency (FEMA)** www.fema.gov

Conditional Probabilistic Approach: SAC/FEMA

- An **empirical method** based on assuming that an engineer uses **ground motions** and a **computational model** of a structure to test the **likelihood** that a building will **perform as intended (?)** over the **period of interest**

System Performance

Ex. 1: Is the structure capable of remaining **fully operational** in a **frequent** earthquake?

Hazard Levels (Return Period)	System Performance			
	Fully Operational	Operational	Life Safety	Near Collapse
Frequent (43 years)	●	○	○	○
Occasional (72 years)	△	●	○	○
Rare (475 years)	◆	△	●	○
Very rare (949 years)		◆	△	●

- : unacceptable performance
- : basic safety objective
- △: essential hazardous objective
- ◆: safety critical objective

Ex. 2: Is the structure capable of **surviving** a **very rare** earthquake?

Conditional Probabilistic Approach: SAC/FEMA



- Can be considered as a **special application** of the more general **PEER PBEE** framework (*to be discussed later!*)
 - Complete consideration of **uncertainty and probability**
 - Performance assessment **not with decision variables** (DV)
 - Performance assessment considering
 - Intensity Measure (IM)
 - Engineering Demand Parameter (EDP)
 - Capacity of the Engineering Demand Parameter (ECP)
 - DV can be interpreted as a *binary* indicator of achieving the performance level:
 - **0**: unacceptable performance
 - **1**: acceptable performance

Conditional Probabilistic Approach: SAC/FEMA

Motivation for Consideration of Uncertainty

Traditional earthquake design (TED) philosophy:

- Prevent damage in low-intensity EQ (**50% in 50 years**)
 - Limit damage to repairable levels in medium-intensity EQ (**10% in 50 years**)
 - Prevent collapse in high-intensity EQ (**2% in 50 years**)
-
- ❑ If an engineer would accept that the world is **deterministic**, then in the case that he/she observes a structure **not collapsing for the 2% in 50 years event**, he/she could conclude that the probability of **global collapse of the building would certainly be less than 2% in 50 years**

 - ❑ There are many **sources of uncertainty** in this problem that need to be taken into account for a realistic assessment of the collapse probability of this building

 - ❑ These **uncertainties** will probably make the **probability of global collapse much higher than 2% in 50 years**

Conditional Probabilistic Approach: SAC/FEMA



Types and Sources of Uncertainty



Alea (Latin) = Dice

Aleatory uncertainty (*randomness*): The uncertainty inherent in a nondeterministic (stochastic, random) phenomenon.

Examples: The location and the magnitude of the next earthquake and the intensity of the ground shaking generated at a given site

Epist (Greek): Knowledge

Epistemic uncertainty: The uncertainty attributable to incomplete knowledge about a phenomenon that affects our ability to model it.

Examples: The definition of parameters and rules of a constitutive model for concrete

Conditional Probabilistic Approach: SAC/FEMA



Background

Total probability theorem:

Given n mutually exclusive events* A_1, \dots, A_n whose probabilities sum to 1.0, then the probability of an arbitrary event B :

$$p(B) = p(B|A_1)p(A_1) + p(B|A_2)p(A_2) + \dots + p(B|A_n)p(A_n)$$

$$p(B) = \sum_i [p(B|A_i) p(A_i)]$$

Conditional probability of B given the presence of A_i

Probability of A_i

*Occurrence of any one of them automatically implies the non-occurrence of the remaining $n-1$ events

Conditional Probabilistic Approach: SAC/FEMA



Background

- Calculate the probability of exceedance (POE), **P**, of the i^{th} value of EDP

$$P(\text{EDP}^i) = \sum_m P(\text{EDP}^i | \text{IM}_m) p(\text{IM}_m)$$

Computational model & simulations Hazard curve

- Calculate the probability (**p**) of EDP^i
for $i = 1 : \# \text{ of EDP values}$
 $p(\text{EDP}^i) = P(\text{EDP}^i)$ if $i = \# \text{ of EDP values}$
 $p(\text{EDP}^i) = P(\text{EDP}^i) - P(\text{EDP}^{i+1})$ otherwise

Conditional Probabilistic Approach: SAC/FEMA



Background

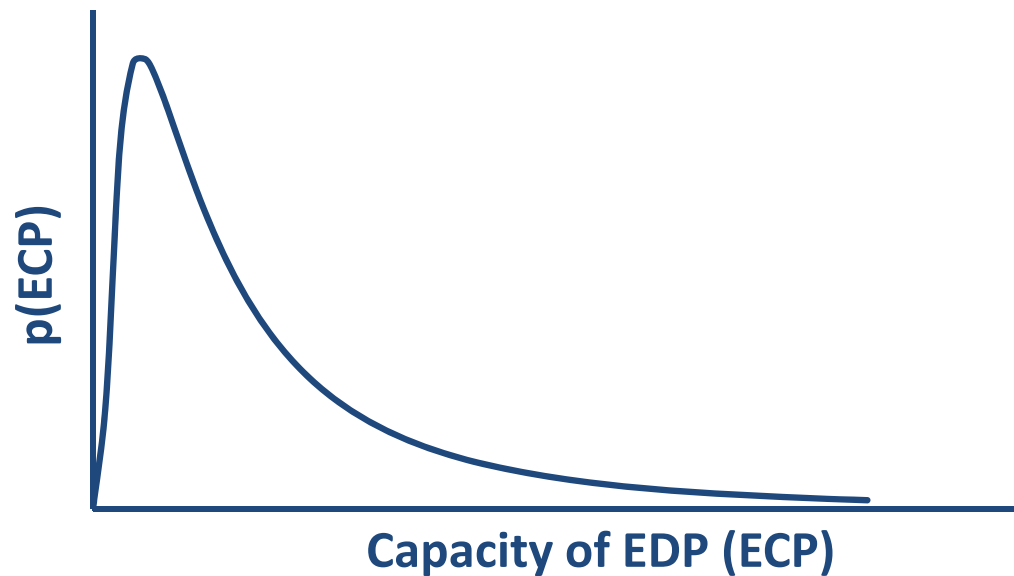
- If an engineer was sure that the structure would fail its performance level when it reached a certain *limiting* EDP value (EDP^L), then the probability of not meeting that performance level (p_{fPL}) would be $P(EDP^L)$
- However, the engineer cannot be sure about the above issue, since there is **uncertainty** in the **corresponding capacity limit**
- Theoretically, every value of EDP has a **finite likelihood of making a structure to fail a performance level**
- Uncertainty in the capacity of an EDP (**ECP**) should be considered for the calculation of p_{fPL}
- Considering the uncertainty in capacity: p_{fPL} is defined as the probability of ECP being smaller than EDP [$p(ECP < EDP)$]
- Same uncertainty is considered in a different format in **Damage Analysis** stage of PEER PBEE framework

Conditional Probabilistic Approach: SAC/FEMA



Background

Uncertainty in capacity: Capacity of EDP that corresponds to a Performance Level (PL) is represented with a probability distribution



Conditional Probabilistic Approach: SAC/FEMA

Background

Table 6-7 Modeling Parameters and Numerical Acceptance Criteria for Nonlinear Procedures—
Reinforced Concrete Beams

Modeling Parameters ³			Acceptance Criteria ³							
			Plastic Rotation Angle, radians							
PR	Plastic Rotation Angle, radians	Residual Strength Ratio	Performance Level							
			Primary		Secondary					
			IO	LS	CP	LS	CP			
Conditions			a	b	c	IO	LS	CP	LS	CP
i. Beams controlled by flexure ¹										
$\frac{\rho - \rho'}{\rho_{bal}}$	Trans. Reinf. ²	$\frac{V}{b_w d \sqrt{f'_c}}$								
≤ 0.0	C	≤ 3	0.025	0.05	0.2	0.010	0.02	0.025	0.02	0.05
≤ 0.0	C	≥ 6	0.02	0.04	0.2	0.005	0.01	0.02	0.02	0.04
≥ 0.5	C	≤ 3	0.02	0.03	0.2	0.005	0.01	0.02	0.02	0.03
≥ 0.5	C	≥ 6	0.015	0.02	0.2	0.005	0.005	0.015	0.015	0.02
≤ 0.0	NC	≤ 3	0.02	0.03	0.2	0.005	0.01	0.02	0.02	0.03
≤ 0.0	NC	≥ 6	0.01	0.015	0.2	0.0015	0.005	0.01	0.01	0.015
≥ 0.5	NC	≤ 3	0.01	0.015	0.2	0.005	0.01	0.01	0.01	0.015
≥ 0.5	NC	≥ 6	0.005	0.01	0.2	0.0015	0.005	0.005	0.005	0.01

