Broadband dynamic rupture modeling with fractal fault roughness, frictional heterogeneity, viscoelasticity and topography: the 2016 Mw 6.2 Amatrice, Italy earthquake

Taufiqurrahman Taufiqurrahman¹, Alice-Agnes Gabriel¹, Thomas Ulrich², Lubica Valentova³, and Frantisek Gallovič⁴

¹Ludwig-Maximilians-Universität München ²Ludwig-Maximilians-Universität ³Charles University, Faculty of Mathematics and Physics, Dept. of Geophysics ⁴Charles University, Faculty of Mathematics and Physics, Department of Geophysics

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Abstract

Advances in physics-based earthquake simulations, utilizing high-performance computing, have been exploited to better understand the generation and characteristics of the high-frequency seismic wavefield. However, direct comparison to ground motion observations of a specific earthquake is challenging. We here propose a new approach to simulate data-fused broadband ground motion synthetics using 3D dynamic rupture modeling of the 2016 Mw6.2 Amatrice, Italy earthquake. We augment a smooth, best-fitting model from Bayesian dynamic rupture source inversion of strong-motion data (<1 Hz) with fractal fault roughness, frictional heterogeneities, viscoelastic attenuation, and topography. The required consistency at long periods allows us to quantify the role of dynamic source heterogeneities, such as the 3D roughness drag, from observational broadband seismic waveforms. We demonstrate that 3D data-constrained fully dynamic rupture synthetics show good agreement with various observed ground-motion metrics up to \sim 5 Hz and are an important avenue towards non-ergodic, physics-based seismic hazard assessment.

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4	T. Taufiqurrahman ¹ , AA. Gabriel ¹ , T. Ulrich ¹ , L. Valentová ² , F. Gallovič ²					
5	¹ Department of Earth and Environmental Sciences, Faculty of Geosciences, Ludwig Maximilian					
6	University, Munich, Germany					
7	² Department of Geophysics, Faculty of Mathematics and Physics, Charles University, Prague,					
8	Czech Republic					
9	Corresponding author:					
10	Taufiq Taufiqurrahman (<u>taufiqurrahman@outlook.de</u>)					
11	Alice-Agnes Gabriel (gabriel@geophysik.uni-muenchen.de)					
12	Key Points:					
13	• We propose a novel approach to design data-driven, 3D physics-based broadband					
14	dynamic rupture scenarios from Bayesian dynamic inversion					
15	• Our synthetics fit observations in terms of velocity and accelerations waveforms, as well					
16	as Fourier-amplitude-spectra up to $\sim 5 \text{ Hz}$					
17	• Analyzing the role of earthquake modeling ingredients highlights the importance of					
18	dynamic source heterogeneity for broadband ground-motion					

20 Abstract

Advances in physics-based earthquake simulations, utilizing high-performance 21 computing, have been exploited to better understand the generation and characteristics of the 22 high-frequency seismic wavefield. However, direct comparison to ground motion observations of 23 a specific earthquake is challenging. We here propose a new approach to simulate data-fused 24 broadband ground motion synthetics using 3D dynamic rupture modeling of the 2016 Mw6.2 25 26 Amatrice, Italy earthquake. We augment a smooth, best-fitting model from Bayesian dynamic rupture source inversion of strong-motion data (<1 Hz) with fractal fault roughness, frictional 27 heterogeneities, viscoelastic attenuation, and topography. The required consistency to match long 28 29 periods allows us to quantify the role of small-scale dynamic source heterogeneities, such as the 3D roughness drag, from observational broadband seismic waveforms. We demonstrate that 3D 30 data-constrained fully dynamic rupture synthetics show good agreement with various observed 31 32 ground-motion metrics up to ~5 Hz and are an important avenue towards non-ergodic, physicsbased seismic hazard assessment. 33

34 Plain Language Summary

Models of earthquakes are used to better understand the origin and features of strong seismic shaking using supercomputers. But the connection of such computer simulations with actual measurements is complex. This study suggests a new way to use observations, computer models, and physics to model details of the damaging ground shaking recorded during the 2016 Amatrice, Italy, earthquake. We start from an earlier, relatively low-resolution earthquake model that matches seismograms at low periods (<0.5-1Hz), derived using many Monte Carlo simulations. We carefully enhance this earthquake model by adding roughness to the slipping fault, smaller-scale variations of the frictional resistance of this fault to earthquake slip, mountains and basins scattering the seismic waves, and the effects of anelastic damping of wave amplitudes while they propagate. We show that we now also match the seismic waves at frequencies up to 5 Hz, while not losing the good match at long periods. This result is important to better understand how hazardous earthquakes in specific regions may be. Our modeling indicates which ingredients are required in computer simulations to generate realistic ground motions for physics-based seismic hazard assessment.

