

Nonmodel-based framework for rapid seismic risk and loss assessment of instrumented steel buildings

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Abstract

This paper proposes a nonmodel-based framework for estimating story-based engineering demand parameters (EDPs) in instrumented steel frame buildings with steel moment-resisting frames (MRFs). The proposed framework utilizes a wavelet-based damage-sensitive feature and basic building geometric information to infer the building damage state at a given seismic intensity. The story-based EDPs are predicted with a reasonable accuracy compared to those predicted from rigorous nonlinear response history analyses that typically require the explicit use of a nonlinear building model. The efficiency of the proposed framework is demonstrated through a number of illustrative examples including actual instrumented steel frame buildings that experienced the 1994 Northridge earthquake in Los Angeles. It is shown that if the building content is known the proposed framework can facilitate building-specific seismic risk and loss assessment within minutes after an earthquake provided that the recorded floor absolute acceleration histories at discrete locations along the height of the building are accessible. The nonmodel-based framework is also extended at the city-scale through the development of generalized earthquake-induced damage and loss maps for the same earthquake event. The same framework can facilitate the decision-making for effective pre-disaster measures for earthquake disaster risk management of building assets.

Keywords: Rapid seismic risk assessment, Instrumented steel buildings, Wavelet analysis, Earthquake-induced loss assessment, Generalized loss map, City-scale simulation, Disaster-risk management

1. Introduction

City-scale safety assessment in the aftermath of an earthquake is a challenging problem with vast socio-economic consequences. In that sense, visual inspections [1–4] have been historically employed. Such inspections may take months to complete and therefore earthquake-induced losses due to downtime can be considerable [1–4]. A number of researchers have utilized model-based approaches for the earthquake-induced structural and non-structural damage assessment (e.g., [5–11]). Although such approaches predict reasonably well story-based engineering demand parameters (EDPs) they typically require the explicit use of nonlinear building models and subsequently an appreciable

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8 time investment for the model validation. Therefore, such approaches cannot be used in the context of community
9 resilience in which a rapid seismic risk assessment of building assets is necessary.

10 Vibration-based condition assessment [12–18] is an interesting alternative compared to *model-based* approaches
11 conditioned that ambient vibration monitoring data is available and/or a seismic instrumentation program has been
12 already established within the earthquake-affected region [19–21]. A challenge in this case could be that a densely
13 arrayed sensing system may be required [22, 23]. Noh et al. [17, 18] proposed EDP indicators for *nonmodel-based*
14 seismic vulnerability assessment of steel frame buildings by observing the changes in wavelet energies at a particular
15 scale over time. In a more recent study, Hwang and Lignos [24] demonstrated that the wavelet-based damage-sensitive
16 features (DSFs) can facilitate the seismic vulnerability assessment of buildings with fairly low instrumentation density.
17 However, the challenge to establish the relationship between the wavelet energy shifting at a particular building
18 frequency and the story-based building EDP estimates still remains. These EDPs are essential to facilitate earthquake-
19 induced risk and loss assessment at a given seismic intensity.

20 In this paper, we propose a framework that facilitates the rapid seismic risk and loss assessment of instrumented
21 steel frame buildings. The proposed framework is based on a nonmodel-based approach that combines concepts
22 from structural health monitoring and performance-based earthquake engineering. The efficiency of the proposed
23 framework is evaluated in three parts. The first part illustrates comparisons between the proposed framework and
24 computationally intensive state-of-the-art approaches in predicting story-based building EDPs at a given seismic in-
25 tensity. The second part utilizes the proposed framework to conduct a rapid seismic risk and loss assessment of an
26 instrumented steel frame building that experienced the 1994 Northridge earthquake. The final part of the paper de-
27 scribes how the proposed framework can be extended at a city-scale in order to facilitate regional earthquake-induced
28 loss assessment for emergency-response operations in the aftermath of an earthquake and pre-disaster measures for
29 earthquake risk management.

30 **2. Proposed framework for performance-based rapid assessment of steel frame buildings**

31 This section presents a framework for performance-based rapid seismic vulnerability assessment of instrumented
32 steel frame buildings. The emphasis is currently on steel frame buildings with MRFs as their lateral load-resisting
33 system. Figure 1 illustrates schematically the main components of the proposed framework. The first step involves
34 the collection of basic information from the instrumented building of interest including the number of stories and
35 the building height. In terms of recorded data, the absolute acceleration at the roof of the building is only required.
36 Referring to Figure 1, the proposed framework requires the identification of the dynamic properties of the instrumented
37 building inferred to its undamaged state. This is further elaborated in Section 2.1.

38 Once the dynamic properties of the instrumented building are known, a refined wavelet-based DSF [17, 24] is
39 employed. The wavelet-based DSF is extracted from the measured absolute acceleration at the building roof. This is
40 discussed in detail in Section 2.2. Referring to Figure 1, the third component of the proposed framework includes the

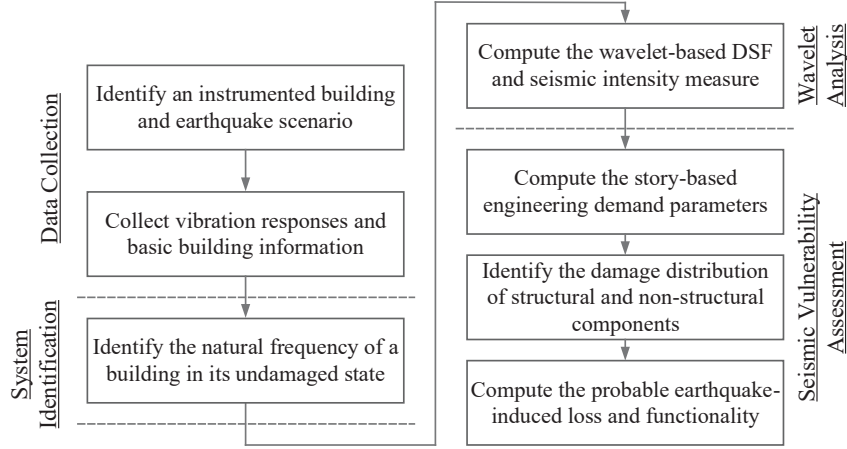


Figure 1: Flowchart of the proposed framework for rapid seismic risk and loss assessment of instrumented steel frame buildings.

41 mapping of the computed DSFs with story-based EDPs along the height of the instrumented building. This mapping
 42 is achieved through multivariate regression equations that relate the wavelet-based DSFs and basic geometric building
 43 characteristics with story-based EDPs at a given seismic intensity (see Section 2.3). Finally, the earthquake-induced
 44 economic losses are computed. Referring to Figure 1, depending on the density of the instrumented buildings within
 45 an earthquake-prone region a generalized loss map can be generated through the use of the geographic information
 46 system (GIS). The subsequent sections provide specific details of the main components of the proposed framework
 47 for seismic risk and loss assessment of instrumented steel frame buildings.

48 2.1. System identification

49 To compute the wavelet-based DSFs, the first natural frequency, f_1 of the undamaged state of a building is re-
 50 quired. For this purpose, the use of ambient vibrations obtained before the earthquake occurs is preferred if available.
 51 Output-only system identification methods can be utilized because of the absence of a measured input excitation [13].
 52 Otherwise, the last portion of the measured vibration recorded during a strong motion is considered as the vibration
 53 response of the undamaged instrumented building. The last portion is determined as the number of data points to
 54 provide best frequency resolution (0.097 Hz) for distinctive peaks in the power spectrum density diagram (PSD) es-
 55 timation by applying a fast Fourier transform with a 1024-point Hann window [25]. In this paper, the autoregressive
 56 with exogenous term (ARX) method [14] is utilized for this purpose. The ARX model uses least squares to estimate
 57 the dynamic properties of a multi-degree-of-freedom (MDF) system from recorded absolute acceleration data in the
 58 discrete time domain. This model is mathematically defined as follows,

$$\sum_{i=0}^M \mathbf{A}_i \mathbf{y}(n-i) = \sum_{i=0}^M \mathbf{B}_i \mathbf{x}(n-i) + \mathbf{e}(n) \quad (1)$$

59 in which M is the model order of the ARX model; $\mathbf{x}(n)$ and $\mathbf{y}(n)$ are the p -dimensional input and q -dimensional
60 output vectors, respectively; $\mathbf{e}(n)$ is the residue error vector; and \mathbf{A}_i and \mathbf{B}_i are $p \times p$ and $q \times p$ coefficient matrices of
61 the autoregressive (AR) polynomial and exogenous (X) input. The model in Eq. (1) may be re-written as follows,

$$\mathbf{y}(n) = -\sum_{i=1}^M \mathbf{A}_i \mathbf{y}(n-i) + \sum_{i=0}^M \mathbf{B}_i \mathbf{x}(n-i) + \mathbf{e}(n) = \Phi^T(n) \cdot \Theta + \mathbf{e}(n) \quad (2)$$

62 in which

$$\Phi^T(n) = \begin{bmatrix} -\mathbf{y}(n-1) & \cdots & -\mathbf{y}(n-M) & \mathbf{x}(n-1) & \cdots & \mathbf{x}(n-M) \end{bmatrix} \quad (3)$$

$$\Theta = \begin{bmatrix} \mathbf{A}_1 & \cdots & \mathbf{A}_M & \mathbf{B}_1 & \cdots & \mathbf{B}_M \end{bmatrix}^T \quad (4)$$

63 The parameter matrix, Θ can be estimated based on the least square method as follows,

$$\arg \min_{\Theta} J(\Theta) = \arg \min_{\Theta} \|\mathbf{y}(n) - \Phi^T(n) \cdot \Theta\|^2 \quad (5)$$

64 The AR coefficient and X input matrices are used to formulate the system matrix of equations. The dynamic
65 properties of a MDF system are estimated by eigenvalue decomposition for the system matrix [14].

66 Due to random noise, it is common that spurious modes are induced [12, 14]. In this case, a stable mode is
67 estimated by changing the ARX model order. A stabilization diagram [12, 14] is typically used for this purpose.
68 From this diagram, stabilization occurs when the relative differences of the dynamic properties identified using two
69 different model orders are not more than 5%, 10%, and 5% for the natural frequencies, the damping ratios, and
70 the modal assurance criterion (MAC) of mode shapes [12] (i.e., convergence thresholds), respectively. Figure 2
71 illustrates the stabilization diagram for the y loading direction of a 4-story steel frame building with moment-resisting
72 frames (MRFs) tested at the E-Defense facility [26, 27] estimated based on the single-input/three-output ARX method.
73 Referring to Figure 2, a relatively solid vertical line represents true modes. Moreover, the circled symbols represent
74 the values that were converged to the thresholds for the frequency, MAC and damping ratio.