FEMA-356

- If $PR < 0.01$ radians → PL = IO
- If $0.01 < PR < 0.02$ → PL = LS
- If $0.02 < PR < 0.025$ → PL = CP

No uncertainty in capacity



- PL = IO → $p_{fPL} = P(PR=0.01)$
- PL = LS → $p_{fPL} = P(PR=0.02)$
- PL = CP → $p_{fPL} = P(PR=0.025)$

Uncertainty in capacity



- PL = IO → $p_{fPL} \neq P(PR=0.01)$
- PL = LS → $p_{fPL} \neq P(PR=0.02)$
- PL = CP → $p_{fPL} \neq P(PR=0.025)$

Reminder: **p**: probability, **P**: probability of exceedance (POE)

Conditional Probabilistic Approach: SAC/FEMA

Background

- Calculate the probability of exceedance (POE), P , of the i^{th} value of EDP

$$P(\text{EDP}^i) = \sum_m P(\text{EDP}^i | \text{IM}_m) p(\text{IM}_m)$$

Computational model
& simulations

Hazard curve

- Calculate the probability (p) of EDP^i

for $i = 1 : \#$ of EDP values

$$p(\text{EDP}^i) = P(\text{EDP}^i) \quad \text{if } i = \# \text{ of EDP values}$$

$$p(\text{EDP}^i) = P(\text{EDP}^i) - P(\text{EDP}^{i+1}) \quad \text{otherwise}$$

From slide 10

- Calculate the probability of **not meeting a performance level** (p_{fPL})

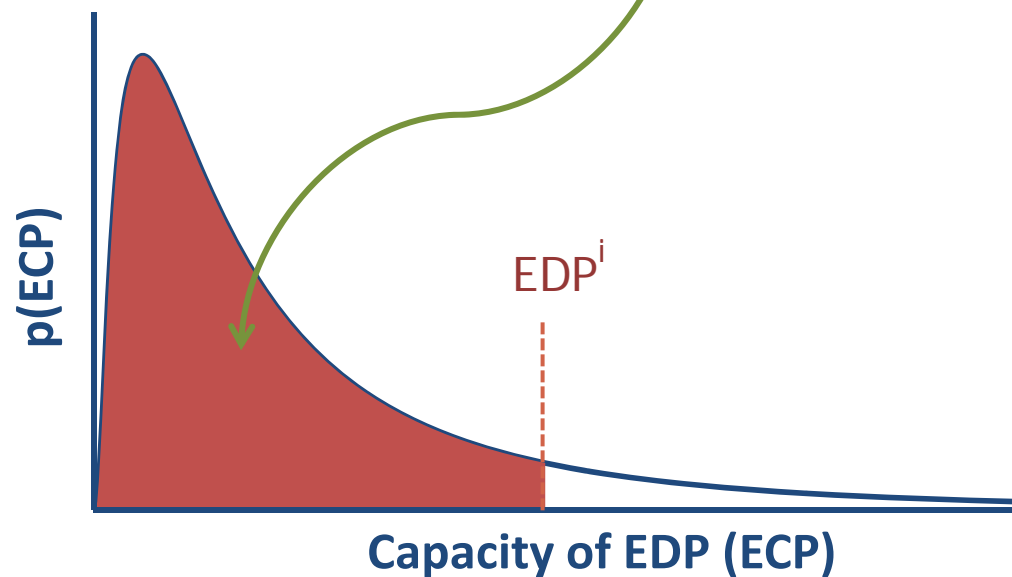
$$p_{\text{fPL}} = p[\text{ECP} \leq \text{EDP}] = \sum_i p(\text{ECP} \leq \text{EDP}^i | \text{EDP}^i) p(\text{EDP}^i)$$

Conditional Probabilistic Approach: SAC/FEMA

Background

- Calculate the probability of **not meeting a performance level** (p_{fPL})

$$p_{fPL} = p[ECP \leq EDP] = \sum_i p(ECP \leq EDP^i | EDP^i) p(EDP^i)$$



Conditional Probabilistic Approach: SAC/FEMA



Application Formats

- Approach requires large number of numerical simulations
- Computational effort introduced by the probability equations
- Two theoretically equivalent (**with some practical differences**) formats to reduce the computational burden:
 - Mean Annual Frequency (MAF) Format: A simple, closed-form evaluation of seismic risk (**involving hazard, exposure, & vulnerability**)
 - Demand and Capacity Factored Design (DCFD) Format: A check of whether the building satisfies the selected limit-state requirements

Conditional Probabilistic Approach: SAC/FEMA

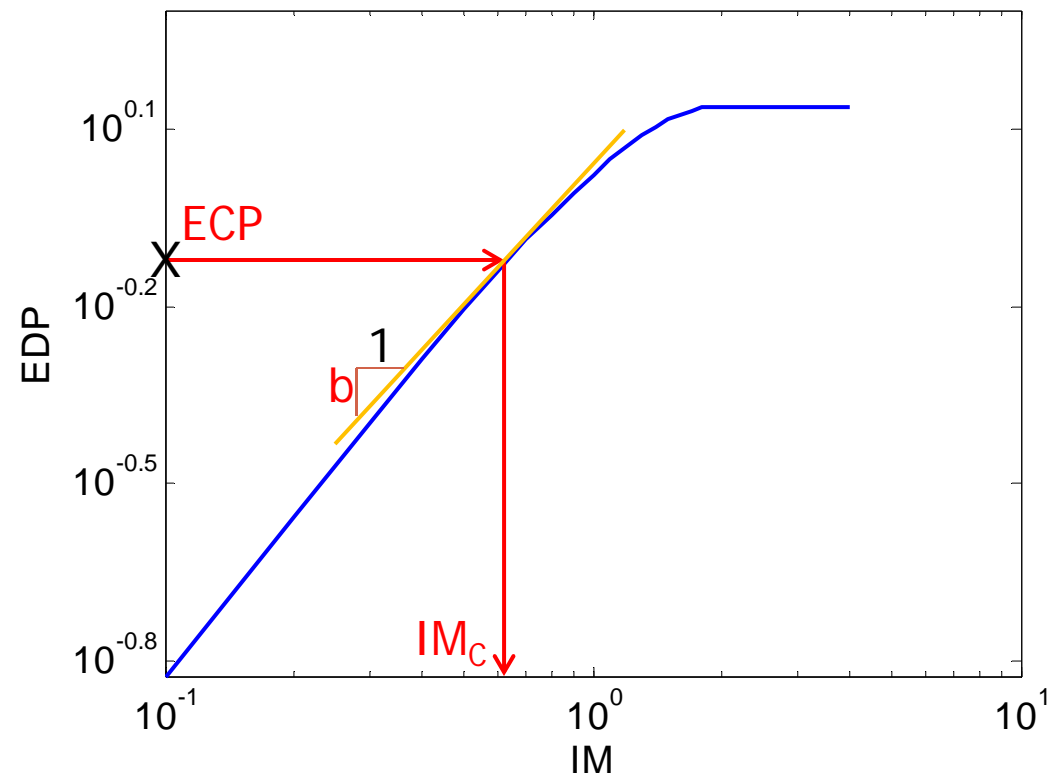


Application Formats: MAF

Mean annual frequency of not meeting a certain performance level PL: λ_{fPL}

$$\lambda_{fPL} = H(IM_C) \exp\left[\frac{1}{2} \frac{k^2}{b^2} (\beta_{DT}^2 + \beta_{CT}^2)\right]$$

IM_C : Value of IM that causes the structure to reach the EDP capacity (ECP) associated with the onset of the limit-state corresponding to the performance level PL.



