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## The pan-European Engineering Strong Motion (ESM) flatfile: consistency check via residual analysis --Manuscript Draft--

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<b>Corresponding Author:</b>	Dino Bindi, PhD Helmholtz Centre Potsdam GFZ German Research Centre For Geosciences Potsdam, GERMANY
<b>Corresponding Author Secondary Information:</b>	
<b>Corresponding Author's Institution:</b>	Helmholtz Centre Potsdam GFZ German Research Centre For Geosciences
<b>Corresponding Author's Secondary Institution:</b>	
<b>First Author:</b>	Dino Bindi, PhD
<b>First Author Secondary Information:</b>	
<b>Order of Authors:</b>	Dino Bindi, PhD Sreeram Reddy Kotha Graeme Weatherill Giovanni Lanzano Lucia Luzi Fabrice Cotton
<b>Order of Authors Secondary Information:</b>	
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<b>Abstract:</b>	<p>We present the results of a consistency check performed over the flatfile extracted from the Engineering Strong Motion (ESM) database. The flatfile includes 23014 recordings from 2179 earthquakes in the magnitude range from 3.5 to 7.8 that occurred since the 1970s in Europe and Middle East, as presented in the companion article by Lanzano et al. (2018a). The consistency check is developed by analyzing different residual distributions obtained from ad-hoc ground motion prediction equations for the absolute spectral acceleration (SA), displacement (SD) and Fourier amplitude spectra (FAS). Only recordings from earthquakes shallower than 40 km are considered in the analysis. The between-event, between-station and event-and-station corrected residuals are computed by applying a mixed-effect regression. We identified those earthquakes, stations, and recordings showing the largest deviations from the GMPE median predictions, and also evaluated the statistical uncertainty on the median model to get insights on the applicable magnitude-distance ranges and the usable period (or frequency) range. We observed that robust median predictions are obtained up to 8s for SA and up to 20 Hz for FAS, although median predictions for <math>M_w \geq 7</math> show significantly larger uncertainties with 'bumps' starting above 5s for SA and below 0.3 Hz for FAS. The between-station variance dominates over the other residual variances, and the dependence of the between-station residuals on logarithm of <math>V_s30</math> is well-described by a piece-wise linear function with period-dependent slopes and hinge velocity around 580 m/s. Finally, we compared the between-event residuals obtained by considering two different sources of moment magnitude. The results show that, at long periods, the between-event terms from the two regressions have a weak correlation and the overall between-event variability is dissimilar, highlighting the</p>

	importance of magnitude source in the regression results .
<b>Suggested Reviewers:</b>	Carlo Cauzzi, Dr Eidgenossische Technische Hochschule Zurich carlo.cauzzi@sed.ethz.ch International expert on Ground Motion Prediction Equation and strong motion data analysis
	Laurentiu Danciu laurentiu.danciu@sed.ethz.ch International expert about the subject
	abdullah sandikkaya abdullahsandikkaya@hacettepe.edu.tr International expert about the topic
	Benedikt Halldorsson skykkur@hi.is International expert about the topic
	Susana Vilanova susana.vilanova@ist.utl.pt
<b>Response to Reviewers:</b>	our replies are uploaded as attachments to the manuscript.

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## The pan-European Engineering Strong Motion (ESM) flatfile: consistency check via residual analysis

Bindi D. (1), Kotha S-R.(1), Weatherill G.(1), Lanzano G.(2), Luzi L.(2), and F. Cotton (1,3)

5 (1) German Research Centre for Geosciences GFZ, Potsdam, Germany

(2) Istituto Nazionale di Geofisica e Vulcanologia INGV, Milan, Italy

(3) University of Potsdam, Potsdam, Germany

### Abstract

10 We present the results of a consistency check performed over the flatfile extracted from the Engineering Strong Motion (ESM) database. The flatfile includes 23014 recordings from 2179 earthquakes in the magnitude range from 3.5 to 7.8 that occurred since the 1970s in Europe and Middle East, as presented in the companion article by Lanzano et al. (2018a). The consistency check is developed by analyzing different residual distributions obtained from ad-hoc ground motion prediction equations for the  
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20 to get insights on the applicable magnitude-distance ranges and the usable period (or frequency) range. We observed that robust median predictions are obtained up to 8s for SA and up to 20 Hz for FAS, although median predictions for  $M_w \geq 7$  show significantly larger uncertainties with ‘bumps’ starting above 5s for SA and below 0.3 Hz for FAS. The between-station variance dominates over the other residual variances, and the dependence of the between-station residuals on logarithm of  $V_{s30}$  is well-  
25 described by a piece-wise linear function with period-dependent slopes and hinge velocity around 580 m/s. Finally, we compared the between-event residuals obtained by considering two different sources of moment magnitude. The results show that, at long periods, the between-event terms from the two regressions have a weak correlation and the overall between-event variability is dissimilar, highlighting the importance of magnitude source in the regression results .

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## Introduction

Most applications in Engineering Seismology and Earthquake Engineering deal with the analysis of strong-motion data. In Europe, the dissemination of uniformly processed strong-motion waveforms accompanied by event and station metadata progressed since the 1980s (Ambraseys, 1990; Ambraseys and Bommer, 1991; Ambraseys et al., 2000; Ambraseys et al 2004). The two most recent examples are the Reference Database for Seismic Ground-motion Prediction in Europe-RESORCE (Akkar et al., 2014), developed in the framework of the Seismic Hazard HARMONIZATION in Europe project (SHARE; [www.share.eu.org](http://www.share.eu.org)), and the Engineering Strong Motion (ESM) database (Luzi et al., 2016) developed in the framework of the Network of European Research Infrastructures for Earthquake Risk Assessment and Mitigation project (NERA). For many applications aimed at developing and testing Ground-Motion Prediction Equations (GMPEs), ground-motion intensity measures of engineering interest (e.g., peak ground acceleration-PGA, peak ground velocity-PGV, response spectra, Housner intensity, etc) are extracted or computed from the waveforms, and disseminated along with the relevant event and station metadata in the form of parametric tables - often referred to as flatfiles. Examples of recent flatfiles include the Pacific Earthquake Engineering Research Center (PEER) flatfile NGA-West2 (Ancheta et al., 2014) and regional flatfiles like those for Japan (Dawood et al., 2016) and Chile (Bastias and Montalva., 2016).

In this work, we focus on the flatfile generated from the ESM data-base, as described in detail in a companion paper (Lanzano et al., 2018a). In order to support the development of ground-motion models from the ESM flatfile, we perform a consistency check over the data and metadata from the perspective of the ground-motion variability, to report possible anomalies that a GMPE developer might encounter. To accomplish this task, we first calibrate a set of GMPEs for spectral acceleration (SA), spectral displacement (SD) and Fourier amplitude spectra (FAS), considering simple models describing the distance and magnitude scaling. Then, the random-effects and residual distributions obtained by performing a mixed-effects regression (Bates et al., 2015) are analyzed to detect anomalous records, events and stations. Issues that could be of relevance for ground-motion models, such as those related to the selected magnitude scale, and to ground-motion and path regionalization are discussed as well.

## Data selection for developing the reference model

60 In this study, we use the 2018 version of the ESM flatfile (Lanzano et al., 2018b). The characteristics of  
the ESM flatfile are described in Lanzano et al. (2018a) whereas the waveform processing is detailed in  
Puglia et al. (2018). For developing the reference ground-motion model, we select only data at distances  
shorter than 300 km, generated by earthquakes with at least two recordings, and occurring at depths  
shallower than 40 km. The lower limit of hypocentral depth is set to select only the Active Shallow  
65 Crustal Earthquakes, whereas the deeper continental and subduction events are left out of this study  
considering their relatively complex phenomenon requiring ad-hoc studies. For SA and SD, each record  
used in the GMPE regression for period T should satisfy the usable period criterion  $T \leq 0.8/f_{HP}$ , where  $f_{HP}$   
is the high-pass corner frequency of the filter applied in the waveform processing stage. For FAS  
computed at frequency f, only spectral amplitudes in the range  $1.2 f_{HP} \leq f \leq f_{LP}/1.2$ , where  $f_{LP}$  is the low-  
70 pass corner-frequency of the filter, are considered in deriving the GMPE. The distributions of the corner-  
frequencies of the filter applied to all recordings are provided in Lanzano et al. (2018b). In particular,  
most of the  $f_{LP}$  frequencies lie in the range 30-40 Hz. Regarding the station selection, we only select  
stations installed at surface (sensor depth <10 m) avoiding borehole installations (identified by  
*housing\_code* equal to "WEL") and stations installed at building floor (*installation\_code* equal to "BF").  
75 For our consistency check, we also select sensors with *location\_code* equal to "00".

Regarding the magnitude, the flatfile includes multiple magnitude scales and estimates. In particular, the  
local magnitude (ML) and two choices for the moment magnitude (Mw), namely the moment magnitude  
listed in the ESM database (hereinafter referred to as  $Mw^{ESM}$ ) and the moment magnitude from the  
EMEC catalog (Grünthal et al., 2012; hereinafter referred to as  $Mw^{EMEC}$ ). In particular, while  $Mw^{ESM}$  is  
80 computed from moment tensor solutions provided by different sources selected according to an  
hierarchical schema (Lanzano et al., 2018b),  $Mw^{EMEC}$  is based on both moment tensor solutions and  
empirical regional conversions from other magnitude scales. The two Mw scales are compared in Figure  
1a, where the best fit line is  $Mw^{EMEC} = 0.23 + 0.95Mw^{ESM}$ , with the residual standard deviation equal to  
0.16. Since neither  $Mw^{ESM}$  nor  $Mw^{EMEC}$  is available for all events, we derive an empirical relationship  
85 between ML and Mw in order to extend the residual analysis to as large a number of events as possible.  
These relations are not proposed for applications different from those performed in the context of this  
study. The best fit lines shown in Figure 1 (right) are:

$$Mw^{ESM} = 6.81 - 1.67 ML + 0.26 ML^2 \quad (\pm 0.3) \quad (1)$$

$$Mw^{EMEC} = 6.18 - 1.34 ML + 0.23 ML^2 \quad (\pm 0.3) \quad (2)$$

## GMPE functional form

Mixed-effects regressions (Brillinger and Preisler, 1985; Bates et al., 2015) are performed for the natural-logarithm of SA, SD and FAS considering simple parametric functional forms describing the median scaling with distance and magnitude:

95

$$\log Y = e_1 + F(M_w) + G(R_{hypo}) + RE + \epsilon \quad (3)$$

$$F(M_w) = \begin{cases} b_1(M_w - M_{ref}) & \text{if } M_w \leq M_H \\ b_1(M_H - M_{ref}) + b_3(M_w - M_H) & \text{otherwise} \end{cases} \quad (4)$$

$$100 \quad G(R_{hypo}) = \begin{cases} c_{1A} \log(R_{hypo}) & \text{if } R_{hypo} \leq R_H \\ c_{1A} \log(R_H) + c_{1B} \log\left(\frac{R_{hypo}}{R_H}\right) + c_3 R_{hypo} & \text{otherwise} \end{cases} \quad (5)$$

where in equation (1) the random effect (RE) are

$$RE = \delta B_e + \delta S2S_s + \delta C2C_c \quad (6)$$

In equations (4) and (5), the reference and hinge magnitude and distances are set to  $M_{ref}=4.5$ ,  $M_H=6$ ,  
 105  $R_H=15\text{km}$ , respectively. The country-to-country random effect ( $\delta C2C_c$ ), where  $c$  indicates country, is applied either on the offset term ( $e_1$ ) or to  $c_3$  (that is, two different regressions are performed). The other random-effects applied to the offset are the customary between-event ( $\delta B_e$ ) and between-station ( $\delta S2S_s$ ) residuals, where subscripts  $e$  and  $s$  are event and station indices. Residual  $\epsilon$  in equation (3) represents the event- and station-corrected record-to-record variability. The standard deviations of  $\epsilon$ ,  
 110  $\delta S2S_s$ ,  $\delta C2C_c$ , and  $\delta B_e$  random distributions are indicated with  $\phi_0$ ,  $\phi_{S2S}$ ,  $\phi_{C2C}$  and  $\tau$ , respectively. The regressions for SA and SD are performed over the orientation independent (rotD50) ground-motion measure (Boore, 2010), while the geometric-mean of the horizontal components is used for FAS GMPEs.

For the GMPEs regressed on SA considering  $\delta C2C_c$  on the offset  $e_1$ , Figure 2 shows the period-dependent coefficients along with their standard-errors. Regarding the scaling with distance, coefficient  $c_{1B}$

115 increases from about -2 at  $T < 0.1s$  to about -1.3 for  $T > 0.5s$  while coefficient  $c_{1A}$  is maximum around  $T = 0.1s$  and decreases below -2 for  $T > 4s$ . The term  $c_3$  is negative for short periods and turns positive for  $T \geq 1.5s$ . The positive values of  $c_3$  suggest to investigate further segmentation of the distance scaling component to capture the effects of late-arrivals on long period ground-motion at distances beyond 80km (e.g., Lanzano et al., 2016; Kotha et al., 2018).

120 Regarding the scaling with magnitude, although large uncertainties affect  $b_3$  (describing the magnitude scaling above the hinge magnitude), its values are always smaller than  $b_1$ , suggesting weaker scaling at larger magnitudes. The trend of the coefficients suggest that regressions above 8s are unstable due to a strong reduction in the number of data usable in GMPE regressions (Lanzano et al., 2018a). For example, the number of recordings used for the regressions at 0.01, 4, 8, and 10 s are 18859, 16222, 9664 and  
125 2767, respectively. The asymptotic standard error ( $\sigma_\mu$ ) on the median predictions, quantifying the within-model statistical estimate of epistemic uncertainty (e.g., Al Atik and Youngs, 2014 ; Bindi et al., 2017a), due to the combined effect of limited data availability and complexity of implemented functional form (in terms of extent of GMPE parametrization) are shown in Figure 3. For  $M_w \leq 6.5$ ,  $\sigma_\mu$  is smaller than 0.2 and it shows a weak dependency on magnitude for  $M_w < 6$ . For  $M_w < 6.5$ ,  $\sigma_\mu$  reaches the minimum values  
130 around  $T = 2s$  while the sharp increase above  $T = 8s$  indicates that regressions are not well-constrained for these periods. At  $M_w > 6.5$ ,  $\sigma_\mu$  increases rapidly, exceeding 0.25 at  $M_w = 7.5$ . Above magnitude 7,  $\sigma_\mu$  shows a bump around  $T = 5s$ , suggesting that at periods longer than 4s GMPE median predictions for  $M_w > 7$  are uncertain. The dependence of  $\sigma_\mu$  on distance (here not shown) is negligible because the data appears to constrain very well the distance scaling component of the GMPE (small errors on  $c_{1A}$ ,  $c_{1B}$ ,  $c_3$  in  
135 Figure 2).

## Results

The results are presented in terms of event- and station-corrected residuals ( $\varepsilon$ ), between-event ( $\delta B_e$ ) and  
140 between-station ( $\delta S_2 S_3$ ) distributions evaluated for SA with regional attributes  $\delta C_2 C_c$  on  $e_1$ .

### Event- and station-corrected residuals $\varepsilon$

The event- and station-corrected residuals for  $T=0.01s$  and  $T=8s$  are shown in Figure 4 as function of the recording Id (top), of magnitude (middle) and of distance (bottom). In the middle and bottom frames, the average residuals computed over magnitude and distance bins (red error bars) are shown as well. Please note that the number of recordings used in GMPE regression decrease towards long periods. The residuals do not show any significant trend either with distance or with magnitude. The number of residuals (in absolute value) above 3 times  $\phi_0$  is 60 at  $T=0.01s$  and 20 at  $T=4s$ , as shown in Tables 1 and 2, respectively. A further check of the waveforms does not reveal any peculiarities in the recordings except for the records [IT-2016-0001, IV.INTR], which is characterized by high noise level at high frequencies, as shown in Figure 5.

### **Between-event random effects $\delta B_e$**

The between-event random-effects for  $T=0.01s$  and  $T=8s$  are shown in Figures 6a and 6b, respectively.  $\delta B_e$  does not show significant trend for  $M > 4.5$ , while  $\delta B_e$  increases with decreasing magnitude for  $M < 4.5$ . Unlike the functional form implemented in this study, an additional segmentation of  $F(M_w)$  at  $M_w = 4.5$  (e.g, Campbell and Bozorgnia 2014, Kotha et al. 2018) would allow different scaling for small, moderate and large magnitude bins, removing the trend observed for small magnitudes. We note here that moment magnitudes below 4 are not converted from local magnitude.

Differences in observed SAs based on Style-of-Faulting (SoF) are widely debated in GMPE development. For the dataset analyzed in this study, the box and whisker plots corresponding to  $\delta B_e$  dependence on SoF do not show any systematic differences. Since the model included a random-effect at the country level (Italy-IT, Turkey-TK, Romania-RO, Greece-GR, other countries-Others), a dependence of  $\delta B_e$  on the country grouping is not expected, as shown by the box and whisker plots in Figures 6a and 6b. Events with  $\delta B_e$  larger than 3 times  $\tau$ , in absolute value, are listed in Table 3 and 4 for  $T=0.01$  and  $8s$ , respectively.

### **Between-station random-effects $\delta S2S_s$**

The between-station random-effects for  $T=0.01s$  and  $T=8s$  are shown in Figures 7a and 7b, respectively. At short periods,  $\delta S2S_s$  shows a trend with  $V_{s30}$ , as expected ( $\log(V_{s30})$  is commonly used as proxy for site amplifications). The best-fit model obtained through a break-point regression (Muggeo, 2003) is

described by a piece-wise linear function with break-point at  $\log_{10}(Vs30)=(2.78 \pm 0.09)$  (i.e.,  $Vs30=600$  m/s) and slopes  $(-0.6 \pm 0.2)$ ,  $(-1.7 \pm 0.40)$ . A weaker relation exists between  $\delta S2S_s$  and  $Vs30WA$  (which is an estimate of  $Vs30$  derived from the topographic gradient, see Lanzano et al., 2018b). At long periods, the trend with measured  $Vs30$  is strong only for  $Vs30 < 600$  m/s, while considering  $Vs30WA$  a correlation over a broader velocity range is observed. For  $T=8s$ , the break-point is located at  $\log_{10}Vs30=(2.77 \pm 0.09)$ , and the slopes are  $(-1.6 \pm 0.2)$  and  $(-0.3 \pm 0.2)$ .

When split among the different countries (box and whiskers plots in Figure 7), no residual offset is detected. Often the station  $Vs30$  distributions (mean and range) differ from region-to-region depending on the local geology over which the networks reside. As a result, systematic region-dependent offsets could be expected. In our analysis, since all the country-to-country variability is accounted by the  $\delta C2C_c$  random-effects on  $e_1$ , we observe no differences in mean  $\delta S2S_s$  among the countries. Nevertheless, it is worth of further investigation whether the scaling of  $\delta S2S_s$  with  $Vs30$  is identical across various regions or not. Kotha et al. (2016) for instance, observed a similar (regionally) unbiased  $\delta S2S_s$  for Italy and Turkey, but with significantly different scaling with  $Vs30$ . Stations with  $\delta S2S_s$  larger than 3 times  $\phi_{S2S_s}$ , in absolute value, are listed in Table 5 and 6 for  $T=0.01$  and 8 s, respectively.