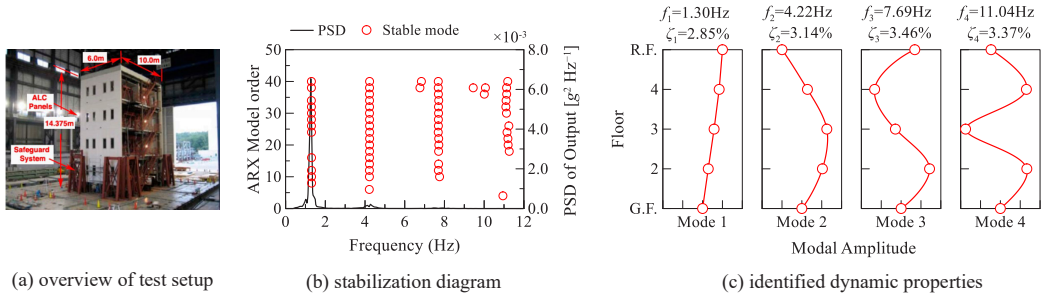


Figure 2: Stabilization diagram of the y loading direction of the 4-story steel frame building with MRFs tested at E-Defense facility.

2.2. Wavelet-based damage-sensitive features

In order to develop an approximate method for rapid earthquake vulnerability assessment of steel frame buildings with MRFs, a nonmodel-based approach is employed. In particular, wavelet-based DSFs are utilized as proposed in [17]. The wavelet-based DSFs are computed based on the absolute acceleration response history recorded at the building roof. The DSFs are then interpreted as story-based EDP indicators by monitoring the change in wavelet-based DSFs at a given seismic intensity. This section briefly describes the theoretical background of the wavelet-based DSF that is utilized in this paper. Given a scale parameter $a > 0$, and time shift parameter b , the continuous wavelet transform can be mathematically described as follows,

$$C(a, b) = \int_{-\infty}^{\infty} f(t) \frac{1}{\sqrt{a}} \psi^* \left(\frac{t-b}{a} \right) dt \quad (6)$$

in which $f(t)$ is the response history data (i.e., the absolute acceleration time history in this paper); $\psi(t)$ is the mother wavelet function (the Morlet wavelet basis function [28] is used as a mother wavelet due to its resemblance to earthquake pulse); and $*$ is the complex conjugate. A set of basis functions, which are termed as daughter wavelets, is established by continuously dilating and translating the mother wavelet function, $\psi(t)$. The continuous wavelet transform coefficients, $C(a, b)$ are then obtained by convoluting the basis functions and recorded absolute acceleration history data, $f(t)$ at the building roof.

Noh et al. [17] introduced the wavelet-based DSFs as structural damage indicators, which are defined as the ratio of the wavelet energy at the first-mode natural frequency of the building over time to the total wavelet energy. Hwang and Lignos [24] refined the wavelet-based DSF for cases that higher mode contributions become considerable. Therefore, the wavelet-based DSFs as proposed in [24] are utilized as follows,

$$DSF = 1 - \frac{\sum_{i=1}^3 E_{\text{scale}(f_i)}}{E_{\text{tot}}} \quad (7)$$

in which $E_{\text{scale}(f_i)}$ is the wavelet energy at a scale corresponding to the i th natural frequency of the building under consideration. Referring to Eq. (7), the DSF values represent how the distribution of vibration energies at the natural frequencies of a building changes while the structural damage progresses. Therefore, the DSF values range between 0 (representing no structural damage) and 1 (representing severe structural damage) as suggested in [17]. The DSF is a more rational damage indicator than the peak absolute acceleration at the building roof that more-or-less saturates once the building enters into the inelastic regime. The wavelet energy $E_{\text{scale}(f_i)}$ shows how the vibration energy of the acceleration response data is distributed over time given at a particular scale (i.e., the scale corresponding to i th natural frequency). By assuming that the natural frequencies of the building are well separated, the second- and third-mode natural frequencies are approximated at $3f_1$ and $5f_1$, respectively [29]. The wavelet energy, $E_{\text{scale}(f_i)}$ can then be computed as follows,

$$E_{\text{scale}(f_i)} = \sum_{b=1}^K |C(f_i, b \times \Delta t)|^2 \quad (8)$$

103 Referring to Eq. (7), the total wavelet energy, E_{tot} of the recorded absolute acceleration response history is defined
 104 as follows,

$$E_{\text{tot}} = \sum_{i=1}^3 E_{\text{scale}(f_i)} + E_{\text{scale}(0.5 \times f_1)} + E_{\text{scale}(2 \times f_1)} + E_{\text{scale}(4.5 \times f_1)} \quad (9)$$

105 Hwang and Lignos [24] conducted further validations on the potential use of wavelet-based DSFs for structural
 106 damage identification based on full-scale shake table experiments on steel frame buildings with steel MRFs and
 107 concentrically braced frames (CBFs). They found that when we consider the wavelet energies at scales that correspond
 108 to the first three natural frequencies of the respective building is typically enough to capture the higher mode effects on
 109 the building seismic response. The extra energy terms at $0.5/2/4.5 \times f_1$ in Eq. (9) are explicitly considered in order
 110 to capture a wavelet energy shift due to decrease in the first three natural frequencies of the respective building due to
 111 the onset of structural damage. Hwang and Lignos [24] also found that the mapping of a DSF value to the maximum
 112 EDP across all stories of a building is possible at a given seismic intensity when the absolute floor acceleration history
 113 at the roof of a building is recorded. This is illustrated in Figure 3 for the 4-story steel frame building with MRFs
 114 that was tested at the E-Defense shake table through collapse [26, 27]. The building was subjected to the JR Takatori
 115 ground motion from the 1995 Hygoken-Nanbu earthquake. Referring to Figure 3, the DSF values of the 4-story steel
 116 MRF building are plotted with respect to the four discrete seismic intensities. While the steel MRF building remained
 117 elastic (i.e., 20% and 40% of the JR Takatori record), the corresponding DSF values were on the order of 0.15 or less.
 118 For a design-basis seismic event, the peak story drift ratios were on the order of 2% along the height of the building;
 119 the corresponding DSF values were on the order of 0.20 to 0.30. For larger drifts, DSF values attained 1.0. Figure
 120 3 suggests that while the story-based EDPs of the building increase the corresponding DSF value tends from zero to
 121 one. The EDP-DSF relation is further explored in Section 2.3.2.

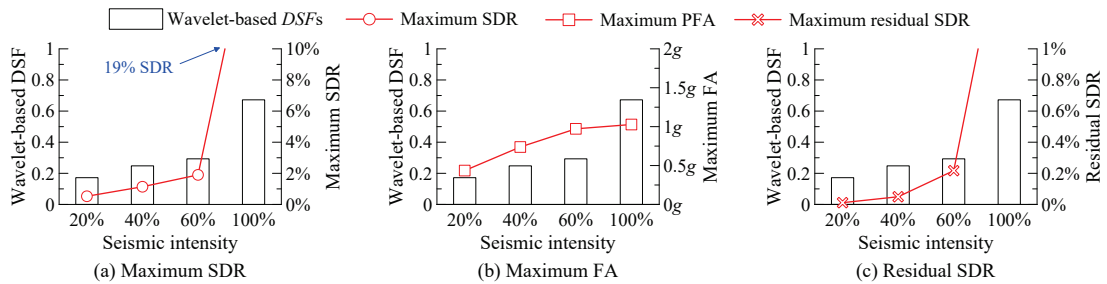


Figure 3: Wavelet-based DSF values in the y loading direction of the 4-story steel frame building with MRFs tested at E-Defense facility.

122 *2.3. Approximate method for computing story-based EDPs*

123 Figure 4 illustrates schematically how we developed the approximate method that maps the wavelet-based DSF
 124 with story-based EDPs at a given seismic intensity. Three phases of analysis are conducted: (i) nonlinear response
 125 history analyses (NRHAs) that employ nonlinear building models of representative steel MRF buildings to obtain
 126 simulated story-based EDPs and the roof absolute acceleration response histories of the respective building over a
 127 wide range of seismic intensities; (ii) wavelet analysis to obtain the wavelet-based DSFs (see Section 2); and (iii)
 128 stepwise multivariate linear regression analysis for the development of story-based EDP predictive models. The
 129 story-based EDPs of interest are the peak story drift ratio (SDR), residual SDRs and peak floor absolute accelerations
 130 (PFAs).

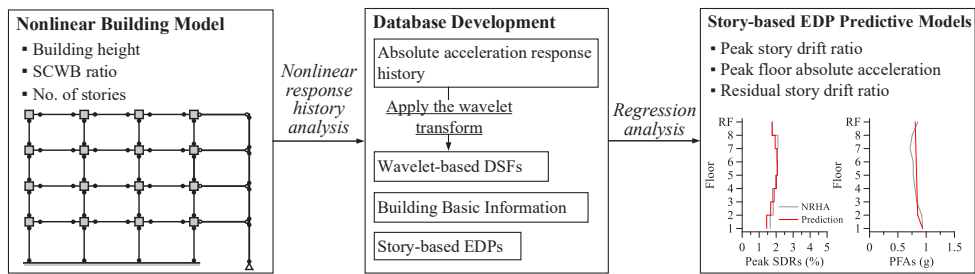


Figure 4: Flowchart for the development of the approximate method for story-based EDP computations.

131 *2.3.1. Building response database development*

132 In order to populate the best-suited damage indicators (i.e., wavelet-based DSFs) discussed later on in Section
 133 3.2 a wide range of archetype steel frame buildings with MRFs is considered. In particular, the archetypes range
 134 from 1 to 20 stories. Their steel MRFs are designed with three strong-column/weak-beam (SCWB) ratios of 1.0 (i.e.,
 135 code-based design), 1.5 and 2.0 in accordance with current seismic provisions in North America [30, 31]. Figure 5
 136 illustrates a plan view and elevation of a representative 8-story steel frame building with perimeter MRFs. Detailed
 137 information about the design details of the archetypes can be found in [32–34].

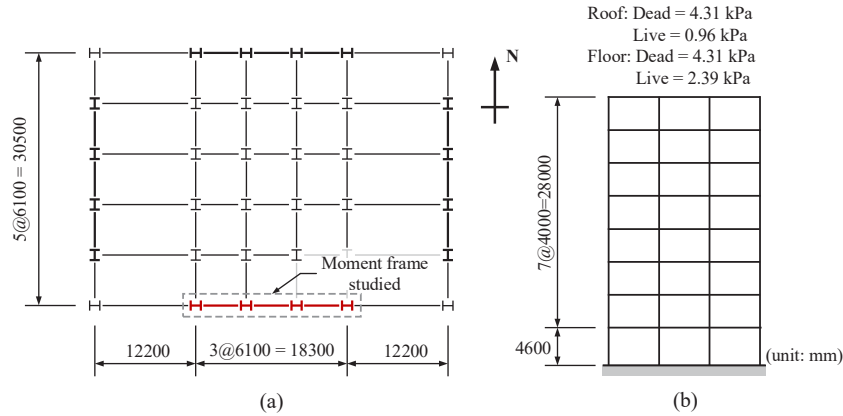


Figure 5: Typical 8-story archetype steel frame building: (a) plan view; and (b) elevation.

