

# Characterizing seismic hazard to distributed systems using efficient simulation techniques

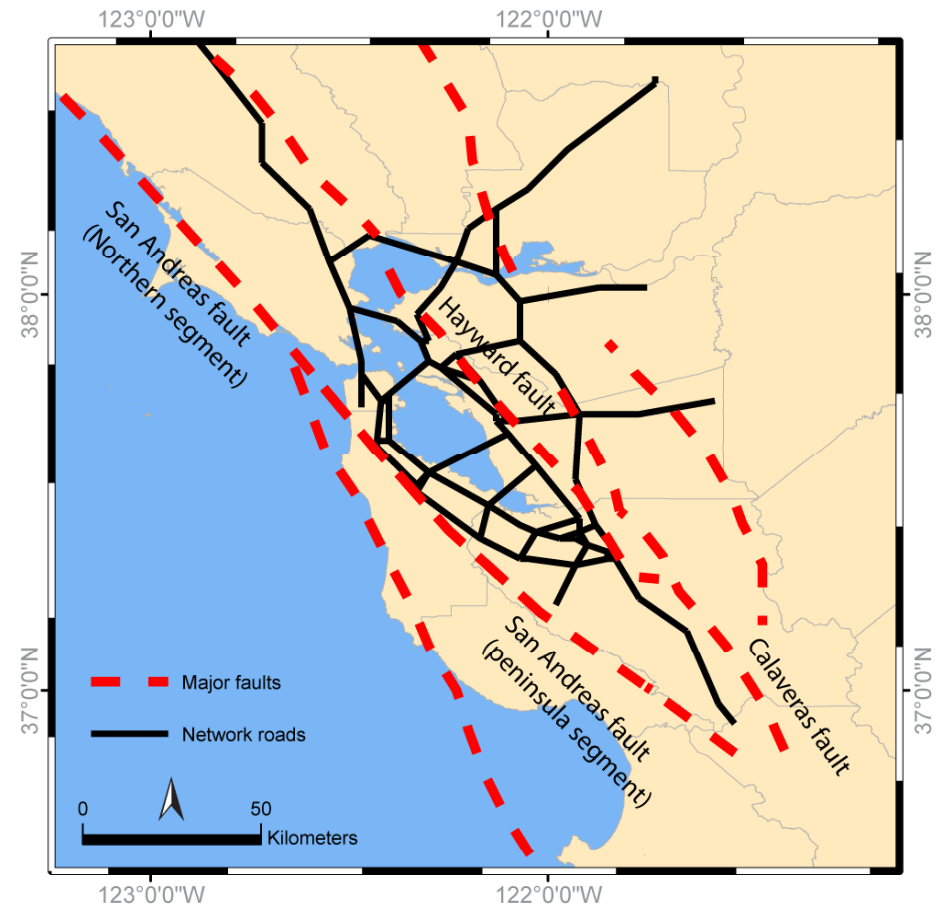
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## Motivation

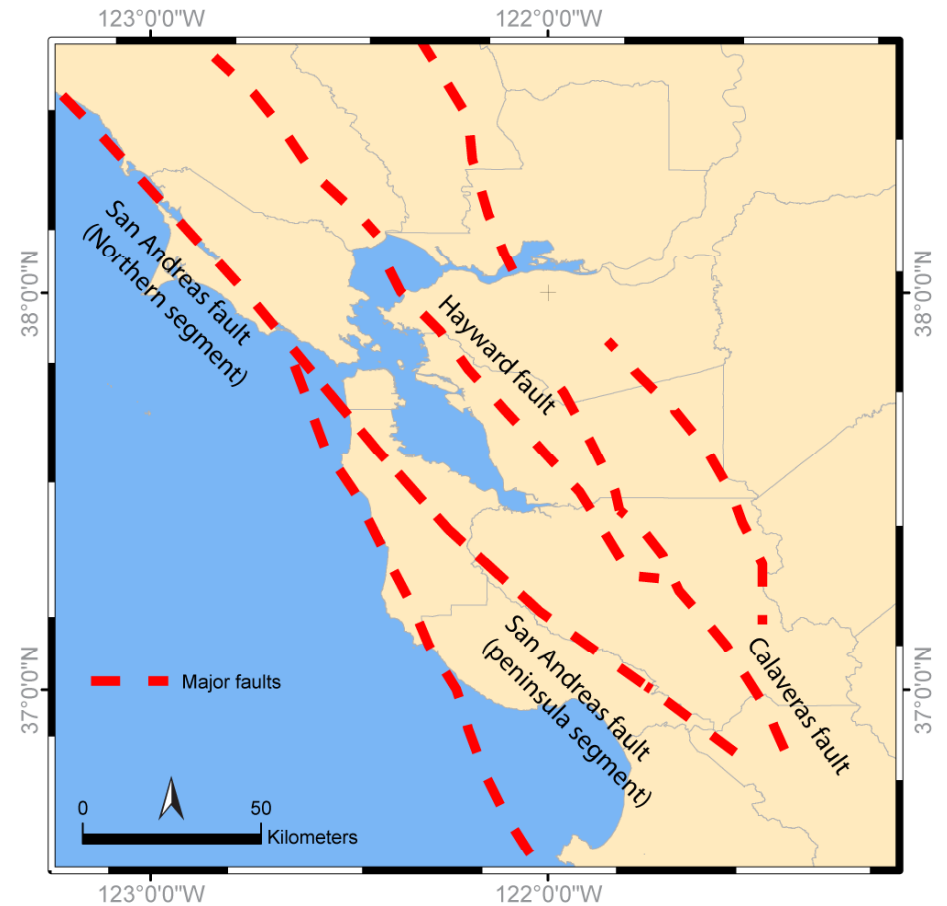
- We are interested in assessing seismic risk to distributed transportation networks
- The spatial extent of these systems is a challenge
- Spatial correlation of ground motion intensities is an important consideration for these analyses



Bay area interstate highways

## Background: seismic sources

Both this work and traditional Probabilistic Seismic Hazard Analysis require seismic source characterization