Conditional Probabilistic Approach: SAC/FEMA

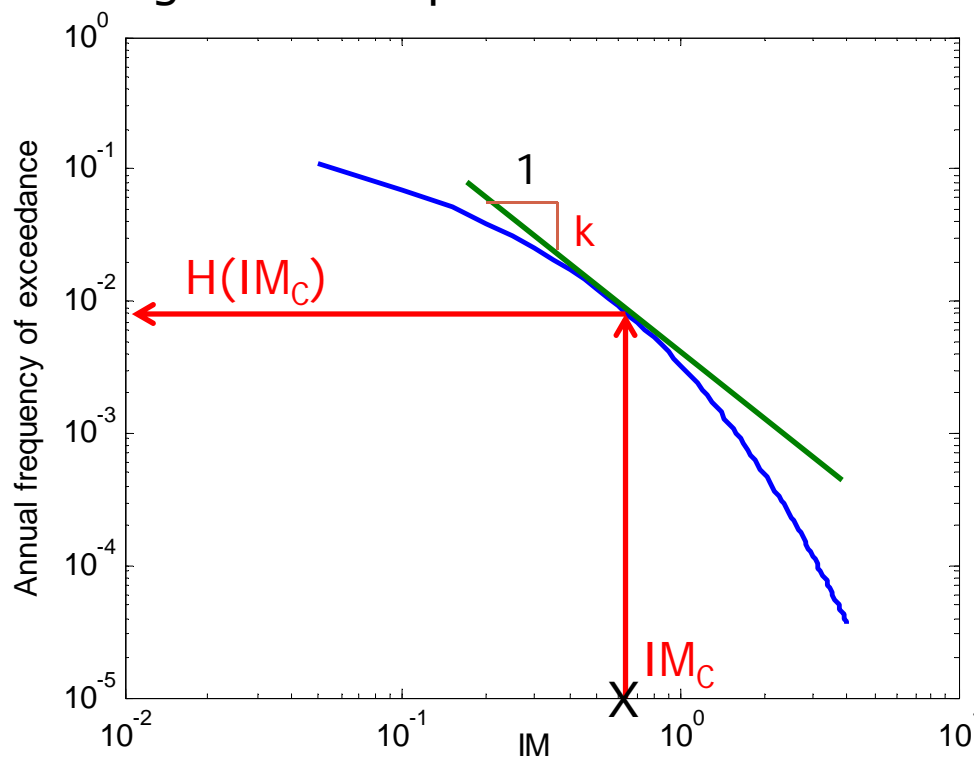


Application Formats: MAF

Mean annual frequency of not meeting a certain performance level PL: λ_{fPL}

$$\lambda_{fPL} = H(IM_C) \exp\left[\frac{1}{2} \frac{k^2}{b^2} (\beta_{DT}^2 + \beta_{CT}^2)\right]$$

$H(IM_C)$: Value of seismic hazard at IM_C



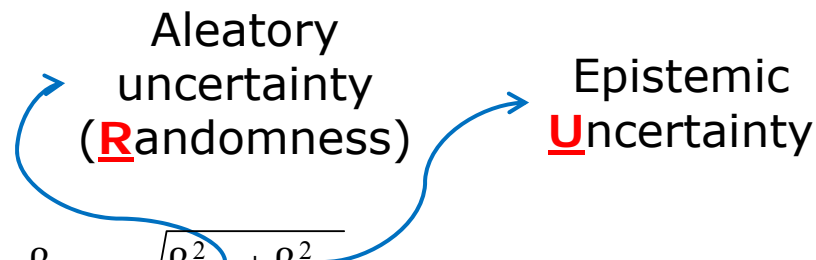
Conditional Probabilistic Approach: SAC/FEMA



Application Formats: MAF

Mean annual frequency of not meeting a certain performance level PL: λ_{fPL}

$$\lambda_{fPL} = H(IM_C) \exp \left[\frac{1}{2} \frac{k^2}{b^2} (\beta_{DT}^2 + \beta_{CT}^2) \right]$$



β_{DT} : Dispersion in **D**emand → $\beta_{DT} = \sqrt{\beta_{DR}^2 + \beta_{DU}^2}$

β_{CT} : Dispersion in **C**apacity → $\beta_{CT} = \sqrt{\beta_{CR}^2 + \beta_{CU}^2}$

Dispersion: Standard dev. of the log of the data

Aleatory Uncertainty

β_{DR} : Variability observed in structural response (**D**emand) from record-to-record

β_{CR} : Natural variability observed in tests to determine the EDP capacity (**E**CP) of a structural or non-structural component

Epistemic Uncertainty

β_{DU} : Uncertainty in modeling and analysis methods for estimating **d**emand

β_{CU} : Incomplete knowledge of the structure for estimating **c**apacity

Conditional Probabilistic Approach: SAC/FEMA



Application Formats: MAF

Mean annual frequency of not meeting a certain performance level PL: λ_{fPL}

$$\lambda_{fPL} = H(IM_C) \exp \left[\frac{1}{2} \frac{k^2}{b^2} (\beta_{DT}^2 + \beta_{CT}^2) \right]$$



Probability of not meeting a certain performance level PL: p_{fPL}

$$p_{fPL} = 1 - \exp(-\lambda_{fPL} t) \quad t: \text{considered time period [years]}$$

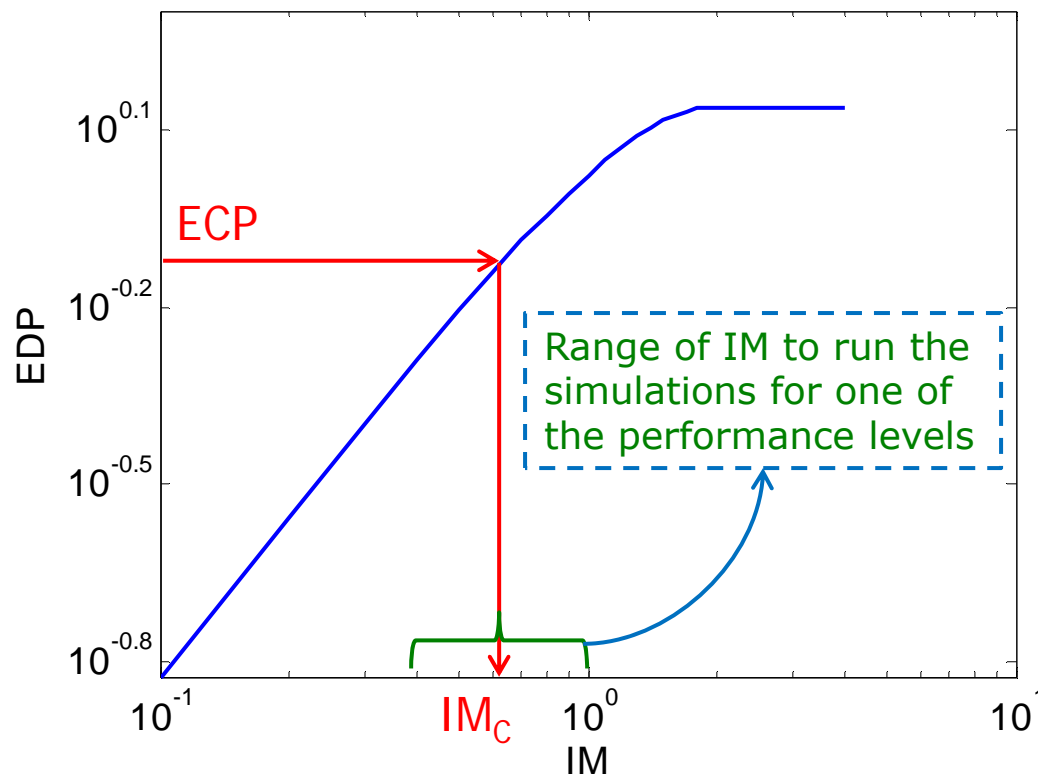
Conditional Probabilistic Approach: SAC/FEMA



Application Formats: MAF

Advantage:

- Time history simulations **do not need** to be conducted **for all IM values**
- It **may be sufficient** to conduct the simulations for an **estimated range of IM** which covers the ECP values of the considered **performance levels**



Conditional Probabilistic Approach: SAC/FEMA



Application Formats: DCFD

- A **check** of whether a certain **performance level** has been **met or violated**
- Resembles the familiar **Load and Resistance Factor Design (LRFD)** of modern design codes
- Unlike the MAF format, it **cannot provide** an estimate of the **annual frequency** of exceeding a given performance level

Conditional Probabilistic Approach: SAC/FEMA

Application Formats: DCFD

- FC: Factored capacity corresponding to the **Performance Level**
- FD_λ : Factored demand evaluated at the **Hazard Level**
- ECP_m : Median **EDP capacity** for the considered **Performance Level**
- $EDP_{m\lambda}$: Median **demand** evaluated at the IM level corresponding to λ

		System Performance			
		Fully Operational	Operational	Life Safety	Near Collapse
Hazard Levels (Return Period)	Frequent (43 years)	●	○	○	○
	Occasional (72 years)	△	●	○	○
	Rare (475 years)	◆	△	●	○
	Very rare (949 years)		◆	△	●

○: unacceptable performance
●: basic safety objective
△: essential hazardous objective
◆: safety critical objective

A performance objective:

Satisfy a **Performance Level** under a given **Hazard Level**

λ represents the **annual frequency of exceedance associated with the Hazard Level**

$$FC \geq FD_\lambda \Rightarrow \varphi \cdot ECP_m \geq \gamma \cdot EDP_{m\lambda},$$

φ = Uncertainty Factor (~ Strength Reduction Factor),

γ = Uncertainty Factor (~ Load Amplification Factor)

