

Cloud Analysis revisited again: What should we do with the "collapse" cases?

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The 42nd Risk, Hazard & Uncertainty Workshop, Hydra, 22-25 June 2016



The Road Map

- Setting the scene: Y variable, Risk Integral, Fragility
- Cloud method in one slide
- The Bayesian diversion
- Considering the collapse cases
- The logistic treatment
- Achieving the softening effect on the median and the percentiles
- The [inevitable] comparison with IDA ...

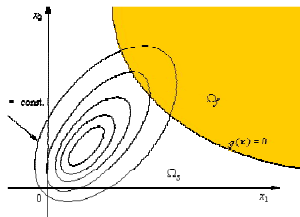


Performance-based Structural Risk Assessment: The main players

$$\lambda_{ls} = \int P(Y_{ls} > 1 | IM) \cdot |d\lambda(IM)|$$

Risk IM Fragility Hazard

- ✓ $\lambda(IM)$ is the mean annual rate of exceeding a given IM
- ✓ $P(Y_{ls} > 1 | IM)$ is the structural fragility
- ✓ λ_{ls} is the mean annual rate of exceeding a prescribed limit state



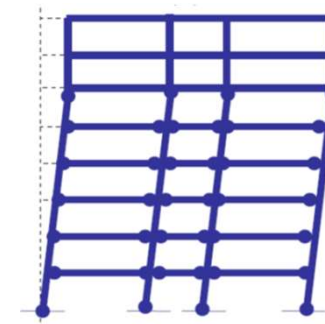
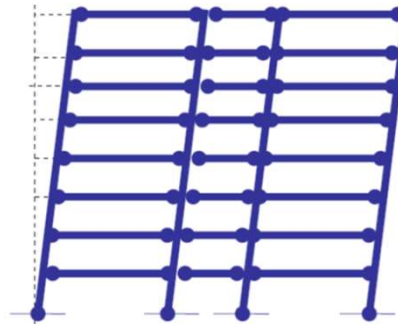
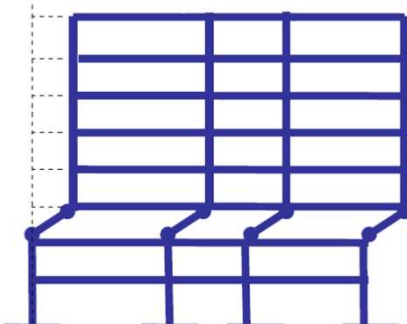
Setting the scene: The structural performance variable

- *global mechanisms*
- Ultimate rotation
- Shear capacity
- Joint safety checking

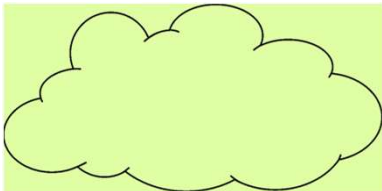
$$Y = \max_{i=1, N_{mech}} \left(\min_{j \in I_i} \frac{D_j}{C_j} \right)$$

"weaker" mechanism
"stronger" component

- N_{mech} number of mechanisms
- I_i indexes of components in the i -th mechanism



Jalayer, F., Franchin, P. and Pinto, P.E., 2007. A scalar damage measure for seismic reliability analysis of RC frames. *Earthquake Engineering & Structural Dynamics*, 36(13), pp.2059-2079.

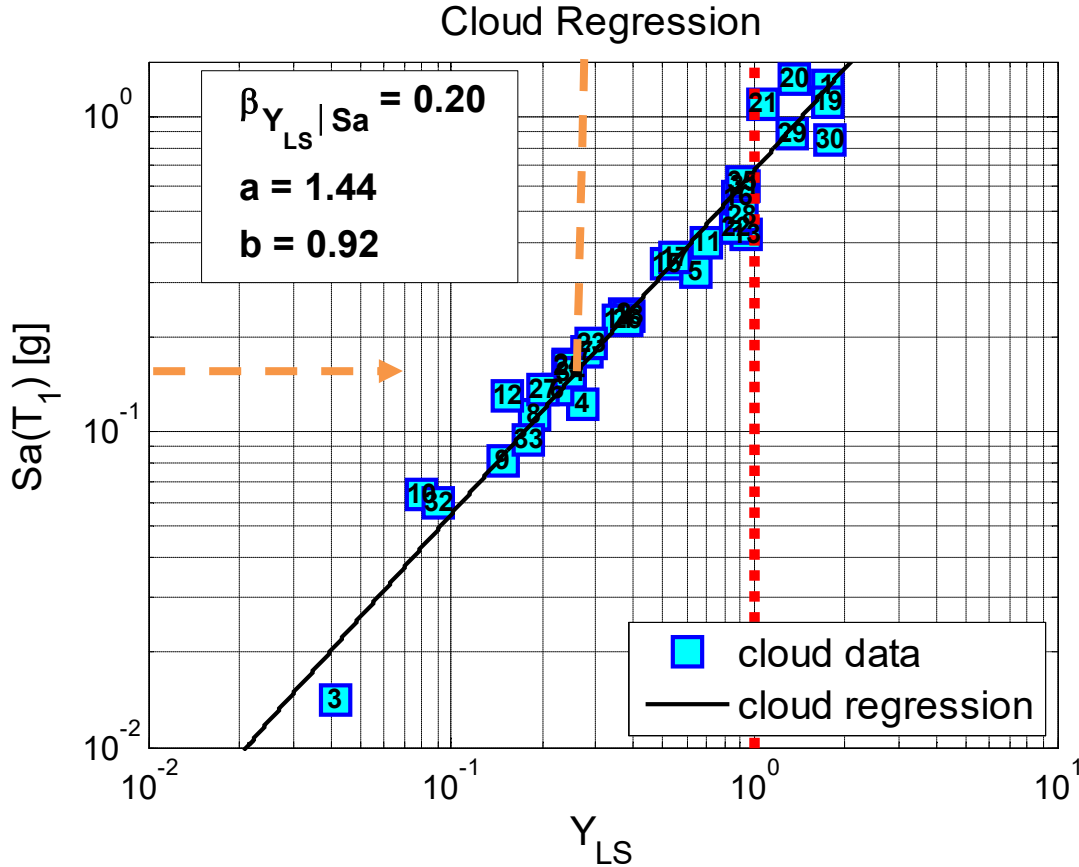


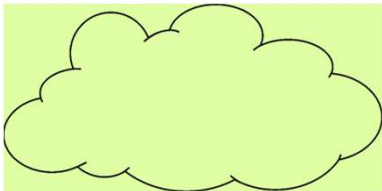
Cloud Method in One Slide

$$\log \eta_{Y|S_a} = \log a + b \log S_a$$

$$\beta_{Y|S_a} = \sigma_{\log Y|S_a} = \sqrt{\frac{\sum_{i=1}^n (\log Y_i - \log \eta_{Y|S_a_i})^2}{n-2}}$$

$$P(Y_{LS} > 1 | S_a) = \Phi\left(\frac{\log \eta_{Y|S_a}}{\beta_{Y|S_a}}\right)$$