### Standard deviations of random-effects

The standard deviations of the random-effects are summarized in Figure 8. The between-station variability  $\phi_{S2S_s}$  dominates the total aleatory variability of the dataset ( $\sigma$ ): it shows a peak around 0.10 s and it decreases above 2s. The peak at  $T = 0.1s$  is observed in several other studies as well (e.g. Zhao et al. 2016, Cauzzi and Faccioli 2017; Kotha et al. 2018), indicating a high variability of short period site-response in response spectra domain. The between-event standard deviation  $\tau$  has a peak located around 2s, otherwise almost constant with a small bump around 0.1 s;  $\tau$  shows a suspicious drop at  $T=9s$  and 10s, probably due to the strong reduction in the usable recordings. The event- and station-corrected variability ( $\phi_0$ ) is close to  $\tau$  for  $T < 0.5s$ , while for longer periods it is smaller than the between-event variability. Finally, the between-country variability  $\phi_{C2C}$  (region-to-region variability of offset  $e_1$ ) is small (of the order of 0.25) with a trough around 2s. The anti-symmetry of  $\tau$  and  $\phi_{C2C}$  around  $T = 2s$  suggests a trade-off between event-to-event and country-to-country variability in this GMPE regression, suggesting the need of further investigating the formulation of REs in GMPEs.

### Regional differences on anelastic attenuation.

205 An additional regression is performed by applying the country-group random-effect not on the overall offset  $e_1$  in equation 3 but on the anelastic coefficients  $c_3$  of equation 5. Figure 9 shows the obtained random effects  $\delta c_3$ , which can be interpreted as regional adjustment to the generic region-independent “anelastic attenuation” term  $c_3$ . These country-specific adjustments (to  $c_3$ ) are statistically significant with small standard errors (horizontal bar in the caterpillar plot of Figure 9), suggesting considerable  
210 differences in propagation effects across these regions. For instance, since the  $\delta c_3$  value for Romania is not well-constrained with the very few data from that region, its standard error is larger than that for Italy, Greece, and Turkey. For periods between 0.01 and 2 s, Greek and Italy show stronger attenuation than Turkey; for longer periods we recall that  $c_3$  becomes positive.

Since in this case the country-based RE is not applied to the overall offset, the between-events show  
215 some variability across the countries (Figure 10). Note that since  $\delta C_2 C_c$  is no longer applied on  $e_1$ , systematic country-to-country variability in  $\delta B_e$  and  $\delta S_2 S_s$  distributions may now appear, as seen in the between-event distributions of Figure 10. For example, on average the events from Greece (GR) show a higher  $\delta B_e$  compared to those from Turkey (TR) and Italy (IT) (Figure 9).

### 220 $\delta B_e$ obtained using either $M_w^{EMEC}$ or $M_w^{ESM}$ .

Several studies (e.g., Rohades, 1997; Moss, 2011; Kuehn and Abrahamson, 2018) evaluated the impact of the uncertainty in the predictor variables on the aleatory variability. The effect of magnitude uncertainty on  $\delta B_e$  variability has been observed in Europe for the RESORCE flatfile by comparing  $M_w$  values computed following different procedures (Kotha et al., 2016). For the ESM flatfile, we assess the impact  
225 of using  $M_w^{EMEC}$  or  $M_w^{ESM}$  by repeating the regression for each magnitude type, with  $\delta C_2 C_c$  applied on  $e_1$ . The results are compared in Figure 11 in terms of  $\delta B_e$  for  $T=0.1$  and 2.5s. The red dots in Figure 11 indicate recordings for which both  $M_w^{EMEC}$  and  $M_w^{ESM}$  are available, while the other colors indicate earthquakes for which at least one of the two  $M_w$  values is derived from ML. In particular, the green dots are for earthquakes without  $M_w$  (neither EMEC nor ESM). Figure 11 shows that at short periods  $\delta B_e$   
230 for the two regressions show a good correlation although with a large scatter. On the other hand, at

longer periods ( $T=2.5s$  in Figure 11 but results for longer periods are similar) the correlation is very weak (being trivial the correlation shown by the green points). Figure 12 shows the standard deviation  $\tau$  of the  $\delta B_e$  residuals for the two regressions, considering both actual  $M_w$  values and values derived from ML.  $\tau$  for  $M_w^{EMS}$  is higher than the one obtained using  $M_w^{EMEC}$  for periods  $T < 2.0s$  whereas, from  $T=2.0$  to  $8s$ , similar values are obtained for the two regressions.

### Summary for SD and for FAS

The ESM flatfile includes several intensity measure of engineering interest, such as spectral acceleration (SA) and displacement (SD); Fourier amplitude spectra (FAS); peak ground velocity (PGV) acceleration (PGA) and displacement (PGD); Arias intensity, Housner intensity, effective duration, cumulative absolute velocity (CAV). While the check of all those parameters is beyond the aim of this work, the residual analysis have been extended to SD and FAS as well (Bindi et al., 2017b). Figure 13 summarizes the results by comparing the standard deviations of the residual distributions. The main conclusions drawn from SD and FAS analysis agree with those from SA. Regarding the uncertainty on the median model for FAS, the dependency of  $\sigma_{\mu}$  on frequency is shown in Figure 14. The models are robust up to 20 Hz, for higher frequencies the number of available recording decreases due the condition applied to the usable frequency range (see Lanzano et al., 2018b, for the distribution of the low-pass corner frequencies of the filter applied during the processing phase) and the uncertainty becomes very large. Towards low frequencies,  $\sigma_{\mu}$  for  $M_w \geq 6.5$  shows a bump around 0.2 Hz. In general, in agreement with the results obtained for SA,  $\sigma_{\mu}$  significantly increases above magnitude 7.

### Conclusion

We performed residuals analysis over the ESM flatfile (Lanzano et al., 2018b) as complementary information to the flatfile description (Lanzano et al., 2018a), for supporting those studies aiming at developing ground-motion models. The main outcomes of the consistency check performed on recordings from shallow earthquakes (hypocentral depth less than 40 km) can be summarized as follow:

- 1) considering the data population and the selection criteria applied to define the usable period range, robust results for response spectra have been obtained for periods up to 8s, although the median models for  $M_w \geq 7$  are affected by large within-model epistemic uncertainty ( $\sigma_{\mu}$ ) above 5s. Regarding the

260 model for Fourier,  $\sigma_{\mu}$  increases significantly above 20Hz and shows a bump around 0.2 Hz for magnitudes above 7. The epistemic uncertainty for both response and Fourier spectra increases significantly for  $M_w \geq 6.5$ ;

2) the sanity check has been performed considering a simple functional form for describing the magnitude scaling and the scaling with hypocentral distance, although for larger finite-source events the Joyner-Boore or the rupture distances should be considered. Regarding the distance scaling, a piece-wise linear function of logarithm of distance was needed to avoid trends in the residuals at distances shorter than 15km (see also Kotha et al., 2018). Dependence of the hinge distance on magnitude and period is worth future investigation in order to model the magnitude dependence of geometric spreading as well;

3) regional dependencies of the attenuation with distance (so-called apparent anelastic attenuation parameter) already observed in previous studies (e.g., Kotha et al., 2016) are confirmed. In addition, it is now possible to refine the regionalization scheme given the significantly larger amount of data from a few more regions;

4) regarding the scaling of SA with magnitude, it is worth considering a second hinge magnitude at small magnitudes, around 4 (see also Kotha et al 2018);

275 5) the scaling of the between-station random-effects  $\delta S_{2S_s}$  with the logarithm of  $V_{s30}$  is well described by a piece-wise linear function with period-dependent slopes and hinge velocity around 580 m/s;

6) magnitude is one of the two fundamental explanatory variables used to develop ground motion models, and the flatfiles includes information about different magnitude scales extracted from the ESM database. In addition, two different moment magnitude values are provided (when available):  $M_w^{ESM}$  is the moment magnitude as included in ESM without any conversion (a reference for the source is provided);  $M_w^{EMEC}$  is the moment magnitude according to the EMEC catalog (Grünthal et al., 2012). Derived through different procedures, the two magnitudes show an overall agreement (linear correlation) but with large scatter around the mean trend. The between-event residuals  $\delta B_e$  at long periods absorb the uncertainties in the seismic moment estimates: therefore, the weak correlation shown by the  $\delta B_e$  distribution obtained considering the two different moment magnitudes warns about the critical role played by the magnitude selection;

285 7) Records, events, and stations showing standardized residuals larger, in absolute value, than 3 are listed in tables to allow for a comparison with other models.

8) among different intensity measures, Fourier amplitude spectra allow an easier and direct comparison  
290 between components of the observed ground-motion variability and seismological interpretations.

### **Data and Resources**

In this study, we analyzed the ESM flatfile (Lanzano et al, 2018b) with persistent identifier PID:  
11099/ESM\_flatfile\_2018. The flatfile can be downloaded from the ESM webpage: <http://esm.mi.ingv.it>.

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380 **Table 1.** Event- and station-corrected residuals greater than 3 times  $\phi_0$  in absolute value

( $|\varepsilon| > 3\phi_0$ ) at  $T=0.01s$

Event ID	Station	$\varepsilon(0.01s)$	Event ID	Station	$\varepsilon(0.01s)$
EMSC-20101109_0000039	FR.REVF	-1.943414	EMSC-20161129_0000079	IT.AQG	-1.870889
EMSC-20110707_0000048	RA.IRPV	2.168072	EMSC-20161129_0000079	IT.AQV	-1.864027
EMSC-20130125_0000039	IV.BDI	2.24683	EMSC-20161129_0000079	IT.GNU	-1.864027
EMSC-20130829_0000020	KO.GONE	-1.996238	EMSC-20161129_0000079	IT.PSC	-1.912796
EMSC-20140907_0000025	IT.BRM	1.882042	EMSC-20161129_0000079	IT.SPD	-2.056146
EMSC-20140907_0000025	IT.PRT	3.342786	EMSC-20161129_0000079	IT.SULA	-1.839875
EMSC-20141216_0000014	KO.GONE	-1.881109	EMSC-20161129_0000079	IT.SULC	-1.954815
EMSC-20141219_0000039	IT.FOC	-1.853693	EMSC-20161129_0000079	IV.BSSO	-1.938496
EMSC-20141224_0000053	IT.SULC	1.87239	EMSC-20161129_0000079	IV.CERA	-2.019819
EMSC-20150130_0000003	IV.OPPE	1.919609	EMSC-20161129_0000079	IV.MOMA	2.029745
EMSC-20150130_0000003	NI.POLC	-2.251522	EMSC-20161129_0000079	IV.RNI2	-2.023737
EMSC-20150130_0000003	RF.GEPF	-1.924863	EMSC-20161129_0000079	IV.VAGA	-1.965676
EMSC-20150205_0000061	HL.ZKR	2.124032	EMSC-20161201_0000064	IT.CNE	1.943368
EMSC-20150726_0000076	HL.PLG	2.325277	GR-1999-0001	HL.PREA	3.271508
EMSC-20150913_0000012	KO.CAME	2.257234	IT-1987-0003	IT.PNN	-2.084439
EMSC-20151101_0000023	MN.PDG	1.961717	IT-2007-0208	IT.FMG	2.013387
EMSC-20151118_0000047	KO.CAME	-2.911697	IT-2009-0186	IV.SACR	-2.090624
EMSC-20151118_0000094	KO.CAME	-2.740185	IT-2009-0189	IV.SACR	-2.126487
EMSC-20160530_0000085	IV.SACS	2.118076	IT-2010-0003	IV.NRCA	2.243845
EMSC-20160824_0000006	IT.GRD	-1.950702	IT-2011-0029	IT.GAI	-2.075673
EMSC-20160824_0000295	IV.FIAM	1.86564	IT-2012-0002	FR.REVF	2.025614
EMSC-20160824_0000295	IV.RM33	2.219246	IT-2013-0003	IT.PSC	1.974364
EMSC-20160903_0000009	IT.ACC	-1.866856	IT-2013-0003	IV.POFI	2.157508
EMSC-20160903_0000009	IT.CIT	-1.983131	IT-2013-0005	IT.MDN	2.272414
EMSC-20160903_0000009	IT.FOC	2.200381	IT-2013-0010	IV.SNTG	-2.148043
EMSC-20160903_0000009	IV.NRCA	1.919352	IT-2016-0001	IV.INTR	-3.470851
EMSC-20160903_0000009	IV.T1201	-1.887857	ME-1979-0003	EU.BAR	-2.051513

EMSC-20160903_0000009	IV.T1218	-2.022831	TK-1999-0077	TK.8101	1.865905
EMSC-20161027_0000072	IV.SNTG	2.005795	TK-2003-0180	TK.1701	-1.970150
EMSC-20161129_0000079	IT.AQF	-1.928834	TK-2006-0098	TK.1001	-2.058961

**Table 2.** Event- and station-corrected residuals greater than 3 times  $\phi_0$  in absolute value ( $|\varepsilon| > 3\phi_0$ ) at  $T=8s$

Event ID	Station	$\varepsilon(8.0s)$	Event ID	Station	$\varepsilon(8.0s)$
EMSC-20110707_0000048	RA.IRPV	1.996037	IT-2010-0003	IV.NRCA	1.48384
EMSC-20150205_0000061	HL.ZKR	1.78805	IT-2010-0006	IV.NRCA	1.627814
EMSC-20151106_0000016	FR.REVF	-2.236079	IT-2012-0010	TV.MIR01	-1.484999
EMSC-20151106_0000016	FR.SURF	1.709347	IT-2012-0025	IT.FIN0	1.660915
EMSC-20151118_0000094	KO.CAME	-1.846812	IT-2012-0025	IV.T0800	1.737367
EMSC-20161030_0000033	IT.ACC	1.99265	IT-2012-0025	IV.T0811	2.127796
EMSC-20161129_0000079	IT.AQV	-1.492742	IT-2012-0035	IV.CPGN	-1.493063
IT-2009-0009	IT.ANT	-1.557013	IT-2012-0063	IT.SBT	1.630845
IT-2010-0001	IV.NRCA	1.59095	IT-2013-0013	IT.NAS	1.569184
IT-2010-0001	IV.SENI	-1.600929	TK-1999-0077	TK.8101	1.501051

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**Table 3.** Between-event greater than 3 times  $\tau$ , in absolute value ( $|\delta B_e| > 3\tau$ ), at  $T=0.01 s$

Event ID	$\delta B_e(0.01s)$	N records
EMSC-20121203_0000066	1.73	18
EMSC-20150411_0000019	-1.69	10
EMSC-20161102_0000149	-1.66	72
IT-2009-0316	1.97	6

**Table 4.** Between-event greater than 3 times  $\tau$ , in absolute value ( $|\delta B_e| > 3\tau$ ), at  $T=8.0 s$

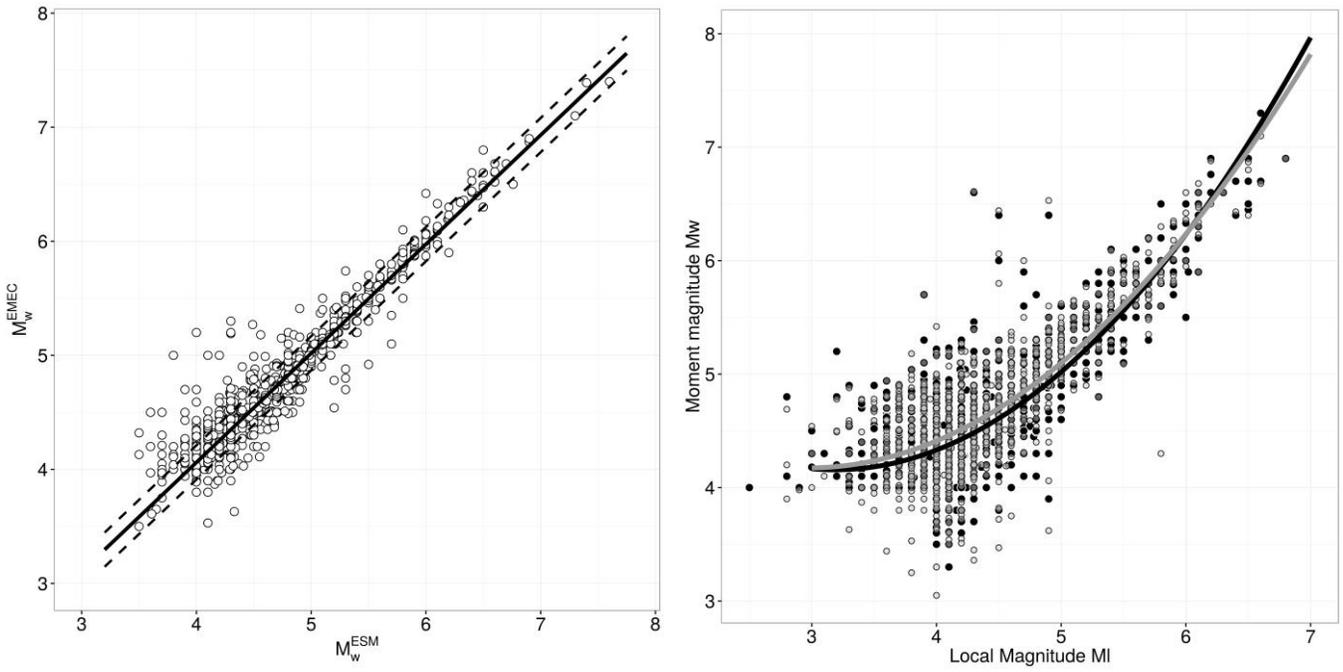
Event ID	$\delta B_e(8.0s)$	N records
EMSC-20121203_0000066	1.65	11
EMSC-20141124_0000030	1.84	5
IT-2010-0003	1.95	3

390 **Table 5.** Between-station greater than 3 times  $\phi_{s2s}$ , in absolute value ( $|\delta S2S| > 3\phi_{s2s}$ ), at T=0.01 s

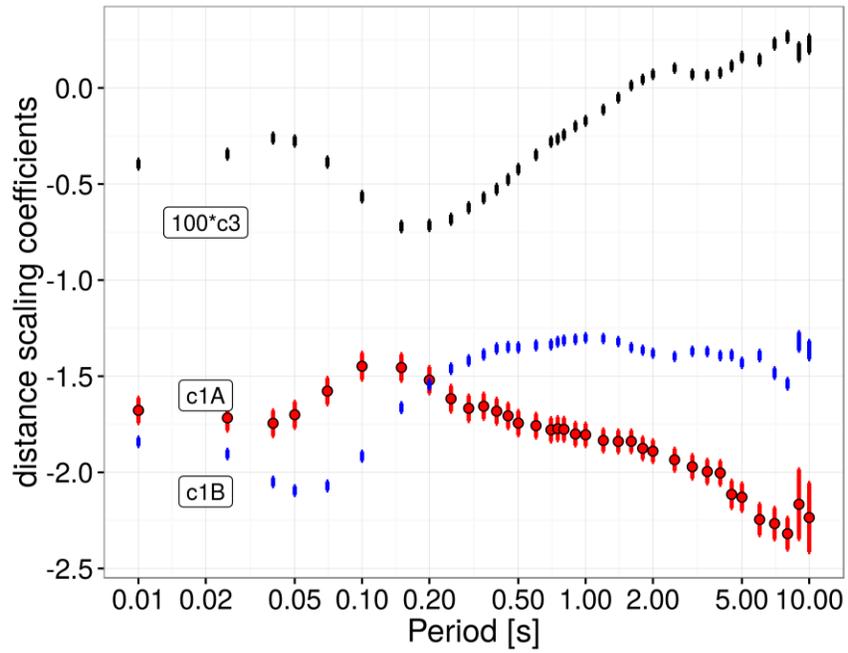
Station	$\delta S2S(0.01s)$	N records
IV.T0203	-2.35	1
HL.KSL	-2.04	42
CR.VELS	2.01	1
RO.JURR	2.04	2
FR.REVF	2.05	5

**Table 6.** Between-station greater than 3 times  $\phi_{s2s}$ , in absolute value ( $|\delta S2S| > 3\phi_{s2s}$ ), at T=8.0 s