49 **1 Introduction**

Simulations of broadband (> 1 Hz) ground motions are of great importance to 50 51 seismologists and the earthquake engineering community. Despite the fact that we often lack detailed knowledge of the subsurface and earthquake source processes at small scales, it is 52 essential to understand the generation and characteristics of the high-frequency seismic 53 54 wavefield coinciding with most buildings' resonance frequencies. Broadband ground motions 55 have been successfully simulated using hybrid techniques (e.g., Graves & Pitarka, 2010; Mai et al., 2010) that combine low-frequency deterministic ground motion synthetics with stochastically 56 generated high-frequency components. However, hybrid synthetic waveforms lack deterministic 57 information at higher frequencies and pose challenges in the realistic parameterization of wave 58 59 propagation and earthquake rupture. Indeed, high-frequency radiation may arise, for instance, from acceleration and deceleration of the rupture front (Madariaga, 1977) caused by fault kinks, 60 segmentation, or roughness (e.g., Shi & Day, 2013; Bydlon & Dunham, 2015), frictional or 61 62 stress heterogeneities (e.g., Ripperger et al., 2008; Valentová et al., 2021) or from off-fault damage (e.g., Okubo et al., 2019; Yamashita, 2000). Additionally, the radiated wavefield is 63

scattered by complex topography and structural heterogeneities (e.g., Imperatori & Mai, 2013;
Takemura et al., 2015; Hartzell et al., 2016; Langer et al., 2019; Pitarka et al., 2021).

Recent advances in high-performance computing and dynamic earthquake rupture 66 modeling allow deterministic 3D regional-scale broadband simulations to resolve frequencies up 67 to 10 Hz (e.g., Heinecke et al., 2014; Rodgers et al., 2020; Savran & Olsen, 2020; Pitarka et al., 68 2021). Such simulations often assume a kinematic, thus predefined, finite earthquake source 69 representation. In distinction, dynamic rupture models offer physically self-consistent 70 descriptions of the earthquake rupture process. Generic dynamic rupture simulations across 71 rough faults (both in 2D or 3D, e.g., Dunham et al., 2011; Shi & Day, 2013; Bydlon & Dunham, 72 2015; Withers et al., 2018; Bruhat et al., 2020) are characterized by highly complex rupture 73 processes translating into ground motion synthetics that can match empirical ground-motion 74 prediction equations (GMPEs). 75

In this study, we investigate 3D effects of fault roughness and regional topography on 76 earthquake source dynamics and broadband ground motions by direct validation against 77 observations. The August 24th, 2016, Mw 6.2 Amatrice earthquake (Chiaraluce et al., 2017; 78 Michele et al., 2020) is the first in the Amatrice-Visso-Norcia earthquake sequence in the 79 80 Central-Northern Apennine system of NW-SE aligned normal faults. It was the sequence's most 81 destructive event, causing extensive damage to surrounding buildings and infrastructure 82 (Michele et al., 2016). The earthquake was recorded by a remarkably dense network of strong 83 ground motion instruments, including 20 near-source stations within a radius of 50 km from the earthquake epicenter (Figure 1, Table S1). The two closest stations, in Amatrice (AMT) and 84 Norcia (NRC), are located only a few kilometers away from the fault. 85