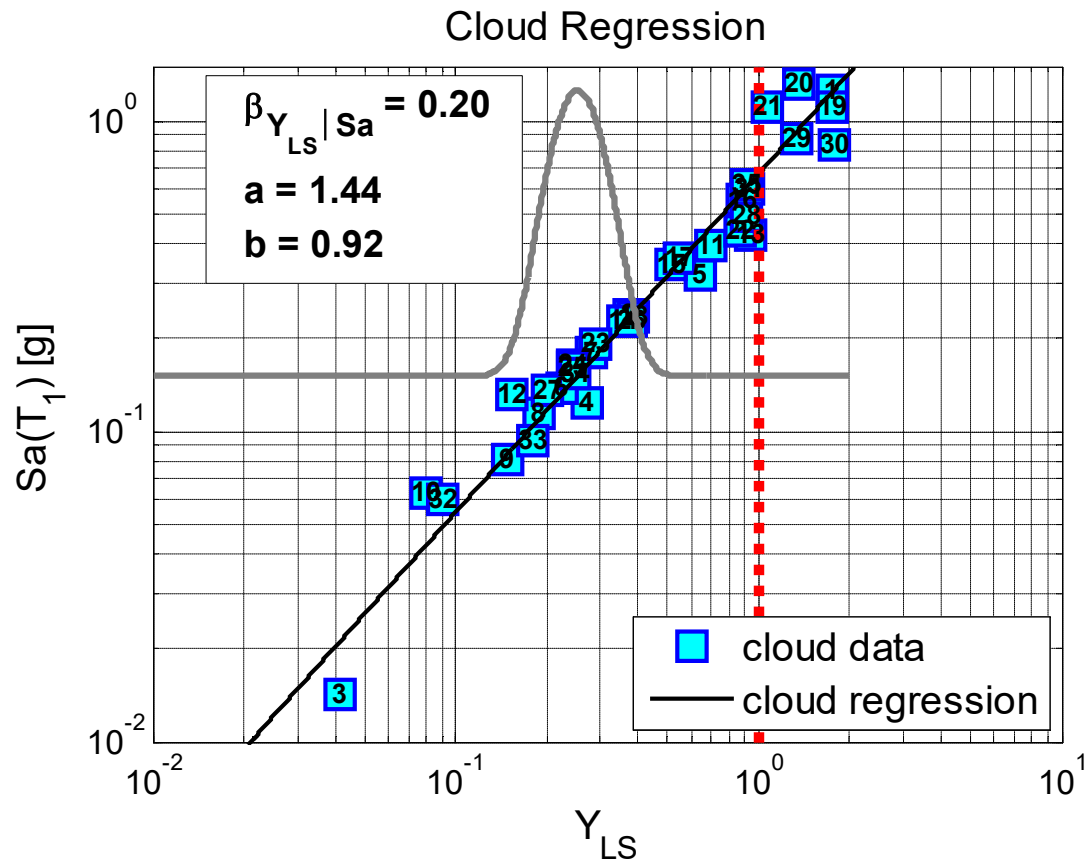


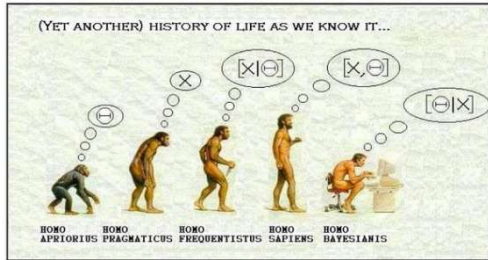
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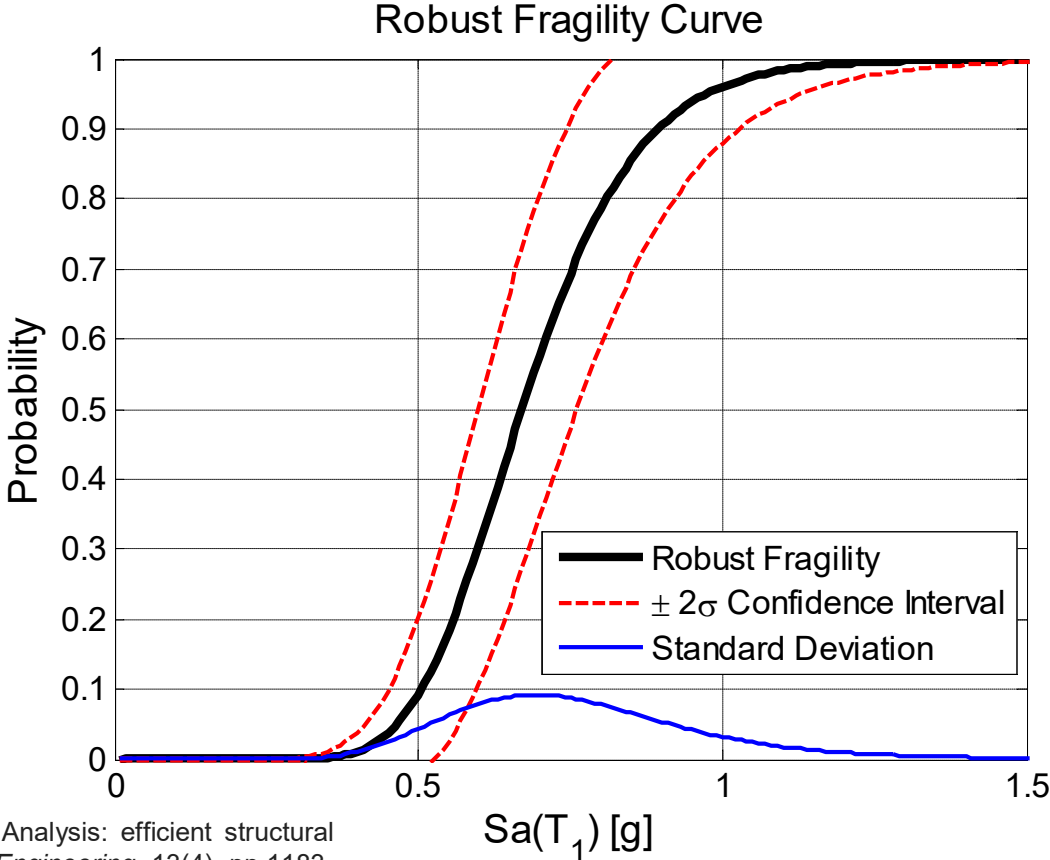
The Bayesian Take: Robust Fragility

Data Fragility Joint PDF

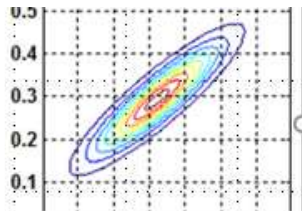
$$R(D) = \int_{\Omega(\chi)} h(\chi) \cdot p(\chi|D) \cdot d\chi$$

Model Parameters

$$h(\chi) = P(Y > 1 | S_a, \chi) = \Phi \left(\frac{\log(\eta_{Y|S_a})}{\beta_{Y|S_a}} \right)$$



Jalayer, F., De Risi, R. and Manfredi, G., 2015. Bayesian Cloud Analysis: efficient structural fragility assessment using linear regression. *Bulletin of Earthquake Engineering*, 13(4), pp.1183-1203.



The posterior distribution for χ

Bi-variate Normal Distribution

$$p(\chi | \mathbf{D}) = p(\log a, b, \beta_{Y|IM} | \mathbf{D}) = \underbrace{p(\log a, b | \beta_{Y|IM}, \mathbf{D})}_{\text{Bi-variate Normal Distribution}} \cdot \underbrace{p(\beta_{Y|IM} | \mathbf{D})}_{\text{Chi-Square Distribution}}$$

Chi-Square Distribution

- ✓ $\chi = [\log a, b, \beta_{Y|IM}]$ is the vector of the parameters (regression coefficients and standard deviation) for the prescribed Log-Normal fragility function.
- ✓ Where $p(\beta_{Y|IM} | \mathbf{D})$ is the marginal posterior distribution of $\beta_{Y|IM}$
- ✓ Where $p(\log(a), b | \beta_{Y|IM}, \mathbf{D})$ is the conditional posterior distribution of $\log(a), b$ given β



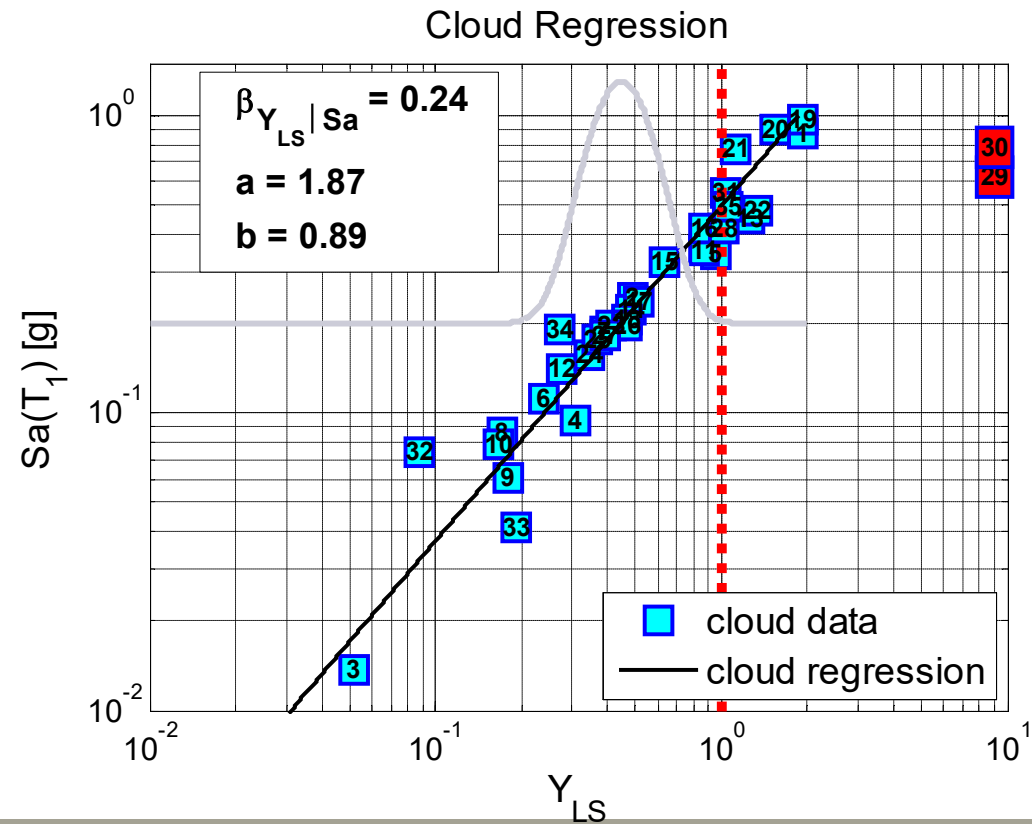
Taking into account the collapse cases

$$P(Y_{LS} > 1 | S_a) = P(Y_{LS} > 1 | S_a, NC)P(NC | S_a) + P(Y_{LS} > 1 | S_a, C)P(C | S_a)$$

$$\log \eta_{Y|NC, S_a} = \log a + b \log S_a$$

$$\beta_{Y|IM} = \sigma_{\log Y|NC, S_a} = \sqrt{\frac{\sum_{i=1}^n (\log Y_i - \log \eta_{Y|NC, S_{a,i}})^2}{n-2}}$$

$$P(Y_{LS} > 1 | S_a, NC) = \Phi\left(\frac{\log \eta_{Y|NC, S_a}}{\beta_{Y|NC, S_a}}\right)$$