138 2.3.2. Nonlinear building models

139 Two-dimensional (2-D) nonlinear model representations of all the archetype MRFs in the east-west (E-W) loading
 140 direction are developed and implemented in the Open System for Earthquake Engineering Simulation (OPENSEES)
 141 Platform [35]. Referring to Figure 5(a), these MRFs are shown in the highlighted dashed box. The MRF steel
 142 beams and columns are modeled with elastic beam-column elements and lumped-plasticity flexural hinges at their
 143 ends as shown in Figure 6(a). The phenomenological model that was developed by Ibarra et al. [36] and refined
 144 and calibrated by [37, 38], is utilized to simulate cyclic and in-cycle strength and stiffness deterioration. Figure 6(b)
 145 illustrates a comparison between the simulated beam moment-chord rotation relations and the one deduced from a
 146 full-scale experiment conducted by Gilton et al. [39]. Referring to Figure 6(a), the nonlinear building model also
 147 considers explicitly the beam-to-column joint panel zone. The Krawinkler model [40] is employed for this purpose.
 148 Second order geometric effects are considered with a fictitious leaning column as shown in Figure 6(a). Damping is
 149 idealized with the Rayleigh model. Two percent damping ratio ($\zeta = 2\%$) is assigned to the first and third modes of
 150 all the nonlinear building models as discussed in Elkady and Lignos [33, 34].

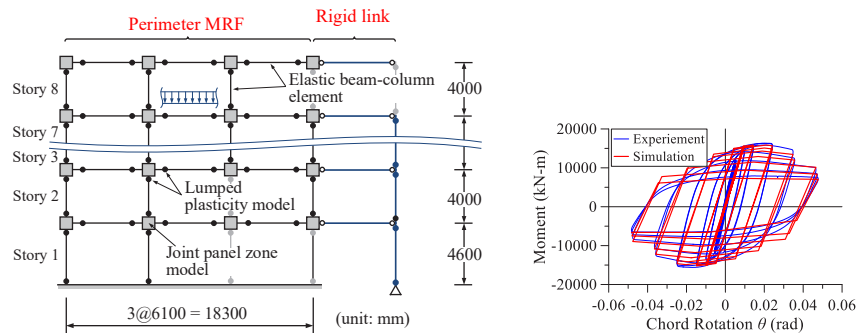


Figure 6: Example of nonlinear building model representation: (a) 2-D numerical model of the 8-story steel frame building; and (b) component deterioration model validation (data from Gilton et al. [39]).

151 Multiple NRHAs (i.e., incremental dynamic analysis, IDA) [41] are performed based on a suite of ground motions
 152 with large moment-magnitude, $6.5 \leq M_w \leq 7$ and short closest-to-fault-rupture distance, $13 \text{ km} < R_{rup} < 40 \text{ km}$ that
 153 represents the seismic hazard at the design location [42]. The story-based EDPs (i.e., peak SDRs, residual SDRs,
 154 and PFAs) are obtained for each ground motion over a wide range of seismic intensities. The wavelet-based DSFs
 155 are determined from the absolute acceleration response histories recorded at the roof of each archetype building as
 156 discussed in Section 2.2. Figure 7 shows the wavelet-based DSFs with respect to the peak SDRs of the 8-story
 157 archetype at stories 1, 2 and 8. In steel MRFs large story drift demands are expected to saturate near the bottom
 158 stories at low probability of occurrence earthquakes [40]. Referring to Figure 7, this is depicted by the employed DSF.
 159 A more rapid slope of the peak SDR – DSF relation indicates that in this particular story the structural damage is very
 160 evident. Same observations hold true for the rest of the story-based EDPs of interest but are not shown herein due to
 161 brevity.

162 The database of nonlinear building responses as well as the mapped wavelet-based DSF values are employed to
 163 develop empirical equations that predict the story-based EDPs of interest given a seismic intensity.

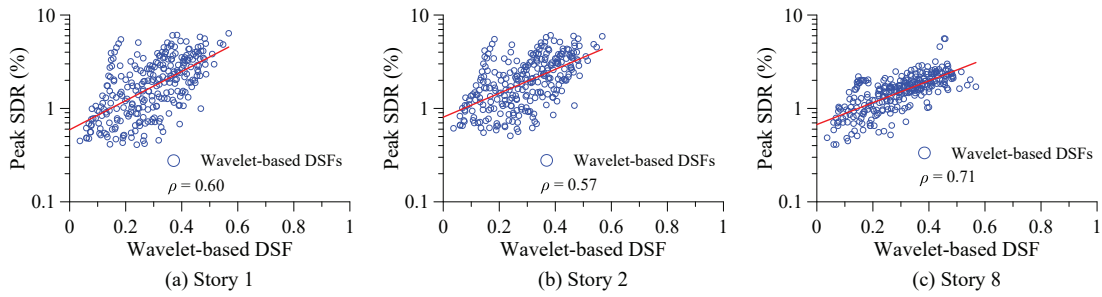


Figure 7: Scatter plots of wavelet-based DSF versus peak SDRs of the 8-story MRF building.

164 2.3.3. Proposed relationships for rapid earthquake-induced damage assessment

165 A stepwise multivariate linear regression analysis [43] is employed to establish *a priori* the relation between the
 166 building responses (i.e., story-based EDPs along the building height) and several predictor variables. These relation-
 167 ships are based on the building response database discussed in Section 2.3.2. The empirical equations consider only
 168 statistically significant predictor variables. Equation (10) represents the general functional form for estimating median
 169 story-based EDPs,

$$\begin{aligned}
 \ln(EDP_i) = & \beta_0 + \beta_1 \cdot \ln(IM) + \beta_2 \cdot (DSF) + \beta_3 \cdot (h_x/H) + \beta_4 \cdot (h_x/H)^2 + \beta_5 \cdot (h_x/H)^3 + \beta_6 \cdot (SCWB) + \beta_7 \cdot (N) \\
 & + \beta_8 \cdot (N)^2 + \varepsilon
 \end{aligned}
 \tag{10}$$

170 in which β_i are the regression constants; ε is the random error (i.e., residual); EDP_i are the corresponding peak SDRs,
 171 residual SDRs and PFAs at level i . Referring to Eq. (10), it was found that the average spectral acceleration S_{avg}

Table 1: Range of predictor variables for peak story-based EDPs based on the building response database.

	S_{avg} (g)	PGA (g)	h_x/H	SCWB	N
Minimum	0.02	0.05	0.05	1.00	2.00
Maximum	1.61	7.17	1.00	2.00	20.00
Mean	0.25	1.17	0.56	1.53	11.31
Standard deviation	0.18	0.97	0.29	0.41	5.67
COV	0.03	0.94	0.08	0.16	32.16

172 proposed by [44–46] provides best estimates of peak and residual SDRs compared to other seismic intensity measures
 173 (IMs) (e.g., pseudo-acceleration (S_a), peak ground acceleration (PGA), peak ground velocity (PGV), Arias intensity
 174 (I_a) [47]) that were examined. The S_{avg} is computed as the geometric mean of 5% damped spectral accelerations
 175 ranging between $0.2T_1$ and $3T_1$ with a uniform interval of 0.01s [46, 48] based on the horizontal component of
 176 excitation in the loading direction of interest; T_1 is the first-mode period of the building under consideration. Similarly,
 177 the PGA was found to provide better estimates for PFAs compared to other IMs that were examined. Notably, the
 178 FEMA P-58 simplified approach [9] utilizes the same IM for computing PFAs along the building height. In order
 179 to compute the S_{avg} and PGA after an earthquake an acceleration sensor should be placed at the ground floor of the
 180 building. If the base motion is not available, the output-only system identification method proposed by Lignos and
 181 Miranda [49] may be used to obtain the input ground motion. Furthermore, a number of alternatives to simplify Eq.
 182 (10) to the extent possible without sacrificing the prediction accuracy were considered. In certain cases, the DSF was
 183 excluded from the regression model; however, the corresponding functional form lead to a 45% under-prediction of
 184 EDPs at seismic intensities associated with a Design Basis and a Maximum Considered Earthquake.

185 Referring to Eq. (10), the wavelet-based DSF is determined from the absolute acceleration response history
 186 recorded at the building roof; h_x is the height above the base of the building to floor level x ; H is the total building
 187 height above the ground; $SCWB$ is the strong-column/weak-beam ratio determined by the year of building construction
 188 and regional seismic provisions; and N is the number of stories of the building under consideration. Equation (10)
 189 includes the term h_x/H to reflect the variability of EDPs as a function of the story. Table 1 summarizes the range of
 190 applicability of Eq. (10). The same table provides the minimum, maximum, mean, standard deviation, and coefficient
 191 of variation (COV) of the building response database.

192 To treat the statistical error and associated uncertainty in the regression model, t - and F -statistics are performed
 193 at a 5% significance level. Tables 2 and 3 summarize the intercepts, the regression coefficients and their p -values in
 194 the t -statistic and the F -statistic, the coefficient of determination R^2 and standard deviation σ_n for buildings with less
 195 than 8 stories and buildings with more than 9 stories, respectively. Referring to Tables 2 and 3, all the considered
 196 predictor variables significantly affect the accuracy of the models. In particular, the p -values in the t -statistic are zero.

197 Referring to Tables 2 and 3, the R^2 of the residual SDR is smaller than the corresponding values for the peak SDRs
 198 and PFAs. The residual SDR is influenced much more than other EDPs by the component modeling parameters and
 199 the record-to-record variability [8, 10, 50–53].

Table 2: Regression coefficients for story-based EDPs of steel frame buildings with less than 8 stories.

Predictor variables	Peak SDR			PFA			Residual SDR		
	Coefficient	<i>t</i> -statistic	<i>p</i> -value	Coefficient	<i>t</i> -statistic	<i>p</i> -value	Coefficient	<i>t</i> -statistic	<i>p</i> -value
$\ln(S_{avg})$	0.69	140.55	0.00	–	–	–	0.57	61.80	0.00
$\ln(PGA)$	–	–	–	0.63	213.30	0.00	–	–	–
<i>DSFs</i>	0.34	11.54	0.00	0.12	5.64	0.00	0.55	3.59	0.00
h_x/H	-1.75	39.38	0.00	-0.06	-9.73	0.00	0.22	2.71	0.01
$(h_x/H)^2$	-1.34	-36.95	0.00	–	–	–	-0.32	-4.82	0.00
$(h_x/H)^3$	–	–	–	–	–	–	–	–	–
<i>SCWB</i>	-0.09	-14.09	0.00	0.08	18.99	0.00	-0.09	-8.01	0.00
<i>N</i>	0.42	36.24	0.00	-0.21	-28.92	0.00	0.42	18.57	0.00
<i>N</i> ²	-0.04	-36.60	0.00	0.02	26.39	0.00	-0.04	-18.52	0.00
Intercept=-4.43, $F=6.24 \times 10^3$, p -value=0, $R^2=0.77$, $\sigma_{ln}=0.29$			Intercept=0.38, $F=1.56 \times 10^4$, p -value=0, $R^2=0.88$, $\sigma_{ln}=0.20$			Intercept=-5.22, $F=1.04 \times 10^3$, p -value=0, $R^2=0.39$, $\sigma_{ln}=0.53$			

Table 3: Regression coefficients for story-based EDPs of steel frame buildings with 9 to 20 stories.