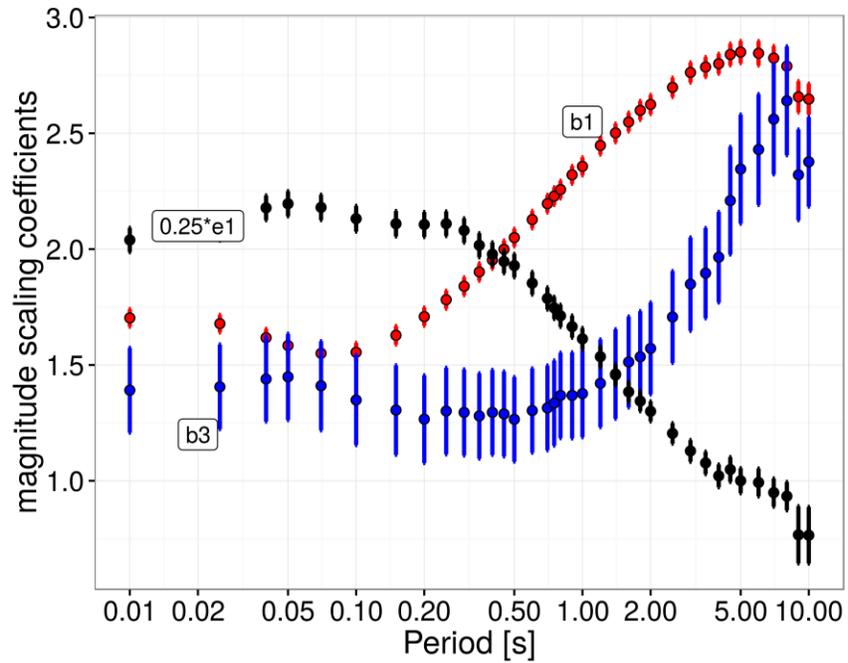
Station	$\delta S2S(8.0s)$	N records
IV.T0501	-1.86	3
FR.REVF	2.29	4
IV.SNTG	-1.74	15
KO.ENZZ	1.70	7



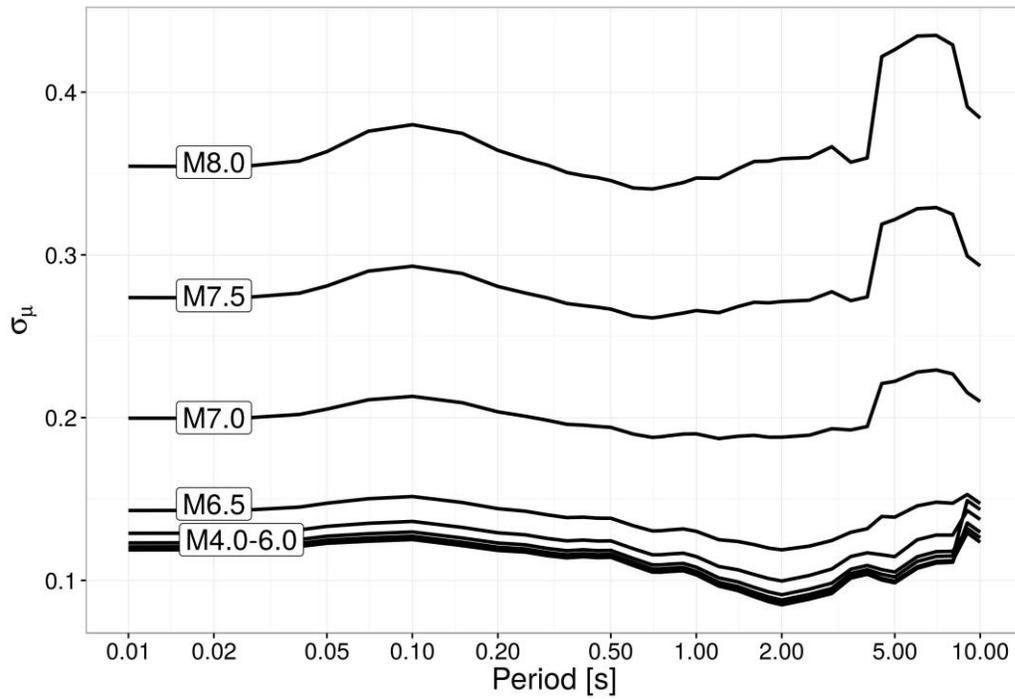
400 **Figure 1.** Left: regression between  $Mw^{ESM}$  and  $Mw^{EMEC}$ ; right: quadratic regressions between  $Mw^{ESM}$  and ML (black line and points) and between  $Mw^{EMEC}$  and ML (gray line and points) (see equations 1 and 2).



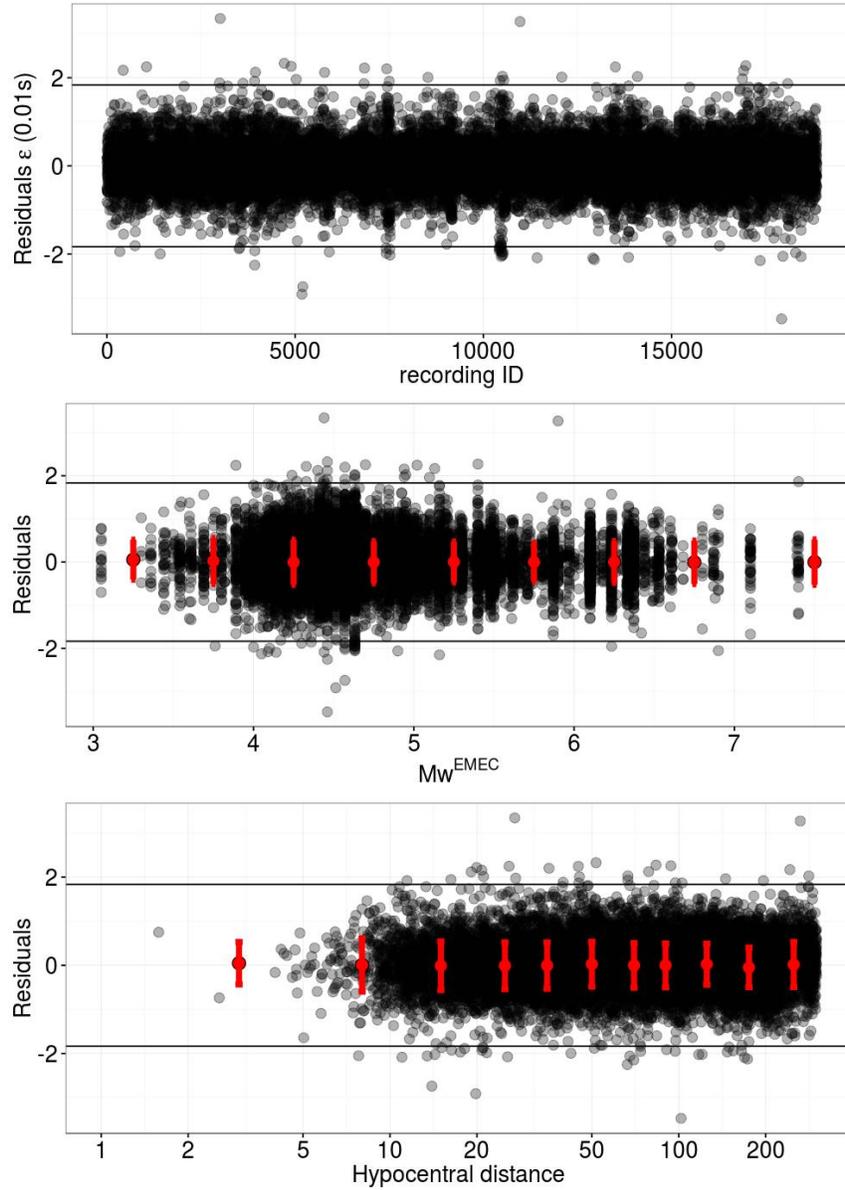
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**Figure 2.** Coefficients (mean and standard error) of the model in equation (4) and (5) for Spectral Acceleration (SA).

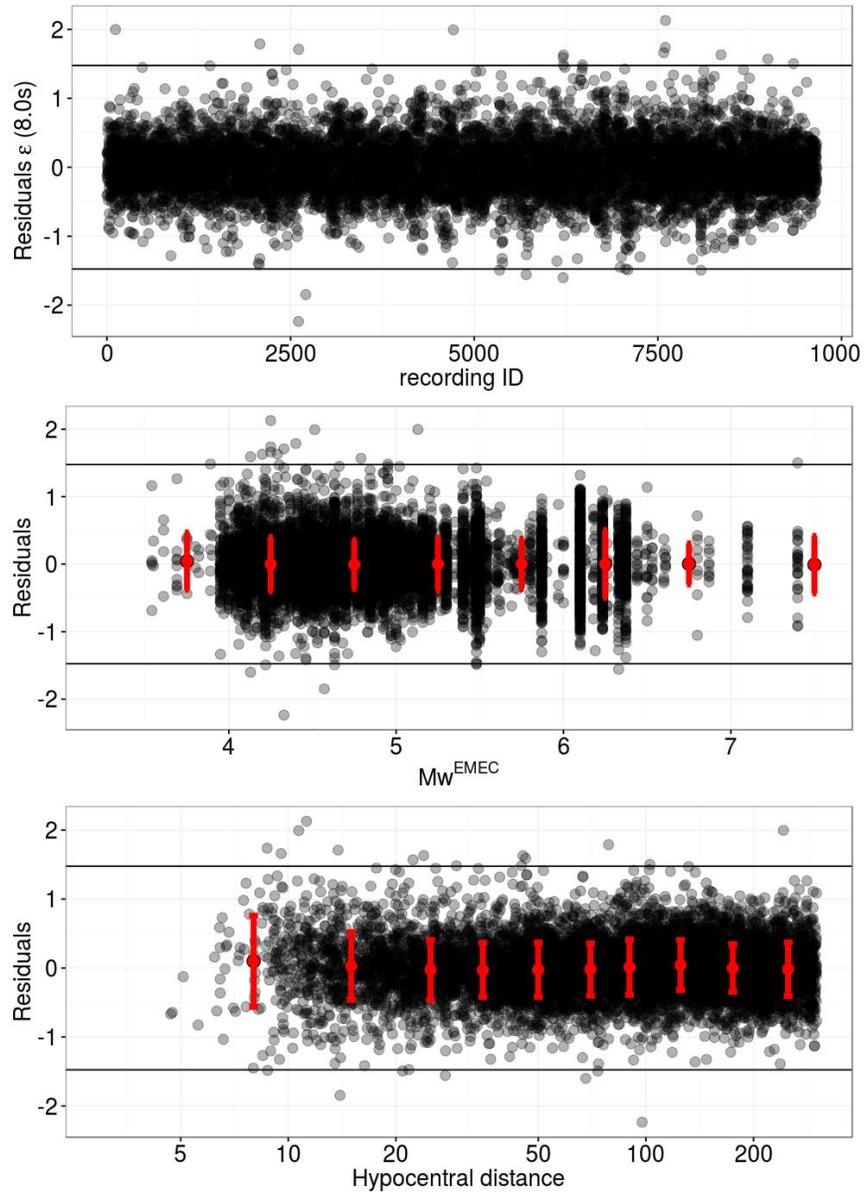


410 **Figure 3.** Period dependence of the asymptotic standard error on the median ( $\sigma_{\mu}$ ) for SA considering different magnitudes and a fixed distance of 30 km.



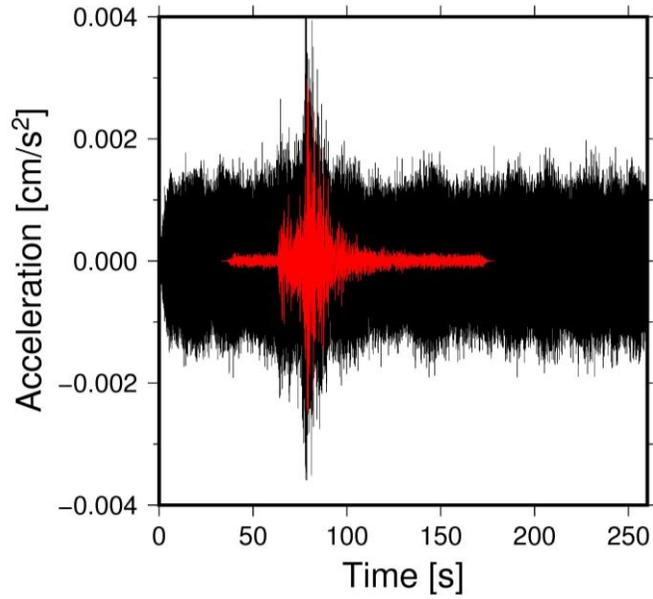
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**Figure 4a.** Residuals  $\varepsilon$  computed for SA at  $T=0.01s$ . Horizontal lines correspond to  $\pm 3\phi_0$ . The average residuals  $\pm 1$  standard deviation computed over magnitude and distance bins are shown as red bars.



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**Figure 4b.** The same as Figure 4a but for  $T=8s$ .



**Figure 5** Record (EW component) of event IT-2016-0001 at station IV.INTR: comparison between raw data (black) and processed acceleration (red). The processed acceleration has been filtered over the bandwidth 0.15-30Hz using a second order a-causal Butterworth filter.

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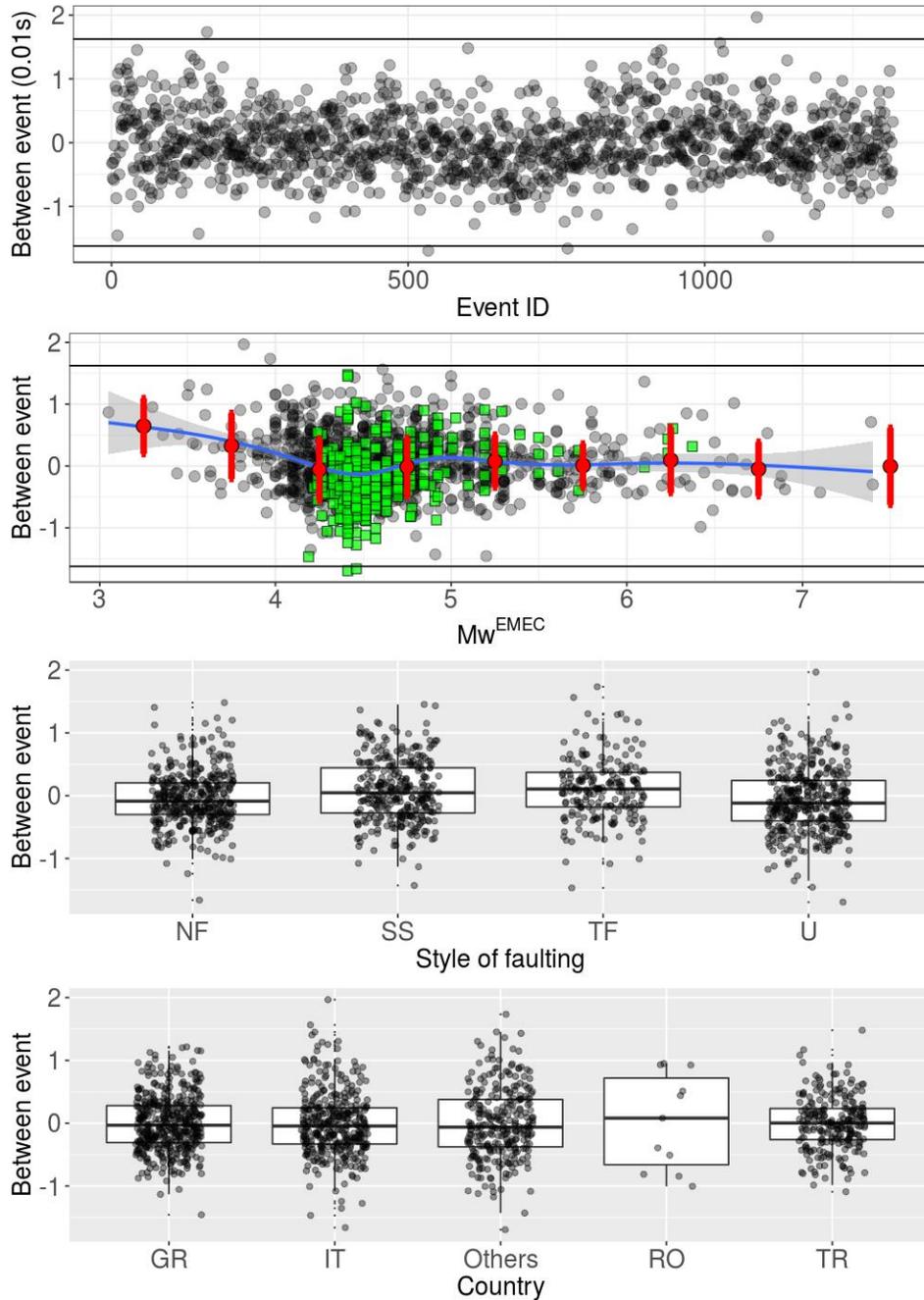


Figure 6a. Between-event residuals  $\delta B_e$  at  $T=0.01$  s for SA. Green squares indicate earthquakes with  
 435  $M_w$  estimated from ML using equation (2). Horizontal lines correspond to  $\pm 3\tau$ . The following  
 abbreviations are used for the style of faulting: NF- normal faulting; SS – strike slip; TF – thrust faulting; U  
 – unknown style of faulting. The country codes are: GR- Greece; IT – Italy; RO – Romania; TR- Turkey;  
 Others – any other country different from the previous ones.