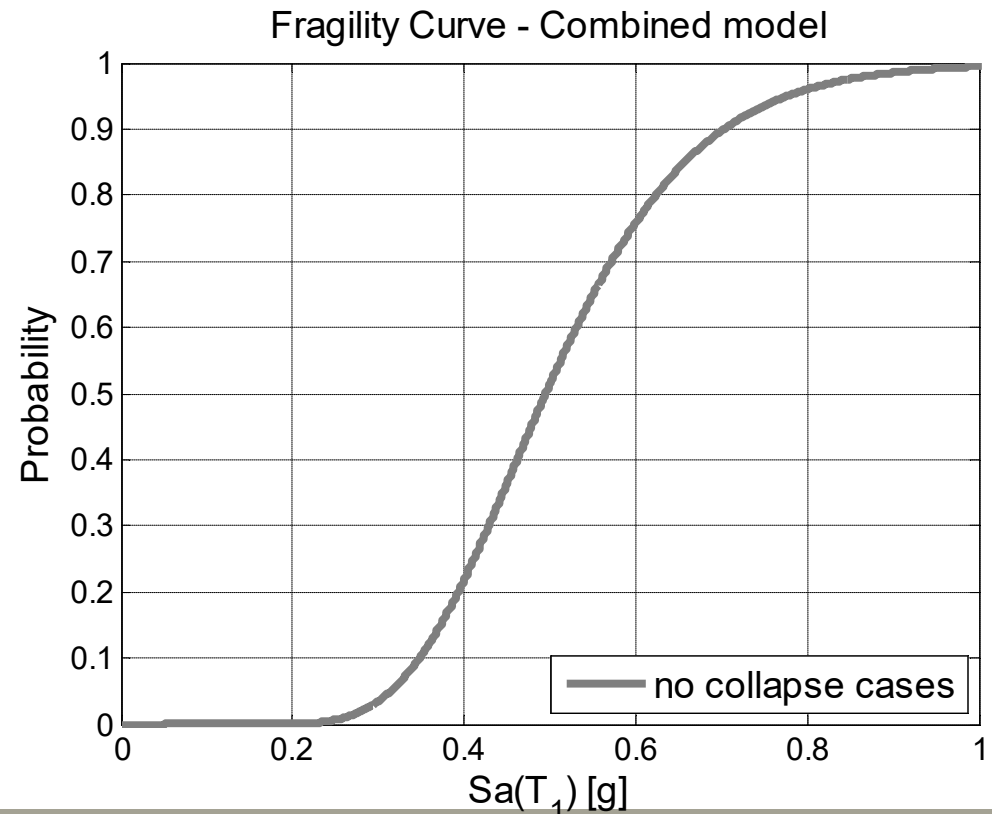
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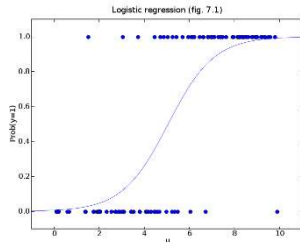
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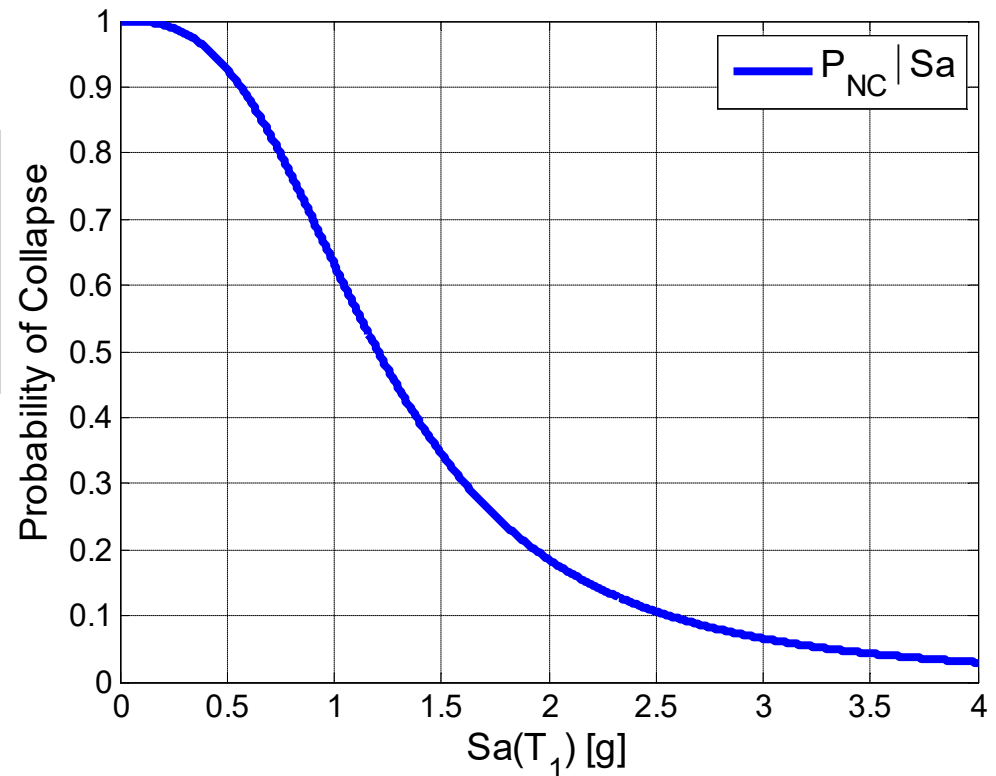


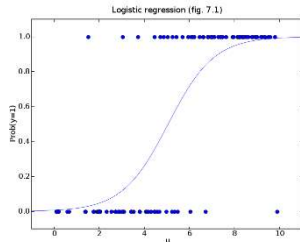
Taking into account the collapse cases:
The logistic regression

$$P(Y_{LS} > 1 | S_a) = P(Y_{LS} > 1 | S_a, NC)P(NC | S_a) + (1 - P(NC | S_a))$$

$$P(NC | IM) = \frac{e^{-(\beta_0 + \beta_1 \cdot S_a)}}{1 + e^{-(\beta_0 + \beta_1 \cdot S_a)}}$$