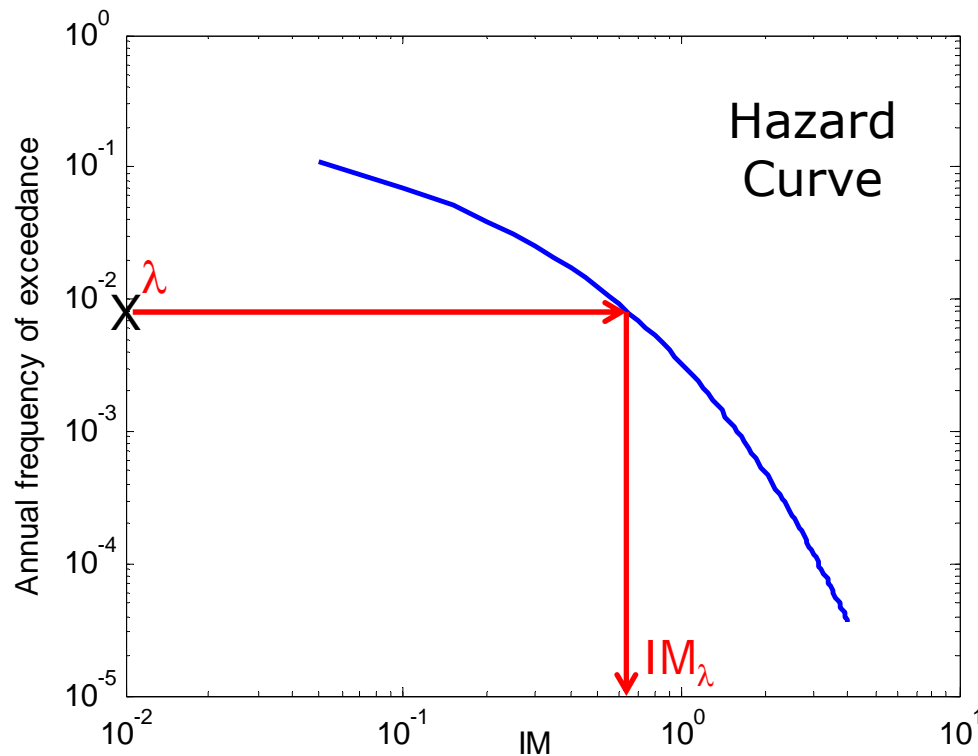
Conditional Probabilistic Approach: SAC/FEMA



Application Formats: DCFD

$$FC \geq FD_\lambda \Rightarrow \varphi \cdot ECP_m \geq \gamma \cdot EDP_{m\lambda}$$

$EDP_{m\lambda}$: Median demand calculated at the **IM value (IM_λ)** corresponding to λ



Conditional Probabilistic Approach: SAC/FEMA



Application Formats: DCFD

$$FC \geq FD_{\lambda} \Rightarrow \phi \cdot ECP_m \geq \gamma \cdot EDP_{m\lambda}$$

$$\phi = \exp\left[-\frac{1}{2} \frac{k}{b} (\beta_{CR}^2 + \beta_{CU}^2)\right] \quad \gamma = \exp\left[\frac{1}{2} \frac{k}{b} (\beta_{DR}^2 + \beta_{DU}^2)\right]$$

Remark:

- **Median** values are considered for **capacity** and **demand**
- **Uncertainty** is considered through the use of ϕ and γ
- **Guarantees** the Performance Objective with a **confidence value greater than 50%**
- **Modifications** have been made in DCFD **to control and increase the confidence level**: Enhanced DCFD (**EDCFD**)

Conditional Probabilistic Approach: SAC/FEMA



Application Formats: EDCFD

$$FC_R \geq FD_{R\lambda} \cdot \exp(K_x \beta_{TU}) \Rightarrow \varphi_R \cdot ECP_m \geq \gamma_R \cdot EDP_{m\lambda} \cdot \exp(K_x \beta_{TU})$$

$$\left. \begin{aligned} \varphi_R &= \exp\left[-\frac{1}{2} \frac{k}{b} \beta_{CR}^2\right] \\ \gamma_R &= \exp\left[\frac{1}{2} \frac{k}{b} \beta_{DR}^2\right] \end{aligned} \right\} \text{Only Aleatory uncertainty}$$
$$\beta_{TU} = \sqrt{\beta_{DU}^2 + \beta_{CU}^2} \left. \vphantom{\beta_{TU}} \right\} \text{Epistemic uncertainty}$$

K_x : Standard normal variate (set of all random variables that obey a given probabilistic law) corresponding to the desired confidence level, α : $K_x = 1.28 \rightarrow \alpha=90\%$; $K_x = 0.00 \rightarrow \alpha=50\%$

EDCFD allows a user-defined **level of confidence** to be incorporated in the assessment.

Differing levels of confidence for:

- Ductile versus brittle modes of failure (larger K_x for brittle)
- Local versus global collapse mechanisms (larger K_x for global)

Conditional Probabilistic Approach: PEER PBEE



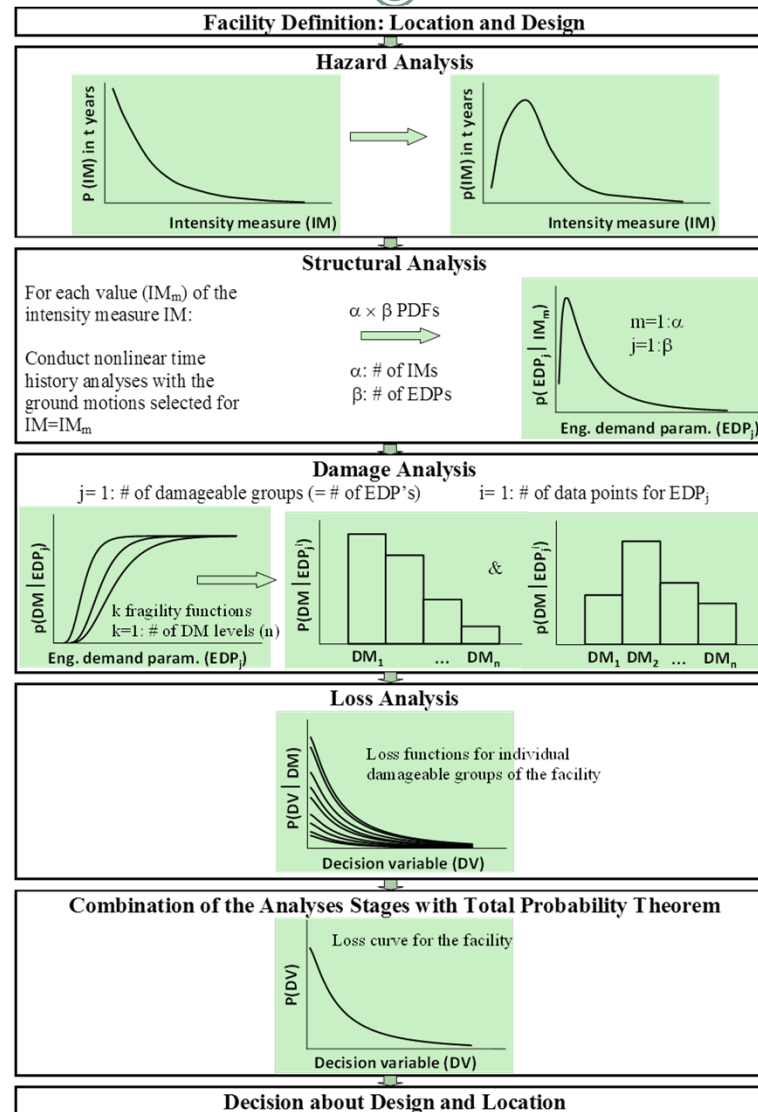
❑ SAC/FEMA

- Complete consideration of **uncertainty and probability**
- Performance assessment **not** with **decision variables** (DV)
- A special application of PEER PBEE framework

❑ PEER PBEE framework

- Complete consideration of **uncertainty and probability**
- Performance assessment with **decision variables** in terms of the **direct interest of various stakeholders**
- Performance assessment considering:
 - Intensity Measure (IM)
 - Engineering Demand Parameter (EDP)
 - Damage Measure (DM)
 - Decision Variable (DV)

Conditional Probabilistic Approach: PEER PBEE



More about PEER PBEE in the afternoon

Unconditional Probabilistic Approach: Introduction

Conditional Probabilistic Approach (CPA)

- Practice-oriented
- Conditioned on IM
- Obtain the $p(\text{IM})$ from hazard curve
- Employ recorded ground motions compatible with IM

Replaced
by

Unconditional Probabilistic Approach (UPA)

- More advanced
- Not conditioned on IM
- Stochastic models to directly describe the random time-series of seismic motion in terms of macro-seismic parameters, e.g. magnitude, distance, ... etc.