Predictor variables	Peak SDR			PFA			Residual SDR		
	Coefficient	<i>t</i> -statistic	<i>p</i> -value	Coefficient	<i>t</i> -statistic	<i>p</i> -value	Coefficient	<i>t</i> -statistic	<i>p</i> -value
$\ln(S_{avg})$	0.52	135.86	0.00	–	–	–	0.32	0.32	0.00
$\ln(PGA)$	–	–	–	0.61	270.96	0.00	–	–	–
<i>DSFs</i>	0.76	33.20	0.00	0.34	20.80	0.00	0.62	16.53	0.00
h_x/H	3.89	44.40	0.00	-0.96	-53.05	0.00	0.78	15.49	0.00
$(h_x/H)^2$	-7.16	-39.03	0.00	0.83	51.46	0.00	-0.99	-21.89	0.00
$(h_x/H)^3$	3.82	34.27	0.00	–	–	–	–	–	–
<i>SCWB</i>	-0.10	-19.77	0.00	0.03	9.68	0.00	-0.07	-8.08	0.00
<i>N</i>	0.11	25.85	0.00	-0.07	-27.82	0.00	0.14	20.32	0.00
<i>N</i> ²	-0.004	-25.11	0.00	0.002	23.44	0.00	-0.005	-19.81	0.00
Intercept=-4.49, $F=5.55 \times 10^3$, p -value=0, $R^2=0.76$, $\sigma_{ln}=0.32$			Intercept=0.37, $F=2.84 \times 10^4$, p -value=0, $R^2=0.89$, $\sigma_{ln}=0.19$			Intercept=-5.51, $F=1.11 \times 10^3$, p -value=0, $R^2=0.23$, $\sigma_{ln}=0.53$			

200 Referring to Figure 8, indicative box whisker plots of the residuals against predictor variables are shown for steel
 201 frame buildings with less than 8 stories; the centerline and the bottom/top of the box represent the median and the
 202 25th/75th percentiles of the residuals against predictor variables. The box whiskers show the 10th/90th percentile
 203 information of the grouped data. From this figure, the box whisker plots do not suggest any explicit dependency of the
 204 residuals on each predictor variable. In particular, the residuals fall within a horizontal band around zero. Furthermore,
 205 the normal probability plot (i.e., Q-Q plot [43]) shows that the residual points lie approximately on a straight line. A
 206 slight deviation from the straight line is only observed in the tails of the Q-Q plot. Therefore, the distribution of the
 207 residuals may be regarded as normal [43] with a zero mean and a standard deviation found from the residual analysis.
 208 Same observations hold true for steel frame buildings with 9 to 20 stories.

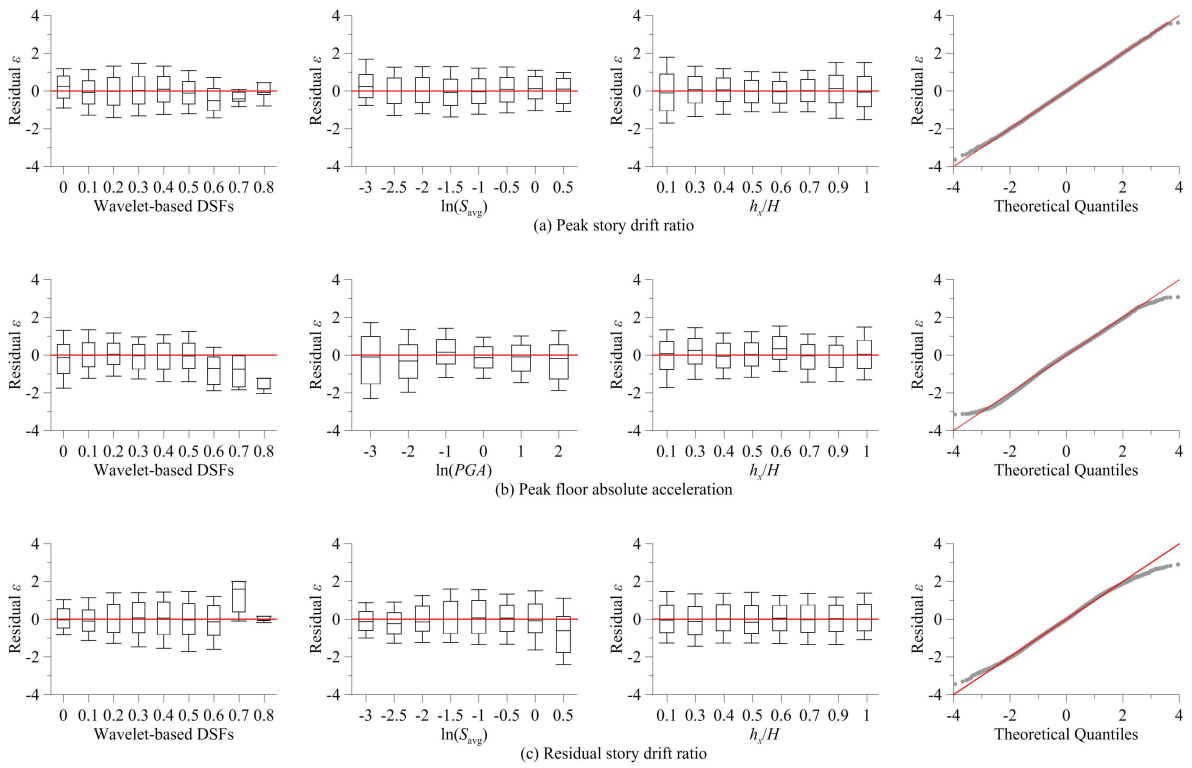


Figure 8: Diagnostic residual plots for story-based EDPs of steel frame buildings with 8 stories or less.

209 The efficiency of Eq. (10) in predicting story-based EDPs is further illustrated for selected seismic intensities
 210 that represent hazard levels of interest to the engineering profession: namely (i) a service-level earthquake (i.e., SLE:
 211 seismic hazard level of 50% probability of exceedance in 50 years); (ii) a design-basis earthquake (i.e., DBE: seismic
 212 hazard level of 10% probability of exceedance in 50 years); and (iii) a maximum considered earthquake (i.e., MCE:
 213 seismic hazard level of 2% probability of exceedance in 50 years). Figures 9 and 10 show the box-whisker plot of
 214 simulated-to-predicted EDP ratios (i.e., residuals) at the three seismic hazard levels of interest for an 8- and a 12-story
 215 steel frame building, respectively. Referring to Figures 9 and 10, the centerline and the box edges indicate the median

216 and the 25th/75th percentiles of the predicted story-based EDPs. The box whiskers extend to the 10th/90th percentiles
 217 of the simulated-to-predicted EDP ratio. From the same figures, Eq. (10) slightly overestimates the peak SDRs and
 218 PFAs in the lower stories of both buildings at the SLE seismic intensity (i.e., approximately 30% to the maximum
 219 relative to NRHA results). It is evident that the predictive equations provide reasonable estimates of peak SDRs and
 220 PFAs along the height of the steel frame buildings at the DBE and MCE seismic intensities.

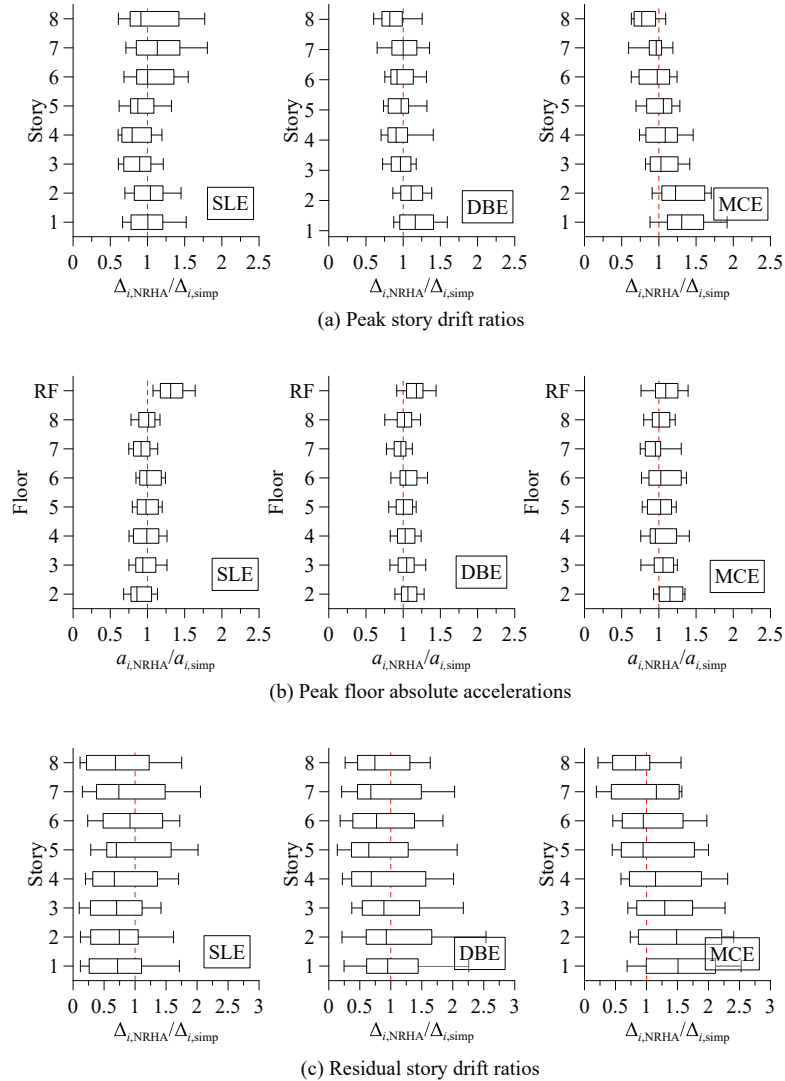


Figure 9: Diagnostic residual plots for the 8-story steel frame building designed with $SCWB \geq 1.0$.

221 Referring to Figures 9(c) and 10(c), Eq. (10) generally tends to overestimate the median residual SDRs (i.e.,
 222 $\Delta_{i,NRHA}/\Delta_{i,simp} \leq 1.0$). This is mainly attributed to the influence of the record-to-record variability and input model
 223 parameters on the residual drift demands along the height of a building [10, 54]. The median values of simulated-to-
 224 predicted residual SDR ratio for steel frame buildings of interest are between 0.5 and 1.75. Same observations hold

225 true for the rest of the cases included in the building response database discussed in Section 2.3.2.