Combined model

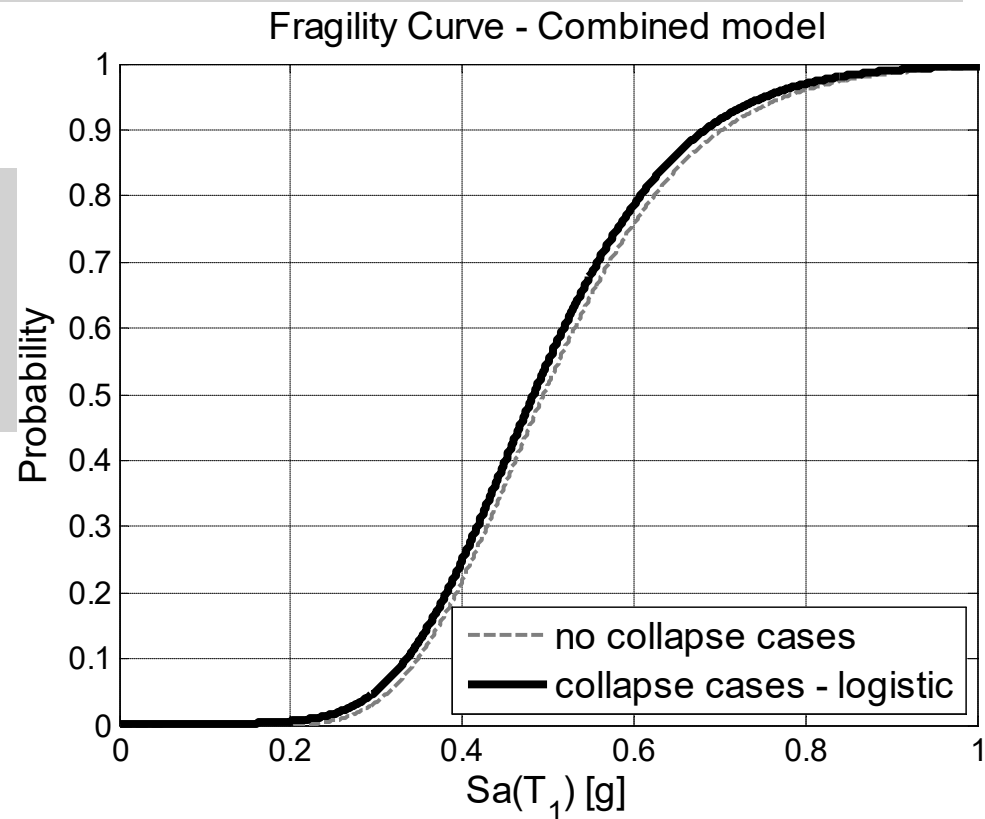




Taking into account the collapse cases:
The logistic regression

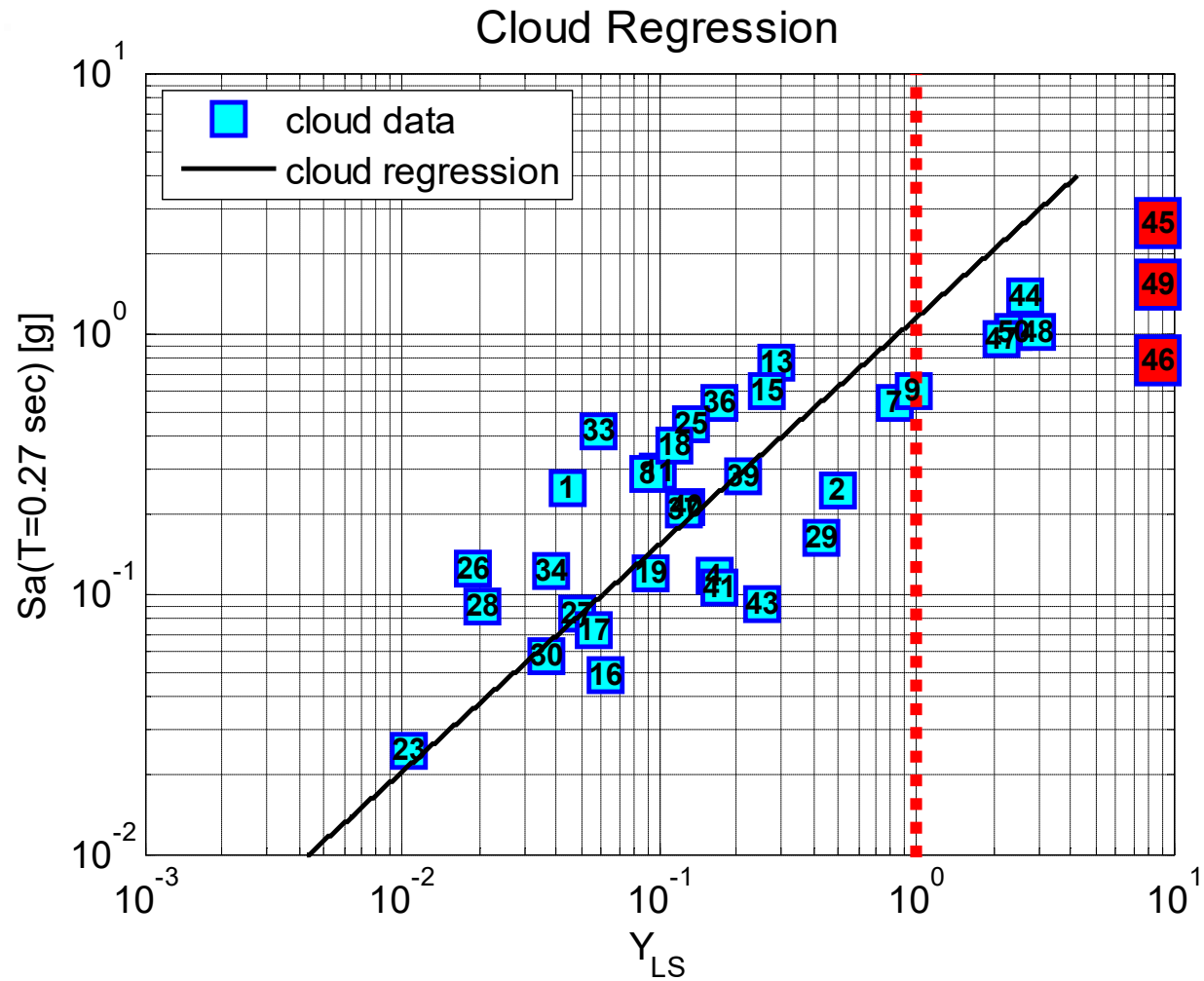
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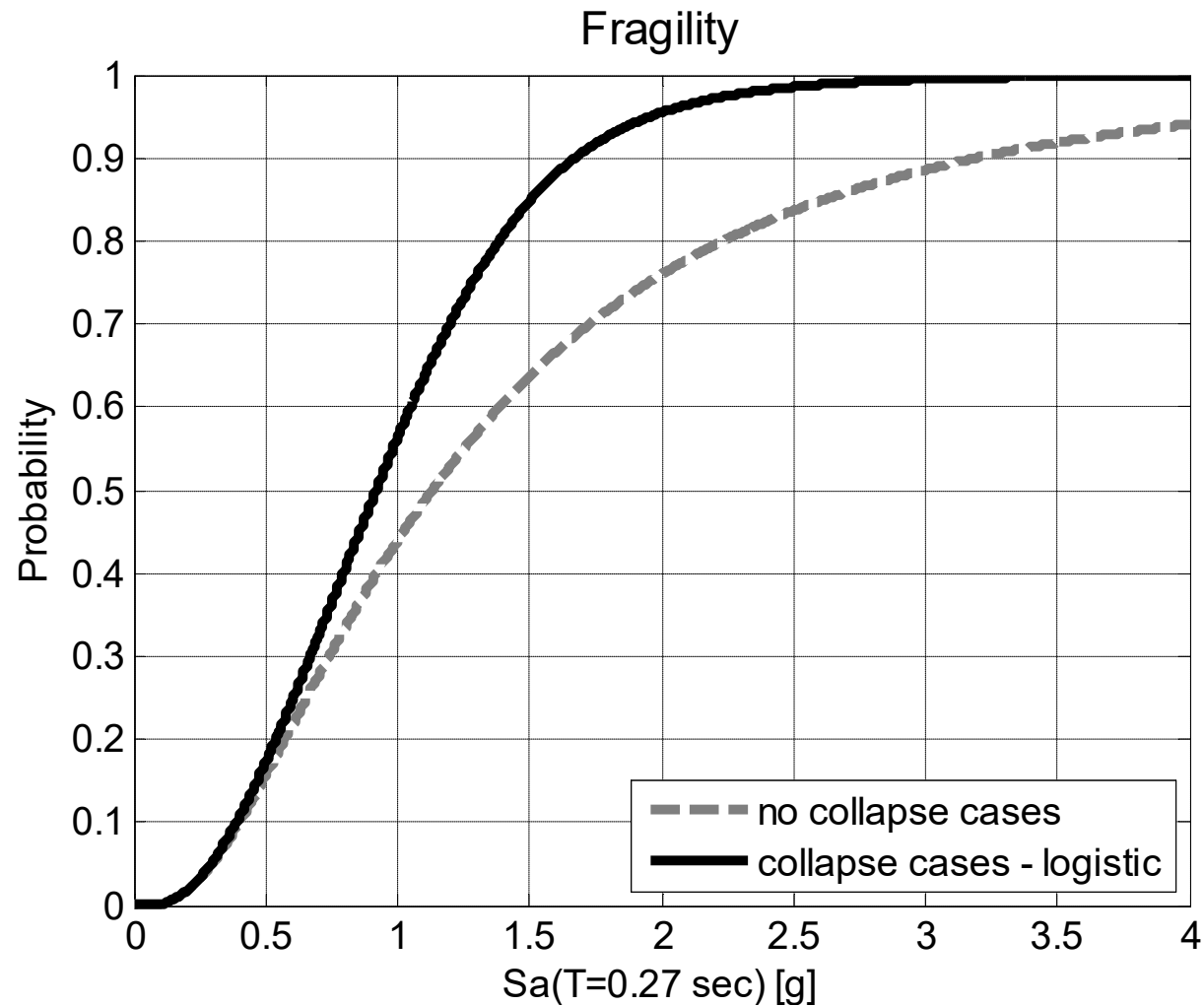


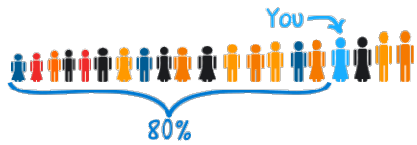
Taking into account the collapse cases: Another example





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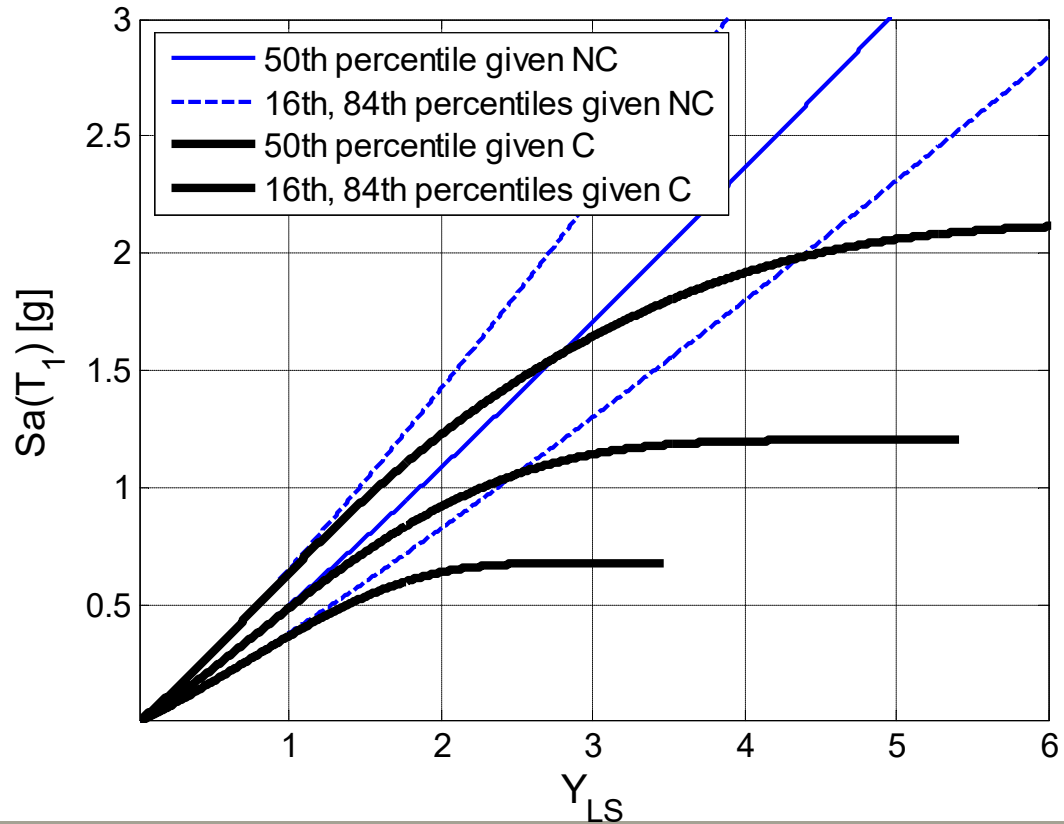