- Synthetic ground motions are employed in UPA
- The main difference with the CPA is in the to description of seismic motion at the site (synthetic motions)
- UPA-related research is mostly conducted up to generation of ground motions

Unconditional Probabilistic Approach: Introduction



- ❑ Methods of Unconditional Probabilistic Approach: Describe the randomness in the problem by a vector of random variables (\mathbf{x}). \mathbf{x} should ideally cover the randomness in:
 - Earthquake source
 - Propagation path
 - Site geology/geotechnical aspects
 - Frequency content of the time-series
 - Structural response and capacity

- ❑ Simulations for \mathbf{x} sampled from its probability distribution, $f(\mathbf{x})$

Unconditional Probabilistic Approach



Simulation Methods

Simulation:

- ❑ A robust way to explore the behavior of systems of any complexity
- ❑ Based on the observation of system response to input

$\mathbf{x} = [x_1 \quad x_2 \quad \dots \quad x_n]^T \Rightarrow f(\mathbf{x})$: probability distribution for \mathbf{x}

- Form a set of inputs of \mathbf{x} from $f(\mathbf{x})$ $\mathbf{x}^{<i> = [x_{1i} \quad x_{2i} \quad \dots \quad x_{ni}]^T$
- Obtain the corresponding outputs
- Determine the distribution of the output through **statistical post-processing**

Unconditional Probabilistic Approach



Simulation Methods: Monte Carlo Simulation (MCS)

- A chosen set of inputs for \mathbf{x} : $\mathbf{x}^{<i>} = [x_{1i} \quad x_{2i} \quad \dots \quad x_{ni}]^T$
- If $\mathbf{x}^{<i>}$ fails in meeting certain performance requirements, then the contribution of $\mathbf{x}^{<i>}$ to the probability of not meeting those performance requirements $(p_f) = f(\mathbf{x}^{<i>})d\mathbf{x}$
- Then $p_f = \int_F f(\mathbf{x})d\mathbf{x}$
F covers all $\mathbf{x}^{<i>}$ that fail in meeting the performance requirements

$$p_f = \int_F f(\mathbf{x})d\mathbf{x} = \int \underbrace{I_f(\mathbf{x})}_{\text{Indicator function}} f(\mathbf{x})d\mathbf{x} = E[I_f(\mathbf{x})]$$
$$\text{Indicator function} = \begin{cases} 1 & \text{if } \mathbf{x} \text{ belongs to } F \\ 0 & \text{otherwise} \end{cases}$$

Unconditional Probabilistic Approach



Simulation Methods: Monte Carlo Simulation (MCS)

$$p_f = \int_F f(\mathbf{x}) d\mathbf{x} = \int I_f(\mathbf{x}) f(\mathbf{x}) d\mathbf{x} = E[I_f(\mathbf{x})]$$

Number of failed simulations

Monte Carlo Simulation: $p_f = E[I_f(\mathbf{x})] \cong \frac{1}{N} \sum_{i=1}^N I_f(\mathbf{x}^{<i>}) = \frac{N_f}{N} = \hat{p}_f$

Number of total simulations

- Obtain samples of $\mathbf{x}^{<i>}$ from the distribution $f(\mathbf{x})$
- Evaluate the performance of the structure for each $\mathbf{x}^{<i>}$
- Determine N_f and \hat{p}_f

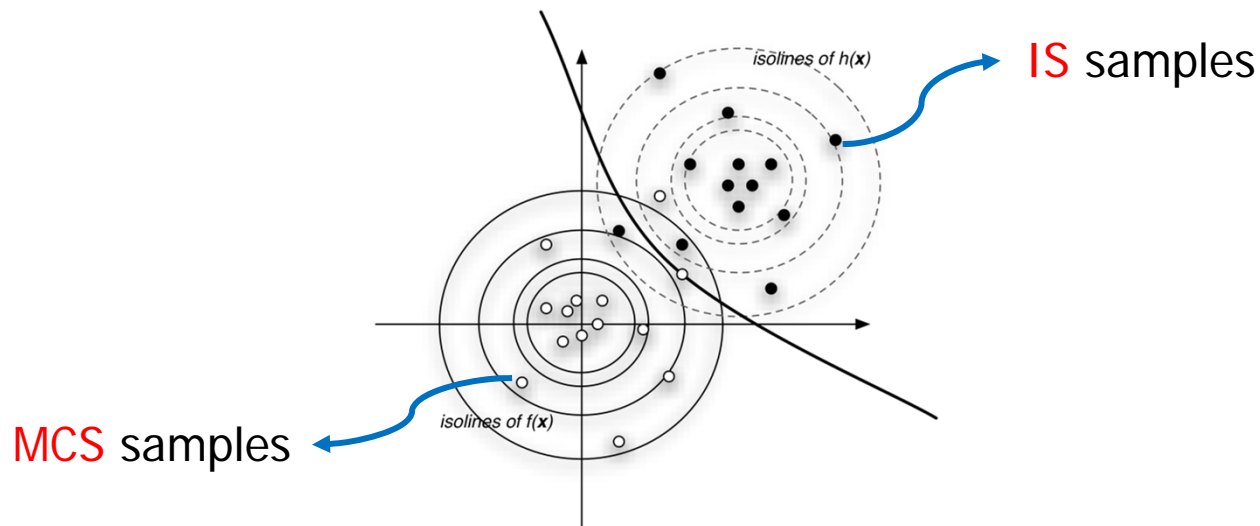
- \hat{p}_f is an **unbiased estimator** of p_f
- Variance of \hat{p}_f around p_f is **proportional to p_f** itself and **decreases with increasing N**

Unconditional Probabilistic Approach



Simulation Methods: Importance Sampling (IS)

- For **very small** values of p_{fi} , N may need to be **substantially large** to obtain a **few outcomes** for N_f
 - A possible solution to **avoid excessive number** of simulations →
- Importance sampling (IS):** Sample according to a more favorable distribution



Unconditional Probabilistic Approach



Simulation Methods: IS w/ K-means Clustering (IS-K) (Jayaram & Baker, 2010)

- For both **MCS** & **IS** methods, some of the samples could be redundant
- IS-K method **identifies** & **combines redundant** samples →
- **Reduces** the number of **simulations** further

In its simplest version, IS-K consists of **five** main steps:

Step 1: Pick (randomly) K samples

Step 2: Calculate the cluster centroids (typically mean of the K samples)

Step 3: Assign each sample to the cluster with the closest centroid

Step 4: Recalculate the centroid of each cluster after the assignments

Step 5: Repeat steps 1 to 3 until no more reassignments (**in step 4**) take place

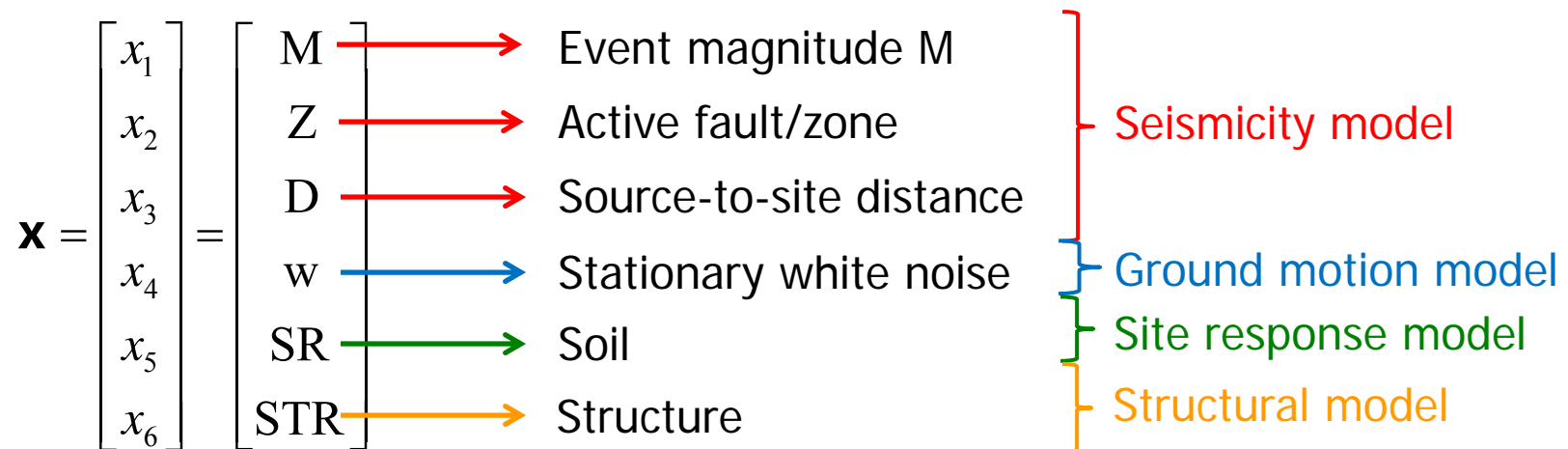
Once all the events are clustered, a single random sample from each cluster is used to represent all samples in that cluster