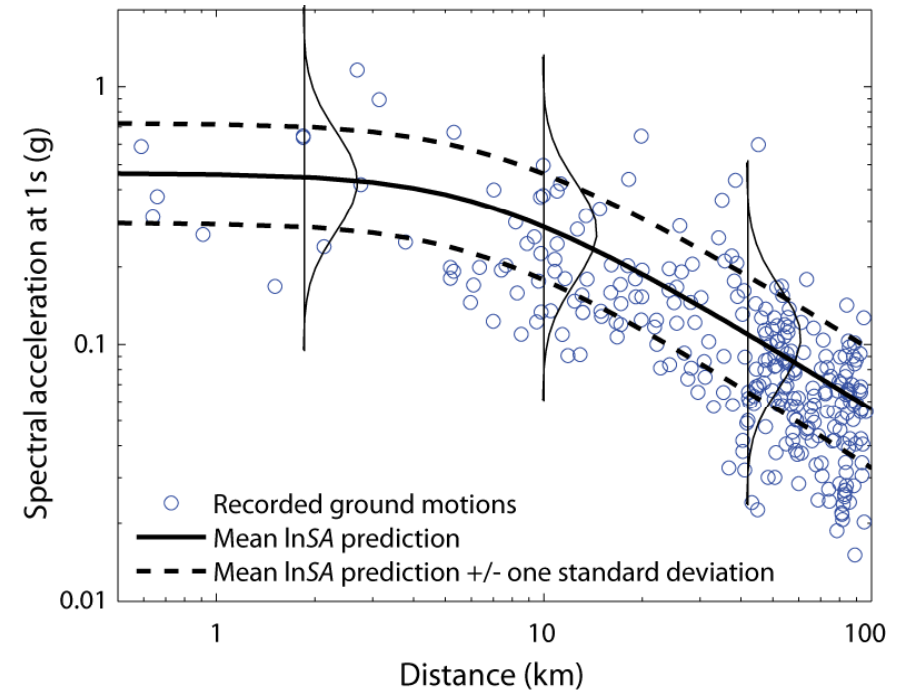


Major Bay area faults

## Background: ground motion prediction

Ground Motion Prediction (“attenuation”) models provide predictions of the distribution of ground motion intensity (e.g., spectral acceleration) as a function of earthquake magnitude, source-to-site distance, etc.

Observed spectral acceleration values from the 1999 Chi-Chi, Taiwan earthquake



Model form:

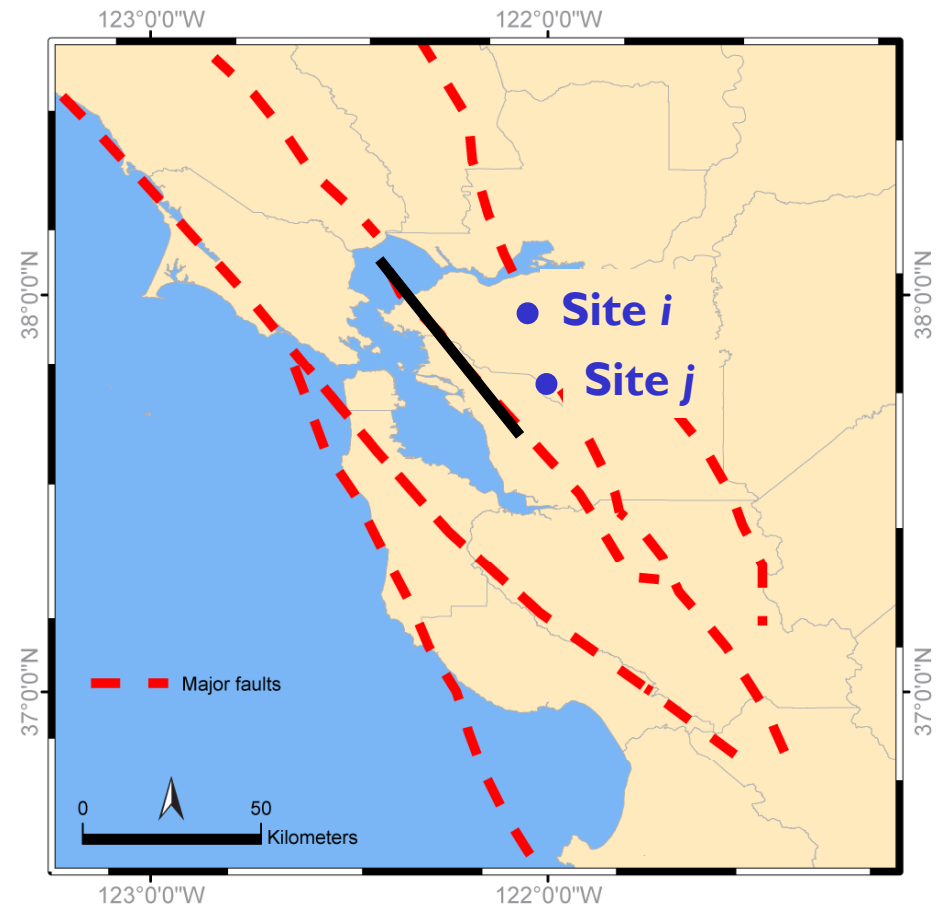
$$\ln Sa_i(T) = \overline{\ln Sa_i(M, R, T, \dots)} + \varepsilon_i + \eta$$

↑ Spectral acceleration at site “i”  
 ↑ Predicted mean (log) spectral acceleration  
 ↑ Intra-event variability at site “i”  
 ↑ Inter-event variability (at all sites)

## Ground motion intensity at two sites

$$\ln Sa_i(T) = \overline{\ln Sa_i(M, R_i, T, \dots)} + \varepsilon_i + \eta$$

$$\ln Sa_j(T) = \overline{\ln Sa_j(M, R_j, T, \dots)} + \varepsilon_j + \eta$$



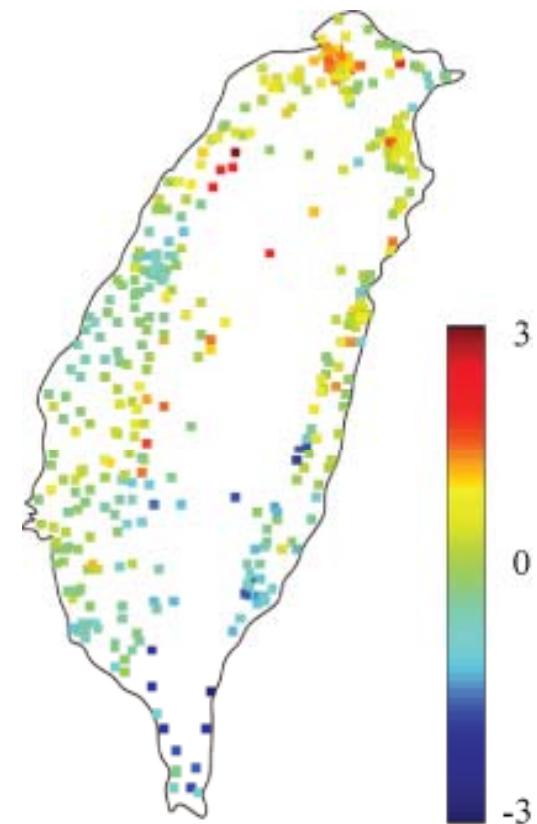
## Observed “residuals” from well-recorded earthquakes