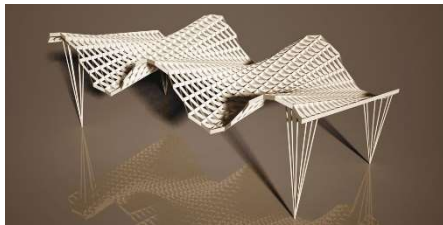


Taking into account the collapse cases: The percentiles

$$y^p = \eta_{Y|IM,NC} \cdot \exp\left(\beta \cdot \Phi^{-1}\left(\frac{p}{P(NC|S_a)}\right)\right)$$

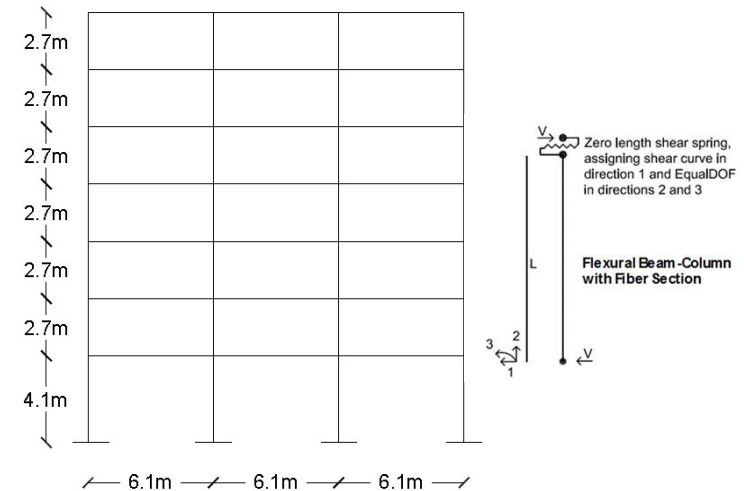
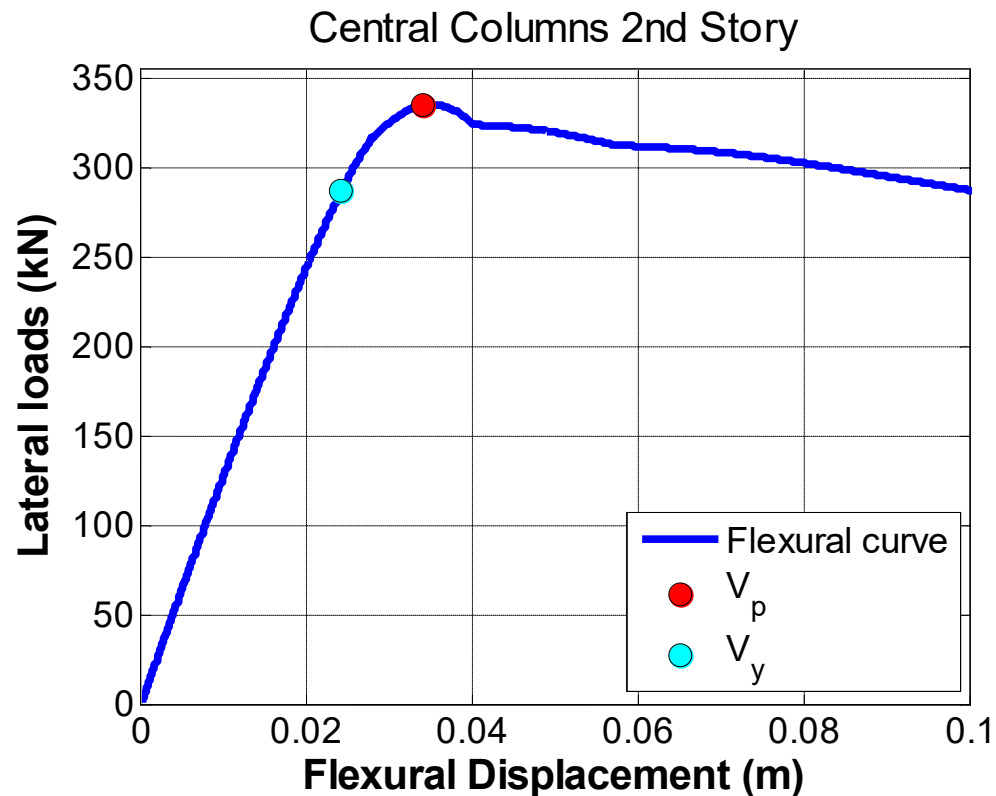
Combined model



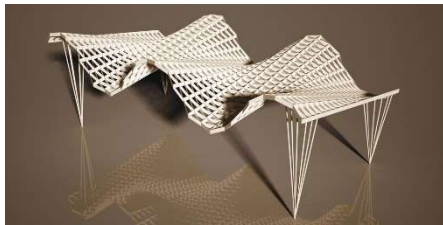


The Structural Model: Van Nuys Hotel Transversal Frame

- ✓ The structure has been re-modelled using OPENSEES and considering the axial-shear-flexural interaction

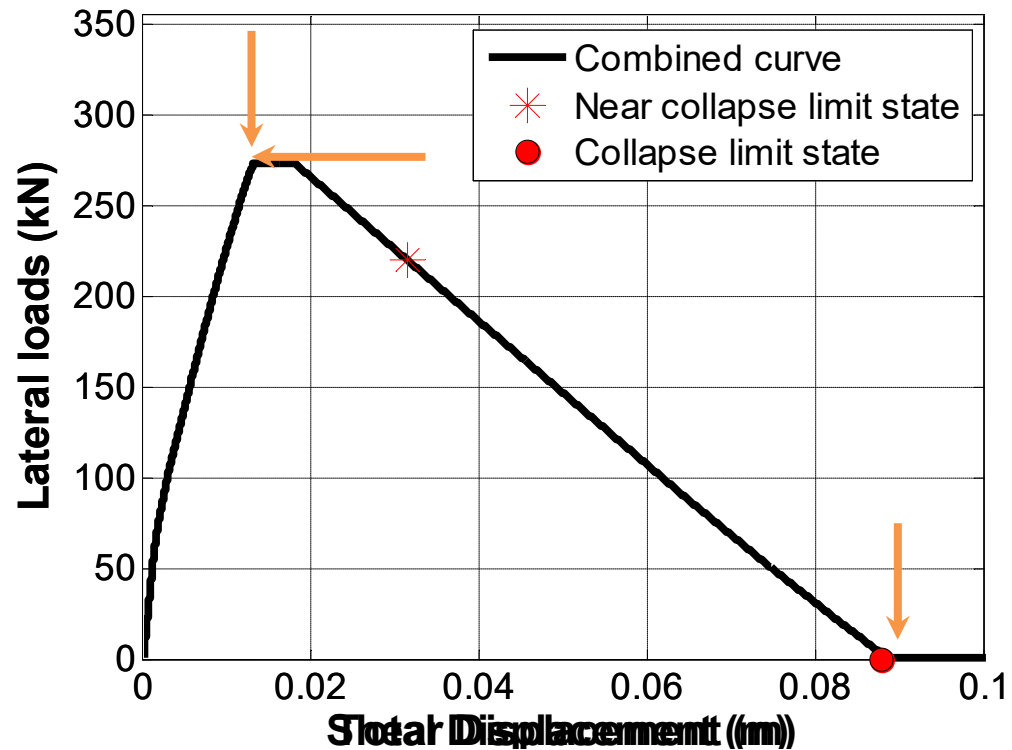


- ✓ The concrete 01 material has been used for modelling the concrete behavior.
- ✓ The longitudinal bars are modeled using the Steel02 with 1% strain hardening.