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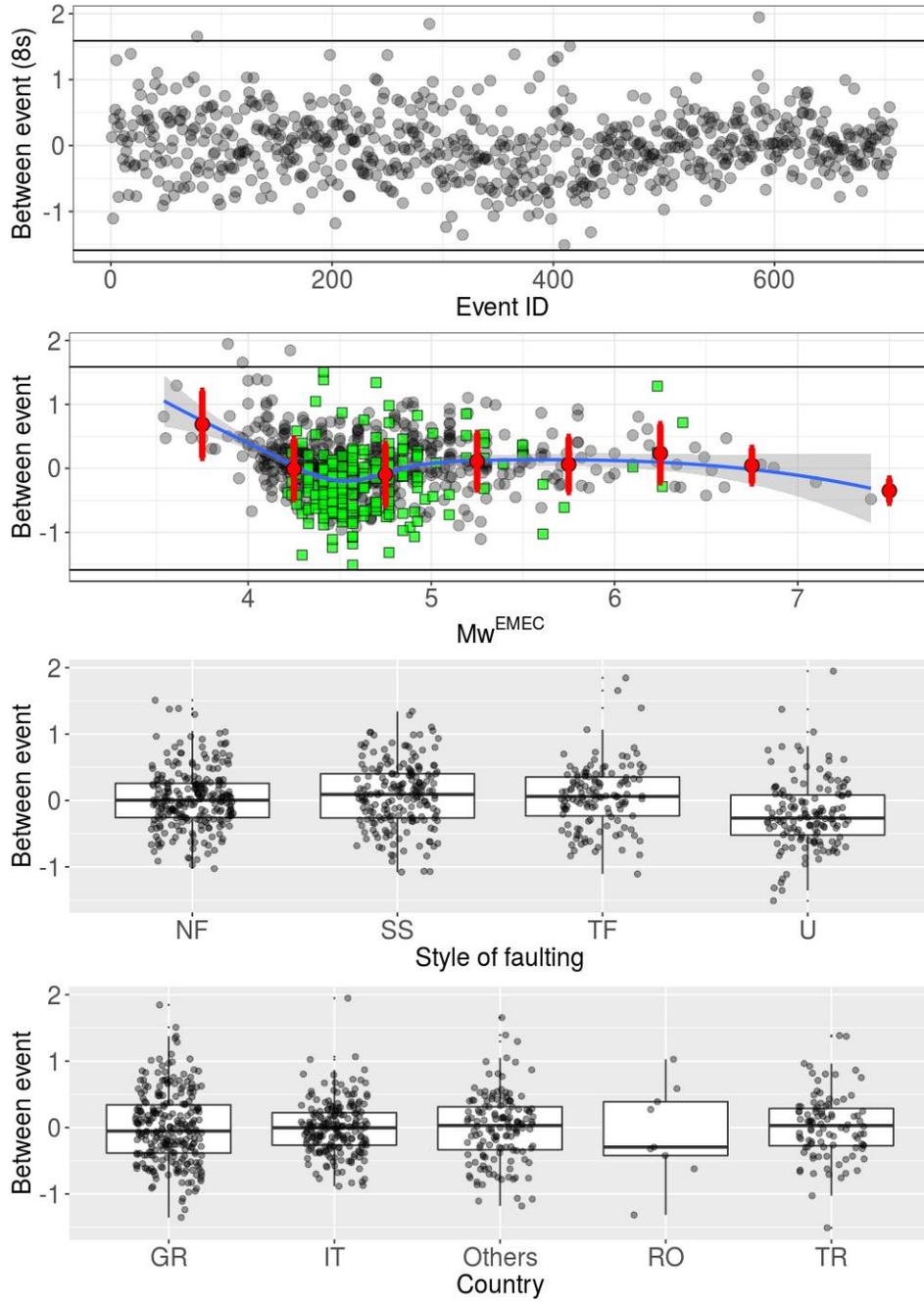


Figure 6b. The same as Figure 6a but for  $T=8$  s

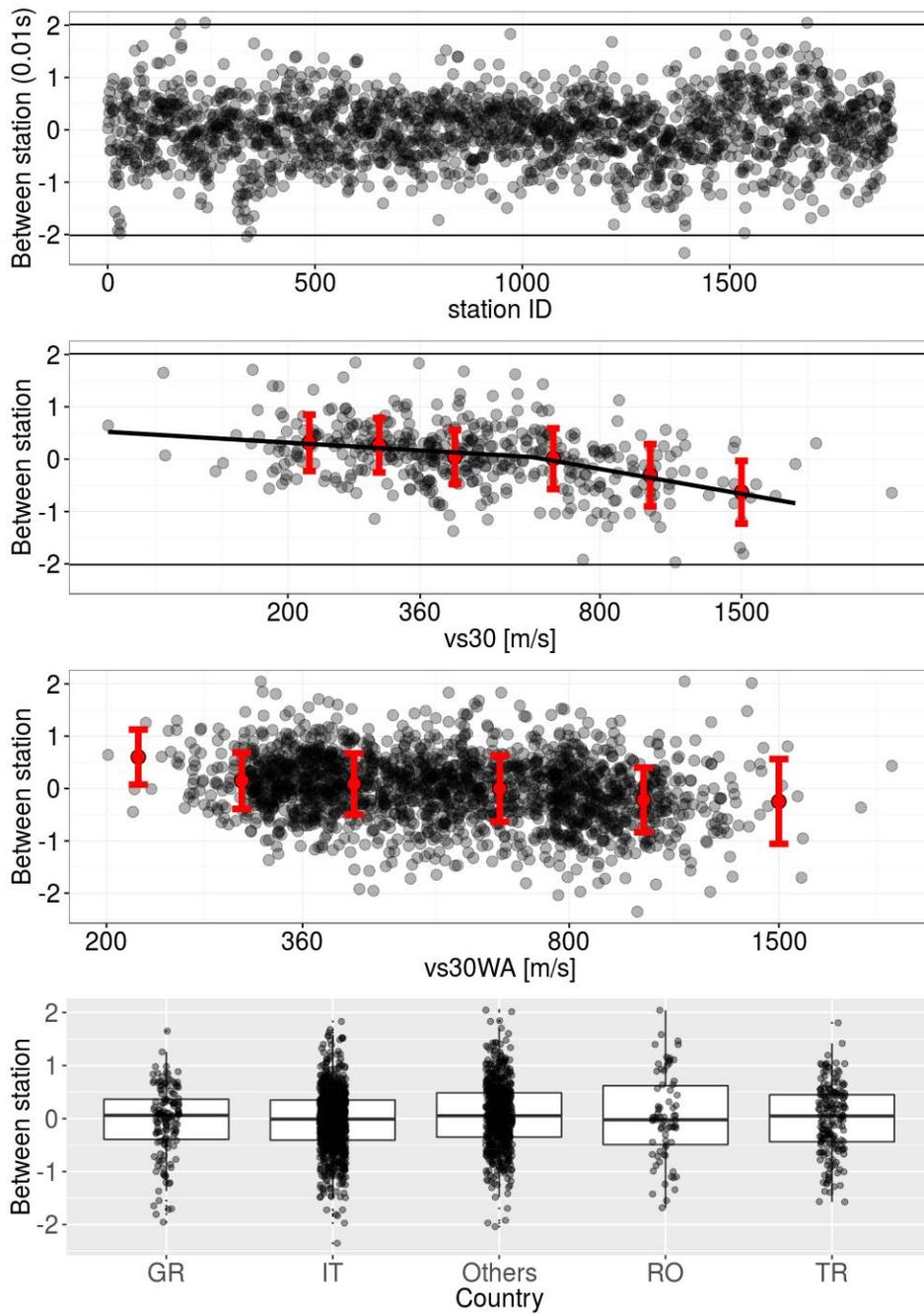
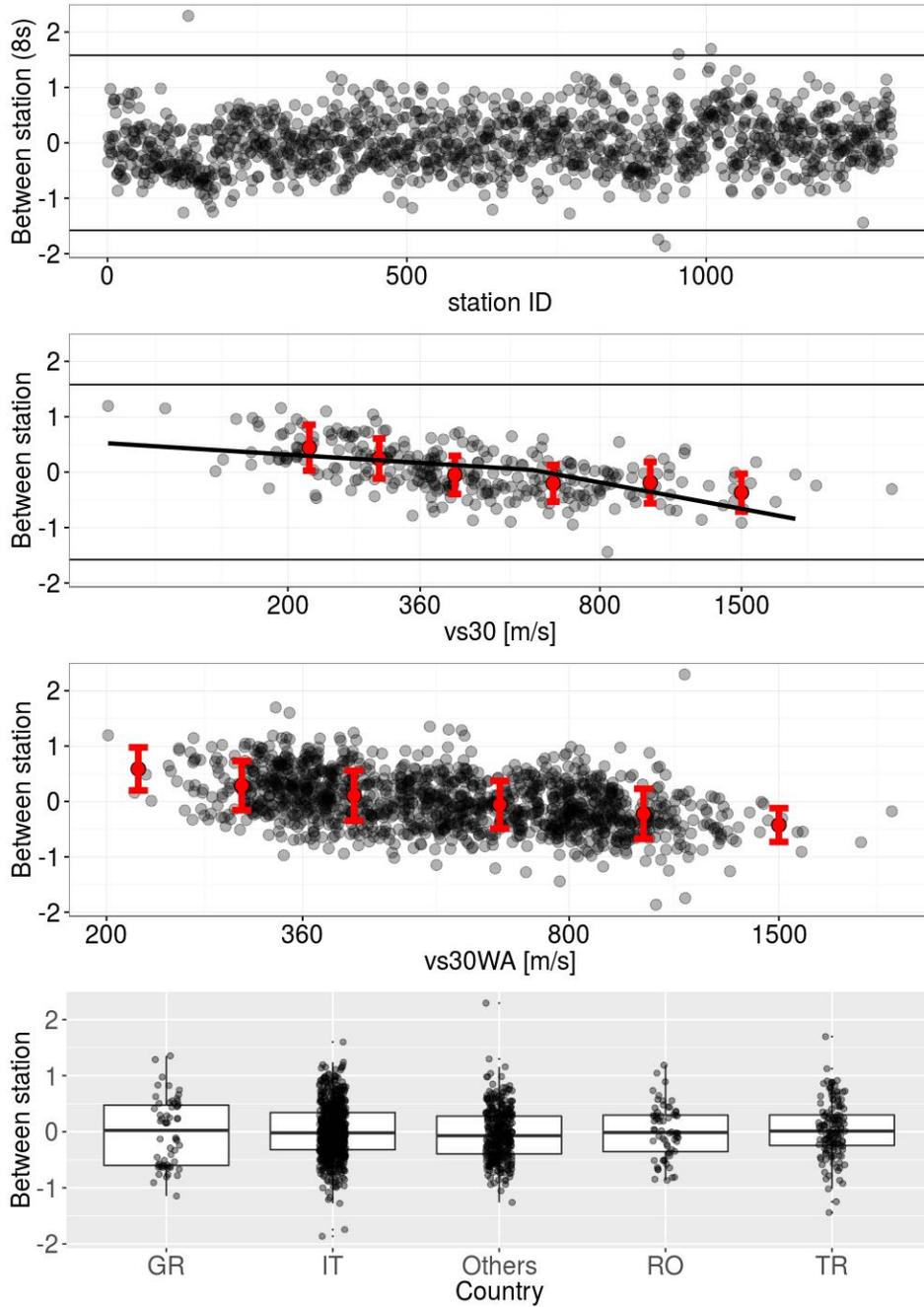


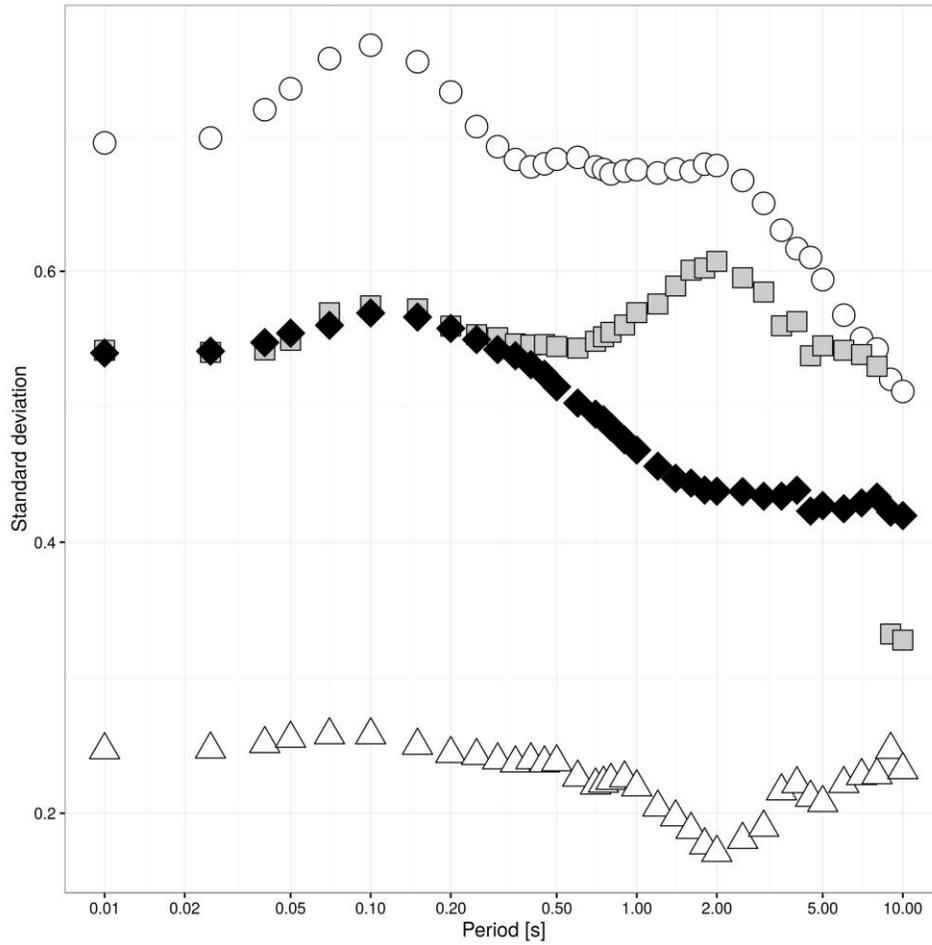
Figure 7a. Between-station residuals  $\delta S_{2S_s}$  at  $T=0.01$  s for SA. Horizontal lines correspond to  $\pm 3\phi_{S_{2S_s}}$ .



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**Figure 7b.** The same as Figure 7a but for  $T = 8$  s.

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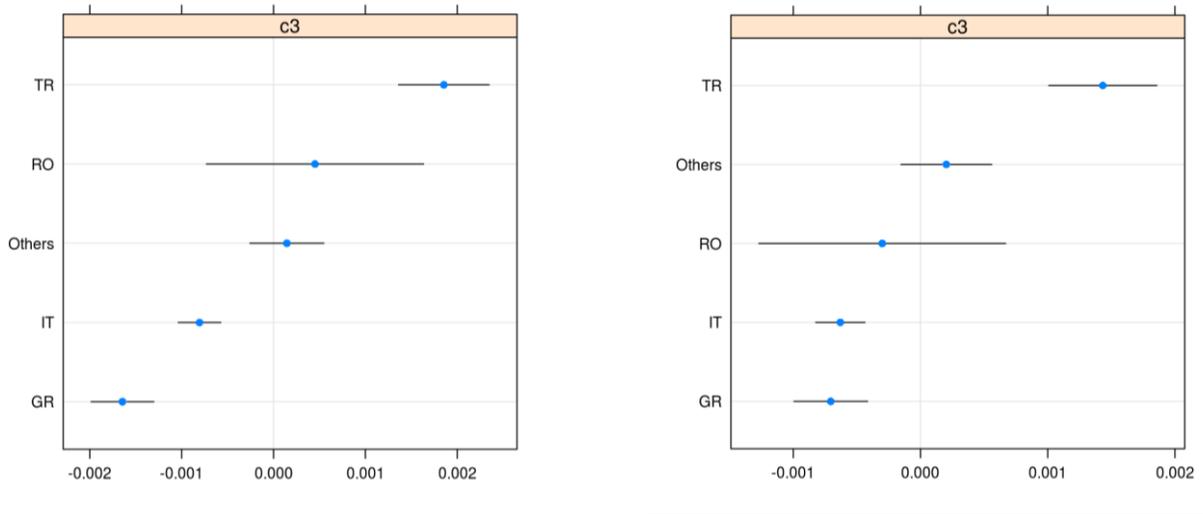
**Figure 8.** Standard deviation of different residual distributions for SA ( $\delta C2C_c$  applied to the offset coefficient  $e_1$ ). White circle: standard deviation of the between-station distribution (i.e.,  $\phi_{s2s}$ ); gray square: standard deviation of the between-event distribution (i.e.,  $\tau$ ); white triangle: standard deviation of the between-country distribution ( $\phi_{c2c}$ ); black diamond: standard deviation of the residual distribution ( $\phi_0$ ).

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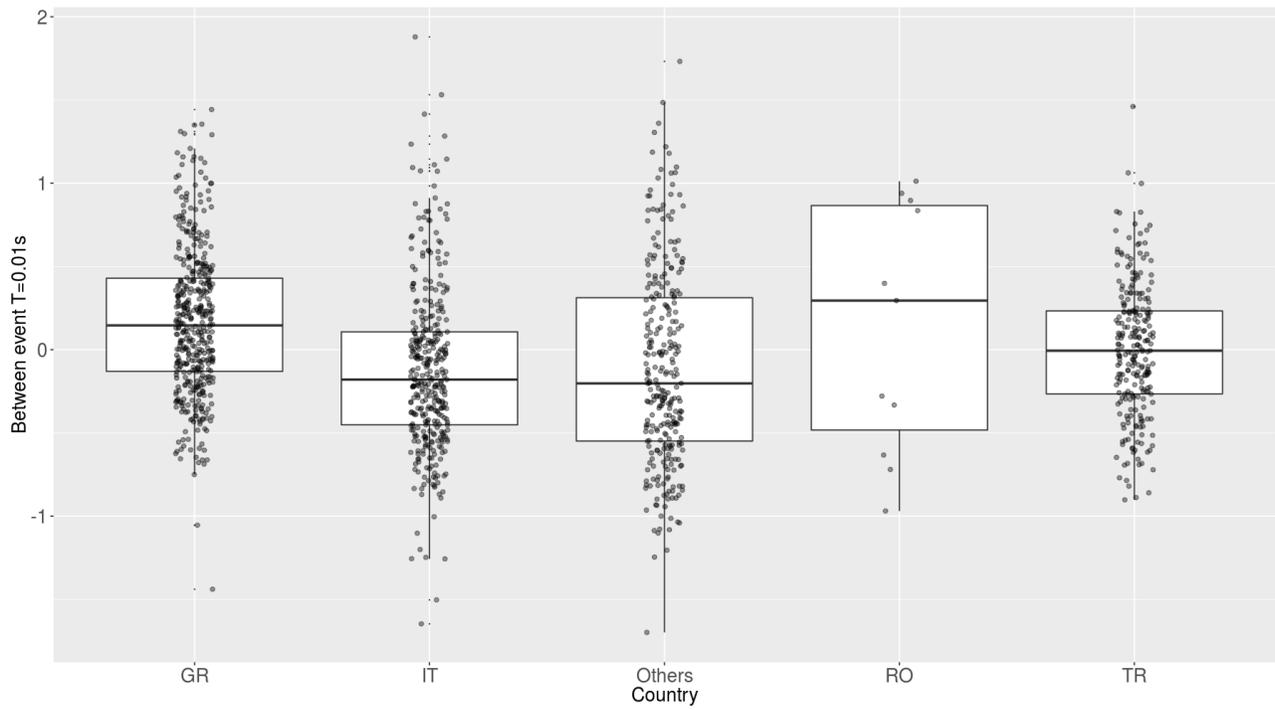
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**Figure 9.** Random-effect on  $c_3$  (aneleptic parameter) depending on the country grouping level for  $T=0.01s$  (left) and  $T=2s$  (right), considering SA. Please, note the different order for the country classes and the different horizontal scale.