Observations of past earthquakes shows that these residuals are correlated at nearby sites, due to

- Common source earthquake
- Similar location to asperities
- Similar wave propagation paths
- Similar local site effects

Note that this correlation is different than ground motion coherence

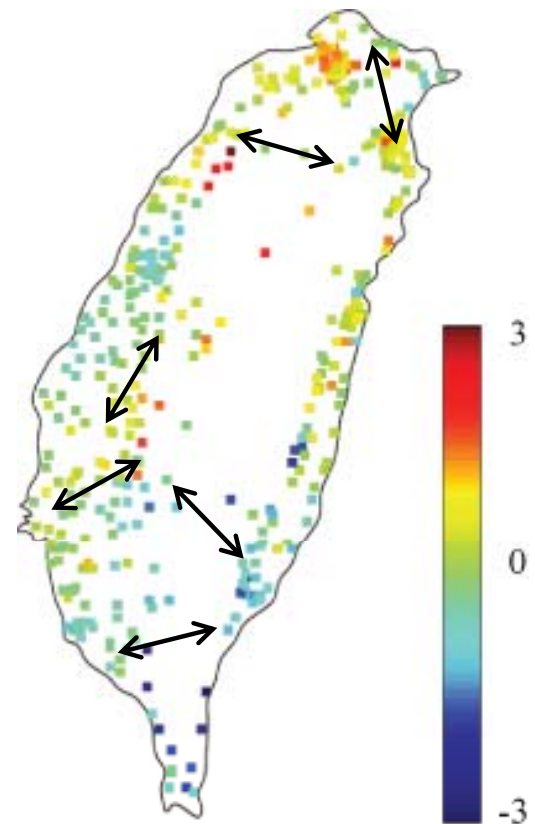
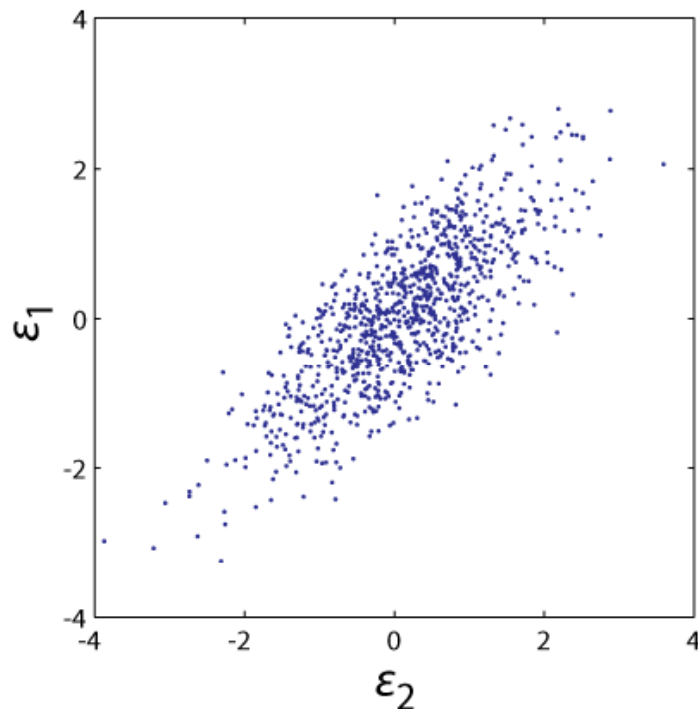
$$\varepsilon_i = \ln Sa_i(T) - \overline{\ln Sa_i(M, R_i, T, \dots)} - \eta$$



PGA  $\varepsilon$ 's from the 1999 Chi-Chi earthquake

## Estimation of correlation (or covariance)

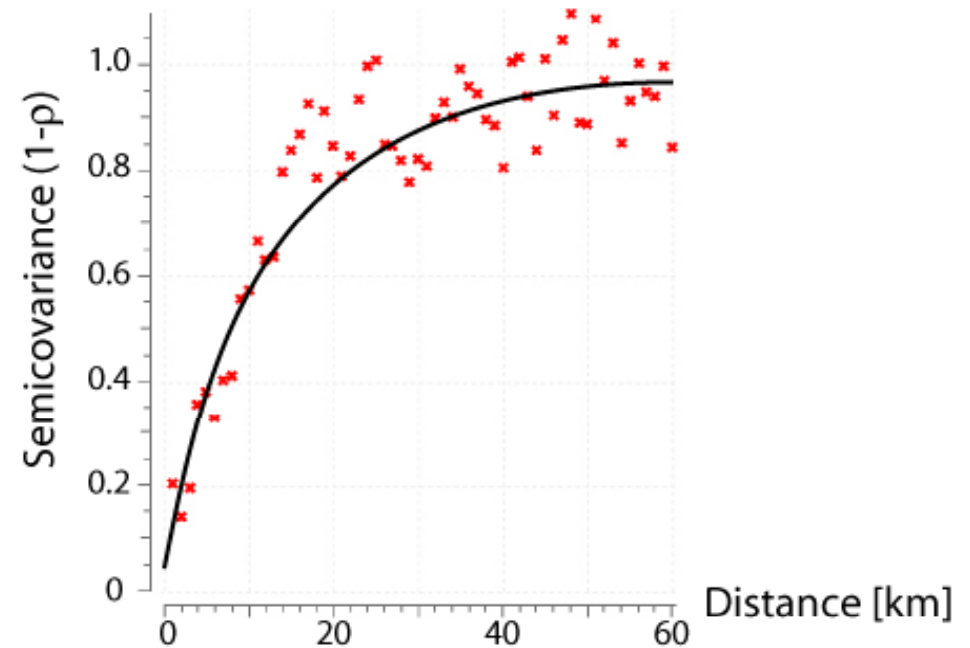
If we assume stationarity and anisotropy of the  $\varepsilon$ 's, we can pool paired observations with comparable distances to estimate a correlation coefficients



## Estimation of correlation from well-recorded earthquakes

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We can turn these observations into a predictive model



Empirical semivariogram for PGA  $\varepsilon$ 's  
from the 1999 Chi-Chi earthquake



