

PEER International Pacific Rim Forum June 16-17, 2021

Incorporation of Uncertainties in Seismic Hazard Characterization Using Numerical Simulations of Ground Motions

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> > June 17, 2021



3-D Simulations of GM in Seismic Hazard

- 3-D Simulations
 - Include path-specific effects
 - Include fault geometry-specific effects
- From a seismic hazard point of view, 3-D simulations are non-ergodic GMMs
 - Not just magnitude, distance, site condition

Implementation of Non-Ergodic GMMs in Seismic Hazard

- For the specific source/site combination:
 - Median ground motion
 - Aleatory variability
 - Epistemic uncertainty in the estimate of the median

Uncertainty Matrix

	Aleatory Variability	Epistemic Uncertainty
Parametric	Multiple realizations of the source Sample pdf for source inputs	For a given simulation method:
	 Rupture dimension Slip distribution Hypocenter location 	Alternative pdfs for the source inputs - mean and std dev
	 Rupture-velocity distribution Rake-angle distribution 	Alternative 3-D velocity models
Modeling	Limitation of the simulation method to match data - variability that can't be explained by	Does the simulation method give the correct median GM?
	the model is treated as aleatory variability	Alternative simulation methods
	Must be estimated empirically - Misfit between data and simulations	Range of median values for different simulation methods for a specific site - using Single method
	for optimized source parameters.	- Evaluate the bias (simulations versus data) and standard error of the bias.

Uncertainty Matrix

	Aleatory Variability	Epistemic Uncertainty
Parametric	σ_{par}	σ_{μ_par}
Modeling	σ_{mod}	σ_{μ_mod}
Total	$\sigma = \sqrt{\sigma_{mod}^2 + \sigma_{par}^2}$	$\sigma_{\mu} = \sqrt{\sigma_{\mu_mod}^2 + \sigma_{\mu_par}^2}$

All Four Elements of uncertainty matrix Required for Hazard Implementation

- Single Realizations
 - Provide examples of the spatial variability, but are not useful for seismic hazard
- Current approach used in CyberShake and LBNL
 - single 3-D velocity model
 - multiple realizations of the source (σ_{par})
 - missing parts
 - modeling aleatory term (σ_{mod})
 - epistemic uncertainty of the 3-D velocity model (σ_{μ_par})
 - epistemic uncertainty of the distribution for source parameters (σ_{μ_par})
 - epistemic uncertainty from different methods (σ_{μ_mod})
- For hazard applications:
 - Missing values will be assumed
 - If you don't want the hazard analysts to pick values, 3-D simulations need to provide the estimates

Estimating Aleatory Elements of Uncertainty Matrix

- Parametric Aleatory term
 - Need to sample the full range of source properties
 - Straightforward, but requires a large number of 3-D simulations
 - Typically need about 50 realizations to get a good estimate of the standard deviation
- Modeling Aleatory term (Missing)
 - Need to compare with observations to quantify limitation of the method
 - Issues
 - May not have good 3-D velocity model for regions with data from large magnitude earthquakes
 - The geotechnical layer is often missing from the 3-D velocity. model
 - Does the misfit represent the limitation of the method or the limitation of the 3-D velocity model?
 - Validation (SCEC BBP) provide estimates of the model misfit from 1-D simulations
 - Can we use modeling aleatory from 1-D model misfits to approximate 3-D modeling aleatory?

Estimating Epistemic Elements of Uncertainty Matrix

- Parametric Epistemic term (missing)
 - Uncertainty in the 3-D velocity model
 - Need method to develop alternative 3-D velocity models
 - Large increase in the number of 3-D simulations needed
 - Uncertainty in the source model inputs
 - Different distributions of the source parameters (e.g. mean and std dev)
 - Can change weights to realizations used for the parametric aleatory term
 - May not need an increase in 3-D simulations
- Modeling Epistemic term (missing)
 - Multiple methods
 - Range of median from alternative simulation methods
 - Different rupture generation methods
 - Different methods for 3-D simulations
 - Increase number of simulations
 - Single method
 - Uncertainty in the bias from comparison with data
 - Part of the validation

Estimating Full Uncertainty Matrix

- Key issue:
 - Need a significant increase in the number of 3-D simulations
- Can we do this in a more efficient manner than Monte Carlo sampling?

Efficient Methods for Increasing Number of Realizations of 3-D simulations

- Objective:
 - Generate a large number of realizations from a small number of available 3-D calculations without using Monte Carlo
- Possible Method Probabilistic Learning on Manifolds (PLoM) by Soize & Ghanem (2016 - 2020)
 - Non-intrusive (no change to the 3-D simulation program)
 - Designed for expensive, large-scale simulations
 - Assumes a limited number of simulations available (50-100)
 - Learns solutions' statistics from small 'training' dataset and physical constraints
 - Efficiently generates many additional 'learned' realizations from learning phase
 - Reconstructs full statistics of the solution efficiently
 - Results are time series, not just response spectra

Planning a Trial Application

- Use 25 realizations for the Hayward events (training data)
- Test implementation of the PLoM method
 - Generate a suite of new realizations using PLoM
- Evaluate the predicted distribution of time series from PLoM with a second set of realizations from the 3-D simulations (Test data)

Summary

- Current sets of 3-D simulations can be used to quantify the parametric aleatory term from multiple realizations of the source
- Most are missing three other uncertainty terms for 3-D simulations
 - Aleatory modeling
 - Ideally, conduct validations 3-D simulations
 - Initially, use 1-D validation results as an estimate
 - Epistemic parametric
 - Requires simulations for alternative 3-D models
 - Can reweight the simulations for alternative pdfs of source parameters
 - Epistemic modeling
 - Different simulation methods set up a common problem for multiple methods to use
 - Single method requires results from validation with data for the uncertainty in mean bias