SELECTION OF GROUND-MOTION PREDICTION EQUATIONS FOR PROBABILISTIC SEISMIC HAZARD ANALYSIS : CASE STUDY OF TAIWAN

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Background

In 2015-June National Center for Research on Earthquake Engineering (NCREE), supported by Tai-Power Company (to response the request of NTTF 2.1 Seismic Reevaluation), launched NUREG/CR– 6372 researches (SSHAC), in understanding and documenting lessons learned from recent PSHAs conducted at the higher SSHAC Levels. (follow the experiences of research on Diablo Canyon Nuclear Power plant)



- Selection of candidate GMPEs for PSHA
- Visualization technique for GMPE selection
- Selection of GMPE common form
- Visualization of model space
- Calculate GMPE weighting

Conclusion









Selection of appropriate GMPEs for PSHA

- 1. Need best estimate of GMPE
 - 2. Consider range of alternative models to characterize the uncertainty in the GMPEs
- **1.** Aleatory uncertainty:

expressing random variability of amplitude about a median prediction equation,

- can be handled in a PSHA by integrating over the distribution of ground-motion amplitude about the median,
- 2. Espistemic uncertainty:

expressing uncertainty concerning the correct value of the median,

• can be handled by considering alternative GMPEs in a logic tree format (must capture **uncertainties in form & amplitude**),

Sensitivity analysis of the proposed weights for GMPEs on the seismic hazard.



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Type of uncertainty

in **GMPEs**

General Feature of Candidate GMPEs

GMPE	GMPE Acronym	Regions	Magnitude Interval	Primary distance	Style of faulting	Site effect	Component	Number of records and events
(Abrahamson, N. A., Silva, W. J., and Kamai, R., 2014)	ASK14	Global	Mw (3.0-8.5)	Rrup (<300km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GMRotI50	15750 and 326
(Boore, D. M., Stewart, J. P ., Seyhan, E., and Atkinson, G. M., 2014)	BSSA14	Global	Mw (3.0-8.5)	Rjb(<300km)	U,SS,NML,REV	Vs30	PGA,PGV,PSA in GMRotI50	~16000 and ~400
(Campbell, K. W., and Bozorgnia, Y., 2014)	CB14	Global	Mw (3.3-8.5)	Rrup (<300km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GMRotI50	15521 and 322
(Chiou, B. S-J., and Youngs, R. R., 2014)	СҮ14	Global	Mw (3.5-8.5)	Rrup (<300km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GMRotI50	12444 and 300
(Idriss, 2014)	ld14	Global	Mw (5-8.5)	Rrup (<150km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GMRotI50	7135 and 160
(Akkar, S., Sandikkaya, M. A., and Bommer, J. J., 2014)	ASB14	EU and ME	Mw (4-7.5)	Rjb (<200km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GM	1041 and 221
(Bindi D., Massa M., Luzi L., Ameri G., Pacor F., Puglia R., and Augliera, P., 2014)	Bi14	EU and ME	Mw(4-7.6)	Rjb (<300km)	U,SS,NML,REV	Vs30	PGA,PGV,PSA in GM	2126 and 365
(Graizer, V., and Kalkan, E., 2015)	GK15	Global	Mw(5.0-8.0)	Rrup (<250Km)	SS,NML,REV	Vs30	PGA,PGV,PSA in GM	2583 and 47
Zhao et al. 2016	Zhao16	Japan	Mw(5.0-7.3)	Rrup(<300km)	FN,SS	Dummy variable	PGA, PSA in GM	6482 and 76 (cr), 47(mum)
Özkan Kale, Sinan Akkar, Anooshiravan Ansari, and Hossein Hamzehloo	Ka15	Turkey and Iran	Mw(4.0-8.0)	Rjb(<200km)	U,SS,NML,RE V	Vs30	PGA, PGV, PSA in GM	670(Tur),528(Ir)
Lin , P.S et al. 2011	Lin11	Taiwan	Mw(5.0-7.6)	Rrup(<240km)	-	no	PGA,PSA in GM	5268 and 52
(Cauzzi, C., Faccioli, E., Vanini, M., and Bianchini, A., 2014)	Ca14	Global	Mw(4.5-7.9)	Rrup (<150km)	U,SS,NML,REV	Vs30	PGA,PGV,PSA in GM	1880 and 98