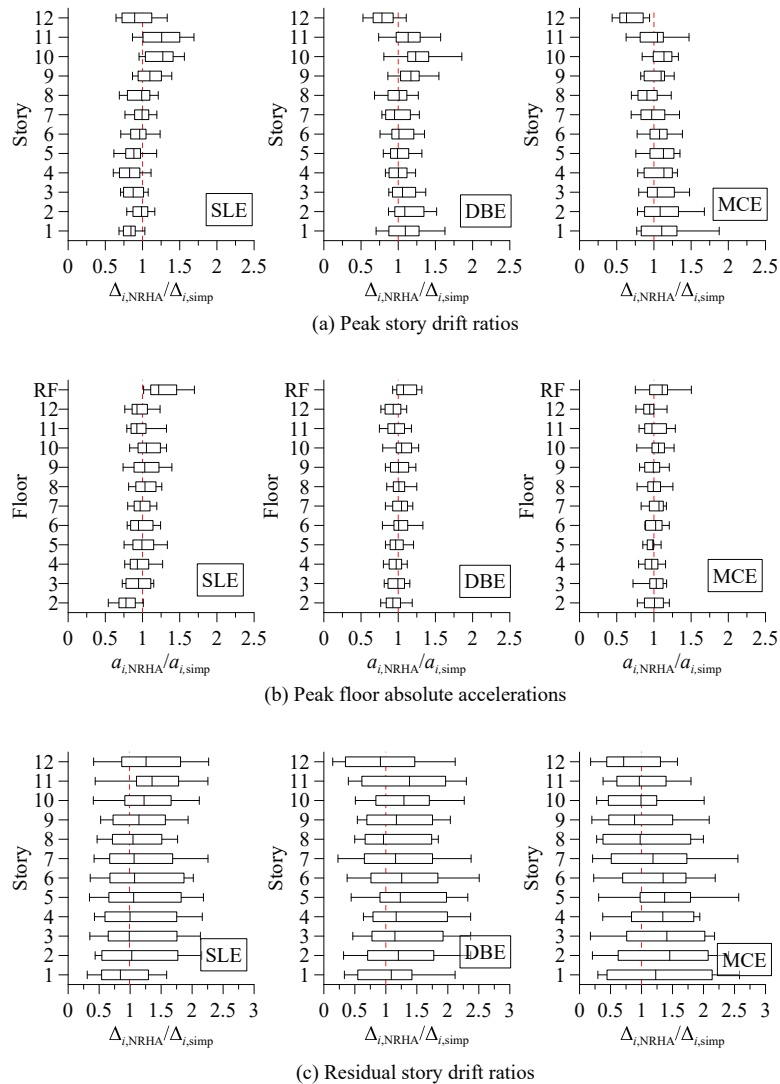


Figure 10: Diagnostic residual plots for the 12-story steel frame building designed with $SCWB \geq 1.0$.

226 2.3.4. Comparisons of proposed nonmodel-based method with available predictive models

227 In this section, the efficiency of the proposed nonmodel-based method in predicting story-based EDPs in steel
 228 frame buildings with MRFs is evaluated with regards to results obtained from rigorous NRHA. For this purpose, the
 229 8-story steel frame building utilized in Section 2.3.3 is employed. This building is subjected to the far-field set of 44
 230 ground motions retrieved from FEMA P695 [55]. For comparison purposes the FEMA P-58 simplified procedure [9]
 231 is also considered. This procedure requires an explicit building model for the story-based EDP computations. This
 232 model should appropriately represent the distribution of mass and stiffness along the height of the building. On the
 233 basis of the FEMA P-58 simplified approach [9] we utilized (i) linear and nonlinear building models; (ii) an elastic

234 analysis based on a first-mode lateral force distribution; and (iii) an estimate of the building's lateral yield strength.
235 This was computed through nonlinear static analysis based on a first-mode lateral load pattern.

236 Figure 11 depicts the predicted peak SDRs along the height of the 8-story steel frame building based on the
237 proposed method in comparison with the median peak SDR demands from NRHA for three seismic hazard levels
238 (i.e., SLE, DBE and MCE) as defined at the design location of interest. In the same figure, we have superimposed the
239 predicted median peak SDRs based on the FEMA P-58 simplified procedure [9]. To facilitate a lower/upper bound
240 analysis, the 16th/84th percentiles of peak SDRs, PFAs and residual SDRs are provided in the same figures.

241 Referring to Figure 11(a), it is evident that the proposed predictive equations provide reasonable estimates of peak
242 SDRs regardless of the seismic intensity. Notably, the only input information that is required is the building height,
243 the employed SCWB ratio and the computed wavelet-based DSF based on the absolute acceleration response history
244 at the roof of the 8-story building. Figure 11(a) suggests that the proposed method provides better estimates of median
245 peak SDR demands compared to those obtained from the FEMA P-58 simplified procedure regardless of the seismic
246 intensity of interest. In particular, at the MCE seismic intensity, the differences of the predicted peak SDRs relative
247 to those determined by NRHA are on average, 12% and 23% based the proposed nonmodel method and the FEMA
248 P-58 simplified procedure, respectively. Referring to Figure 11(a), the FEMA P-58 simplified procedure significantly
249 overestimates the peak SDRs in the upper stories of the 8-story building at the MCE intensity. This approach is not
250 applicable when peak SDRs exceed four times the corresponding yield drift ratio and/or excessive deterioration in
251 strength and stiffness of structural components occurs [9]. On the other hand, the proposed nonmodel-based approach
252 predicts well the peak SDR demands over the building height for the same seismic intensity.

253 Referring to Figure 11(b), the predicted median PFAs along the height of the 8-story steel frame building are
254 shown for the three selected levels of seismic intensity. Superimposed in the same figure is the median PFA demands
255 from NRHA. From this figure, it is found that the proposed method provides reasonable PFA estimates along the
256 height of the building for moderate to severe seismic intensities (i.e., DBE and MCE). At frequently occurring seismic
257 intensities (i.e., SLE), the proposed approach seems to underestimate PFAs by approximate 16%, on average, relative
258 to NRHA results. Similar accuracy is achieved with the FEMA P-58 simplified approach.

259 Similarly, Figure 11(c) compares the predicted residual SDRs along the height of the 8-story steel frame building
260 based on the proposed approach and the FEMA P-58 simplified approach for the three seismic intensities. In the
261 same figure, we have superimposed the median response based on NRHA. Referring to Figure 11(c), the proposed
262 approach tends to slightly overestimate the residual SDRs at the mid-height of the building whereas the FEMA P-58
263 simplified approach tends to underestimate the residual SDRs at the bottom stories of the building. This is not the
264 case at higher seismic intensities associated with low probability of occurrence earthquakes (i.e., MCE). Previous
265 research has identified that residual drifts are highly variable and very sensitive to the earthquake magnitude, distance
266 to the source range, the adopted component hysteretic behavior as well as the analytical model representations of a
267 building [10, 52, 56, 57]. For the aforementioned reasons, it is recommended that a lower/upper bound analysis should
268 be employed based on the 16th/84th percentile of the predicted values of the proposed approach as shown in Figure

269 11(c). In this case, the median response from NRHA is within these two percentiles. Same observations hold true for
 270 buildings with 9 to 20 stories but are not shown here due to brevity.

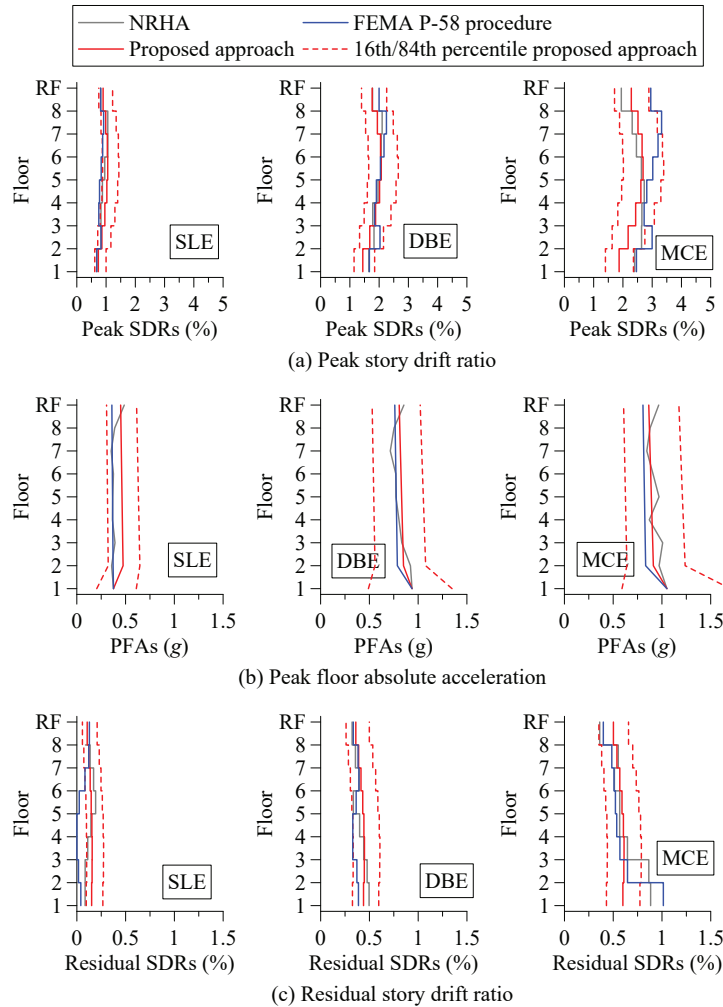


Figure 11: Predicted versus simulated story-based EDPs along the height of the 8-story steel frame building designed with $SCWB \geq 1.0$.

271 3. Application of simplified seismic assessment methodology in instrumented steel frame buildings

272 3.1. Case study instrumented building

273 The proposed approach could be employed for the rapid seismic assessment of instrumented steel frame buildings
 274 with fairly low instrumentation density. In particular, two sensors at each principal axis are only required along the
 275 height of the respective building. Preferably, one sensor should be placed at the base to obtain the peak ground
 276 acceleration; and the second one should be placed at the building roof to obtain the wavelet-based DSF. If sensors in
 277 other locations of the building are available then the output-only system identification technique proposed by Lignos
 278 and Miranda [49] could be employed to obtain the absolute acceleration histories at the base and roof of the building

279 of interest. In that respect, the proposed approach may significantly reduce the cost of installation, operation, and
 280 computation of a large volume of sensors and data.

281 The proposed approach is evaluated with the use of recorded data from an instrumented 15-story steel frame
 282 building that was located in Los Angeles, California (34.058°N, 118.250°W) and experienced the 1994 Northridge
 283 earthquake. Its lateral load resisting system consists of steel MRFs. The building was designed in 1961. Therefore,
 284 capacity design principles were not formally employed. However, the design of similar buildings was mostly governed
 285 by lateral wind loads over seismic loads, and the member properties were determined based on wind demands; there-
 286 fore, they were detailed to behave in a ductile manner [58]. Considering the large column sizes that were typically
 287 employed to satisfy the axial and lateral drift limits in tall buildings, the column flexural strength was not deemed to
 288 be critical [58]. Therefore, the building had a SCWB ratio close to 1.0. The existence of the concrete shear walls up to
 289 the first story of the steel frame building did not seem to influence the seismic behavior of the steel frame building; in
 290 that sense, the lateral load resistance is primarily provided by the steel MRFs. Fifteen accelerometers were installed
 291 at four levels along the height of the building. The recorded data was retrieved from the Center for Engineering
 292 Strong Motion Data (CESMD) operated by the California Department of Conservation's Strong Motion Instrumen-
 293 tation Program (CSMIP) in cooperation with the US Geological Survey (USGS). The building station number was
 294 CSMIP 24569. The plan view and elevation of the building is shown in Figure 12.