Unconditional Probabilistic Approach



Methodology for Seismic Assessment

Vector of random variables \mathbf{x} :



Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Seismicity Model

Seismicity model parameters M, Z and D sampled using simulation

Sampling for M

Monte Carlo Simulation

$$f(m) = \frac{\sum_{i=1}^{n_f} \lambda_i f_i(m)}{\sum_{i=1}^{n_f} \lambda_i}$$

$f_i(m)$: probability distribution of M for the i^{th} fault/source

λ_i : activation frequency for the i^{th} fault/source

(mean annual rate of all events on the source, i.e. events with $M >$ Lower bound M for that source)

n_f : # active faults/sources

Importance Sampling

$$h(m) = \frac{1}{n_m} \frac{f(m)}{\int_{m_k}^{m_{k+1}} f(m) dm}$$

$h(m)$: Sampling density for m lying in the k^{th} partition

n_m : # magnitude intervals (partitions) from m_{\min} to m_{\max}

Importance Sampling and K-means clustering

K-means clustering groups a set of observations into K clusters such that the dissimilarity between the observations within a cluster is minimized

Step 1: Pick (randomly) K samples

Step 2: Calculate the cluster centroids (typically mean of the K samples)

Step 3: Assign each sample to the cluster with the closest centroid

Step 4: Recalculate the centroid of each cluster after the assignments

Step 5: Repeat steps 1 to 3 until no more reassignments take place

Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Seismicity Model

Seismicity model parameters M, Z and D sampled using simulation

Sampling for Z

Given that an earthquake with magnitude $M = m$ has occurred, the probability that the event was generated in the i^{th} source is:

$$p(i|M = m) = \frac{\lambda_i f_i(m)}{\sum_{j=1}^{n_f} \lambda_j f_j(m)}$$

$f_i(m)$: probability distribution of M for the i^{th} fault/source

λ_i : activation frequency for the i^{th} fault/source

n_f : number of active faults/sources

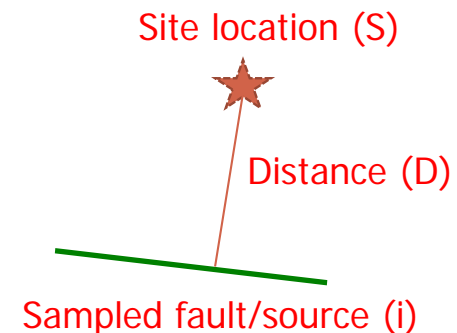
Active zone (Z) is sampled from its discrete probability distribution conditioned on M

Sampling for D

There is no further effort needed to sample D.

It can be determined based on:

- The sampled fault/source
- The deterministic site location (S)



Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Ground Motion Model

Synthetic Ground Motion Models:

- Seismologically-based Models
- Empirical Models

Seismologically-based Models

- Models that are based on the **physical processes** of earthquake generation and propagation
- Such models have reached a **stage of maturity**
- Applied in regions of the world where **data is not sufficient** for a statistical approach to seismic hazard
- Applied also in some regions of the world where **seismic activity is well-known** to (1) check their validity & (2) supplement existing information

Unconditional Probabilistic Approach



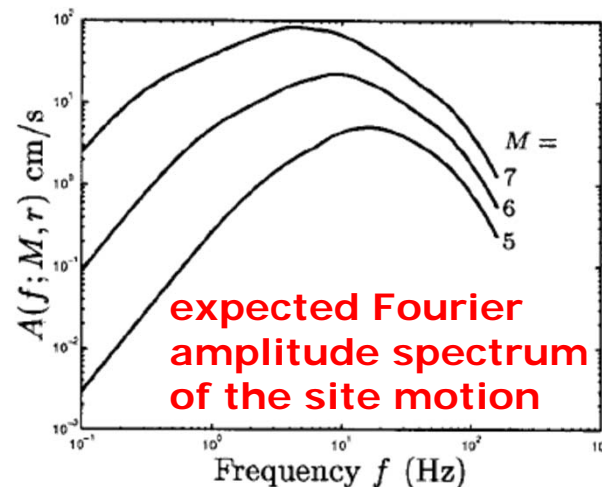
Methodology for Seismic Assessment: Ground Motion Model

Seismologically-based Models (Atkinson & Silva, 2000)

- Acceleration-amplitude Fourier spectrum (or Radiation spectrum)
- Generation of time history

Acceleration-amplitude Fourier spectrum (Au & Beck, 2003, Pinto et al, 2004)

$$A(\mathbf{f}, \mathbf{M}, \mathbf{R}) = A_0(f) \frac{1}{R'} \exp(-\gamma(f)R') \cdot \exp(-\pi f \kappa) V(f) \quad \& \quad A_0(f) = CM_0 (2\pi f)^2 \left[\frac{1-\varepsilon}{1+(f/f_a)^2} + \frac{\varepsilon}{1+(f/f_b)^2} \right] \quad \text{Source spectrum}$$



Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Ground Motion Model

Acceleration-amplitude Fourier spectrum (Au & Beck, 2003, Pinto et al, 2004)

$$A(\mathbf{f}, \mathbf{M}, \mathbf{R}) = A_0(\mathbf{f}) \frac{1}{R'} \exp(-\gamma(\mathbf{f})R') \cdot \exp(-\pi f \kappa) V(\mathbf{f})$$

near-surface attenuation: $\kappa = 0.03$
Anelastic attenuation

$$A_0(\mathbf{f}) = CM_0 (2\pi f)^2 \left[\frac{1-\varepsilon}{1+(f/f_a)^2} + \frac{\varepsilon}{1+(f/f_b)^2} \right]$$

Source spectrum

Geometric spreading factor for direct waves

$$f_a = 10^{2.18-0.496M}, \quad f_b = 10^{2.41-0.408M} \quad \text{Corner frequencies}$$

$$M_0 = 10^{1.5(M+10.7)} \quad \text{Seismic moment}$$

$$C = C_R C_P C_{FS} / (4\pi\rho\beta^3)$$

$$C_R = 0.55 \quad \text{Average radiation pattern for shear waves}$$

$$C_P = 2^{-0.5} \quad \text{Accounts for partition of waves in two horizontal components}$$

$$C_{FS} = 2 \quad \text{Free-surface amplification}$$

$$\rho \ \& \ \beta \quad \text{Density \& shear-wave velocity in the vicinity of the source}$$

$$\varepsilon = 10^{0.605-0.255M} \quad \text{Corner frequencies weighted through this parameter}$$

$$R' = \sqrt{h^2 + R^2} \quad \text{Radial distance between source and site}$$

$$R \quad \text{Epicentral distance}$$

$$h = 10^{-0.05+0.15M} \quad \text{Nominal depth of fault [km] ranging from } \sim 5 \text{ km for } M=5 \text{ to } 14 \text{ km for } M=8$$

$$\gamma(f) = \pi f / (Q\beta), \quad Q = 180f^{0.45} \quad \text{Regional quality factor}$$

$$V(f) \quad \text{Describes the amplification through the crustal velocity gradient (wave passage)}$$

Unconditional Probabilistic Approach

Methodology for Seismic Assessment: Ground Motion Model

Seismologically-based Models

- Acceleration-amplitude Fourier spectrum (or Radiation spectrum)
- Generation of time history