Selection of candidate GMPE

• Selected GMPEs: ASK14, BSSA14, CB14, CY14, Id14, GK15,

ASB14, Bi14, Ca14

Lin11, KAAH15-Turkey, KAAH15-Iran, Zhao16

(total of 13 models)



Use of multiple models with alternative functional forms is required to properly capture uncertainties in **forms** as well as in **amplitude**.



• Selected GMPEs: ASK14, BSSA14, CB14, CY14, Id14, ASB14, Bi14, Ca14,

GK15, Lin11, KAAH15-Turkey, KAAH15-Iran, Zhao16

(total of 13 models)

$$Mix = \sum_{i=1}^{N} w_i GMPE_i(M, R, \theta)$$

- The scenarios for generate synthetic data:
 - M = 5.0, 5.2, 5.4, 5.5, 5.6, 5.8, 6.0, 6.2, 6.4, 6.5, 6.6, 6.8, 7.0, 7.2, 7.4, 7.5, 7.6, 7.8, 8.0 for strike slip and reserve faulting.
 - M = 5.0, 5.2, 5.4, 5.5, 5.6, 5.8, 6.0, 6.2, 6.4, 6.5, 6.6, 6.8, 7.0 for normal faulting.
 - Rx=-200,-150,-100,-85,-70,-65,-60,-55,-50,-45,-40,-35,-30,-28,-26,-24,-22,-20,-18,-16,-15,-14,-12,-10,-8,-6,-5,-4,-2. (foot wall)
 - From fault geometry, Rrup, and Rjb can be calculated.
 - Vs30 = 760 m/s.
 - Dip =90° for strike slip, and dip = 45° for normal and reverse faulting events.
 - Other parameters are set to default (Ztor, W,...)



Reference to the set of 13GMPEs

✓ Add the reference to the set of 13GMPEs:

□ Mix Model (average of all models) : $Mix = \sum_{i=1}^{N} w_i GMPE_i(M, R, \theta)$

Up-Down Scaled models :

Mix + log α , with α = 0.67, 0.8, 1.25, 1.5 \Box S--, S-, S+, S++

□ Magnitude Scaled models :

 $Mix + \beta (M - 6.5)$, with $\beta = -0.4, -0.2, 0.2, 0.4 \square$ M--, M-, M+, M++

Distance Scaled models:

 $Mix + \gamma (R-70)$, with $\gamma = -0.01$, -0.005, 0.005, $0.01 \square$ R--, R-, R+, R++



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Reference to the set of 13GMPEs





Generate Sammon Map

(13 Candidate GMPEs) + (13 Reference models: mix & scale models)

 $GMPE_{i} = f(M, R) \quad \begin{cases} M = 5.0, 5.2, \dots, 7.8, 8.0 \text{ (for SS \& NF)} \\ Rx = -200, -150, \dots, -4, -2. \text{ (foot wall)} \end{cases}$

The simplest technique **for dimensionality reduction** is a straightforward **linear projection**, for example, as in PCA — principal component analysis. (PCA simply maximizes variance)

Non-linear projections may therefore be desirable when analyzing such data.