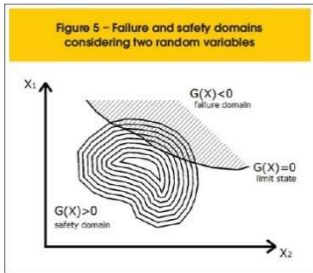


The Structural Model: Van Nuys Hotel Transversal Frame

Central Columns 2nd Story

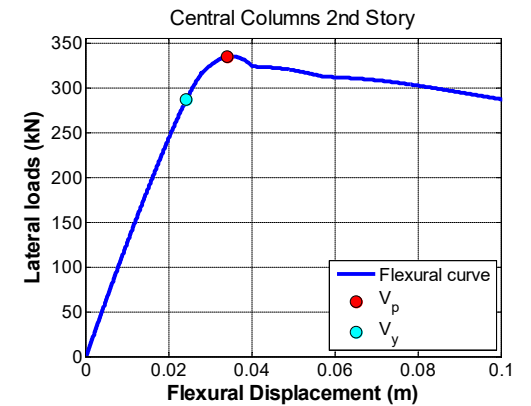
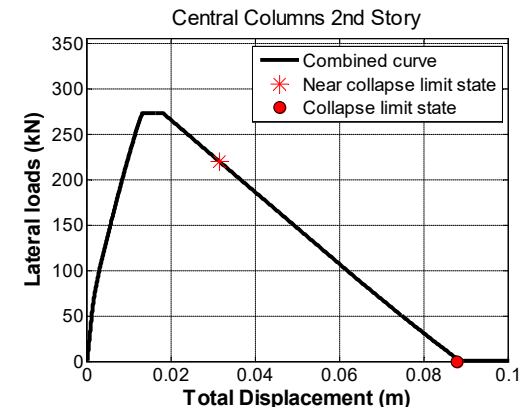


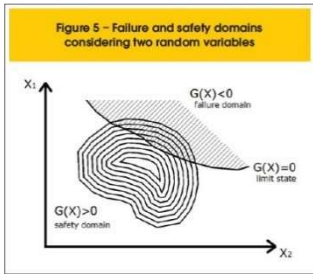
- ✓ The shear strength is modeled using Sezen and Moehle 2004.
- ✓ The shear drift at shear failure is calculated from Gerin and Adebar 2004.
- ✓ The drift at axial failure is calculated from Elwood and Moehle 2003.



The Limit State of Near Collapse

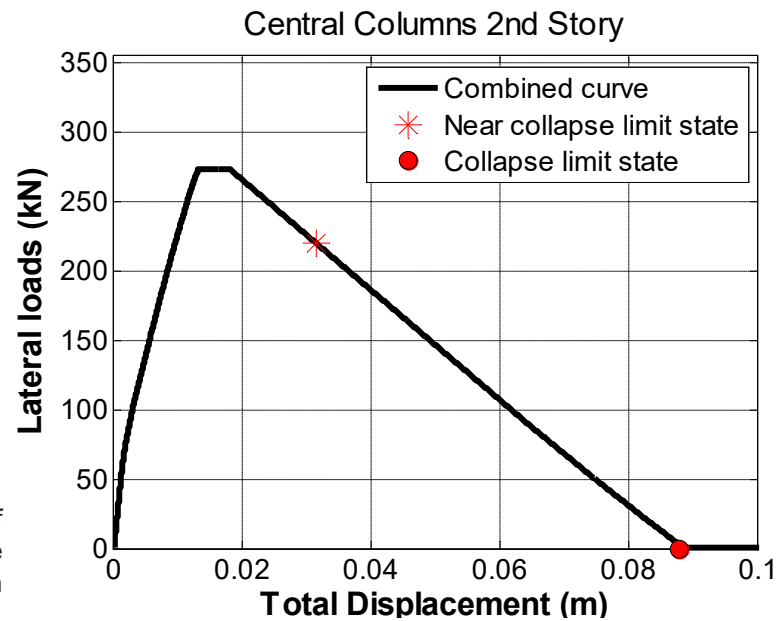
| Failure Type | Component(s) | Definition / Description |
|----------------------|--|--|
| Ductile/Brittle | column / beam | $\theta_{max} > \theta_{ultimate}$ |
| Soft-story mechanism | all columns of one story | $\theta_{max} > \theta_{yielding-flexure}$ |
| Partial mechanism | for a number of adjacent stories: all beams + bottom and top columns | $\theta_{max} > \theta_{yielding-flexure}$ |
| Global mechanism | for the entire building: all beams + base columns | $\theta_{max} > \theta_{yielding-flexure}$ |





The Limit State of Collapse

| Collapse Type | Component(s) | Definition / Description |
|---------------|------------------------------------|------------------------------------|
| Ductile | 50% +1 of the columns of one story | $\theta_{max} > \theta_{ultimate}$ |
| Brittle | 50% +1 of the columns of one story | $\theta_{max} > \theta_{axial}$ |

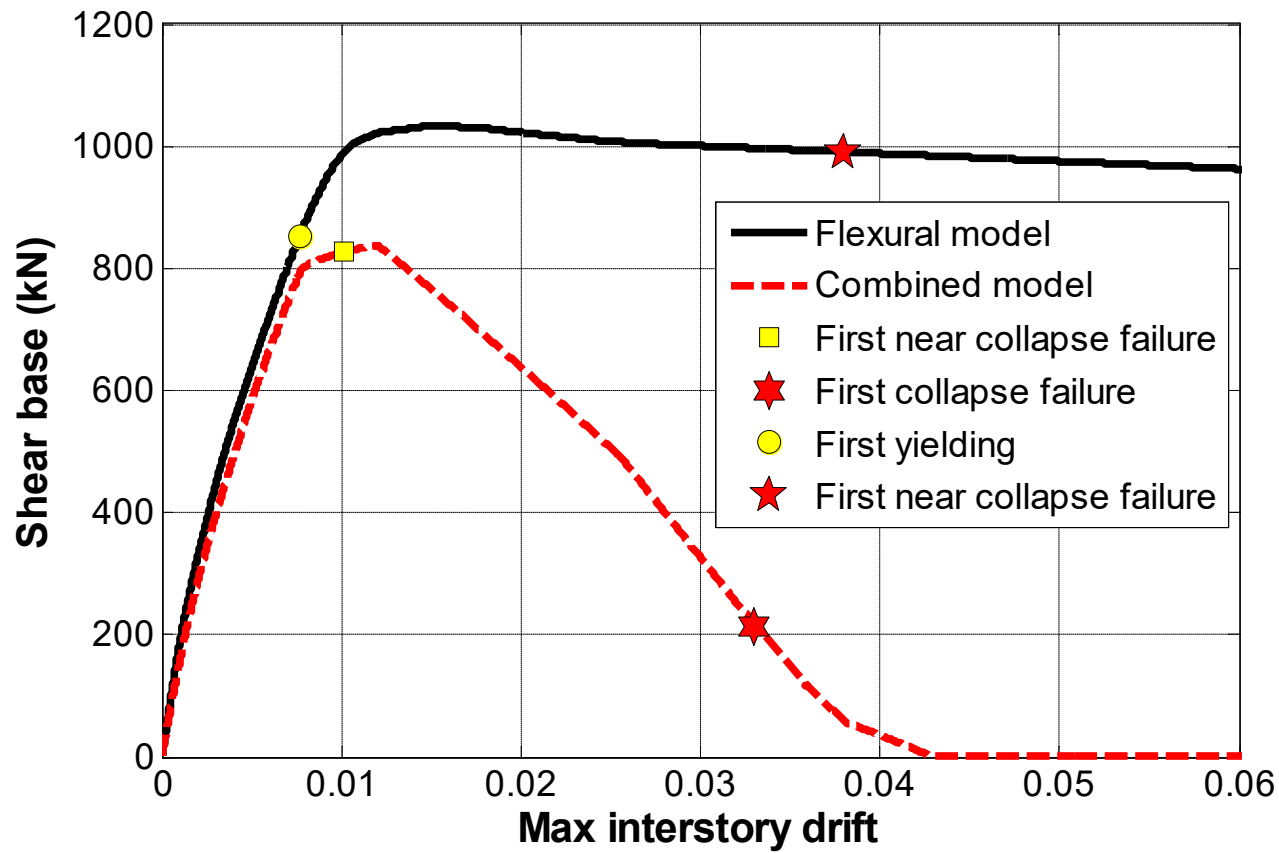


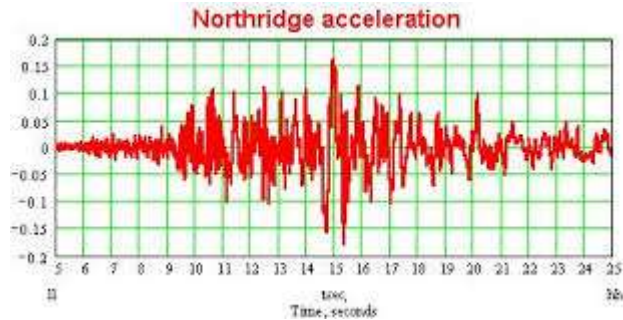
Galanis, P.H. and Moehle, J.P., 2012. Development of collapse indicators for older-type reinforced concrete buildings. In Proceedings of the 15th World Conference on Earthquake Engineering (WCEE).