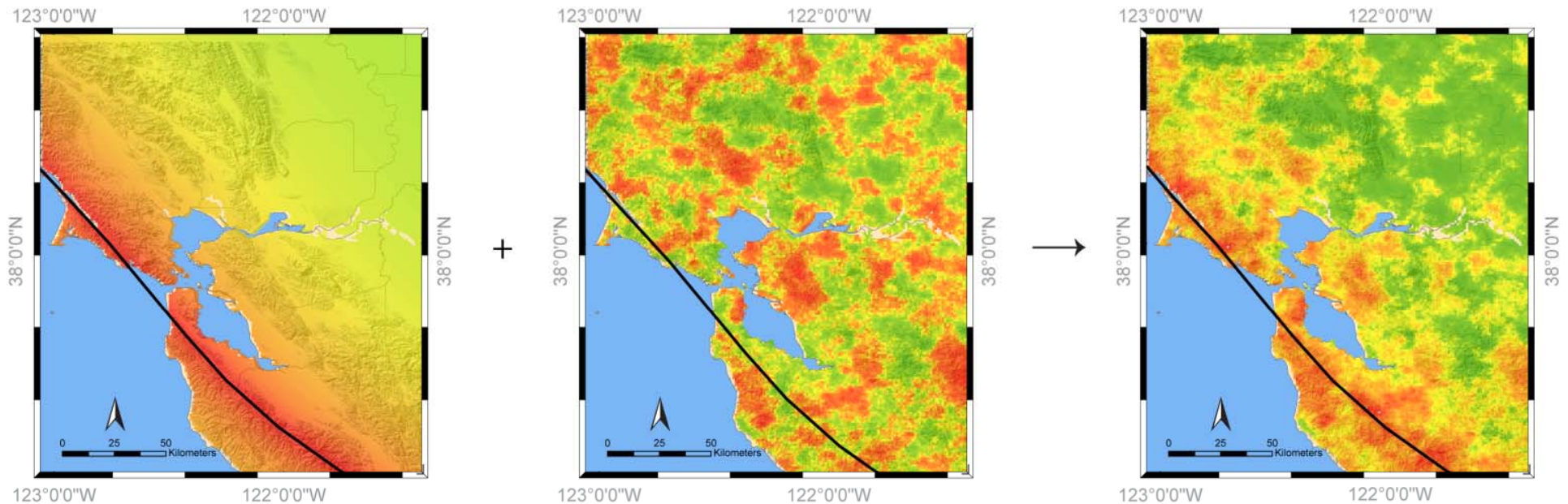
## Application: risk to lifeline systems

$$\ln Sa_i(T) = \overline{\ln Sa_i(M, R, T, \dots)} + \varepsilon_i + \eta$$

$$e^{\overline{\ln Sa_i(M, R, T, \dots)}}$$

$$\varepsilon_i$$

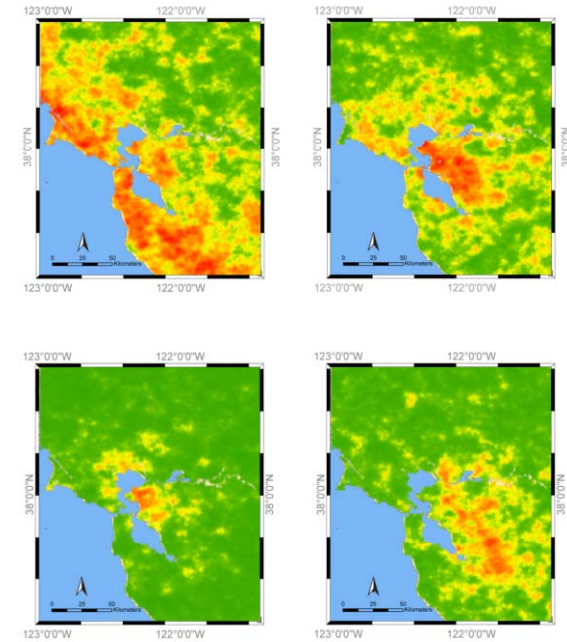
$$Sa_i(T)$$



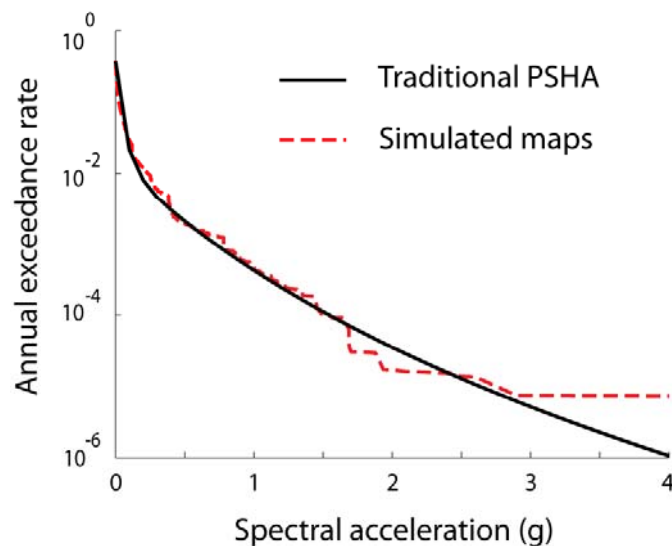
We can't produce PSHA-like maps of hazard, but we can use Monte Carlo simulations to produce a set of ground motion scenarios and associated probabilities

## An aside: comparison with single-site PSHA

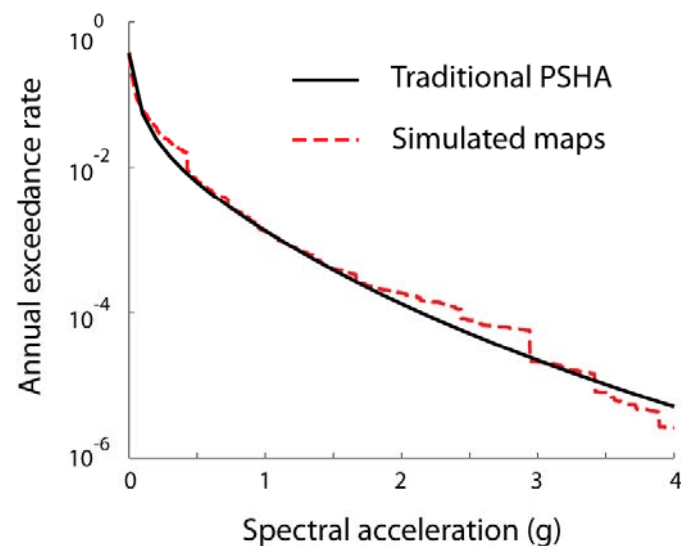
If we look at the observed  $S_a$  values from these simulations at any single site, they match the distribution from traditional single-site PSHA