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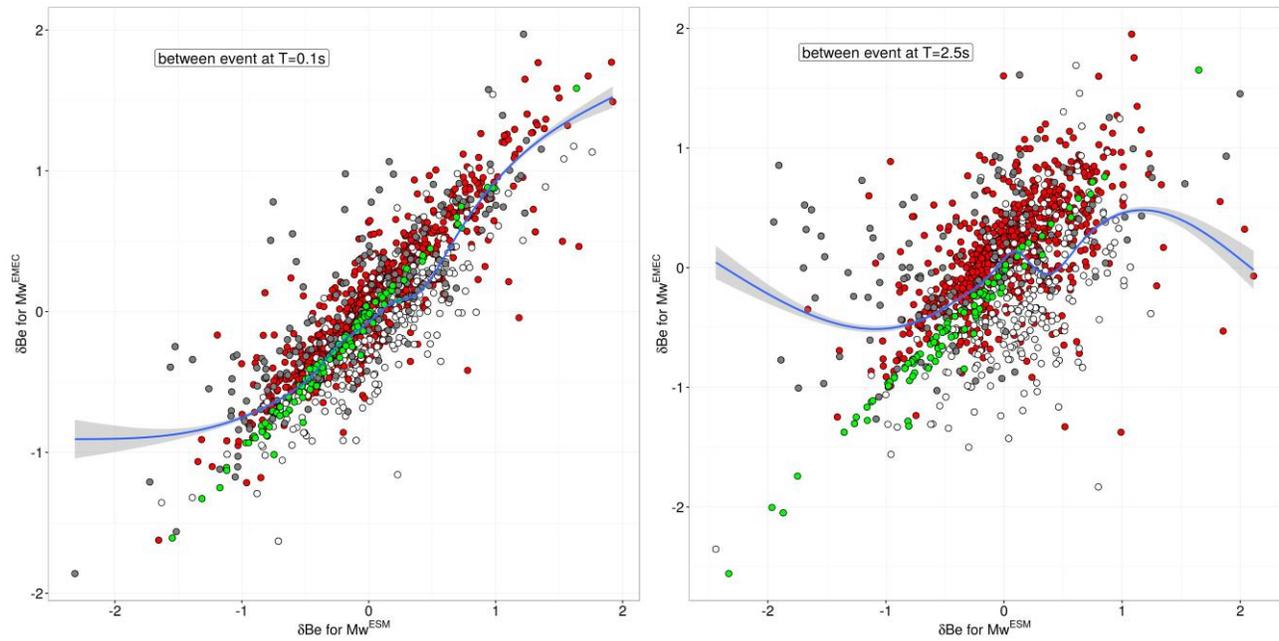
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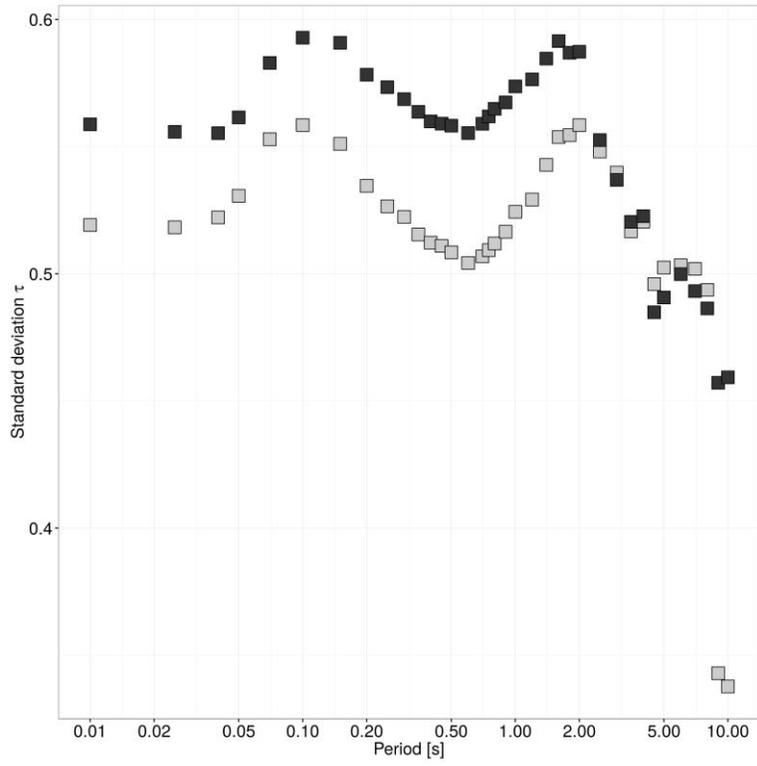
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**Figure 10.** Between-events  $\delta B_e$  for SA at  $T=0.01s$  applying the country-based random-effect  $\delta C2C_c$  to the anelastic attenuation parameter  $c_3$ .

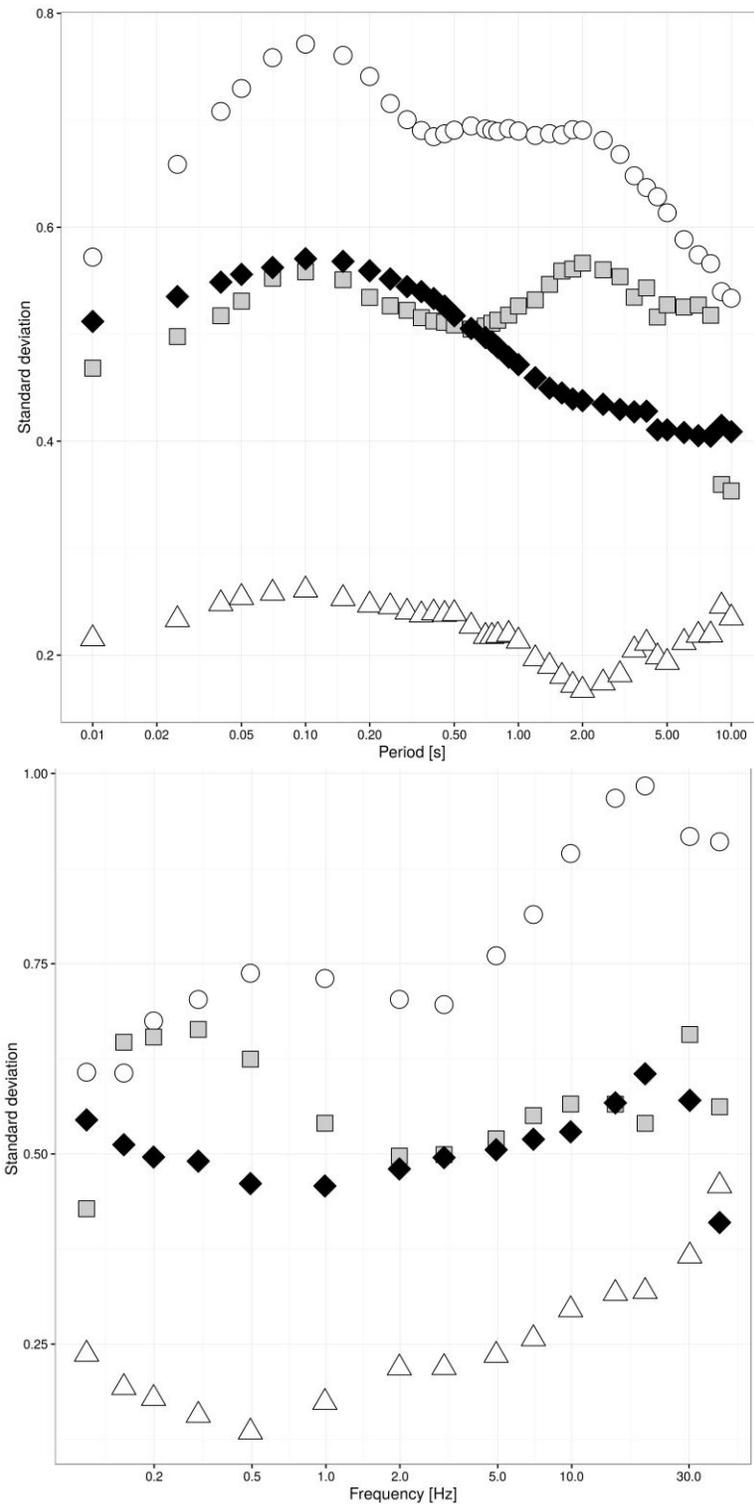
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**Figure 11.** Comparison between  $\delta Be$  obtained using  $M_w$  from EMEC and ESM, considering  $T=0.1s$  (left) and  $2.5s$  (right). The colors represent different selections on magnitude. In particular, only red represents recordings with both  $M_w^{EMEC}$  and  $M_w^{ESM}$  available, whereas the other colors indicate conversions from ML to  $M_w$ . In particular, the green dots represent earthquakes for which both  $M_w^{ESM}$  and  $M_w^{EMEC}$  are derived from the ML (and therefore showing high correlation). The blue curves are the results of a local polynomial regression (LOESS), giving an idea about the degree of correlation between the two  $\delta Be$  estimates.

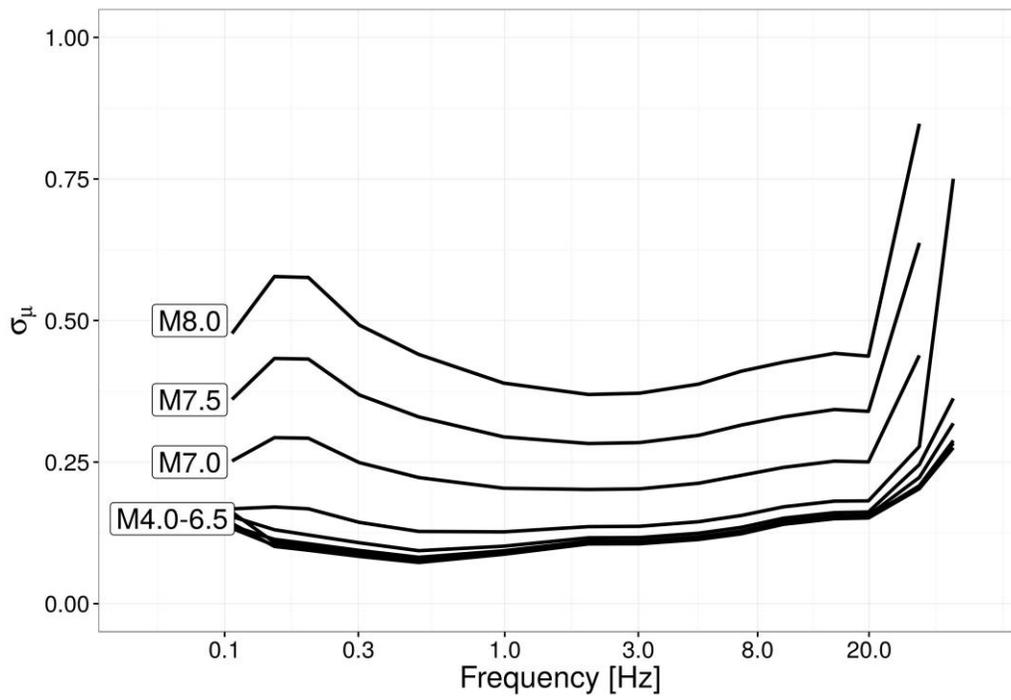


540 **Figure 12.** Between-event standard deviation  $\tau$  for the regression using  $Mw^{EMEC}$  (gray) and  $Mw^{ESM}$  (black)



**Figure 13.** Standard deviation of residual distributions for spectral displacement-SD (top), according to period, and for Fourier Amplitude Spectra-FAS (bottom), according to frequency. White circle: standard deviation of the between-station distribution ( $\phi_{S2S}$ ); gray square: standard deviation of the between-event distribution ( $\tau$ ); white triangle: standard deviation of the between country distribution ( $\phi_{C2C}$ ); black diamond: standard deviation of the residual distribution ( $\phi_0$ ).

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550 **Figure 14.** Frequency dependence of the asymptotic standard error on the median ( $\sigma_{\mu}$ ) for FAS considering different magnitudes and a distance of 30km .

[Click here to view linked References](#)

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## The pan-European Engineering Strong Motion (ESM) flatfile: ~~sanityconsistency~~ check via residual analysis

Bindi D. (1), Kotha S-R.(1), Weatherill G.(1), Lanzano G.(2), Luzi L.(2), and F. Cotton (1,3)

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(1) German Research Centre for Geosciences GFZ, Potsdam, Germany

(2) Istituto Nazionale di Geofisica e Vulcanologia INGV, Milan, Italy

(3) University of Potsdam, Potsdam, Germany

### 10 Abstract

We present the results of a ~~sanityconsistency~~ check performed over the flatfile extracted from the Engineering Strong Motion (ESM) data~~base~~. The flatfile includes 23014 recordings from 2179 earthquakes in the magnitude range from 3.5 to 7.8 ~~that~~ occurred since ~~the~~ 1970s in Europe and Middle East, as presented in the companion article by Lanzano et al. (2018a). The ~~sanityconsistency~~ check is developed by analyzing different residual distributions obtained ~~by developing from~~ ad-hoc ground motion prediction equations for the ~~absolute s~~Spectral ~~a~~Acceleration (SA), ~~D~~isplacement (SD) and Fourier ~~a~~Amplitude ~~s~~Spectra (FAS). Only recordings from earthquakes shallower than 40 km are considered ~~for in~~ the analysis. The between-event, between-station and event~~and~~-station corrected residuals are computed by applying a mixed-effect regression. We identified those earthquakes, stations, and recordings showing the largest deviations from the ~~predicted GMPE~~ median ~~models predictions~~, and ~~we also~~ evaluated the statistical uncertainty on the median model to get insights ~~about on~~ the applicable magnitude-distance ranges and the usable period (or frequency) range. We observed that robust ~~median predictions mean~~ are obtained up to 8s for SA and up to 20 Hz for FAS, although median predictions for  $M_w \geq 7$  show significantly larger uncertainties with 'bumps' starting above 5s for SA and below 0.3 Hz for FAS. The between-station ~~variability variance~~ dominates over the other residual ~~components variances~~, and the dependence of the between-station residuals on logarithm of  $V_{s30}$  is well~~described~~ by a piece-wise linear function with period-dependent slopes and hinge velocity around 580 m/s. Finally, we compared the between-event residuals obtained by considering two different ~~choices sources of for the~~ moment magnitude. The results show that, at long periods, the between-event ~~terms~~ from the two regressions have a weak correlation and the overall between-event variability is

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different/dissimilar, confirming/highlighting the importance of magnitude source played in the regression results by the implemented moment magnitude values.

## Introduction

35 Most of the applications in Engineering Seismology and Earthquake Engineering deal with the analysis of strong-motion data. In Europe, the dissemination of uniformly processed strong-motion waveforms accompanied by event and station meta-data progressed since the 1980s (Ambraseys, 1990; Ambraseys and Bommer, 1991; Ambraseys et al., 2000; Ambraseys et al 2004). The two most recent examples are the Reference Database for Seismic Ground-motion Prediction in Europe-RESORCE (Akkar et al.,  
40 2014), developed in the framework of the Seismic Hazard HARMONIZATION in Europe project (SHARE; www.share.eu.org), and the Engineering Strong Motion (ESM) data-base (Luzy et al., 2016) developed in the framework of the Network of European Research Infrastructures for Earthquake Risk Assessment and Mitigation project (NERA). For many applications aimed at developing and testing Ground-Motion Prediction Equations (GMPEs), ground-motion intensity measures of engineering interest (e.g.,  
45 peak ground acceleration-PGA, peak ground velocity-PGV, response spectra, Housner intensity, etc) are extracted or computed from the waveforms, and disseminated, along with the relevant event and station metadata, in the form of parametric tables, often referred to as flatfiles. Examples of recent flatfiles include the Pacific Earthquake Engineering Research Center (PEER) flatfile NGA-West-2 (Ancheta et al., 2014) and regional flatfiles like those for Japan (Dawood et al., 2016) and Chile (Bastias and Montalva, 2016).  
50

In this work, we focus on the flatfile generated by from the ESM data-base, as described in details in a companion paper (Lanzano et al., 2018a). In order to support the development of ground-motion models from the ESM flatfile, we perform a sanity consistency check over the data and metadata from the perspective of the ground-motion variability, to report possible anomalies that a GMPE developer might  
55 encounter. To accomplish this task, we first calibrate a set of GMPEs for spectral acceleration (SA), spectral displacement (SD) and Fourier amplitude spectra (FAS), considering simple models describing the distance and magnitude scaling. Then, the random-effects and residual distributions obtained by performing a mixed-effects regression (Bates et al., 2015) are analyzed to detect anomalous records, events and stations. Issues that could be of relevance for ground-motion models, such as those related  
60 to the selected magnitude scale, or and to ground-motion and path regionalization, are discussed as well.

### Data selection for developing the reference model

In this study, we use the 2018 version of the ESM flat-file (Lanzano et al., 2018b). The characteristics of the ESM flatfile are described in Lanzano et al. (2018a) whereas the waveforms processing is detailed in  
65 Puglia et al. (2018). For developing the reference ground-motion model, we select only data at distances shorter than 300 km, generated by earthquakes with at least two recordings, and occurring at depths shallower than 40 km. The lower limit of hypocentral depth is set to select only the Active Shallow Crustal Earthquakes, whereas the deeper continental and subduction events are left out of this study considering their relatively complex phenomenon which requires ad-hoc studies. For SA and SD, each  
70 record used in the GMPE regression for period  $T$  should satisfy the usable period criterion  $T \leq 0.8/f_{HP}$ , where  $f_{HP}$  is the high-pass corner frequency of the filter applied for in the waveform processing the waveform stage. For FAS computed at frequency  $f$ , only spectral amplitudes in the range  $1.2 f_{HP} \leq f \leq f_{LP}/1.2$ , where  $f_{LP}$  is the low-pass corner-frequency of the filter, are considered in deriving the GMPE. The distributions of the corner-frequencies of the filter applied to all recordings are provided in Lanzano et  
75 al. (2018b). In particular, most of the  $f_{LP}$  frequencies lie in the range 30-40 Hz. Regarding the station selection, we only select stations installed at surface (sensor depth <10 m) avoiding borehole installations (identified by *housing\_code* equal to "WEL") and stations installed at building floor (*installation\_code* equal to "BF"). For our sanity consistency check, we also select sensors with *location\_code* equal to "00".

80 Regarding the magnitude, the flatfile includes multiple magnitude scales and estimates. In particular, the local magnitude ( $M_L$ ) and two choices for the moment magnitude ( $M_w$ ), namely the moment magnitude listed in the ESM data-base (hereinafter referred to as  $M_w^{ESM}$ ) and the moment magnitude from the EMEC catalog (Grünthal et al., 2012; hereinafter referred to as  $M_w^{EMEC}$ ). In particular, while  $M_w^{ESM}$  is computed from moment tensor solutions provided by different sources selected according to an  
85 hierarchical schema (Lanzano et al., 2018b),  $M_w^{EMEC}$  is based on both moment tensor solutions and empirical regional conversions from other magnitude scales. The two  $M_w$  scales are compared in Figure 1a, where the best fit line is  $M_w^{EMEC} = 0.23 + 0.95 M_w^{ESM}$ , being with the residual standard deviation equal to 0.16. Since neither  $M_w^{ESM}$  nor  $M_w^{EMEC}$  is available for all events, we derive an empirical relationship between  $M_L$  and  $M_w$  in order to extend the residual analysis to as large as possible a number of events  
90 as possible. These relations are not proposed for applications different from those performed in the context of this study. The best fit lines shown in Figure 1 (right) are:

$$M_w^{ESM} = 6.81 - 1.67 ML + 0.26 ML^2 \quad (\pm 0.3) \quad (1)$$

$$M_w^{EMEC} = 6.18 - 1.34 ML + 0.23 ML^2 \quad (\pm 0.3) \quad (2)$$

## 95 GMPE functional form

Mixed-effects regressions (Brillinger and Preisler, 1985; Bates et al., 2015) are performed for the natural logarithm of SA, SD and FAS considering simple parametric functional forms for describing the median scaling with distance and magnitude:

$$100 \quad \log Y = e_1 + F(M_w) + G(R_{hypo}) + RE + \epsilon \quad (3)$$

$$F(M_w) = \begin{cases} b_1(M_w - M_{ref}) & \text{if } M_w \leq M_H \\ b_1(M_H - M_{ref}) + b_3(M_w - M_H) & \text{otherwise} \end{cases} \quad (4)$$

$$G(R_{hypo}) = \begin{cases} c_{1A} \log(R_{hypo}) & \text{if } R_{hypo} \leq R_H \\ c_{1A} \log(R_H) + c_{1B} \log\left(\frac{R_{hypo}}{R_H}\right) + c_3 R_{hypo} & \text{otherwise} \end{cases} \quad (5)$$

105 where in equation (1) the random effect (RE) are

$$RE = \delta B_e + \delta S_2 S_s + \delta C_2 C_c \quad (6)$$

In equations (4) and (5), the reference and hinge magnitude and distances are set to  $M_{ref}=4.5$ ,  $M_H=6$ ,  $R_H=15\text{km}$ , respectively. The country-to-country random effect ( $\delta C_2 C_c$ ), where  $c$  indicates country, is applied either on the offset term ( $e_1$ ) or to  $c_3$  (that is, two different regressions are performed). The other random-effects applied to the offset are the customary between-event ( $\delta B_e$ ) and between-station ( $\delta S_2 S_s$ ) residuals, where subscripts  $e$  and  $s$  are event and station indices. Residual  $\epsilon$  in equation (3) represents the event- and station-corrected record-to-record variability. The standard deviations of  $\epsilon$ ,  $\delta S_2 S_s$ ,  $\delta C_2 C_c$ , and  $\delta B_e$  random distributions are indicated with  $\phi_0$ ,  $\phi_{S_2 S_s}$ ,  $\phi_{C_2 C_c}$  and  $\tau$ , respectively. The

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115 regressions for SA and SD are performed over the orientation independent (rotD50) component ground-  
motion measure (Boore, 2010), while the geometric ~~at~~-mean of the horizontal components is used for  
FAS GMPEs.