Sammon Mapping:

To minimize the differences between corresponding inter-point distances in the dimension space



Combined the embedded GMPEs and the Mix Model & Scaled Models into the
 [X] matrix considered to be N-dimensional space





Visualization technique-GMPEs Calculation

To construct the
$$[X_{Sammon-map}]_{26x2}$$

Using $[X]_{26xN} = \begin{bmatrix} [\mu_{GMPE}]_{13xN} \\ [\mu_{Scaled}]_{13xN} \end{bmatrix}$
Calculate inter-point distance
and define $\varepsilon_{ij} = \sqrt{(\sum_{k=1}^{N} ([X_{ik}] - [X_{jk}])^2)/N}$
to construct $[A_{1x1}]_{26x26} = \begin{bmatrix} 0 & \varepsilon_{X1-2} & \cdots & \varepsilon_{X1-26} \\ \varepsilon_{X2-1} & 0 & \cdots & \varepsilon_{X2-26} \\ \vdots & \vdots & \vdots \\ \varepsilon_{X26-1} & \varepsilon_{X26-2} & \cdots & 0 \end{bmatrix}_{26x26}$
 \longrightarrow Min. $E = \frac{1}{\sum_{i < j} \varepsilon_{ij}} \sum_{i < j} \frac{(\varepsilon_{ij} - \delta_{ij}^{map})^2}{\varepsilon_{ij}}$
Determine : Δ_{PCA}^{Map} to construct the Sammon map $[X_{Sammon-map}]_{26x26}$



Visualization technique-GMPEs Calculation

$$\varepsilon_{ij} = \sqrt{\left(\sum_{k=1}^{N} \left(\left[GMPE_{ik}\right] - \left[GMPE_{jk}\right]\right)^{2}\right)/N}$$

$$\left[\Delta_{[X]}\right]_{26x26} = \begin{bmatrix} 0 & \varepsilon_{X1-2} & \cdots & \varepsilon_{X1-26} \\ \varepsilon_{X2-1} & 0 & \cdots & \varepsilon_{X2-26} \\ \cdots & \cdots & \ddots & \vdots \\ \varepsilon_{X26-1} & \varepsilon_{X26-2} & \cdots & 0 \end{bmatrix}_{26x26}$$

$$\overline{\Delta}_{ij}^{GMPE} = \sqrt{w_{k}} \sum_{k}^{N} \left(GMPE_{ik} - GMPE_{jk}\right)^{2}}$$

$$w_{k} = 0.5 \left(DEAGG\left(M_{k}, R_{k}\right) + \frac{1}{NS}\right)$$





GMPE Distribution



The considered GMPE models are not adequate to fully capture the range of epistemic uncertainty because of the existing gaps among models.



Model A - based on rupture distance, R_{RUP}

$$SA_{Rrup-based} = \theta_{1} - \frac{\theta_{8}^{2}R_{rup} + \theta_{9}^{2}Z_{tor}}{\theta_{2}M + \theta_{3}(M-5.5)} - \exp(\theta_{11})F_{NML} + \exp(\theta_{10})F_{REV} + (\theta_{5} + \theta_{6}(M-5))\ln(\sqrt{R_{rup}^{2} + \theta_{7}^{2}}) + \left\{ \begin{array}{l} \theta_{2}M & for \ M < 5.5 \\ \theta_{2}M + \theta_{3}(M-5.5) & for \ 5.5 \le M \le 6.5 \\ \theta_{2}M + \theta_{3}(M-5.5) + \theta_{4}(M-6.5) & for \ M > 6.5 \end{array} \right.$$

Model B - based on rupture distance, R_{JB}

$$SA_{Rjb-based} = \theta_{1} - \frac{\theta_{8}^{2}R_{jb} - \exp(\theta_{11})F_{NML}}{\theta_{2}M} + \exp(\theta_{10})F_{REV} + (\theta_{5} + \theta_{6}(M - 5))\ln(\sqrt{R_{jb}^{2} + \theta_{7}^{2}})$$

$$+ \begin{cases} \theta_{2}M & for \ M < 5.5 \\ \theta_{2}M + \theta_{3}(M - 5.5) & for \ 5.5 \le M \le 6.5 \\ \theta_{2}M + \theta_{3}(M - 5.5) + \theta_{4}(M - 6.5) & for \ M > 6.5 \end{cases}$$

(Behave differently on the hanging wall side)





Select common form to fit the synthetic data

GMPE Distribution

Develop common funtional form for each candidate GMPE $\mu(\ln y) = f(M, R, \theta, ...)$

For each GMPE *i*, estimate the set of coefficients θ_i (using senthetic data)

Comparison between the common form GMPE with respect to the candidate GMPE





Example : Fit common form to synthetic data.