The static pushover curve

Static Pushover Curves for two models

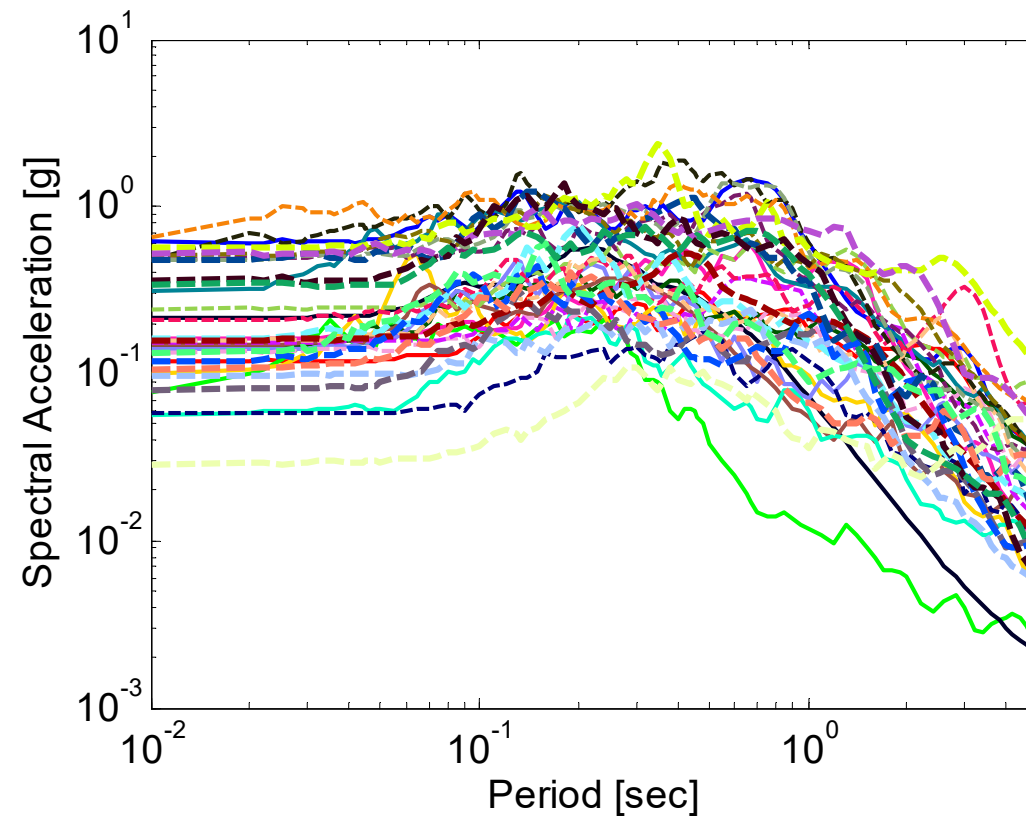


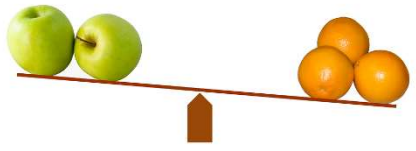


Record Selection

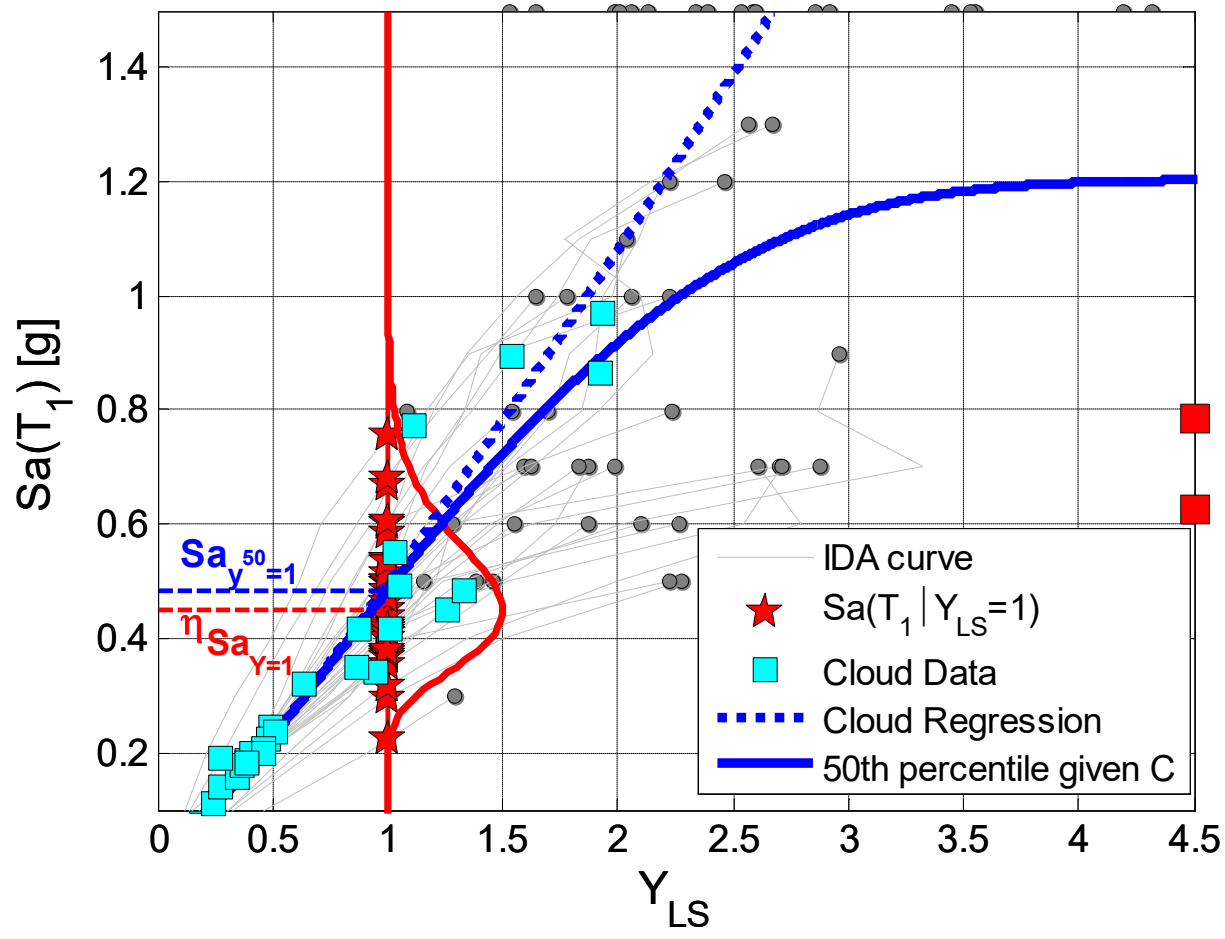
- ✓ A set of 35 records for the California sites.
- ✓ Stiff soil (Geo Matrix types Cand D)
- ✓ Moment magnitude and Joyner-Boor distance in the range:

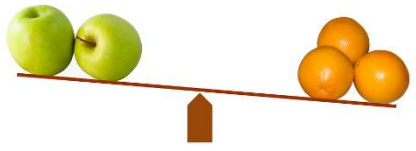
$$5.0 < M < 7.5 \quad 0.1 < R < 115 \text{ km}$$