For the GMPEs regressed on SA considering  $\delta C_2 C_3$  on the offset  $e_1$ , Figure 2 shows the period-dependent  
coefficients along with their standard ~~errors~~. Regarding the scaling with distance, coefficient  $c_{1B}$   
120 increases from about -2 at  $T < 0.1s$  to about -1.3 for  $T > 0.5s$  while coefficient  $c_{1A}$  is maximum around  
 $T = 0.1s$  and decreases below -2 for  $T > 4s$ . The term  $c_3$  is negative for short periods and turns positive for  $T$   
 $\geq 1.5s$ . The positive values of  $c_3$  suggest to investigate ~~a~~ further segmentation of the distance scaling  
component to allow capturing the effects of later ~~arrivals~~ on long period ground ~~motion~~ at distances  
beyond 80km (e.g., Lanzano et al., 2016; Kotha et al., 2018).

125 Regarding the scaling with magnitude, although large uncertainties affect  $b_3$  (describing the magnitude  
scaling above the hinge magnitude), its values are ~~which is~~ always smaller than  $b_1$ , suggesting weaker  
scaling at larger magnitudes. The trend of the coefficients suggest that regressions above 8s are unstable  
due to a strong reduction in the number of data used usable in for GMPE regressions (Lanzano et al.,  
2018a). For example, the number of recordings used for the regressions at 0.01, 4, 8, and 10 s are 18859,  
130 16222, 9664 and 2767, respectively. The statistical uncertainty asymptotic standard error ( $\sigma_\mu$ ) on the  
median predictions, quantifying the within-model statistical estimate of epistemic uncertainty affecting  
the median predictions (e.g., Al Atik and Youngs, 2014 ; Bindi et al., 2017a), due to the combined effects  
of limited data availability and complexity of implemented functional form (in terms of extent of GMPE  
parametrization) are shown in Figure 3. For  $M_w \leq 6.5$ ,  $\sigma_\mu$  is smaller than 0.2 and it shows a weak  
135 dependency on magnitude for  $M_w < 6$ . For  $M_w < 6.5$ ,  $\sigma_\mu$  reaches the minimum values around  $T = 2s$  while  
the sharp increase above  $T = 8s$  indicates that regressions ~~at~~ are not well ~~constrained~~ for these periods.  
~~For At~~  $M_w > 6.5$ ,  $\sigma_\mu$  increases significantly rapidly, exceeding 0.25 at  $M_w = 7.5$ . Above magnitude 7,  $\sigma_\mu$   
shows a bump around  $T = 5s$ , suggesting that at periods longer than 4s GMPE median predictions for  $M_w$   
 $> 7$  are not well constrained uncertain. The dependence of  $\sigma_\mu$  on distance (here not shown) is negligible  
140 because the data appears to constrain very well the distance scaling component of the GMPE (small  
errors on  $c_{1A}$ ,  $c_{1B}$ ,  $c_3$  in Figure 2).

## Results

145 The results are presented in terms of event- and station-corrected residuals ( $\epsilon$ ), between-event ( $\delta B_e$ ) and between-station ( $\delta S2S_e$ ) distributions evaluated for SA with regional attributes  $\delta C2C_e$  on  $e_1$ .

### Event- and station-corrected residuals $\epsilon$

150 The event- and station-corrected residuals for  $T=0.01s$  and  $T=8s$  are shown in Figure 4 as function of the recording Id (top), of magnitude (middle) and of distance (bottom). In the middle and bottom frames, the average residuals computed over magnitude and distance bins (red error bars) are shown as well. Please note that the number of recordings used in GMPE regression decrease towards long periods. The residuals do not show any significant trend neither with distance nor with magnitude. The number of residuals (in absolute value) above 3 times  $\phi_0$  is 60 at  $T=0.01s$  and 20 at  $T=4s$ , as shown in Tables 1 and 2, respectively. A further check of the waveforms does not reveal any peculiarities in the recordings except for the records [IT-2016-0001, IV.INTR], which is characterized by signal-to-noise-ratios small a high noise level at high frequencies, as shown in Figure 5.

### Between-event random-effects $\delta B_e$

160 The between-event random-effects for  $T=0.01s$  and  $T=8s$  are shown in Figures 6a and 6b for  $T=0.01s$  and  $T=8s$ , respectively.  $\delta B_e$  does not show significant trend for  $M > 4.5$ , while  $\delta B_e$  increases with decreasing magnitude for  $M < 4.5$ . Unlike the functional form implemented in this study, an additional segmentation of  $F(M_w)$  at  $M_w = 4.5$  (e.g, Campbell and Bozorgnia 2014, Kotha et al. 2018) would allow different scaling for small, moderate and large magnitude bins, removing the trend observed for small magnitudes. We note here that moment magnitudes below 4 are mostly generated not converted by conversions from local magnitude.

Differences in observed SAs based on Style-of-Faulting (SoF) are widely debated in GMPE development. For the dataset analyzed in this study, the box and whisker plots corresponding to  $\delta B_e$  dependence on SoF do not show any systematic differences. Since the model included a random-effect on at the Ccountry level (Italy-IT, Turkey-TK, Romania-RO, Greece-GR, other countries-Others), a dependence of  $\delta B_e$  on the country grouping is not expected, as shown by the box and whisker plots in Figures 6a and 6b. Events with  $\delta B_e$  larger than 3 times  $\tau$ , in absolute value, are listed in Table 3 and 4 for  $T=0.01$  and  $8s$ , respectively.

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175 **Between-station random-effects  $\delta S_{2S}$ .**

The between-station ~~residuals~~ random-effects for  $T=0.01s$  and  $T=8s$  are shown in Figures 7a and 7b ~~for~~  
 ~~$T=0.01s$  and  $T=8s$~~ , respectively. At short periods,  $\delta S_{2S}$  shows a trend with  $V_{s30}$ , as expected ( $\log_{10}(V_{s30})$   
is commonly used as proxy for site amplifications). The best-fit model obtained through a break-point  
regression (Muggeo, 2003) is described by a piece-wise linear function with break-point at  
180  $\log_{10}(V_{s30})=(2.78 \pm 0.09)$  (i.e.,  $V_{s30}=600$  m/s) and slopes  $(-0.6 \pm 0.2)$ ,  $(-1.7 \pm 0.40)$ . A weaker relation  
exists between  $\delta S_{2S}$  and  $V_{s30WA}$  (which is an estimate of  $V_{s30}$  derived from the topographic  
gradient, *see Lanzano et al., 2018b*). At long periods, the trend with measured  $V_{s30}$  is strong only for  
 $V_{s30} < 600$  m/s, while considering  $V_{s30WA}$  a correlation over a broader velocity range is observed. For  
 $T=8s$ , the break-point is located at  $\log_{10}V_{s30}=(2.77 \pm 0.09)$ , and the slopes are  $(-1.6 \pm 0.2)$  and  $(-0.3 \pm 0.2)$ .

185 When split among the different countries (box and whiskers plots in Figure 7), no residual offset is  
detected. Often the station  $V_{s30}$  distributions (mean and range) differ from region-to-region  
depending on the local geology over which the networks reside. As a result, systematic region-  
dependent offsets could be expected. In our analysis, since all the country-to-country variability is  
accounted by the  $\delta C_{2C}$  random-effects on  $e_1$ , we observe no differences in mean  $\delta S_{2S}$  among the  
190 countries. Nevertheless, it is worth of further investigation whether the scaling of  $\delta S_{2S}$  with  $V_{s30}$  is  
identical across various regions or not. Kotha et al. (2016) for instance, observed a similar (regionally)  
unbiased  $\delta S_{2S}$  for Italy and Turkey, but with significantly different scaling with  $V_{s30}$ . Stations with  
 $\delta S_{2S}$  larger than 3 times  $\phi_{s25}$ , in absolute value, are listed in Table 5 and 6 for  $T=0.01$  and  $8$  s,  
respectively.

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**Standard deviations of random-effects**

The standard deviations of the random-effects are summarized in Figure 8. The between-station  
variability  $\phi_{s25}$  dominates the total aleatory variability of the dataset ( $\sigma$ ): it shows a peak around  $0.10$  s  
and it decreases above  $2s$ . The peak at  $T = 0.1s$  is observed in several other studies as well (e.g. Zhao et  
200 al. 2016, Cauzzi and Faccioli 2017; Kotha et al. 2018), indicating a high variability of short period site-  
response in response spectra domain. The between-event standard deviation  $\tau$  has a peak located  
around  $2s$ , otherwise almost constant with a small bump around  $0.1$  s;  $\tau$  shows a suspicious drop at  $T=9s$

and 10s, probably due to the strong reduction in the usable recordings. The event- and station-corrected variability ( $\phi_0$ ) is close to  $\tau$  for  $T < 0.5s$ , while for longer periods it is smaller than the between-event variability. Finally, the between-country variability  $\phi_{c2c}$  (region-to-region variability of offset  $e_1$ ) is small (of the order of 0.25) with a trough around 2s. The anti-symmetry of  $\tau$  and  $\phi_{c2c}$  around  $T = 2s$  suggests a trade-off between event-to-event and country-to-country variability in this GMPE regression, suggesting the need of further investigating the formulation of REs in GMPEs.

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#### Regional differences on anelastic attenuation.

An additional regression is performed by applying the country-group random-effect not on the overall offset  $e_1$  in equation 3 but on the anelastic coefficients  $c_3$  of equation 5. Figure 9 shows the obtained random effects  $\delta c_3$ , which can be interpreted as regional adjustment to the generic region-independent “anelastic attenuation” term  $c_3$ . The estimation errors on these country-specific adjustments (to  $c_3$ ) are statistically significant with small standard errors (horizontal bar in the caterpillar plot of Figure 9), suggesting considerable differences in propagation effects across these regions. For instance, since the  $\delta c_3$  value for Romania is not well-constrained with the very few data from that region, its standard error is larger than that for Italy, Greece, and Turkey. For periods between 0.01 and 2 s, Greek and Italy show stronger attenuation than Turkey; for longer periods we recall that  $c_3$  becomes positive.

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Since in this case the country-based RE is not applied to the overall offset, the between-events show some variability with across the country class (Figure 10). Note that since  $\delta C2C_e$  is no longer applied on  $e_1$ , systematic country-to-country variability in  $\delta B_e$  and  $\delta S2S_e$  distributions may now appear, as seen in the between-event distributions of Figure 10. For example, on average the events from Greece (GR) show a higher  $\delta B_e$  compared to those from Turkey (TR) and Italy (IT) (Figure 9).

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#### $\delta B_e$ obtained using either $M_w^{EMEC}$ or $M_w^{ESM}$ .

Several studies (e.g., Rohades, 1997; Moss, 2011; Kuehn and Abrahamson, 2018) evaluated the impact of the uncertainty in the explanatory predictor variables on the aleatory variability. The effect of the

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magnitude uncertainty on  $\delta B_e$  variability havehas been observed in Europe for the RESORCE flatfile by comparing Mw values computed following different procedures (Kotha et al., 2016). For the ESM flatfile, we assess the impact of using  $Mw^{EMEC}$  or  $Mw^{ESM}$  by repeating the regression for each magnitude type, with  $\delta C2C_e$  applied on  $e_1$ . The results are compared in Figure 11 in terms of  $\delta B_e$  for  $T=0.1$  and  $2.5s$ . The red dots in Figure 11 indicate recordings for which both  $Mw^{EMEC}$  and  $Mw^{ESM}$  are available, while the other colors indicate earthquakes for which at least one of the two Mw values is derived from M<sub>L</sub>. In particular, the green dots are for earthquakes without Mw (neither EMEC nor ESM). Figure 11 shows that at short periods  $\delta B_e$  for the two regressions show a good correlation although with a large scatter. On the other hand, at longer periods ( $T=2.5s$  in Figure 11 but results for longer periods are similar) the correlation is very weak (being trivial the correlation shown by the green points). Figure 12 shows the standard deviation  $\tau$  of the  $\delta B_e$  residuals for the two regressions, considering both actual Mw values and values derived from M<sub>L</sub>.  $\tau$  for  $Mw^{ESM}$  is higher than the one obtained using  $Mw^{EMEC}$  for periods  $T < 2.0s$  whereas, from  $T=2.0$  to  $8s$ , similar values are obtained for the two regressions.

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#### 245 **Summary for SD and for FAS**

The ESM flatfile includes several intensity measure of engineering interests, such as spectral acceleration (SA) and displacement (SD); Fourier amplitude spectra (FAS); peak ground velocity (PGV) acceleration (PGA) and displacement (PGD); Arias intensity, Housner intensity, effective duration, cumulative absolute velocity (CAV). While the check of all those parameters is beyond the aim of this work, the residual analysis have been extended to SD and FAS as well (Bindi et al., 2017b). Figure 13 summarizes the results by comparing the standard deviations of the residual distributions. The main conclusions drawn from SD and FAS analysis agree with those from SA. Regarding the uncertainty on the median model for FAS, the dependency of  $\sigma_\mu$  on frequency is shown in Figure 14. The models are robust up to 20 Hz, for higher frequencies the number of available recording decreases due the condition applied to the usable frequency range (see Lanzano et al., 2018b, for the distribution of the low-pass corner frequencies of the filter applied during the processing phase) and the uncertainty becomes very large. Towards low frequencies,  $\sigma_\mu$  for  $Mw \geq 6.5$  shows a bump around 0.2 Hz. In general, in agreement with the results obtained for SA,  $\sigma_\mu$  significantly increases above magnitude 7.

#### 260 **Conclusion**

We performed residuals analysis over the ESM flat-file (Lanzano et al., 2018b) as complementary information to the flatfile description (Lanzano et al., 2018a), for supporting those studies aiming at developing ground-motion models. The main outcomes of the sanityconsistency check performed on recordings from shallow earthquakes (hypocentral depth less than 40 km) can be summarized as follow:

265 1) considering the data population and the selection criteria applied to define the usable period range, robust results for response spectra have been obtained for periods up to 8s, although the median models for  $M_w \geq 7$  are affected by large within-model epistemic uncertainty ( $\sigma_\mu$ ) above 5s. Regarding the model for Fourier,  $\sigma_\mu$  increases significantly above 20Hz and shows a bump around 0.2 Hz for magnitudes above 7. The epistemic uncertainty for both response and Fourier spectra increases  
270 significantly for  $M_w \geq 6.5$ ;

2) the sanity check has been performed considering a simple functional form for describing the magnitude scaling and the scaling with hypocentral distance, although for larger finite-source events the Joyner-Boore or the rupture distances should be considered. Regarding the distance scaling, a piece-wise linear function of logarithm of distance was needed to avoid trends in the residuals at distances shorter  
275 than 15km (see also Kotha et al., 2018). Dependence of the hinge distance on magnitude and period is worth of future investigation in order to model the magnitude dependence of geometric spreading as well;

3) regional dependencies of the attenuation with distance (so-called apparent anelastic attenuation parameter) already observed in previous studies (e.g., Kotha et al., 2016) are confirmed. In addition, it is  
280 now possible to refine the regionalization scheme given the significantly larger amount of data from a few more regions;

4) regarding the scaling of SA with magnitude, it is worth considering a second hinge magnitude at small magnitudes, around 4 (see also Kotha et al 2018);

5) the scaling of the between-station residualsrandom-effects  $\delta S_{2S_3}$  with the logarithm of  $V_{vs30}$  is well  
285 described by a piece-wise linear function with period-dependent slopes and hinge velocity around 580 m/s;

6) magnitude is one of the two fundamental explanatory variables used to develop ground motion models, and the flat-files includes information about different magnitude scales as extracted from the ESM data-base. In particularaddition, two different moment magnitude values are provided (when

290 available):  $M_w^{ESM}$  is the moment magnitude as included in ESM without any conversion (a reference for  
the source is provided);  $M_w^{EMEC}$  is the moment magnitude according to the EMEC catalog (Grünthal et  
al., 2012). ~~Being based on~~ Derived through different procedures, the two magnitudes show an overall  
agreement (linear correlation) but with large scatter around the mean trend. The between-event  
residuals  $\delta B_e$  at long periods absorb the uncertainties in the seismic moment estimates: therefore, the  
295 weak correlation shown by the  $\delta B_e$  distribution obtained considering the two different moment  
magnitudes warns about the critical role played by the magnitude selection;

7) ~~tables with rR~~ records, events, and stations showing standardized residuals larger, in absolute value,  
than 3 are listed in tables to allow for a comparison with other models.

8) among different intensity measures, Fourier amplitude spectra allow an easier and direct comparison  
300 between components of the observed ground-motion variability and seismological interpretations.

#### Data and Resources

In this study, we analyzed the ESM flatfile (Lanzano et al, 2018b) with persistent identifier PID:  
11099/ESM\_flatfile\_2018. The flatfile can be downloaded from the ESM webpage: <http://esm.mi.ingv.it>.