R_{RUP}-based



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Example: Fit common form to synthetic data.

R_{IB}-based



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Select common form to fit the synthetic data



(Assumed multivariate normal distribution)



Distribution of GMPE coefficient

$\label{eq:calculate} \begin{array}{c} Calculate \ \mu_{\theta} \ and \ \ \Sigma_{\theta} \\ (from \ the \ fitted \ sets \ of \ coefficients) \end{array}$





GMPE Distribution

Model A: R_{RUP}-based Visualization of models in 2D Model B: R_{IB}-based M = 4.75, 5.25,...,7.75 RJB = 1.25, 1.5, 3.75, ...,87.5 T001s Vs30=760m/s Sof = 0, 1T=.001s 0.5 Convex hull of candidate GMPE In (units) with plus/minus 2_{σAY14} Zhao16 and KAAH-Turkey -0.5 were not included ModelA **Original Candidate GMPEs ModelB** Original Candidate GMPEs+2σ Original Candidate GMPEs-2σ -0.5 0 0.5 In (units) $\ln PSA_{ij} = f_i \left(M_j, R_j, \dots \right) + \alpha \sigma_{AY14} \left(M_j, R_j, F \right)$ Capture the range of epistemic uncertainty. with $\alpha = \{-2, 0, 2\}$



Fitting ellipse to convex hull. Scale up and down by a factor 2, 1.5 and 0.5

Split region covered by four ellipese into many subregion.







Selection of models – Approach 1



 For each subregion, common forms are selected

Select representative model for each subregion.



Visualization of model space: Residual analysis

- Database. (about 900 records from Taiwan area)
 - Selected data that is relevant for the application: M≥5, Rrup≤100km, and 700 ≤Vs30≤ 800m/s.
 - \rightarrow Corrected to Vs30 = 760m/s.
- For each of 2000 sample models.
 - Residual are calculated.
 - These are split into between-event residual and within event residuals, δB and δw_{ij}
 - The log-likelihood is calculated using the eq.7 of Abrahamson and Youngs 1992





 $\mathbf{E}_{ij} = \eta_i + \varepsilon_{ij} \sim \delta \mathbf{B}_{e} + \delta W_{es}$

 $\sigma^2 = \tau^2 + \phi^2$

GMPE Distribution

Scenarios

- M = 4.75, 5.25,...,7.75
- RJB = 1.25, 1.5, 3.75, ...,87.5
- Vs30=760m/s
- Sof = 0, 1
- T=0.001s
- Data corrected to Vs30=760m/s
 Model evaluation.
 - Contour using mean between event residual
 + Candidate GMPEs
 ↓

Overlay the contour map of mean between residual w.r.t. the fitting ellipse







Selection of models – Approach 1

Identify the value of mean between event residual of each form from each split region.





Selection of models – Approach 2

- Discretizing the map using the Voronoi-Diagram based on a set of points (includes median GMPEs, median $\pm 2\sigma_{AY14}$, and the points results from the intersection between contours and ellipses).
- □ This decomposition has the property that an arbitrary point P within the region R{i} is closer to point i than any other points.





the Voronoi-Diagram.

GMPE Distribution

□ Scenarios

- M = 4.75, 5.25,...,7.75
- RJB = 1.25, 1.5, 3.75, ...,87.5
- Vs30=760m/s
- Sof = 0, 1
- T=.001s
- □ Data corrected to Vs30=760m/s
- Model evaluation.
 - Contour based on Log-likelihood value of each model



Log-likelihood



Selection of models – Approach 2





GMPE weightings and cells

- The Representative suite of common Form model
- The candidate GMPE models







Selection of models – Approach 2,



among different sub-region in the Voronoi Diagram



GMPE weightings and cells

The weight for the selected representative model for each cell $1 \sum_{i=1}^{N_i} I$

$$w_i = A_i \frac{1}{N_i} \sum_{j=1}^{i} L_{ji}$$

• *L_{ij}*

- $1/|\mu(\delta B)|$, One over the absolute mean between event residual.
- $1/\mu(\delta B)^2$, One over the squared mean between event residual.
- L, the likelihood.