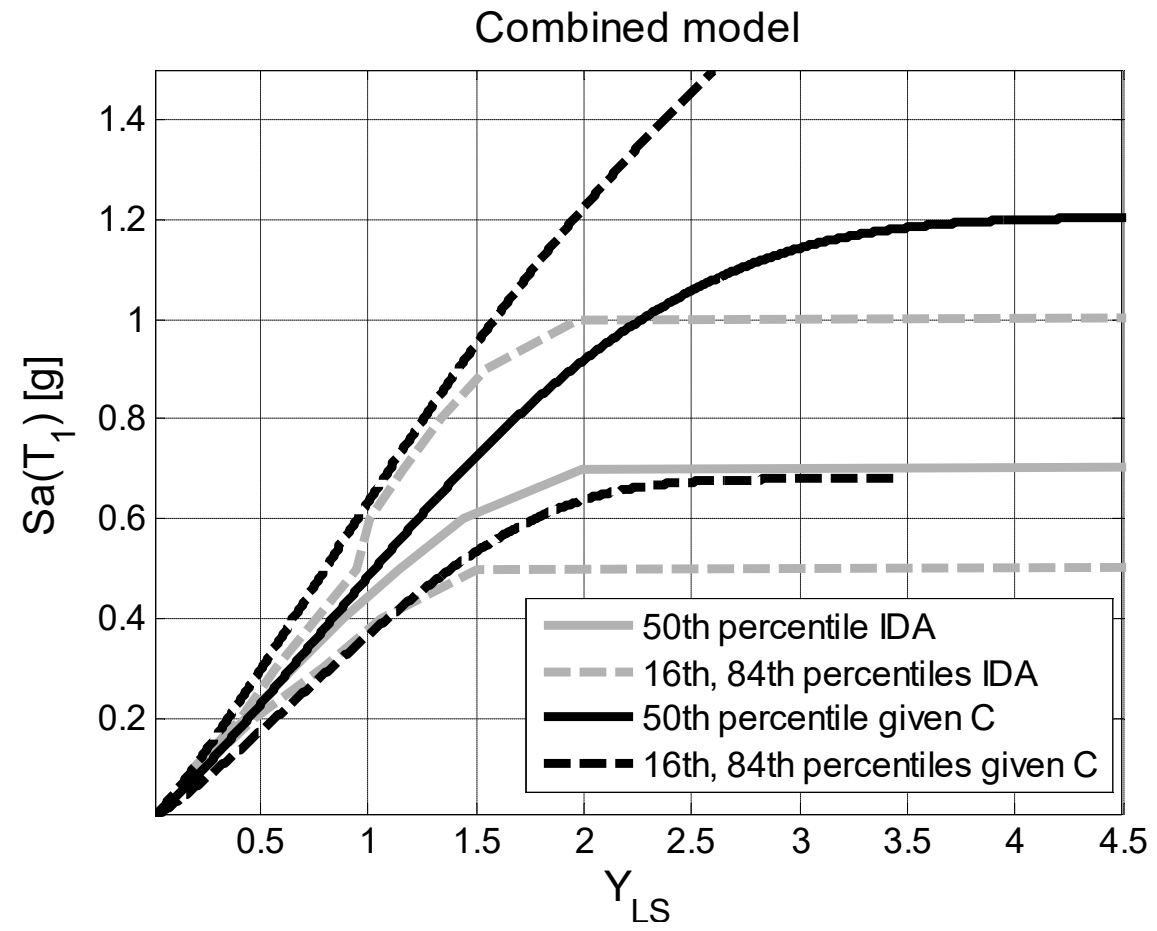


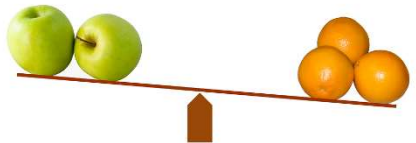
Comparison CLOUD - IDA



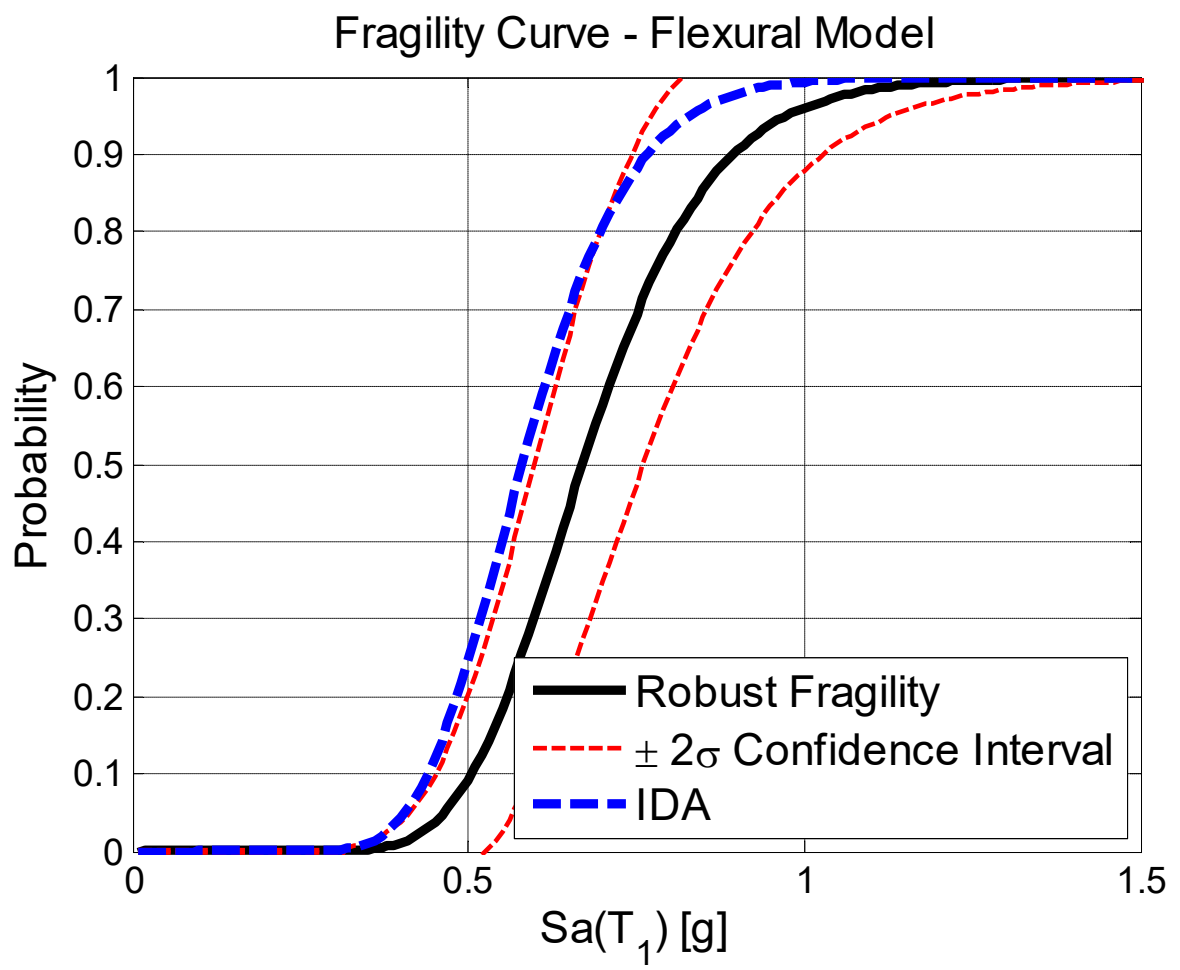


The percentiles, comparison with IDA



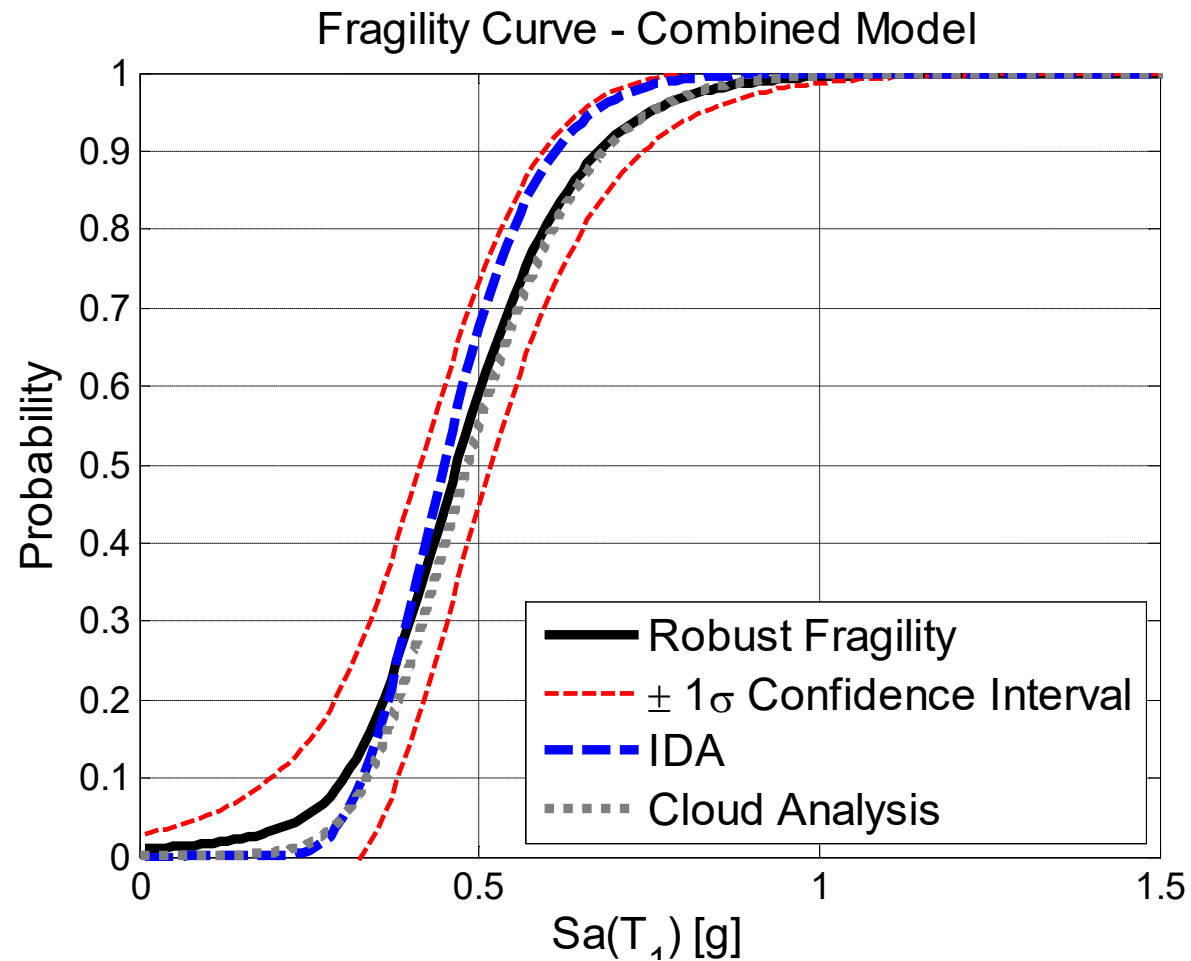


The fragilities, comparison with IDA





The fragilities, comparison with IDA





Some final thoughts

- ✓ Mixing a simple logarithmic regression model and a logistic regression model permits a systematic handling of the collapse cases;
- ✓ Using the cloud with performance-based variable defined based on cut-sets can overcome the need for identifying the collapse cases by setting rather arbitrary thresholds;
- ✓ The cloud method if coupled with careful record selection can lead to reasonable results (in comparison with IDA);
- ✓ This also helps in achieving the famous softening effect in the percentiles of EDP given IM;
- ✓ The bayesian robust fragility estimate helps in defining a confidence interval taking into account the uncertainty in the parameters of the fragility curve;
- ✓ The robust fragility can be calculated also in the case considering the collapses;

Thank You

Special Thanks to:

**Dr. Hossein Ebrahimian
Post-doctoral Researcher
University of Naples Federico II**

and

**Andrea Miano
Ph.D. Candidate
University of Naples Federico II**