86 The source process of the Amatrice event has been imaged using seismic data (Pizzi et al., 2017; Tinti et al., 2016), geodetic data (e.g., Cheloni et al., 2017; Walters et al., 2018), or 87 both (Cirella et al., 2018; Kheirdast et al., 2021), suggesting pronounced source heterogeneities. 88 89 However, kinematic finite-fault inversions are challenged by inherent non-uniqueness (Gallovič & Ampuero, 2015; Mai et al., 2016; Ragon et al., 2018; Shimizu et al., 2020; Tinti et al., 2021). 90 Dynamic source inversions recovering friction parameters and the initial state of fault stress offer 91 a data-driven source description compatible with earthquake physics (e.g., Peyrat & Olsen, 2004) 92 but require a sufficiently simple dynamic rupture model to reduce the computational cost of the 93 forward problem. A Bayesian dynamic inversion using the Parallel Tempering Monte Carlo 94 algorithm (Sambridge, 2013; Gallovič et al., 2019a) was applied to the Amatrice earthquake, 95 utilizing band-pass filtered (between 0.05 and 0.5-1 Hz) strong ground motion data by Gallovič 96 97 et al. (2019b). Assuming a 1D medium with planar topography, the best-fitting model was used to predict ground motions up to higher frequencies than considered in the inversion (up to ~ 5 98 Hz). Yet this approach poorly matched the high-frequency content, presumably most sensitive to 99 100 unresolvable small-scale features of the rupture process.

101 We here propose a new approach to simulate data-fused broadband ground motion synthetics using 3D dynamic rupture modeling. Our starting point is the best-fitting model from 102 the Bayesian dynamic source inversion of the Amatrice earthquake (Figure 1b, hereafter named 103 'reference model'). We self-consistently augment this smooth reference model by adding fault 104 roughness, small-scale frictional heterogeneities, viscoelastic attenuation, and topography 105 yielding realistic high-frequency radiation without disrupting the large-scale characteristics of 106 the reference model. The synthetic near-field ground motions show good agreement with various 107 108 observed ground-motion metrics up to frequencies of ~5 Hz.

109 2 Ingredients for broadband dynamic rupture modeling

We build our model upon Bayesian dynamic rupture inversion of the 2016 Amatrice 110 earthquake following the approach of Gallovič et al. (2019b) with the improved forward solver 111 FD3D TSN (Premus et al., 2020), which was verified in a suite of dynamic rupture benchmarks 112 (Harris et al., 2018). The inversion is performed for a 30 km long and 14 km wide planar fault 113 governed by a slip-weakening friction law (Ida, 1972; Palmer et al., 1973). The dynamic rupture 114 slip rate functions along the fault are convolved with pre-calculated Green's functions 115 representing impulse responses of the medium. In this step, the fault is dipping at 45°, embedded 116 in the 1D velocity structure of Ameri et al. (2012) with a flat free surface. The dynamic models 117 are characterized by three spatially heterogeneous parameters: (i) the initial shear stress along dip 118 τ_{i} , (ii) the friction drop, μ_{s} - μ_{d} , with μ_{s} and μ_{d} the static and dynamic friction coefficient 119 respectively, and (iii) the slip-weakening distance D_c . Yielding occurs when the shear stress τ 120 121 reaches the fault strength $\tau_s = \mu_s \sigma_n$, where σ_n is assumed as linearly depth-dependent normal stress with a gradient of 8.52 MPa/km and a minimum value of 0.1 MPa. The initial along-strike shear 122 stress τ_{strike0} is assumed to be zero. The dynamic friction coefficient μ_d is fixed to 0.4, and 123 frictional cohesion of 0.5 MPa is assumed everywhere on the fault. The best-fitting model from 124 the Bayesian inversion represents the reference model of this study. We then perform high-125 resolution enhanced 3D dynamic rupture simulations using the open-source software package 126 SeisSol (https://github.com/SeisSol/SeisSol), resolving seismic wavefield up to 5 Hz (locally up 127 128 to 10 Hz) within 50 km distance of the fault using an unstructured, statically adaptive mesh 129 consisting of 80 million tetrahedral elements (Figure 1a, Text S1).