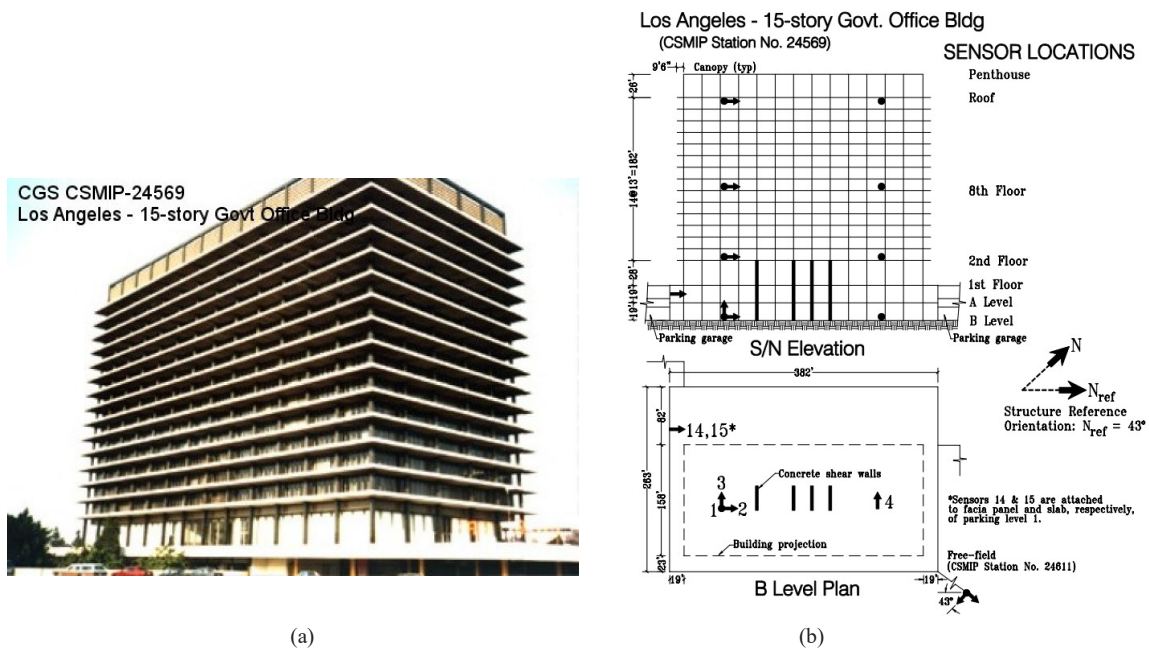


Figure 12: Fifteen-story Government steel frame office building (CSMIP 24569); (a) overview; and (b) plan, and elevation view of the building (images from the US National Center for Engineering Strong Motion Data at <http://strongmotioncenter.org>).

295 3.2. Predicted engineering demand parameters and earthquake-induced economic losses

296 To determine the wavelet-based DSFs to be used in Eq. (10), the first-mode frequency f_1 of the building is
 297 identified in its two orthogonal loading directions based on the input base motion and the output absolute acceleration
 298 response histories recorded at the building roof during the earthquake based on the ARX method [14] (see Section
 299 2.1). Table 4 summarizes the identified natural frequencies of the first two building modes in the two horizontal
 300 loading directions. The equivalent damping ratios, ζ_{eq} for the two modes per loading direction are also identified. The
 301 wavelet-based DSF is then determined from the recorded absolute acceleration response history at the building roof.

Table 4: System identification for the 15-story instrumented steel frame office building in Los Angeles.

Loading direction	Mode	Natural frequency f (Hz)	Equivalent damping ratio, ζ_{eq} (%)
North-South	1st	0.34	2.2
	2nd	0.92	3.8
East-West	1st	0.32	3.4
	2nd	0.90	2.3

302 Figure 13 shows the estimated story-based EDPs of the 15-story building for both loading directions [i.e., North-
 303 South (NS) and East-West (EW) directions]. These EDPs are computed within few seconds based on Eq. (10)
 304 and the corresponding values from Table 3. Referring to Figure 13(c), the recorded PFAs at three floor levels are
 305 superimposed for comparison purposes. It is found that the proposed nonmodel-based approach provides accurate
 306 estimates of the building's PFAs. Referring to Figures 13(a) and 13(b), the proposed approach predicts that the peak
 307 SDR and residual SDR along the height of the same building is 1.30% and 0.35%, respectively. This occurred around
 308 the mid-height of the building in both loading directions of interest. Therefore, the building experienced fairly minor
 309 structural damage due to flexural steel beam yielding. Notably, the FEMA P-58 simplified approach cannot be directly
 310 utilized for the seismic performance assessment of the same building because the building geometry as well as the
 311 material properties of the respective structural components should be known upfront. Furthermore, a considerable
 312 time investment is needed for the nonlinear building model development and validation.

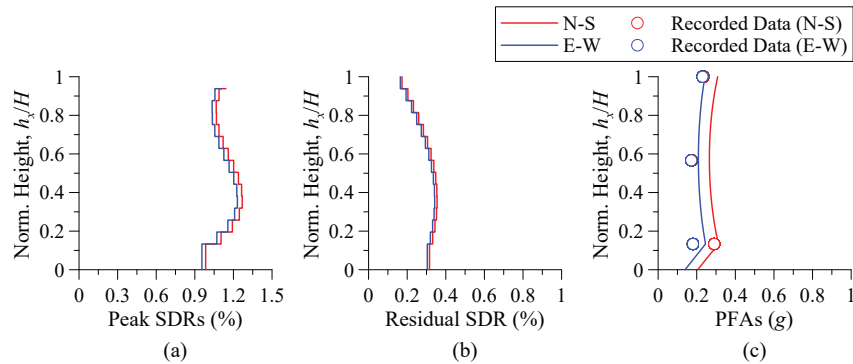


Figure 13: Predicted story-based EDPs for the 15-story Government steel frame office building (CSMIP 24569).

313 The predicted story-based EDPs can be further utilized to conduct a probabilistic building-specific economic loss
 314 assessment. The story-based building-specific loss estimation methodology proposed by Ramirez and Miranda [59]
 315 can be employed for this purpose. The three possible consequences of a building in the aftermath of an earthquake are
 316 considered as follows: (i) collapse does not occur and structural and/or non-structural components shall be repaired
 317 or replaced after the earthquake; (ii) collapse does not occur, but the building may be demolished and rebuilt due
 318 to excessive residual deformations; and (iii) collapse occurs and the building shall be rebuilt. Assuming that these
 319 consequences are mutually exclusive, the expected building losses conditioned on the seismic intensity IM are defined
 320 as follows,

$$E[L_T|IM] = E[L_T|NC \cap R, IM] \cdot P(NC \cap R|IM) + E[L_T|NC \cap D, IM] \cdot P(NC \cap D|IM) + E[L_T|C, IM] \cdot P(C|IM) \quad (11)$$

321 in which $E[L_T|NC \cap R, IM]$ is the expected total repair cost given that collapse does not occur and the building may
 322 be repaired at a given seismic intensity $IM = im$; $E[L_T|NC \cap D, IM]$ is the expected building loss when there is no
 323 collapse but the building may be demolished at a given seismic intensity $IM = im$; $E[L_T|C, IM]$ is the total replacement
 324 cost of the building when collapse occurs at a given seismic intensity $IM = im$, because the building needs to be
 325 replaced in this case. Furthermore, $P(NC \cap R|IM)$ is the probability that the building will not collapse but may be
 326 repaired or replaced conditioned on the seismic intensity $IM = im$; $P(NC \cap D|IM)$ is the probability that the building
 327 will not collapse but it may be demolished because of potentially large residual deformations conditioned on the
 328 seismic intensity $IM = im$; $P(C|IM)$ is the probability of collapse conditioned on the seismic intensity $IM = im$. The
 329 expected total repair cost conditioned on the building not collapsing $E[L_T|NC \cap R, IM]$ is defined as follows,

$$E[L_T|NC \cap R, IM] = \sum_{i=1}^m \sum_{j=0}^n \int_0^{\infty} E[L_{ij}|DS_{ij}] P_{DS_{ij}|EDP} f_{EDP|IM} dEDP \quad (12)$$

330 in which m is the number of damageable components being considered; n is the number of damage states a component
 331 may experience; $E[L_{ij}|DS_{ij}]$ is the mean repair cost for the i th component being in the j th damage state; $P_{DS_{ij}|EDP}$
 332 is the probability of the EDP of interest associated with the i th component being in the j th damage state given an
 333 $EDP = edp$. The probability of having to demolish the building conditioned on the seismic intensity $P(NC \cap D|IM)$ is
 334 modeled by a lognormal distribution with a median $\mu_{D|RSDR}$ of 0.015 radians and a logarithmic standards deviation
 335 $\beta_{\ln D|RSDR}$ of 0.3 as proposed in [59]. More details about the mathematical formulation of the building-specific loss
 336 estimation methodology can be found in [59].

337 In order to reliably quantify the earthquake-induced losses of the instrumented building, the authors adopted a
 338 library of fragility curves of building components from FEMA P-58 [9]. Some of these curves were further refined
 339 by [54, 57]. The employed component fragility curves for the list of damageable components of the instrumented
 340 steel frame building are all listed in Table 5. Note that the suspended ceiling system in the instrumented building
 341 was assumed to be a typical US style with acoustic tiles, that is composed with both vertical and lateral supports,

342 grid members, boundary wall molding, and diagonal brace wires [60, 61]. The fragility curves for the elevator were
 343 assumed to be those of hydraulic elevators that were mostly installed in California in 1976 or later.

344 Although the building plan view was known (see Figure 12), its detailed architectural layout was not possible to
 345 be retrieved. The authors approximated the densities of various non-structural components and building content as
 346 discussed in Bradley et al. [65] in which the quantity of each non-structural component per square meter is determined
 347 according to the Mitrani-Reiser [66] methodology.

348 The fragility curves developed by Ramirez et al. [62] were adopted for pre-Northridge beam-to-column moment
 349 connections. An example of such curves is shown in Figure 14(a) for steel beams in pre-Northridge beam-to-column
 350 connections made of A36 steel. Referring to Figure 13(a), the probability of beam yielding at a peak SDR of 1.30% is
 351 more than 70%. The probability of premature fracture is slightly above 20% for the same peak SDR. For comparison
 352 purposes, the fragility curves of typical post-Northridge fully-restrained beam-to-column connections [67] are also
 353 shown in Figure 14(a). In this case, the likelihood of ductile fracture due to low-cycle fatigue is practically zero.

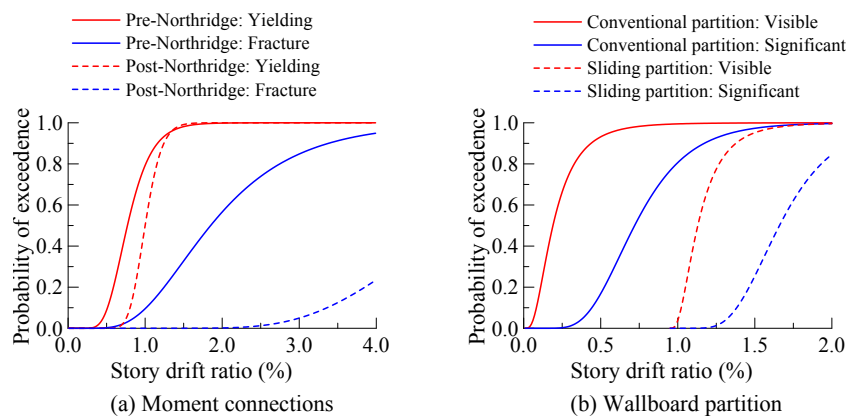


Figure 14: Fragility curves for typical pre- and post-Northridge fully-restrained beam-to-column connections and conventional and sliding gypsum wallboard partitions.