Site A:

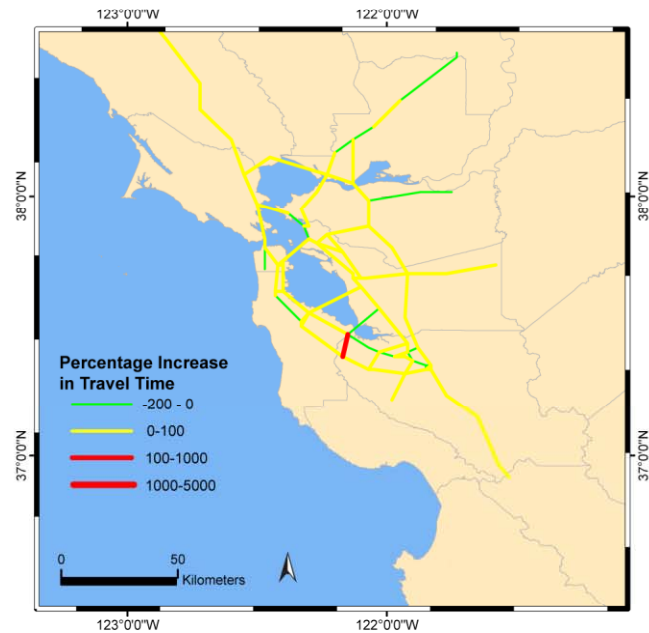
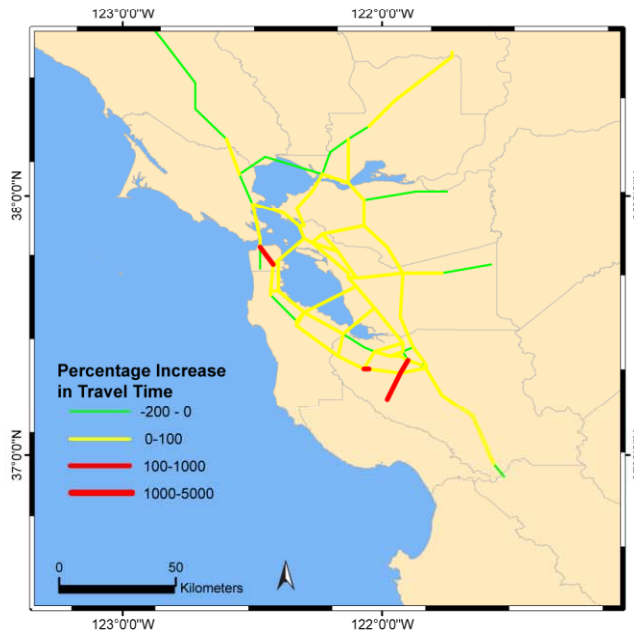
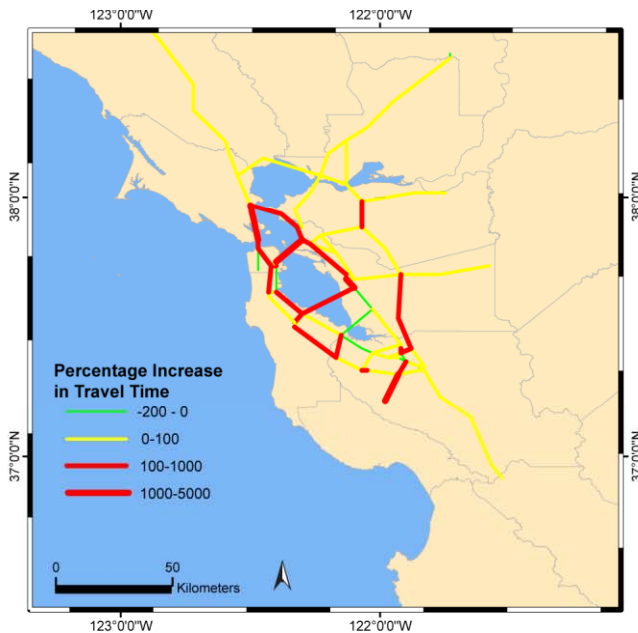
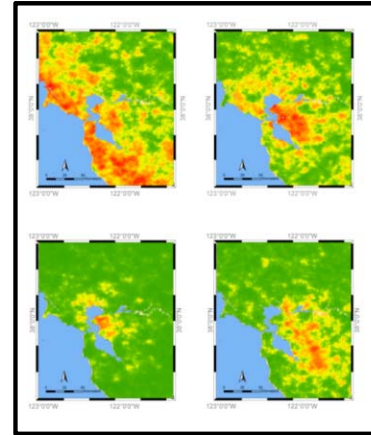


Site B:



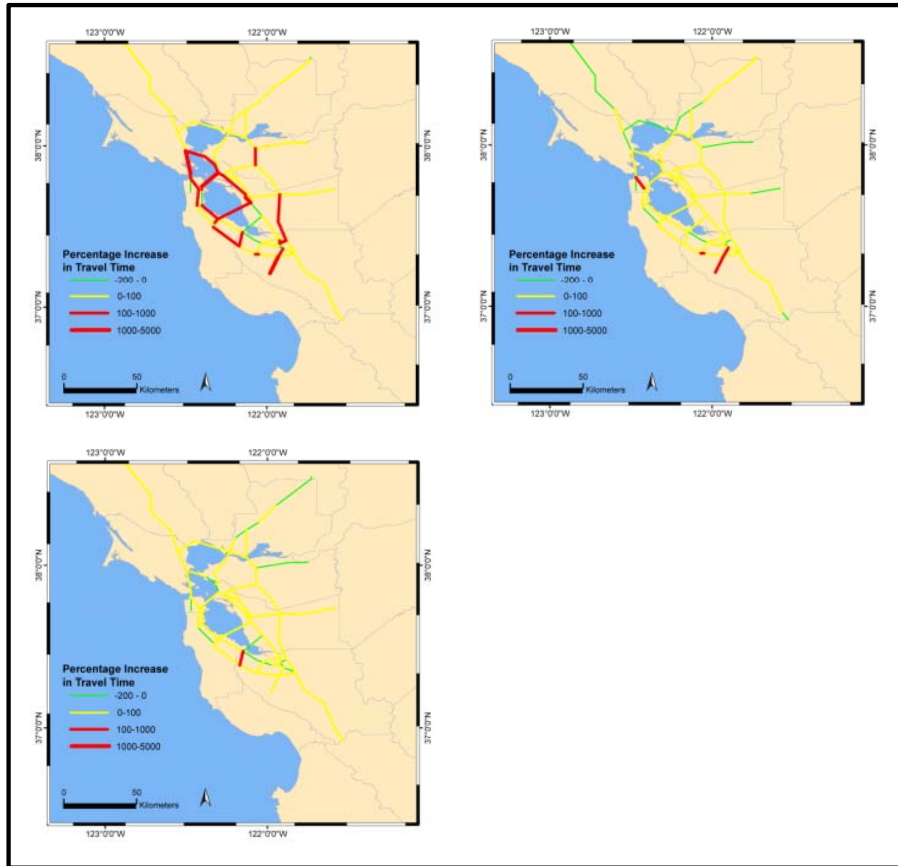
## Application: risk to lifeline systems