130 2.1 Fault roughness

131 The reference model's dynamic parameters are first bilinearly interpolated from their 1.75 km along-dip and 2.3 km along-strike reference resolution into a denser 25 m sampled grid 132 (see Figure 2, column a). Next, we adapt the fault morphology to adhere to a band-limited self-133 similar (Hurst exponent H=1) fractal surface. The amplitude-to-wavelength ratio α of natural 134 faults ranges between 10⁻⁴ and 10⁻² (Power & Tullis, 1991), and we here use $\alpha = 10^{-2}$ allowing 135 comparability to earlier studies (Shi and Day, 2013; Fang & Dunham, 2013; Withers et al., 2018; 136 Bruhat et al., 2020). The fractal surface wavelengths are band-limited between λ_{\min} and λ_{\max} . 137 Choosing λ_{\min} =200 m balances resolution requirements and computational cost for our setup and 138 aligns with previous 3D fault roughness studies (Shi and Day, 2013; Fang & Dunham, 2013). 139 Our choice of $\lambda_{max}=2$ km is motivated by the ~2 km spatial resolution of the dynamic parameters 140 in the reference model. 141

142 2.2 3D roughness drag and heterogeneous initial stresses

143 Shear and normal stresses are dynamically perturbed by fault roughness during rupture 144 propagation. The resulting 'roughness drag' (Dunham et al., 2011), an additional shear resistance 145 to slip, was derived for a 1D rough fault in a 2D quasi-static boundary perturbation analysis by 146 Fang and Dunham (2013) as

147
$$\tau_{\rm drag}^{2\rm D} = 8 \ \pi^3 \alpha^2 \, \mathrm{G}^* \Delta / \lambda_{\rm min} \qquad (1)$$

with Δ the fault slip, λ_{\min} the minimum roughness wavelength, and $G^* = G/(1 - v)$ where G and v are shear modulus and Poisson's ratio, respectively.

To preserve the overall characteristics of the reference scenario while incorporating fault roughness, we compensate the roughness drag in the initial loading by increasing the reference initial shear tractions as $\tau_{dip} = \tau_{dip0} + \tau_{drag}^{3D}$ and $\tau_{strike} = \tau_{strike0} + \tau_{drag}^{3D}$. We thus attempt to empirically approximate the roughness drag (following Dunham et al., 2011, but for the first time in 3D) as

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$$\tau_{\rm drag}^{3D} = C \, \tau_{\rm drag}^{2D} ,$$
 (2)

where *C* is a dimensionless coefficient. In Text S2, we demonstrate that *C* depends on the minimum roughness wavelength λ_{min} and the number of elements *n* resolving λ_{min} , calculating τ_{drag}^{2D} using characteristics of the reference model slip distribution. For our choice of n = 4 elements to resolve $\lambda_{min} = 200$ m, we obtain *C* of ~0.44. The average value of τ_{drag}^{3D} across the rupture area is ~1.4 MPa.

We account for the 3D roughness drag while preserving the smooth reference initial stress distribution by loading the rough fault with a heterogeneous regional stress tensor: we first adapt the smooth reference initial fault loading to balance roughness drag, then expose the now rough fault to the adapted loading (Text S3). As a result, the broadband model features roughness-induced small-scale fluctuations of the initial shear and normal tractions (Figure 2b), consisting of both releasing and restraining slopes that bring the fault closer and farther from failure, respectively (Figure S2).

168 2.3 Frictional heterogeneity

We perturb the smooth variation of the reference characteristic frictional slip weakening distance D_c^{0} , the spatially most variable dynamic parameter in the reference Bayesian dynamic inversion. The relative standard deviation of D_c is on the order of 50% (Gallovič et al., 2019b), highlighting its importance as a proxy for unaccounted geometrical and geological features. We use a band-limited fractal distribution. We prescribe $D_c=\max(0.14 \text{ m}, D_c^{0}(1+\epsilon))$, where 0.14 m is the minimum value of D_c^{0} , and ε follows a fractal distribution of amplitude-to-wavelength ratio $\alpha = 10^{-4}$ generated from a different random seed than the one used for the fault roughness.

176 2.4 Topography and viscoelasticity

In our broadband dynamic rupture, the flat free surface used in the inversion is superseded by high-resolution topography data sampled to 150 m resolution (Farr et al., 2007). The modeled 3D domain spans 300×300 km horizontally and extends to a depth of 150 km to avoid any undesired reflections from the (imperfectly) absorbing boundaries. We incorporate the 1D velocity model, with $V_p = 1.86V_s$, and viscoelastic attenuation, with $Q_p = 2Q_s$, inferred by Ameri et al. (2012), see Table S2.