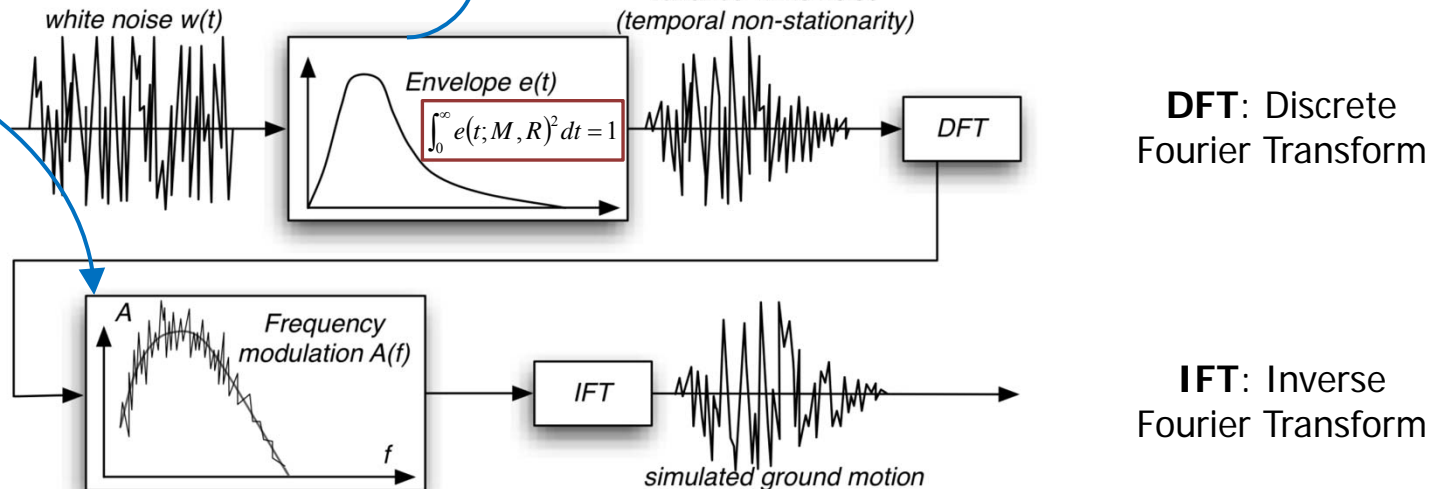
Generation of time history

Dependence on M & R introduced through α_3
 α_1 : Normalizing factor \rightarrow envelope has **unit energy**
 $U(t)$: Unit-step function

$$e(t; M, R) = \alpha_1 t^{\alpha_2 - 1} \exp(-\alpha_3 t) U(t)$$

amplitude-modulated unit-variance white noise (temporal non-stationarity)

multiply



Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Ground Motion Model

Synthetic Ground Motion Models:

- Seismologically-based Models
- Empirical Models

Empirical Models

- Models consist of parameterized **stochastic** (random) **process** models
- Developed by observing that **ground motions** possess **stable statistical nature** given **earthquake** and **site characteristics** (**M**, **R**, & **soil type**)
- This observation led to the idea of considering the ground motion **acceleration time-series** as samples of **random processes**

Unconditional Probabilistic Approach

Methodology for Seismic Assessment: Ground Motion Model

Synthetic Ground Motion Models:

- Seismologically-based Models
- Empirical Models

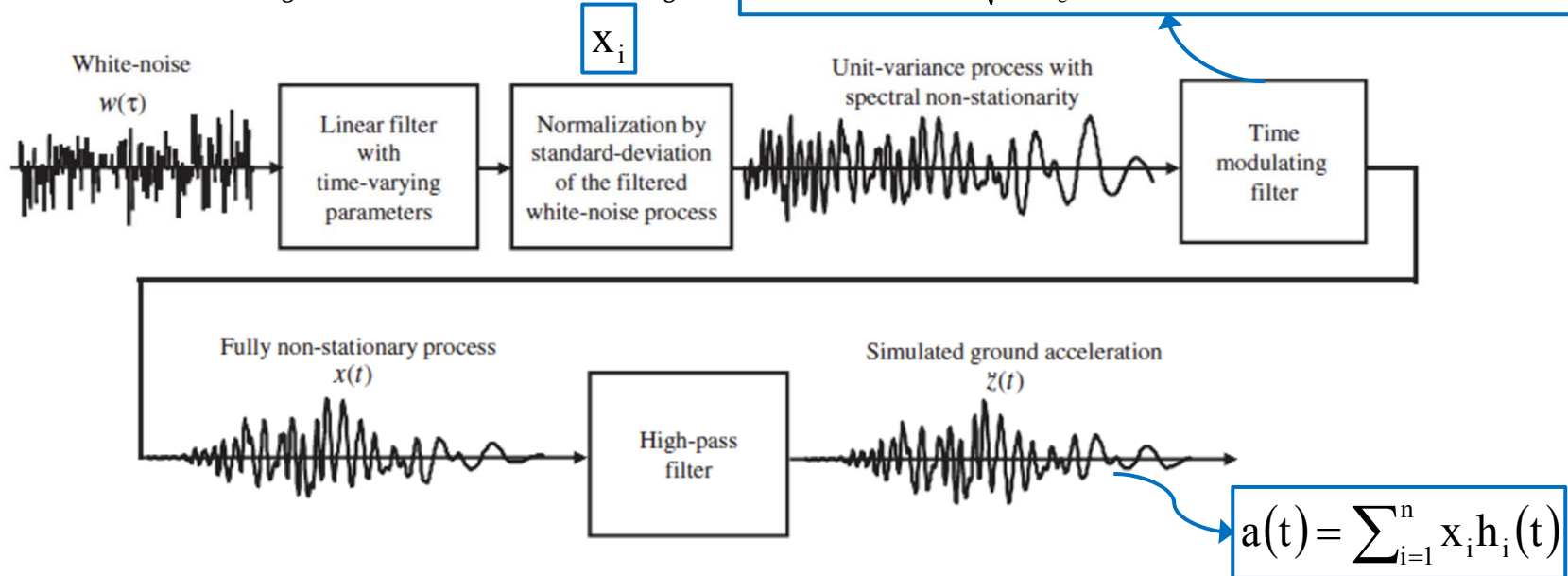
$$\omega_g = \omega_g(t)$$

$$\xi_g = \xi_g(t)$$

Empirical Models (Rezaeian & Der Kiureghian, 2010)

The filter **IRF** (Impulse Response Function) is the acceleration IRF of a linear SDOF oscillator of natural frequency ω_g and damping ratio ξ_g

$$h_i(t) = h(t - \tau_i) = \frac{\omega_g}{\sqrt{1 - \xi_g^2}} \exp[-\xi_g \omega_g (t - \tau_i)] \sin[\omega_g \sqrt{1 - \xi_g^2} (t - \tau_i)]$$



Unconditional Probabilistic Approach



Methodology for Seismic Assessment: Site-Response Model

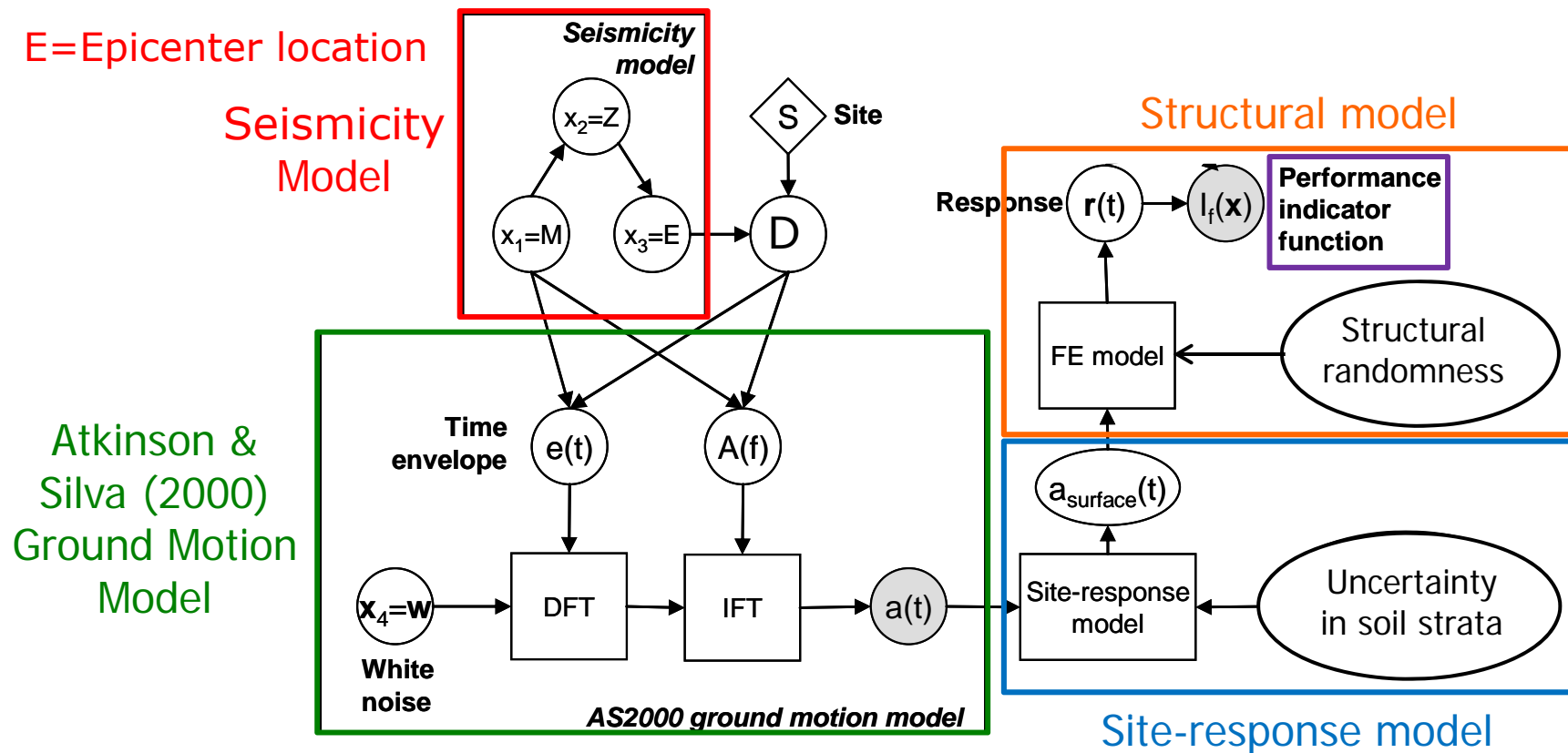
- Ground motion model determines ground motion time history for **bedrock**
- A site response model is used to obtain input motion to the structure **at the surface**
- Model the **soil strata and corresponding stiffness, strength, & damping properties**, e.g. a one-dimensional nonlinear, or equivalent linear, model)
- The strata thicknesses and properties possess **uncertainty**