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390 **Table 1.** Event- and station-corrected residuals greater than 3 times  $\phi_0$  in absolute value

( $|\varepsilon| > 3\phi_0$ ) at T=0.01s

Event ID	Station	$\varepsilon(0.01s)$	Event ID	Station	$\varepsilon(0.01s)$
EMSC-20101109_0000039	FR.REVF	-1.943414	EMSC-20161129_0000079	IT.AQG	-1.870889
EMSC-20110707_0000048	RA.IRPV	2.168072	EMSC-20161129_0000079	IT.AQV	-1.864027
EMSC-20130125_0000039	IV.BDI	2.24683	EMSC-20161129_0000079	IT.GNU	-1.864027
EMSC-20130829_0000020	KO.GONE	-1.996238	EMSC-20161129_0000079	IT.PSC	-1.912796
EMSC-20140907_0000025	IT.BRM	1.882042	EMSC-20161129_0000079	IT.SPD	-2.056146
EMSC-20140907_0000025	IT.PRT	3.342786	EMSC-20161129_0000079	IT.SULA	-1.839875
EMSC-20141216_0000014	KO.GONE	-1.881109	EMSC-20161129_0000079	IT.SULC	-1.954815
EMSC-20141219_0000039	IT.FOC	-1.853693	EMSC-20161129_0000079	IV.BSSO	-1.938496
EMSC-20141224_0000053	IT.SULC	1.87239	EMSC-20161129_0000079	IV.CERA	-2.019819
EMSC-20150130_0000003	IV.OPPE	1.919609	EMSC-20161129_0000079	IV.MOMA	2.029745
EMSC-20150130_0000003	NI.POLC	-2.251522	EMSC-20161129_0000079	IV.RNI2	-2.023737
EMSC-20150130_0000003	RF.GEPF	-1.924863	EMSC-20161129_0000079	IV.VAGA	-1.965676
EMSC-20150205_0000061	HL.ZKR	2.124032	EMSC-20161201_0000064	IT.CNE	1.943368
EMSC-20150726_0000076	HL.PLG	2.325277	GR-1999-0001	HL.PREA	3.271508
EMSC-20150913_0000012	KO.CAME	2.257234	IT-1987-0003	IT.PNN	-2.084439
EMSC-20151101_0000023	MN.PDG	1.961717	IT-2007-0208	IT.FMG	2.013387
EMSC-20151118_0000047	KO.CAME	-2.911697	IT-2009-0186	IV.SACR	-2.090624
EMSC-20151118_0000094	KO.CAME	-2.740185	IT-2009-0189	IV.SACR	-2.126487
EMSC-20160530_0000085	IV.SACS	2.118076	IT-2010-0003	IV.NRCA	2.243845
EMSC-20160824_0000006	IT.GRD	-1.950702	IT-2011-0029	IT.GAI	-2.075673
EMSC-20160824_0000295	IV.FIAM	1.86564	IT-2012-0002	FR.REVF	2.025614
EMSC-20160824_0000295	IV.RM33	2.219246	IT-2013-0003	IT.PSC	1.974364
EMSC-20160903_0000009	IT.ACC	-1.866856	IT-2013-0003	IV.POFI	2.157508
EMSC-20160903_0000009	IT.CIT	-1.983131	IT-2013-0005	IT.MDN	2.272414
EMSC-20160903_0000009	IT.FOC	2.200381	IT-2013-0010	IV.SNTG	-2.148043
EMSC-20160903_0000009	IV.NRCA	1.919352	IT-2016-0001	IV.INTR	-3.470851
EMSC-20160903_0000009	IV.T1201	-1.887857	ME-1979-0003	EU.BAR	-2.051513

EMSC-20160903_0000009	IV.T1218	-2.022831	TK-1999-0077	TK.8101	<u>1.865905</u>
EMSC-20161027_0000072	IV.SNTG	2.005795	TK-2003-0180	TK.1701	<u>-1.970150</u>
EMSC-20161129_0000079	IT.AQF	-1.928834	TK-2006-0098	TK.1001	<u>-2.058961</u>

**Table 2.** Event- and station-corrected residuals greater than 3 times  $\phi_0$  in absolute value ( $|\varepsilon| > 3\phi_0$ ) at T=8s

Event ID	Station	g(8.0s)	Event ID	Station	g(8.0s)
EMSC-20110707_0000048	RA.IRPV	1.996037	IT-2010-0003	IV.NRCA	1.48384
EMSC-20150205_0000061	HL.ZKR	1.78805	IT-2010-0006	IV.NRCA	1.627814
EMSC-20151106_0000016	FR.REVF	-2.236079	IT-2012-0010	TV.MIRO1	-1.484999
EMSC-20151106_0000016	FR.SURF	1.709347	IT-2012-0025	IT.FIN0	1.660915
EMSC-20151118_0000094	KO.CAME	-1.846812	IT-2012-0025	IV.T0800	1.737367
EMSC-20161030_0000033	IT.ACC	1.99265	IT-2012-0025	IV.T0811	2.127796
EMSC-20161129_0000079	IT.AQV	-1.492742	IT-2012-0035	IV.CPGN	-1.493063
IT-2009-0009	IT.ANT	-1.557013	IT-2012-0063	IT.SBT	1.630845
IT-2010-0001	IV.NRCA	1.59095	IT-2013-0013	IT.NAS	1.569184
IT-2010-0001	IV.SENI	-1.600929	TK-1999-0077	TK.8101	1.501051

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**Table 3.** Between-event greater than 3 times  $\tau$ , in absolute value ( $|\delta B_e| > 3\tau$ ), at T=0.01 s

Event ID	$\delta B_e(0.01s)$	<u>N records</u>
EMSC-20121203_0000066	1.73	<u>18</u>
EMSC-20150411_0000019	-1.69	<u>10</u>
EMSC-20161102_0000149	-1.66	<u>72</u>
IT-2009-0316	1.97	<u>6</u>

**Table 4.** Between-event greater than 3 times  $\tau$ , in absolute value ( $|\delta B_e| > 3\tau$ ), at T=8.0 s

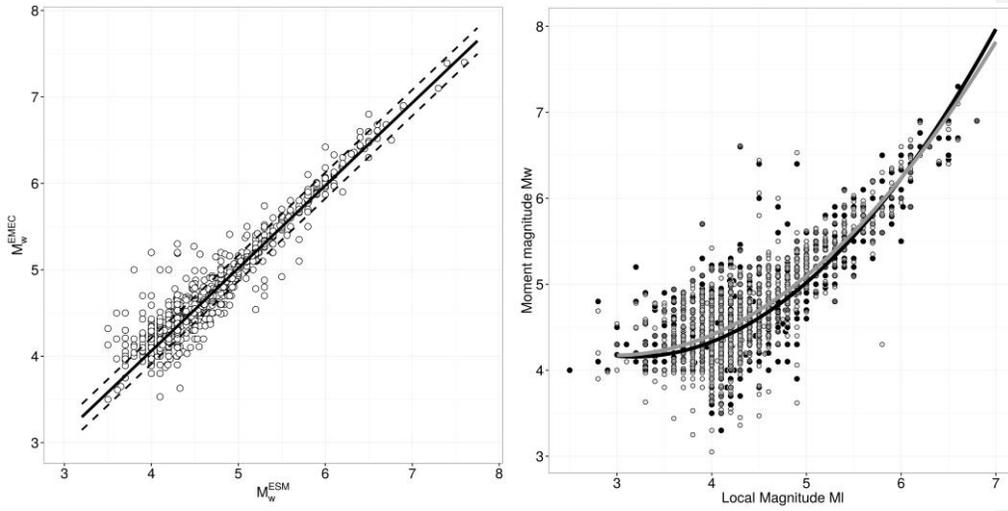
Event ID	$\delta B_e(8.0s)$	<u>N records</u>
EMSC-20121203_0000066	1.65	<u>11</u>
EMSC-20141124_0000030	1.84	<u>5</u>
IT-2010-0003	1.95	<u>3</u>

400 **Table 5.** Between-station greater than 3 times  $\phi_{s2s}$ , in absolute value ( $|\delta S2S| > 3\phi_{s2s}$ ), at T=0.01 s

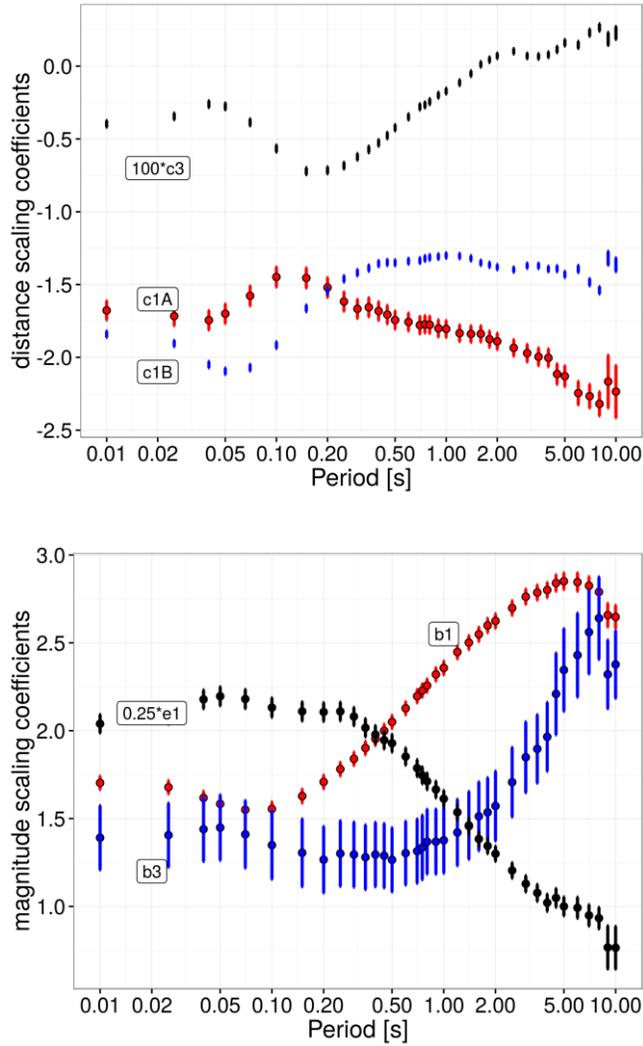
Station	$\delta S2S(0.01s)$	<u>N records</u>
IV.T0203	-2.35	<u>1</u>
HL.KSL	-2.04	<u>42</u>
CR.VELS	2.01	<u>1</u>
RO.JURR	2.04	<u>2</u>
FR.REVF	2.05	<u>5</u>

**Table 6.** Between-station greater than 3 times  $\phi_{s2s}$ , in absolute value ( $|\delta S2S| > 3\phi_{s2s}$ ), at T=8.0 s

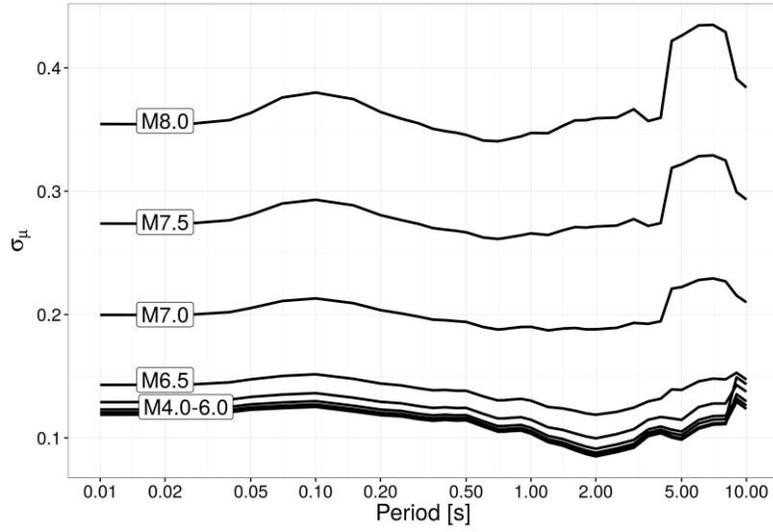
Station	$\delta S2S(8.0s)$	<u>N records</u>
IV.T0501	-1.86	<u>3</u>
FR.REVF	2.29	<u>4</u>
IV.SNTG	-1.74	<u>15</u>
KO.ENZZ	1.70	<u>7</u>



410 **Figure 1.** Left: regression between  $Mw^{ESM}$  and  $Mw^{EMEC}$ ; right: quadratic regressions between  $Mw^{ESM}$  and  $M_L$  (black line and points) and between  $Mw^{EMEC}$  and  $M_L$  (gray line and points) (see equations 1 and 2).

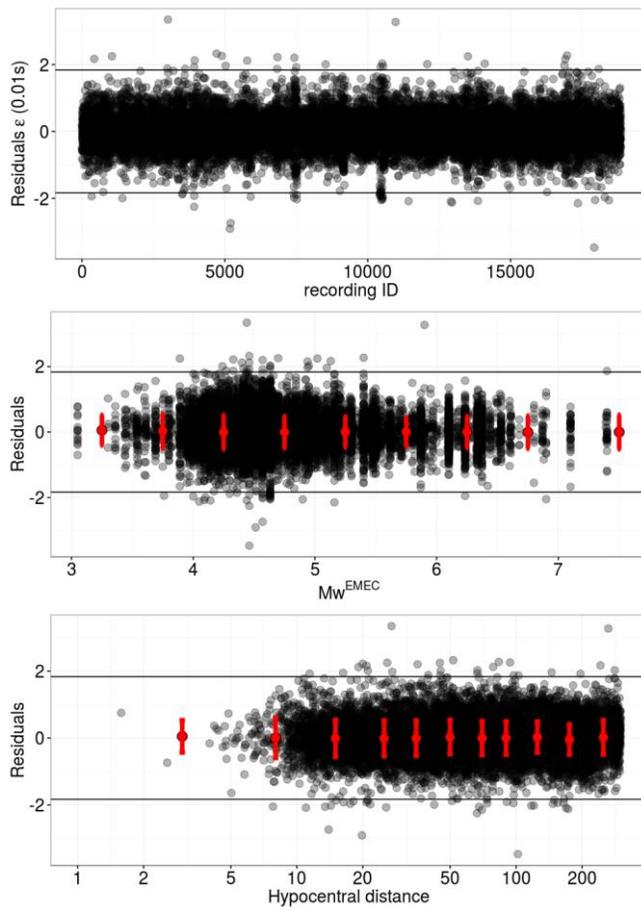


**Figure 2.** Coefficients (mean and standard error) of the model in equation (4) and (5) for Spectral Acceleration (SA).



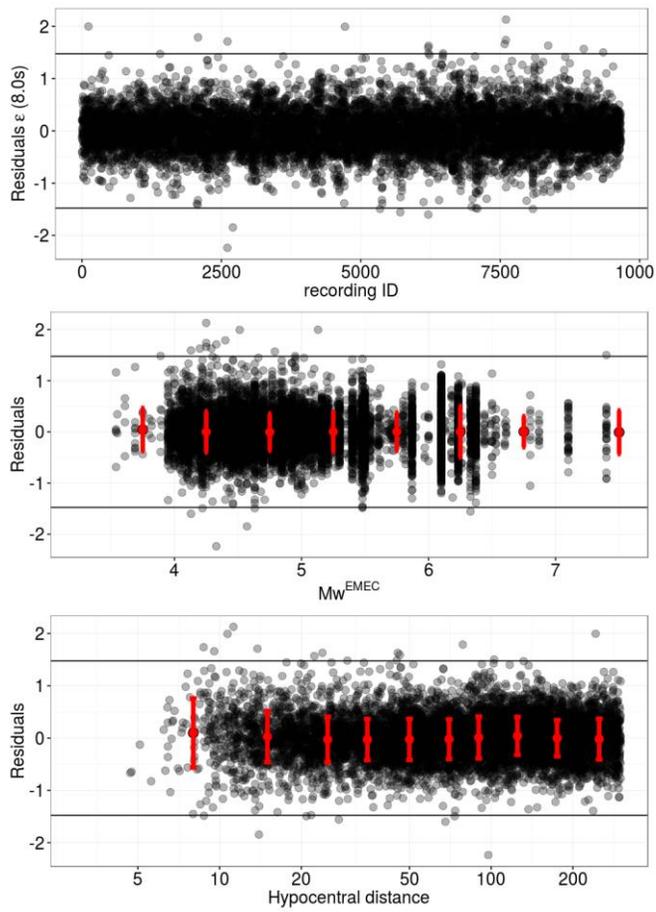
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**Figure 3.** Period dependence of the epistemic uncertainty on asymptotic standard error on the median ( $\sigma_{\mu}$ ) for SA considering different magnitudes and a fixed distance of 30 km.



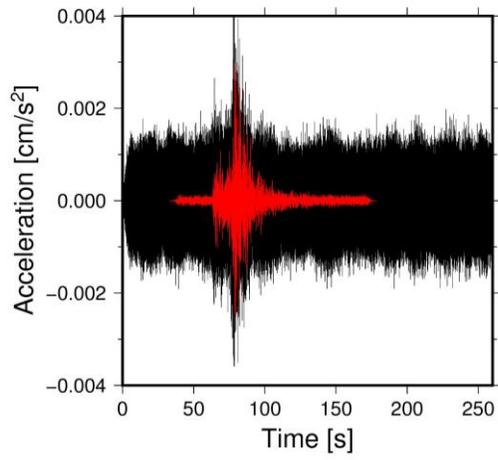
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**Figure 4a.** Residuals  $\varepsilon$  computed for SA at  $T=0.01s$ . Horizontal lines correspond to  $\pm 3\phi_0$ . The average residuals  $\pm 1$  standard deviation computed over magnitude and distance bins are shown as red bars.

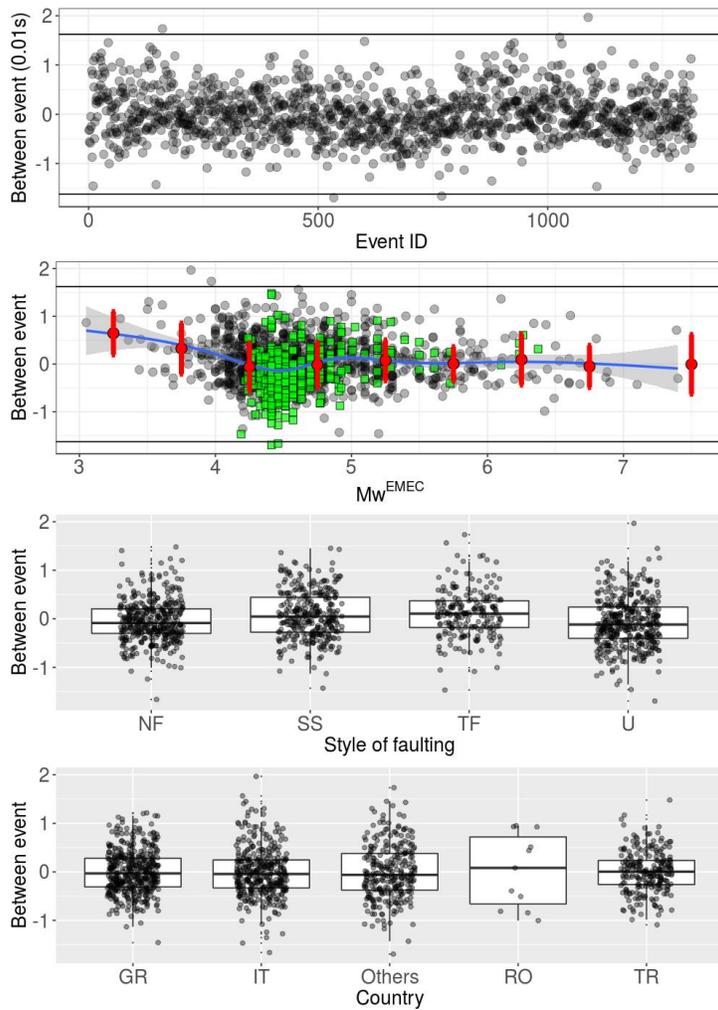


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**Figure 4b.** The same as Figure 4a but for  $T=8s$ .



440 **Figure 5** Record (EW component) of event IT-2016-0001 at station IV.INTR: comparison between raw data (black) and processed acceleration (red). The processed acceleration has been filtered over the bandwidth 0.15-30Hz using a second order a-causal Butterworth filter.



445 Figure 6a. Between-event residuals  $\delta B_e$  at  $T=0.01$  s for SA. Green squares indicate earthquakes with  $M_w$  estimated from ML using equation (2). Horizontal lines correspond to  $\pm 3\tau$ . The following abbreviations are used for the style of faulting: NF- normal faulting; SS – strike slip; TF – thrust faulting; U – unknown style of faulting. The country codes are: GR- Greece; IT – Italy; RO – Romania; TR- Turkey; Others – any other country different from the previous ones.