GMPE weightings and cells

- $1/|\mu(\delta B)|$, One over the absolute mean between event residual.
- $1/\mu(\delta B)^2$, One over the squared mean between event residual.
- L, the likelihood.



- Capture uncertainties in form & amplitude
- Conduct sensitivity analysis from the proposed weights on the seismic hazard





- Method on the selection of GMPE for PSHA is introduced (The above-mentioned method had been used in "Diablo Canyon SSHAC Level-3 Report").
- 2. The calculated weighting value depends not only on the mean between event residual or likelihood value, but also depend on the area of each cell or the way of mapping is partitioned.
- 3. The proposed method can generate the quantitative value on selecting and ranking of GMPE model and provides information for experts on the judgment of weighting factor in seismic hazard calculation



Thank you for your attention !

Questions ?



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A Procedure of Logic Tree for GMPE

(to capture uncertainties in **form** as well as **amplitude**)





Select common form to fit the synthetic data

GMPE Distribution



In order to capture the correlation between the different coefficients θ , the common form is also fitted to the interpolated ground motions from the candidate GMPEs

$$Interp\left(\ln SA(M, R, Vs30)\right) = w_a \ln\left(SA_i\right) + w_b \ln\left(SA_j\right) \text{ with } i \neq j$$
$$\{w_a, w_b\} = \left\{\frac{1}{3}, \frac{2}{3}\right\}, \left\{\frac{1}{2}, \frac{1}{2}\right\}, \left\{\frac{2}{3}, \frac{1}{3}\right\}$$



Develop simple GMPE

• Simple functional form fit to Taiwan data:

 $\ln psa(T 0.01) = c_1 + c_{1a}FRV + c_{1b}\Delta Ztor + c_2M + c_3M^2 + c_4\log(R + c_5\exp(c_6M))$ $\Delta Ztor = Ztor_i - \overline{Z}tor_{CY14}$

- Ztor is limited to 20km because we adopt Ztor-M relation of CY14.
- Method: Maximum regression using mixed effect model

A STABLE ALGORITHM FOR REGRESSION ANALYSES LISING THE RANDOM EFFECTS MODEL

Bulletin of the Seismological Society of America, Vol. 82, No. 1, pp. 505-510, February 1992	c1	-6.3535
BY N. A. ABRAHAMSON AND R. R. YOUNGS	c1a	0.0855
Model bias and variability		0.0455
$- \tau = 0.4061$	C2	2.0335
	C3	-0.03826
$-\phi = 0.5838.$	C4	-2.0609
	C5	0.388
	C6	0.6268
\checkmark		

Use this number to construct the log-likelihood contour



Selection of models – Approach 2



η	Log-likelihood
-1.023	-440.448
-1.0788	-399.806
-0.3595	-363.047
-0.7301	-419.606
-0.5533	-488.515
-0.6589	-474.253
-0.4225	-389.43
-0.5426	-390.071
-0.4222	-376.565
-0.2608	-323.921
-0.1877	-374.395
-0.1211	-304.776
-0.1739	-326.772
-0.3189	-335.055
-0.1726	-324.308
-0.0631	-295.312
-0.1776	-391.671
-0.1535	-330.265
-0.1453	-310.945
-0.076	-299.143
-0.1569	-329.054
-0.1148	-316.892
-0.2214	-325.172
-0.2319	-347.509





Distribution of GMPE weighting

• $1/|\mu(\delta B)|$, One over the absolute mean between event residual.