354 Figure 14(b) illustrates the fragility curves of conventional gypsum wallboard partitions [63] employed in this pa-
 355 per. The fragility curves of improved partitions [63] (i.e., sliding wallboard partitions that accommodate the expected
 356 SDR demand) are also shown in the same figure for comparison purposes. Referring to Figure 14(b), at a given SDR
 357 of 0.5%, the probability that the conventional wallboard partitions are in the minor damage state (i.e., visible) is over
 358 70%. In this case the wallboard partitions can be repaired by means of patching, re-taping, sanding and painting the
 359 gypsum wallboard [63]. On the other hand, the probability of having visible damage in the improved partitions is zero
 360 for the same SDR. Table 5 also summarizes the repair cost per damage state for each building component including
 361 the respective fragility curve obtained from prior studies [9, 56, 62–64].

362 Figure 15 shows the expected earthquake-induced losses of the 15-story Government steel frame office building.
 363 The expected losses are normalized with respect to the total replacement cost of the building. Note that the total

Table 5: Fragility and cost estimates for the 15-story Government steel frame office building.

Assembly description	Damage state	Unit	Fragility parameters			Repair cost
			EDP	x_m	β	x_m (\$)
Columns base ($W < 223\text{kg/m}$) ([9])	Crack initiation	EA	SDR	0.04	0.40	19,224
	Crack propagation			0.07	0.40	27,263
	Fracture			0.10	0.40	32,423
Columns base ($223\text{kg/m} < W \leq 446\text{kg/m}$) ([9])	Crack initiation	EA	SDR	0.04	0.40	20,082
	Crack propagation			0.07	0.40	29,395
	Fracture			0.10	0.40	36,657
Columns base ($W > 446\text{kg/m}$) ([9])	Crack initiation			0.04	0.40	21,363
	Crack propagation	EA	SDR	0.07	0.40	32,567
	Fracture			0.10	0.40	41,890
Column splices ($W < 223\text{kg/m}$) ([9])	Crack initiation	EA	SDR	0.04	0.40	9446
	Crack propagation			0.07	0.40	11,246
	Fracture			0.10	0.40	38,473
Column splices ($223\text{kg/m} < W \leq 446\text{kg/m}$) ([9])	Crack initiation	EA	SDR	0.04	0.40	10,246
	Crack propagation			0.07	0.40	13,012
	Fracture			0.10	0.40	42,533
Column splices ($W > 446\text{kg/m}$) ([9])	Crack initiation	EA	SDR	0.04	0.40	11446
	Crack propagation			0.07	0.40	14,812
	Fracture			0.10	0.40	47,594
Column ($\leq W27$) ([9])	LB	EA	SDR	0.03	0.30	16,033
	LTB			0.04	0.30	25,933
	Fracture			0.05	0.30	25,933
Column ($\geq W30$) ([9])	LB	EA	SDR	0.03	0.30	17,033
	LTB			0.04	0.30	28,433
	Fracture			0.05	0.30	28,433
Pre-Northridge moment connection (A36 steel, one-sided, $\leq W27$) ([62])	Yielding	EA	SDR	0.0077	0.32	0
	Fracture	EA		0.0185	0.47	11,980
Pre-Northridge moment connection (A36 steel, one-sided, $\geq W30$) ([62])	Yielding	EA	SDR	0.0077	0.32	0
	Fracture	EA		0.0185	0.47	12,313
Pre-Northridge moment connection (A36 steel, two-sided, $\leq W27$) ([62])	Yielding	EA	SDR	0.0077	0.32	0
	Fracture	EA		0.0185	0.47	16,653
Pre-Northridge moment connection (A36 steel, two-sided, $\geq W30$) ([62])	Yielding	EA	SDR	0.0077	0.32	0
	Fracture	EA		0.0185	0.47	16,653
Shear tab connections ([9])	Yielding	EA	SDR	0.04	0.40	12,107
	Partial tearing			0.08	0.40	12,357
Corrugated slab (90mm steel; 100mm overlay) ([56])	Complete separation			0.11	0.40	12,307
	Crack initiations	m^2	SDR	0.00375	0.13	180
	Crushing near column			0.01	0.22	330
	Shear stud fracture			0.05	0.35	570

Table 5: Fragility and cost estimates for the 15-story Government steel frame office building (continued).

Assembly description	Damage state	Unit	Fragility parameters			Repair cost
			EDP	x_m	β	x_m (\$)
Wallboard partition ([63])	Visible	6m ²	SDR	0.0019	0.65	90
	Significant			0.0072	0.38	530
Wallboard partition finish ([63])	Visible	6m ²	SDR	0.0019	0.65	90
	Significant			0.0072	0.38	250
Exterior glazing ([64])	Crack	pane	SDR	0.04	0.36	440
	Fallout			0.046	0.33	440
Suspended ceiling ($A > 232\text{m}^2$) ([9])	5% tiles dislodge	232m ²	PFA (g)	0.35	0.40	3542
	30% tiles dislodge			0.55	0.40	29,337
	Collapse			0.80	0.40	55,200
Automatic sprinklers ([64])	Fracture	3.66m	PFA (g)	0.32	1.40	900
Elevator ([9])	Failure	EA	PGA (g)	0.50	0.28	3180

EDP, engineering demand parameter; LB, local buckling; LTB, lateral-torsional buckling; SDR, story drift ratio (unitless); PFA, peak floor absolute acceleration (g); PGA, peak ground acceleration (g); x_m , median value; β , lognormal standard deviation.

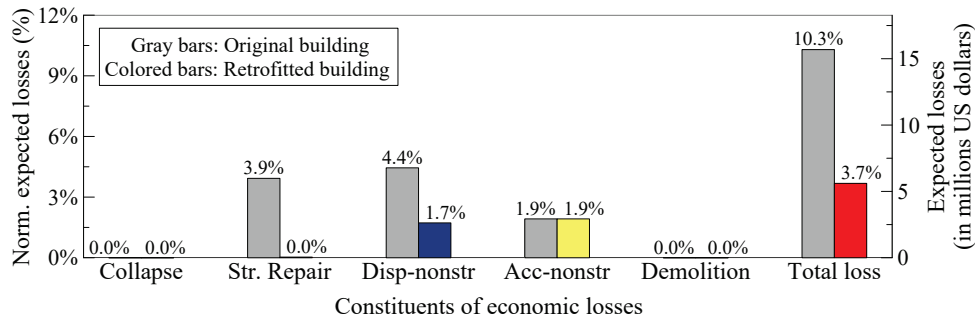


Figure 15: Normalized expected losses for 15-story Government steel frame office building (CSMIP 24569).

364 replacement cost of the building is determined for a given calendar year (i.e., 1994). Referring to Figure 15, the
 365 expected losses due to repairs slightly exceed 10% of the total replacement cost of the building. These losses are
 366 further disaggregated into structural/non-structural component repairs, building demolition, and collapse. For the
 367 given seismic intensity at the site of interest, losses due to collapse and demolition become negligible. This is to be
 368 expected given the amplitude of peak and residual SDRs along the height of the building [see Figure 13(a)]. However,
 369 drift-sensitive non-structural component repairs seem to be the major contributor to the expected building losses. On
 370 the other hand, the expected losses due to acceleration-sensitive non-structural component repairs are approximately
 371 2.0% of the total replacement cost of the building. This seems to be a reasonable estimate based on the recorded
 372 maximum PFAs (i.e., 0.29g) along the height of the building. These PFA demands are much lower than the minimum
 373 seismic force limit (i.e., 0.5g) on acceleration-sensitive non-structural components at the design site [31].

374 Referring to Figure 15, the estimated losses due to structural damage are on the order of 4.4% of the total re-

375 placement cost of the building. Although the peak SDRs along the height of the building did not exceed 1.3% in
 376 both loading directions, repairs due to structural damage are primarily driven by the increased likelihood of premature
 377 fracture of pre-Northridge beam-to-column connections [62, 68]. To better “digest” these numbers, Figure 15 shows
 378 the normalized expected losses of the same building if it were to be built today with fully-restrained beam-to-column
 379 connections that utilize beams with reduced beam section (RBS) [67] and gypsum wallboard partitions that allow for
 380 sliding [63]. In this case, it is evident that the expected economic losses of the instrumented steel building would be
 381 reduced by a factor of 3, if it would be retrofitted with RBS connections and improved gypsum wallboard partitions.

382 3.3. Rapid Seismic Assessment at a “City-scale”

383 The proposed framework discussed in Section 2 offers the opportunity to conduct a rapid seismic risk and loss
 384 assessment of instrumented buildings at a “city-scale” for a given earthquake scenario. This concept is explored
 385 further by utilizing recorded data from an array of instrumented steel frame buildings with MRFs that experienced
 386 the 1994 Northridge earthquake. The recorded data were available through the CSMIP stations for the city of Los
 387 Angeles. Although the data is fairly scarce due to the density of the instrumented buildings at that time, the intention
 388 of the authors at this stage is to illustrate the concept of the generalized damage and expected loss maps at the city
 389 level. The accuracy of these maps can be further refined by either populating the number of instrumented buildings
 390 and/or by combining the proposed framework with other currently available tools that facilitate the city-scale rapid
 391 seismic assessment [16, 69–72].

392 Figures 16(a)–(c) show a generalized damage map for Los Angeles based on the maximum story-based EDP
 393 estimates that were computed along the height of the instrumented steel frame buildings that experienced the 1994
 394 Northridge earthquake. The maps are developed with the use of ArcGIS (release version 10.3) [73]. In regions that
 395 instrumented data were not available, the contour maps were developed with a multivariate (spatial) interpolation. It
 396 was found that the inverse distance weighting (IDW) method [74] provides rational results in this case. This method
 397 assigns maximum values of EDPs to unknown points with a weighted average of the values available at the known
 398 points in the map. The mathematical form of the IDW method [74] is defined as follows,

$$z_{x,y} = \frac{\sum_{i=1}^n z_i w_i}{\sum_{i=1}^n w_i} \quad (13)$$

399 in which $z_{x,y}$ is the value to be estimated at the location point (x,y) ; and z_i represents the control value for the i th
 400 sample point. The weight w_i determines the relative importance of the individual control point z_i in the interpolation
 401 process as follows,

$$w_i = d_{x,y,i}^{-\beta} \quad (14)$$

402 in which $d_{x,y,i}$ is the distance between $z_{x,y}$ and z_i ; and β is a user-defined exponent. In this study, the exponent β was
 403 assumed to be 2.0 as suggested in [73]. Equation (13) can be rewritten as follows,

$$z_{x,y} = \frac{\sum_{i=1}^n z_i d_{x,y,i}^{-\beta}}{\sum_{i=1}^n d_{x,y,i}^{-\beta}} \quad (15)$$

404 Referring to Figures 16(a)–(c), the area around the epicenter suffered the most structural and non-structural damage
 405 age given the distribution of peak SDRs [see Figure 16(a)] and PFAs [see Figure 16(b)]. Notably, near the epicenter the
 406 peak SDRs and peak PFAs were on the order of 1.5% and 0.8g, respectively. From the same figure, in the South-East
 407 of the city the distribution of the peak SDRs and PFAs were on the order of 0.5% and 0.4g, respectively, indicating that
 408 the expected structural and non-structural damage would be fairly minimal in the same region. Referring to Figure
 409 16(c), the distribution of residual SDRs was 0.45% or less; therefore, building demolition would not be a critical concern
 410 throughout Los Angeles. It is understood that the maps shown in Figures 16(a)–(c) can provide a first estimate of
 411 the post-earthquake safety of a city. These maps can be produced within minutes after the earthquake. In that sense,
 412 the proposed framework can be employed for city-scale management in the aftermath of an earthquake.