For each map of spectral accelerations, we can predict the resulting damage and disruption to our network

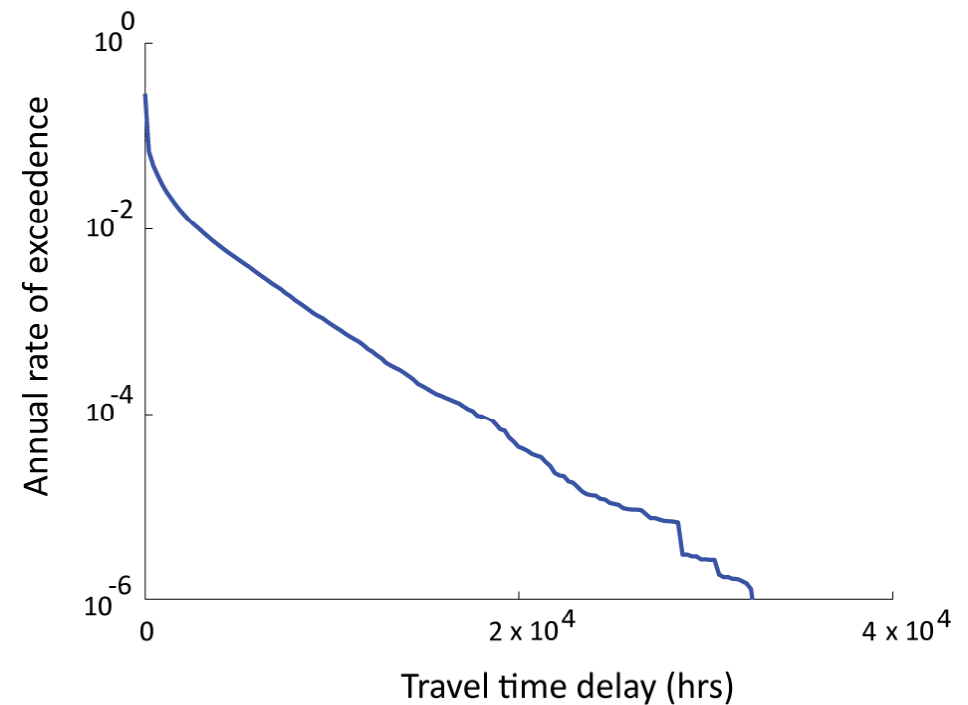


*Illustrative results using a simplified highway network model*

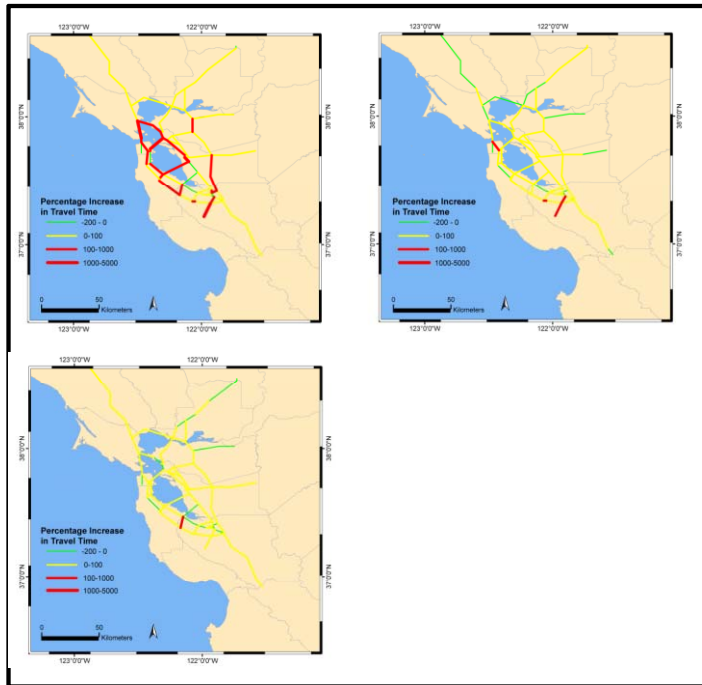
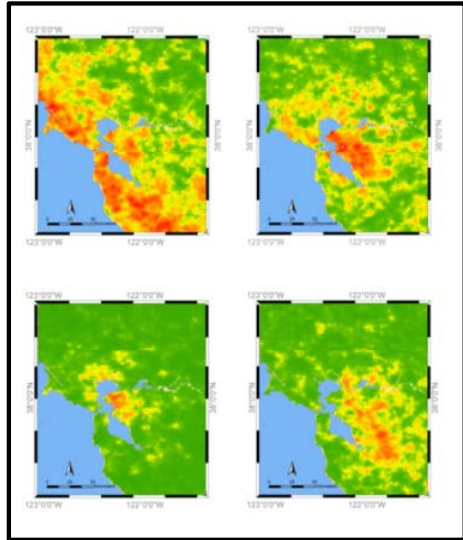
## Application: risk to lifeline systems



If we aggregate all these disruptions, we can compute a rate of disruption exceedance