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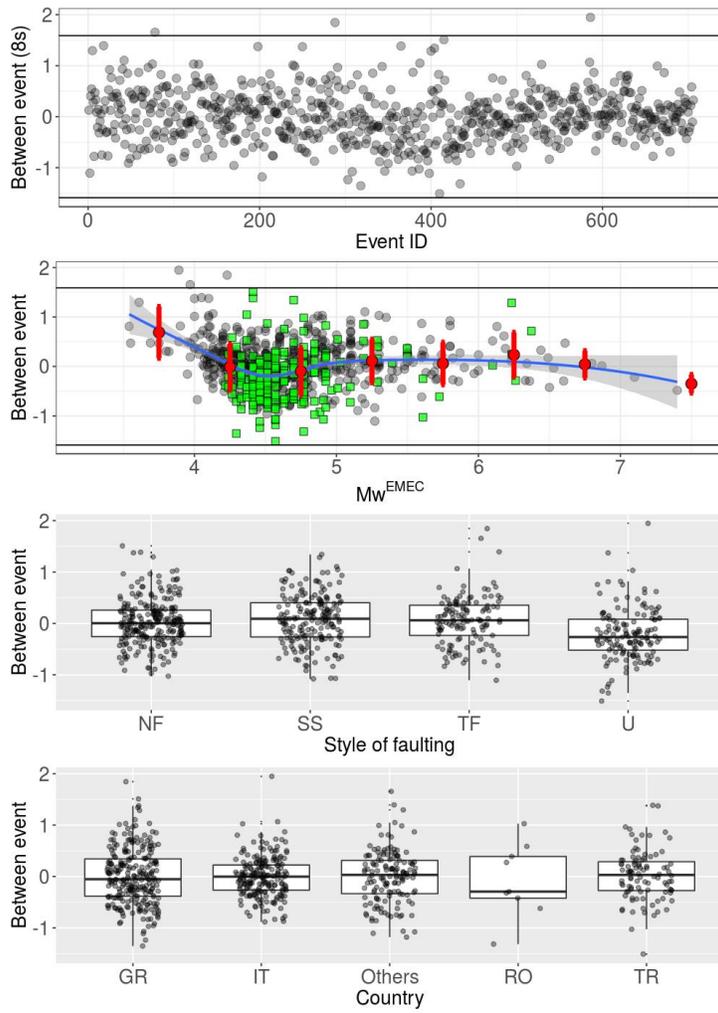


Figure 6b. The same as Figure 6a but for  $T=8$  s

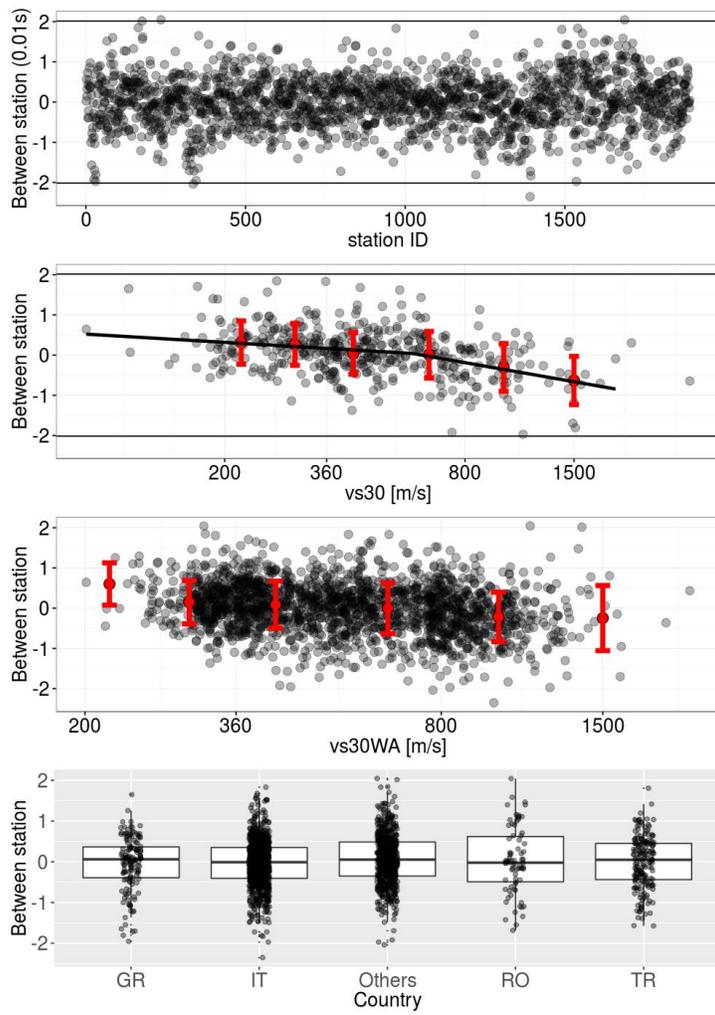
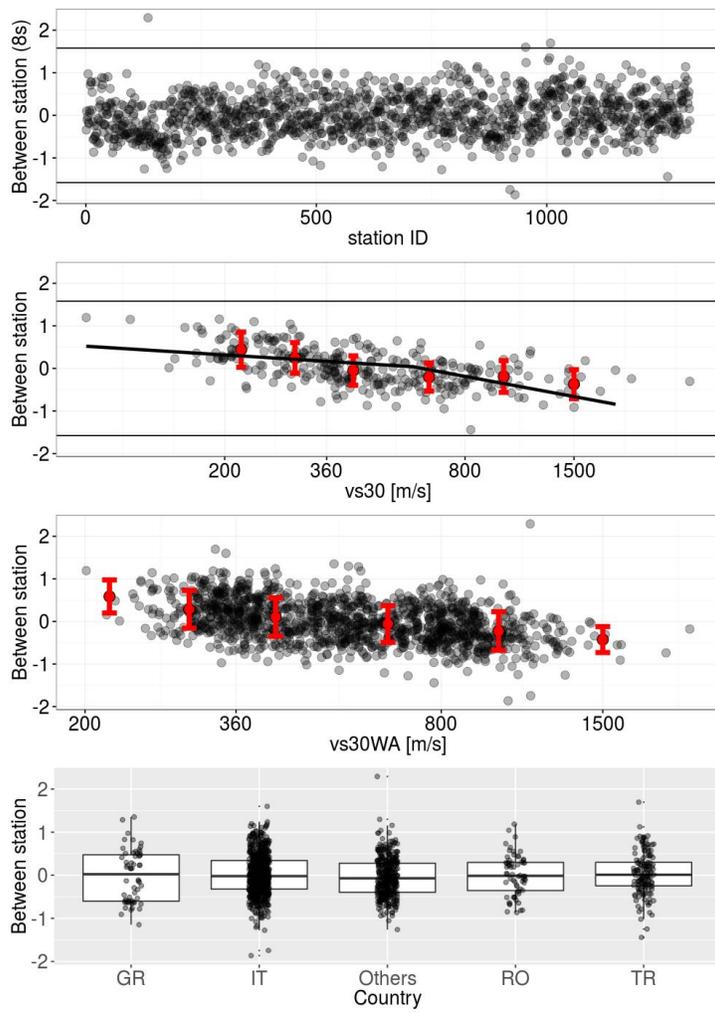
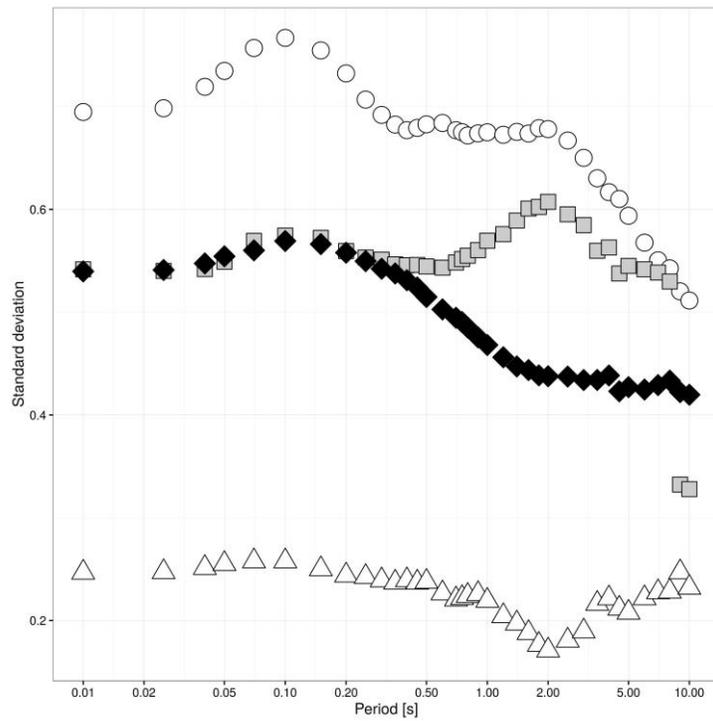


Figure 7a. Between-station residuals  $\delta S_{2S_2}$  at  $T=0.01$  s for SA. Horizontal lines correspond to  $\pm 3\phi_{S_{2S_2}}$ .



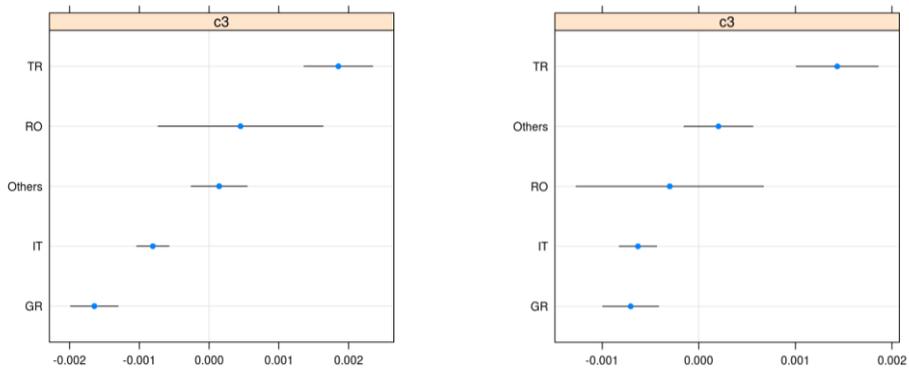
**Figure 7b.** The same as Figure 7a but for  $T=8\text{ s}$ .



470 **Figure 8.** Standard deviation of different residual distributions for SA ( $\delta C2C_e$  applied to the offset  
 475 coefficient  $e_1$ ). White circle: standard deviation of the between-station distribution (i.e.,  $\phi_{s2s}$ ); gray  
 square: standard deviation of the between-event distribution (i.e.,  $\tau$ ); white triangle: standard deviation  
 480 of the between-country distribution ( $\phi_{C2C}$ ); black diamond: standard deviation of the residual  
 distribution ( $\phi_0$ ).

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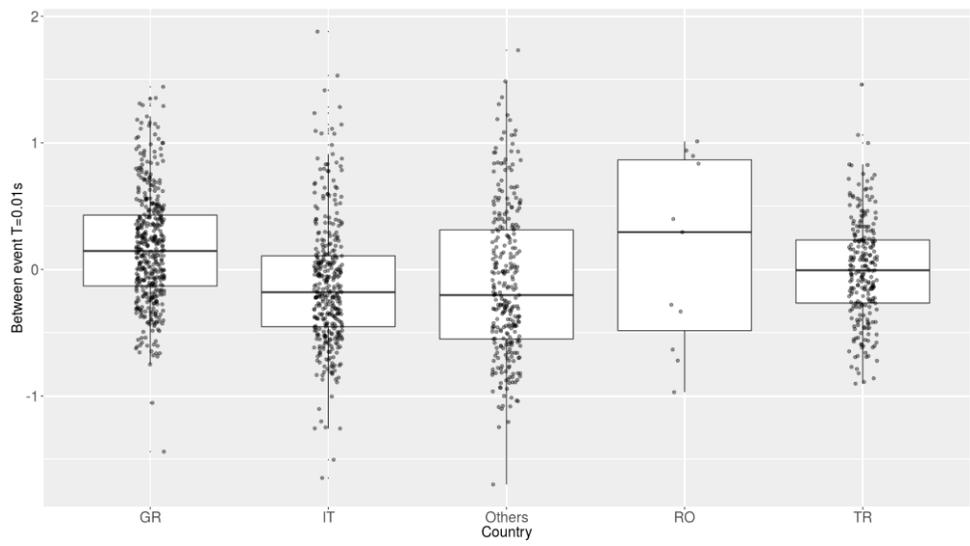
**Figure 9.** Random-effect on  $c_3$  (aneleptic parameter) depending on the country grouping level for  $T=0.01s$  (left) and  $T=2s$  (right), considering SA. Please, note the different order for the country classes and the different horizontal scale.

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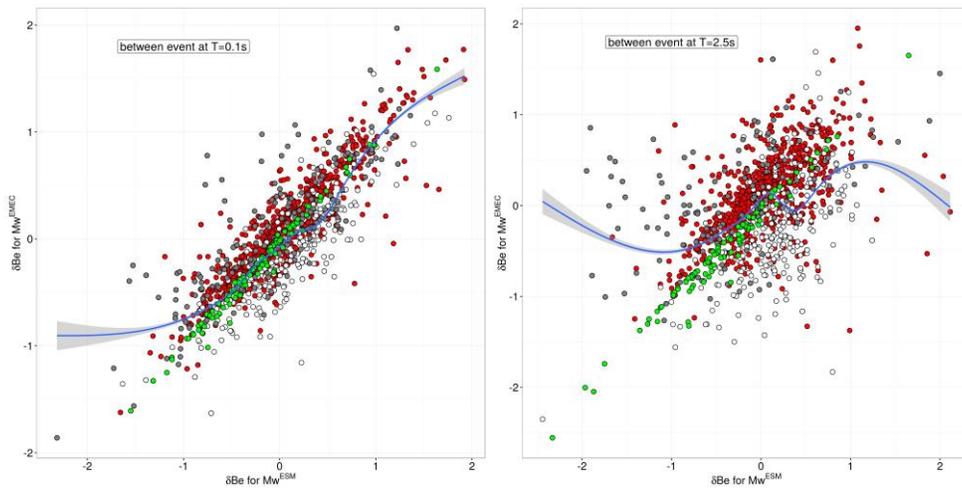
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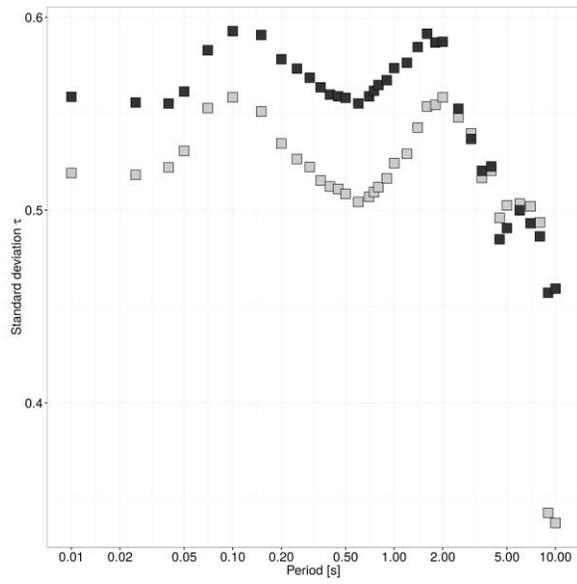


**Figure 10.** Between-events  $\delta B_e$  for SA at  $T=0.01s$  applying the country-based random-effect  $\delta C_{2C_\xi}$  to the anelastic attenuation parameter  $c_s$ .

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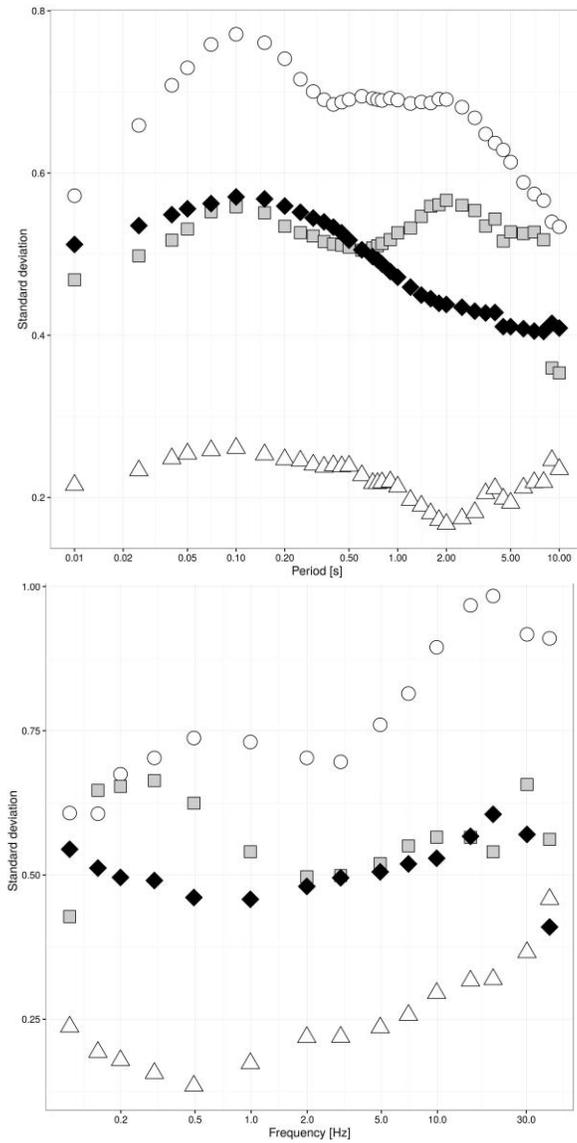


540 **Figure 11.** Comparison between  $\delta Be$  obtained using Mw from EMEC and ESM, considering T=0.1s (left)  
 and 2.5s (right). The colors represent different selections on magnitude. In particular, only red  
 represents recordings with both  $Mw^{EMEC}$  and  $Mw^{ESM}$  available, whereas the other colors ~~may indicate~~  
 conversions from  $M_L$  to Mw. In particular, the green dots represents earthquakes for which both  $Mw^{ESM}$   
 and  $Mw^{EMEC}$  are derived from the local magnitude  $M_L$  (and therefore showing high correlation). The blue  
 545 curves are the results of ~~the trend analysis (LOESS) with high order a local~~ polynomial regression  
 (LOESS), giving an idea about the degree of correlation between the two  $\delta Be$  estimates.

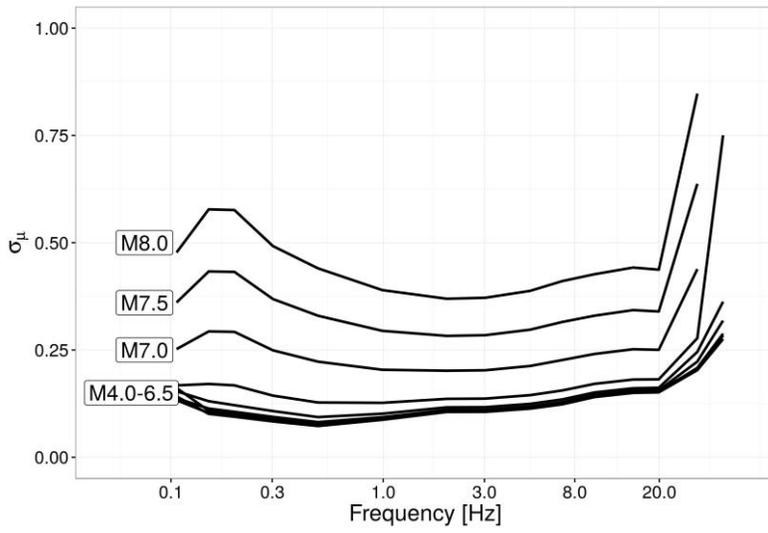


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**Figure 12.** Between-event standard deviation  $\tau$  for the regression using  $Mw^{EMEC}$  (gray) and  $Mw^{ESM}$  (black)



555 **Figure 13.** Standard deviation of residual distributions for spectral displacement-SD (top), ~~as function of~~ according to period, and for Fourier Amplitude Spectra-FAS (bottom), ~~according to as function of~~ according to as function of frequency. White circle: standard deviation of the between-station distribution ( $\phi_{s2s}$ ); gray square: standard deviation of the between-event distribution ( $\tau$ ); white triangle: standard deviation of the between country distribution ( $\phi_{c2c}$ ); black diamond: standard deviation of the residual distribution ( $\phi_0$ ).



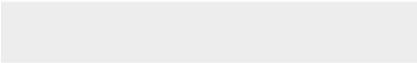
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**Figure 14.** Frequency dependence of the ~~epistemic uncertainty~~ asymptotic standard error on the median ( $\sigma_\mu$ ) for FAS considering different magnitudes and a distance of 30km .

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