Methodology for Seismic Assessment: Structural Model

- Finite element model which determines the **response of the structure**
- Both the structure itself, and the response-model implemented in the analysis software, are affected by **uncertainty** (**more in the afternoon**)

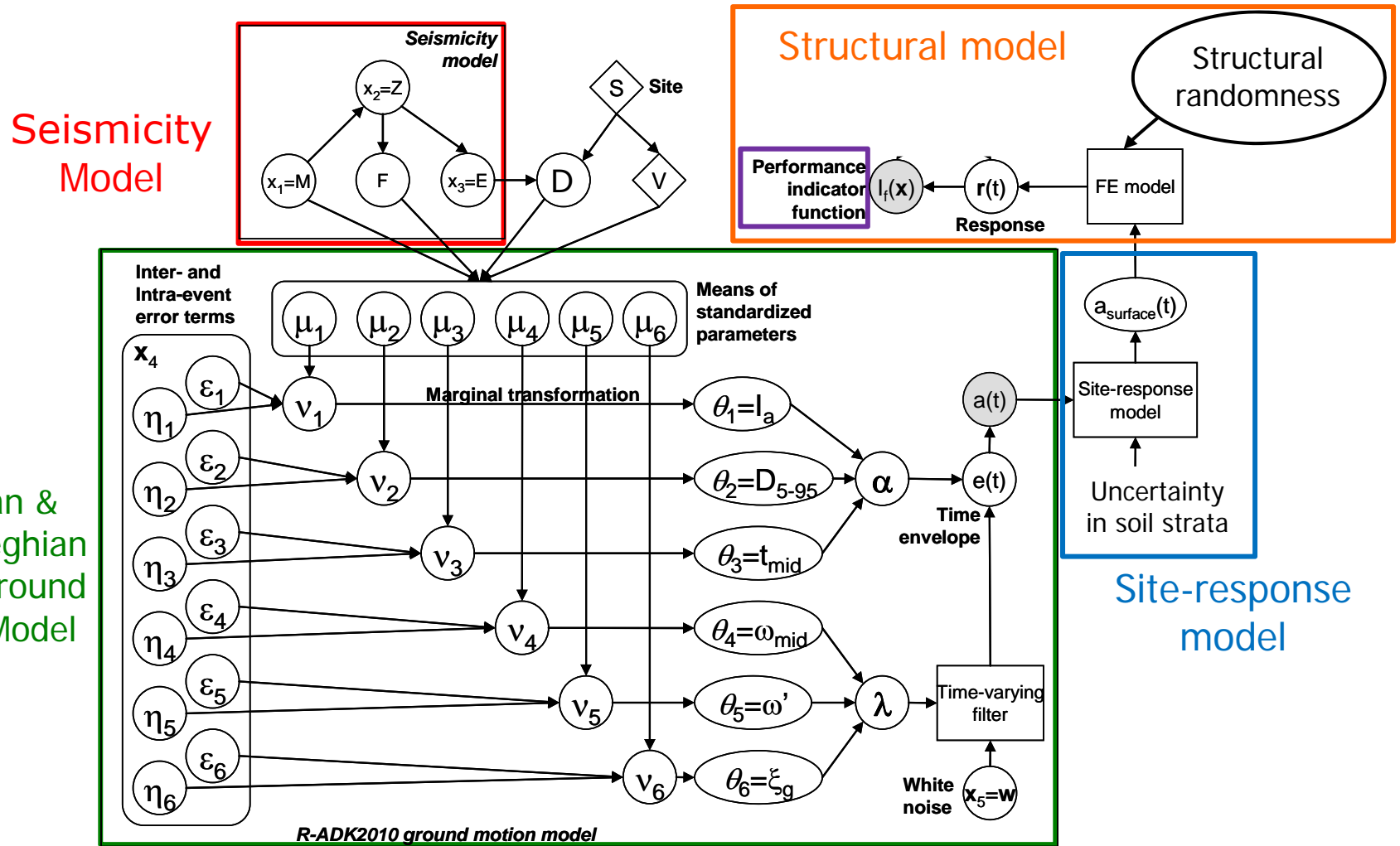
Unconditional Probabilistic Approach

Methodology for Seismic Assessment: Flowchart



Unconditional Probabilistic Approach

Methodology for Seismic Assessment: Flowchart

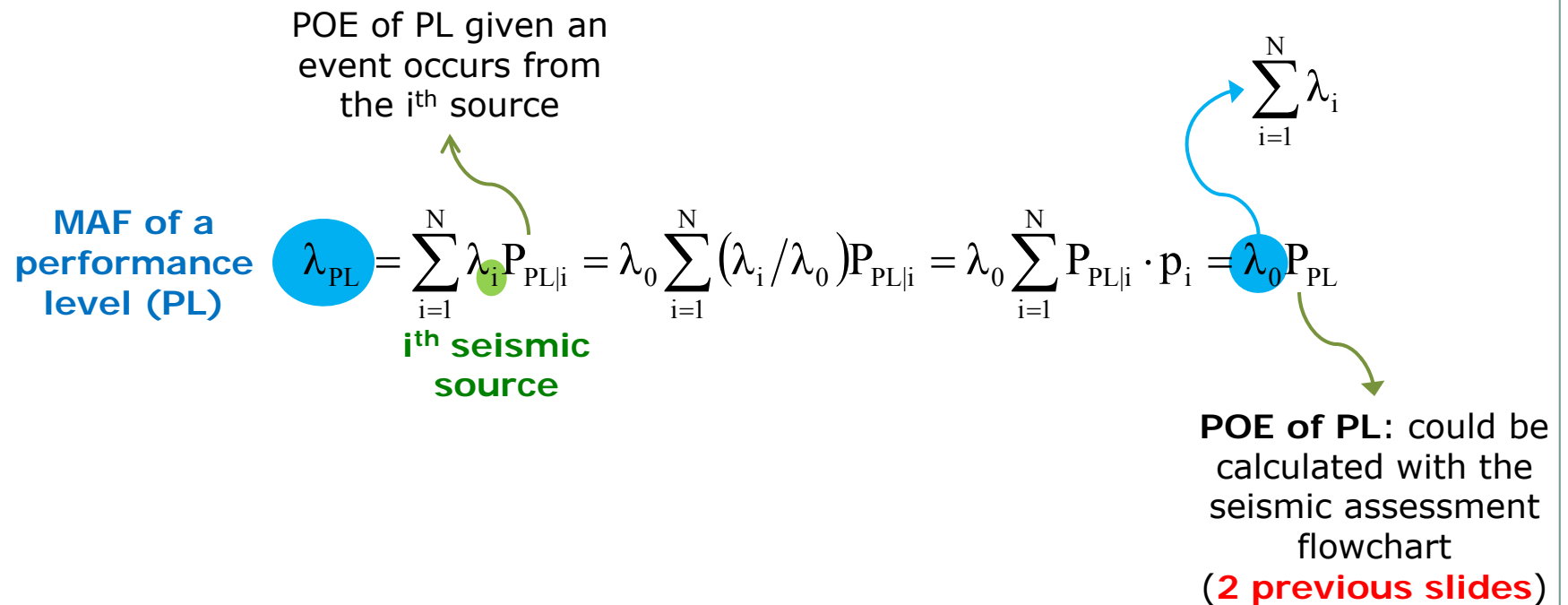


Unconditional Probabilistic Approach



Methodology for Seismic Assessment:

Application to the Estimation of a Structural Mean Annual Frequency



λ_i : activation frequency for the i^{th} fault/source

(mean annual rate of all events on the source, i.e. events with $M >$ Lower bound M for that source)



Thank you