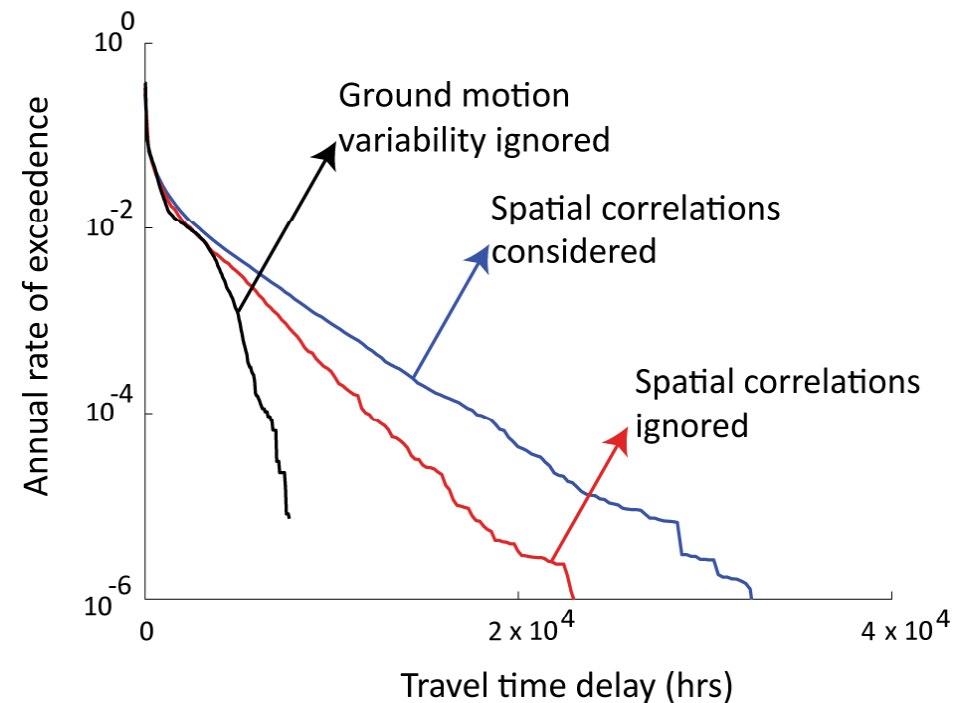


## Consideration of correlations has a significant impact here

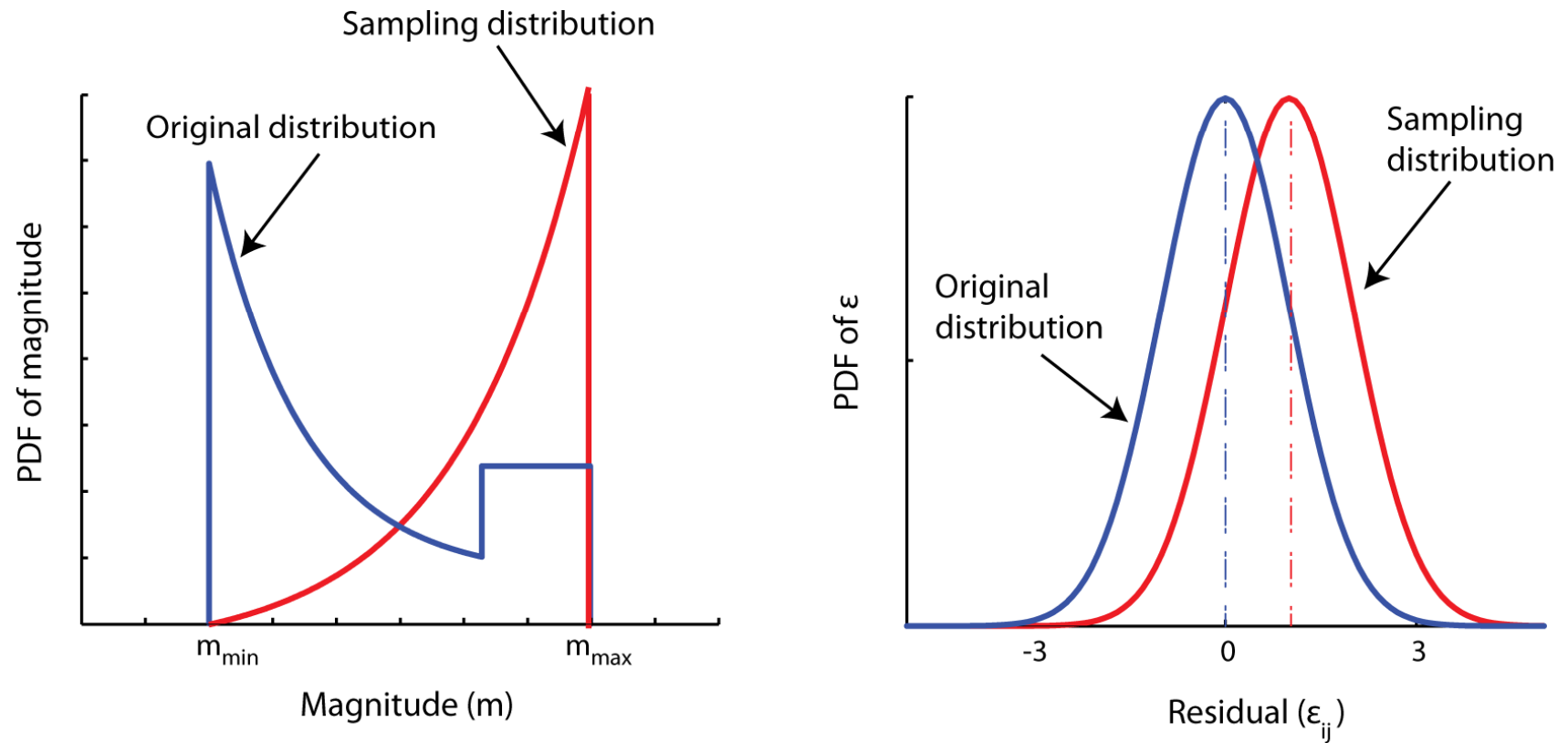


We can repeat this exercise omitting the  $\varepsilon$  correlation, to see the impact of this correlation

$$\ln Sa_i(T) = \overline{\ln Sa_i(M, R, T, \dots)} + \varepsilon_i + \eta$$



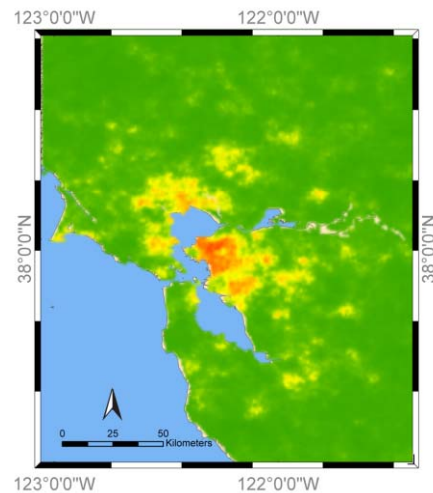
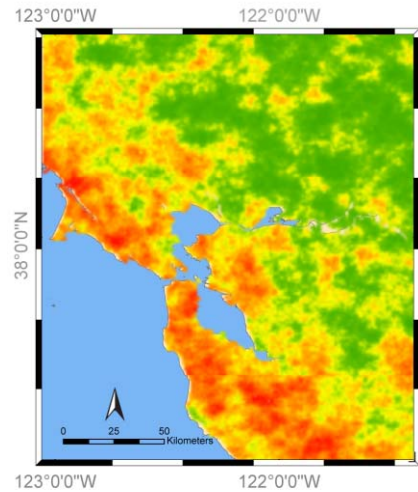
# Improving simulation efficiency: importance sampling