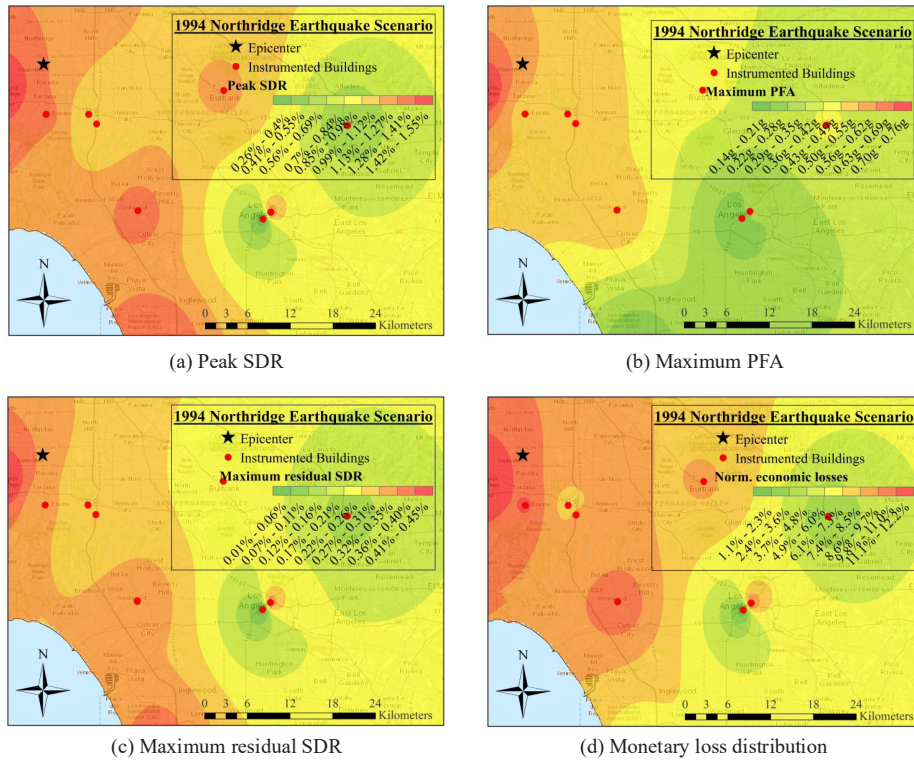


Figure 16: “City-scale” generalized damage and expected loss maps for Los Angeles after the 1994 Northridge earthquake.

413 The computed story-based EDPs shown in Figures 16(a)–(c) can be further utilized to develop a generalized
 414 expected loss map for the same region. This map is shown in Figure 16(d). In this figure, the expected losses due
 415 to building repairs were computed as discussed in Section 3.2 and they were normalized with respect to the total

416 replacement cost of the respective building. Referring to Figure 16(d), steel frame buildings with MRFs located near
 417 the epicenter experienced monetary losses on the order of 12% of their total replacement cost due to damage in drift-
 418 and acceleration-sensitive non-structural components. From the same figure, the expected losses are dominated by
 419 repairs due to premature fracture in pre-Northridge beam-to-column connections. Referring to Figure 16(d), such
 420 failures were fairly minimal in the South-East of the city considering the magnitude of the peak SDRs and PFAs in
 421 this region. The expected losses for the city of Los Angeles after the earthquake are further disaggregated in Figure
 422 17. In particular, Figures 17(a) and 17(b) illustrate the repairs needed due to premature fracture of pre-Northridge
 423 beam-to-column connections and damage in conventional gypsum wallboard partitions.

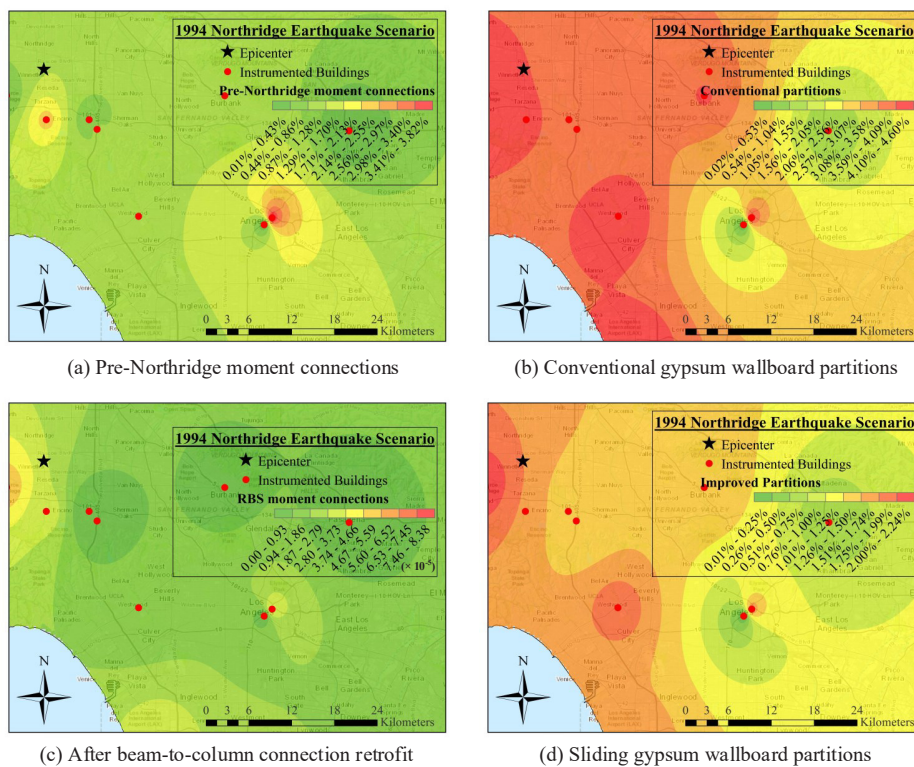


Figure 17: Loss disaggregation maps for the city of Los Angeles after the 1994 Northridge earthquake; (a) losses due to repairs in pre-Northridge beam-to-column moment connections; (b) losses due to repairs in conventional gypsum wallboard partitions; (c) losses due to repairs in retrofitted beam-to-column moment connections; and (d) losses due to repairs in sliding gypsum wallboard partitions.

424 The generalized loss maps shown herein can be easily employed for the computation of the expected losses at a
 425 city-scale for a given earthquake scenario such that proper pre-disaster measures can be prioritized by stakeholders
 426 and building owners. In particular, the proposed framework is utilized to compute the disaggregated losses if the
 427 same buildings were to be retrofitted prior to the same seismic event. In particular, if the beam-to-column connections
 428 would be rehabilitated such that they could behave as standard post-Northridge beam-to-column connections then
 429 the expected losses due to moment connection repairs would be nearly zero as shown in Figure 17(c). Similarly, the

430 expected losses due partition repairs would be reduced by a factor of 2.5, on average, if sliding gypsum wallboard
431 partitions would be installed prior to the seismic event.

432 **4. Limitations of the proposed framework**

433 This paper proposes a nonmodel-based framework for the rapid seismic risk and loss assessment of instrumented
434 steel frame buildings. The potential applicability of this framework for city-scale risk assessment is also investigated.
435 This section summarizes limitations of the proposed framework that can provide the basis for further research. In
436 particular, the proposed framework can only provide information in a global sense. The exact location of structural
437 damage within a story is not possible to be traced. The examined instrumented steel frame buildings were all fairly
438 symmetric and therefore torsional effects were not deemed to be critical. This issue should be further investigated.
439 The relationship between story-based EDPs and wavelet-based damage sensitive features should be defined for other
440 lateral load resisting systems such as concentrically braced frames as well as shear wall structures. Furthermore, soil-
441 foundation-structure interaction (SFSI) effects were not considered. Prior studies on SFSI effects suggest that their
442 influence on story-based EDP demands may be appreciable depending on the soil type [75–77]. This issue deserves
443 more attention in future research studies.

444 **5. Summary and conclusions**

445 This paper proposes a nonmodel-based framework for rapid seismic risk and loss assessment of instrumented
446 steel frame buildings in the aftermath of an earthquake. The proposed framework utilizes a wavelet-based damage-
447 sensitive feature (DSF) and minimal information to infer the damage state of a building at a given seismic intensity.
448 The wavelet-based DSF is able to trace the changes in the building's seismic response as verified with experimen-
449 tal data from shake table experiments. The proposed framework predicts with relatively good accuracy story-based
450 engineering demand parameters (EDPs) that are typically used within a performance-based earthquake engineering
451 framework for building specific earthquake-induced loss assessment. In particular, story-based EDPs of interest in-
452 clude the peak story drift ratios (SDRs), residual SDRs and peak floor absolute accelerations (PFAs) along the height
453 of a building.

454 The efficiency and potential of the proposed framework in computing story-based EDPs at a given seismic intensity
455 is evaluated through a number of illustrative applications including code-compliant archetype buildings with MRFs.
456 It is found that the nonmodel-based framework provides better estimates of peak SDRs and PFAs regardless of the
457 seismic intensity of interest compared to the FEMA P-58 simplified approach that utilizes a detailed numerical model
458 representation of the building of interest. Although the proposed framework predicts reasonably well the residual
459 SDRs for moderate seismic events, large discrepancies are observed between predicted and simulated median residual
460 SDRs at seismic intensities with low probability of occurrence. In this case, lower/upper bound analysis should be
461 considered.

462 The potential use of the proposed framework for rapid seismic risk and loss assessment at a city-scale is illus-
463 trated with a grid of instrumented steel frame buildings that experienced the 1994 Northridge earthquake through the
464 development of so-called generalized damage and loss maps. These maps are generated with the geographic infor-
465 mation system (GIS) and can be available within minutes after the seismic event. The same maps suggest that the
466 primary contributors to the expected losses in the instrumented buildings were the repairs due to premature fracture
467 of pre-Northridge beam-to-column connections and the repairs due to conventional gypsum wallboard partition dam-
468 age. The proposed nonmodel based framework is utilized to examine the efficiency of a retrofit scenario in which
469 pre-Northridge beam-to-column connections and wallboard partitions were upgraded to meet today's seismic perfor-
470 mance standards. It is shown that the proposed framework can effectively serve for pre-disaster risk management.
471 Summary remarks on how to improve the proposed framework are also provided.

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