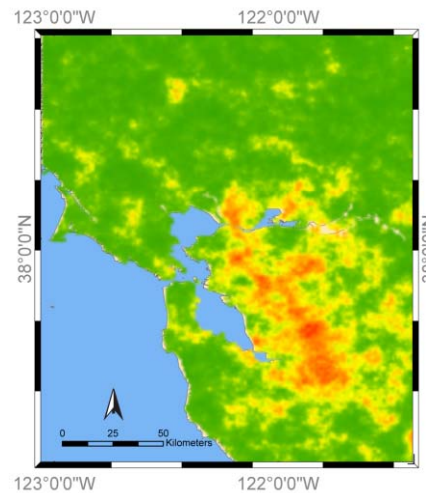
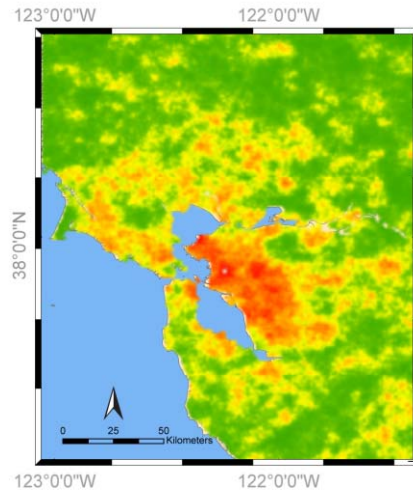
If done correctly, this reduces our computational expense by  $\sim 2$  orders of magnitude



## Improving simulation efficiency: K-means clustering



Group these  
into a cluster

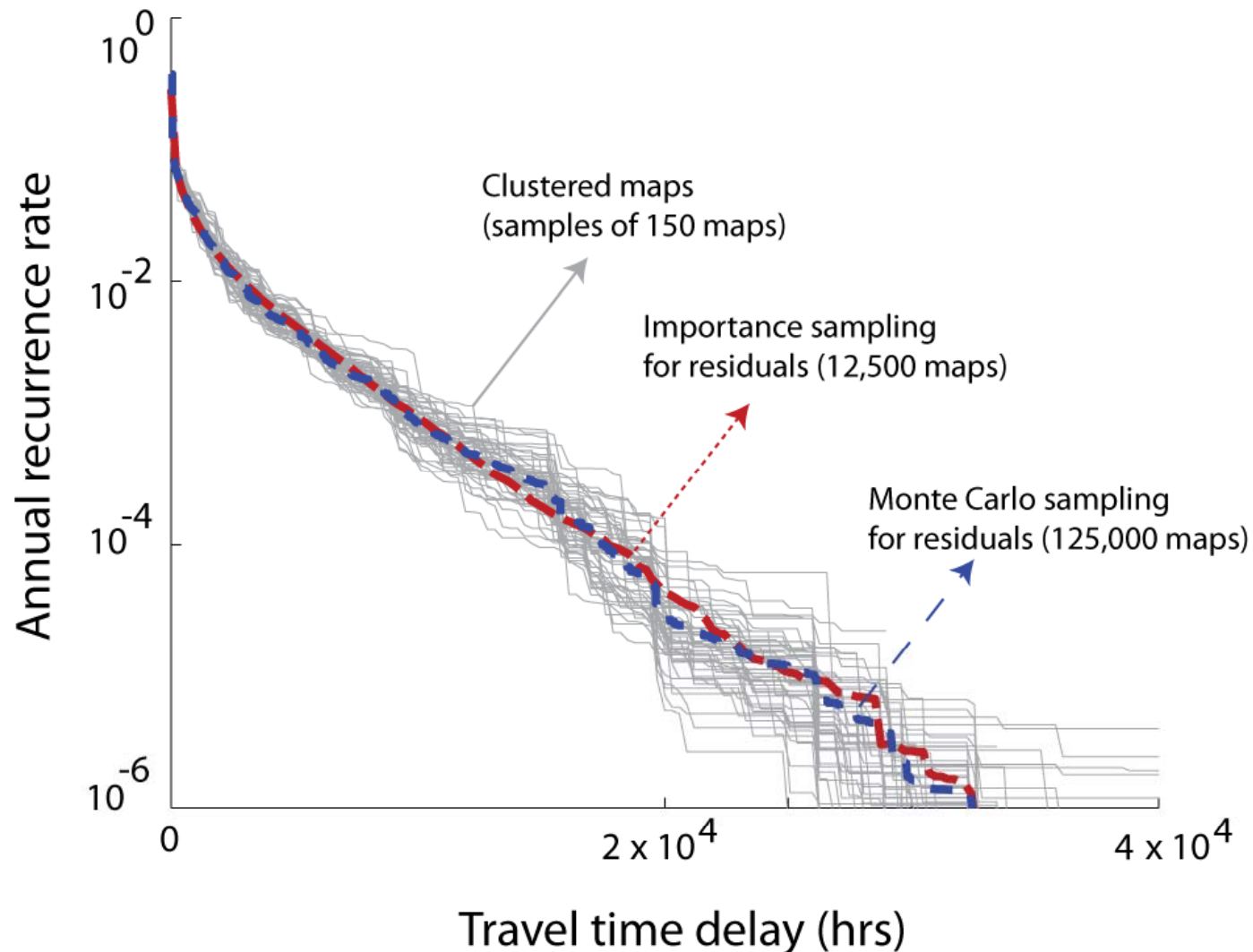


Select a *single* representative  
from the cluster to represent *all*  
simulations in the cluster

This can reduce computational  
expense by another ~2 orders  
of magnitude

## Loss estimation using more efficient techniques

These techniques reduce our computational expense without biasing our estimates of disruption





## Conclusions and current status

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- We have used well-recorded earthquakes to measure spatial correlation of spectral acceleration values
- Using this model, we can generalize traditional PSHA to characterize ground motion intensities at many sites
- This characterization can not be done analytically, but efficient sampling and clustering make simulation tractable
- We assessed a simplified transportation network to demonstrate that this approach is unbiased with respect to basic Monte Carlo simulation
- We now hope to use this approach to study more realistic models distributed infrastructure