183 **3 Broadband rupture dynamics and ground-motion validation**

184 3.1 Rupture dynamics

We compare the broadband dynamic rupture model, incorporating fault roughness, smallscale D_c variation, and topography to the reference model with a planar fault, a flat free surface, and smoothly varying initial conditions, in terms of fault slip (Figures 1b, c), slip rate space-time evolution (Figures 2c, d), and moment rate release (Figure S3).

Figures 1b,c and 2c,d demonstrate similar large-scale slip evolution. The seismic moment of the broadband model is 2.8 x 10^{18} Nm, corresponding to M_w=6.24, which is comparable to the reference model with 2.6 x 10^{18} Nm seismic moment (M_w=6.20). We highlight that both models recover the remarkably weak and slow nucleation phase (Tinti et al., 2016; Gallovič et al., 2019b), as was also inferred for the Norcia earthquake (Tinti et al., 2021). The nucleation is followed by bilateral rupture, which is slower towards the NW than towards the SE in both models. At smaller scales, the broadband model features decoherence of rupture fronts (Shi & Day, 2013). Locally fluctuating rupture speeds are due to acceleration and deceleration at releasing and restraining slopes, heterogeneous initial shear and normal traction, and D_c heterogeneity. Peak slip rates are increased by ~15% in the broadband model, while both models feature pulse-like ruptures, and rise time remains largely unaffected.

Comparisons of moment rate releases (Figure S3a), moment rate spectra (Figure S3b), and the second time-derivative of moment rate releases (Figure S3c) illustrate the effects of the fault roughness, heterogeneous loading, and D_c on the high-frequency rupture radiation. While the two distinct episodes of moment rate release are well recovered, its first peak is about 20% higher in the broadband model than the reference model (Figure S3a), reflecting the required increase in negative strength excess in the nucleation region.

206 3.2 Ground motions

Figure 3 compares the observed three-component velocity and acceleration waveforms recorded at the 20 strong-motion stations (Figure 1, Luzi et al., 2016) with synthetics from the broadband dynamic rupture model. The overall agreement in terms of waveform shape, duration, and amplitudes is good. Analogous plots for the reference model and the rough fault model without topography are shown in Figures S4 and S5, respectively.

To better appreciate the role of the fault roughness and topography, Figure 4 compares EW velocity and acceleration waveforms and Fourier Amplitude Spectra (FAS) of three models the reference, the broadband rough fault model with topography, and the broadband rough fault model without topography - with observations at selected stations. All other components and stations are shown in Figures S6-S12. The synthetic waveforms of the broadband models match long-period data (0.05-0.5 Hz) equally well as the reference model. Contrarily, the reference

model provides waveforms clearly depleted at high frequencies. A general trend is that both, 218 fault roughness and topography, enhance waveforms at high frequencies, although not to the 219 same extent at all stations. The increase in high-frequency content in the broadband waveforms 220 221 without topography (green) is clearly limited in duration compared to the same model incorporating topography (red). High-frequency ground motions are amplified early-on by fault 222 roughness, while topography-induced scattered waves prolong their duration. The combination 223 of both effects is most pronounced in the central and SE part of the hanging wall region (see, 224 e.g., stations PZI1, LSS, and SPD in Figure 4 and S7, or MSC, AMT, and ANT in Figure S6, S9, 225 S10, and S12). At some stations (e.g., stations FOS and ASP), our broadband synthetic spectra 226 are improved but yet underestimating the observed spectra at frequencies higher than 1 Hz. 227

Animations of the three components of the velocity wavefield for the reference and broadband models are shown in Movies S2-S7. They illustrate how seismic waves are both reflected and scattered upon propagating across sharp topographical features like mountains and hills, which explains the prolonged duration of the seismic signal for several receivers (e.g., stations LSS & SPD, Figure 4). Viscoelastic attenuation is important to capture the decay of seismic reverberations caused by topography scattering (Figure S13).

Figures S14 and S15 quantify the fit of the synthetic ground-motions of the broadband and reference models with observations using Goodness-of-fit (GOF) metrics (Olsen & Mayhew, 2010), including peak ground velocity and displacement, spectral acceleration, Fourier amplitude spectra, energy duration, and cumulative energy (Text S4). The broadband model with topography fits the observations better (GOF 45-65) than the reference model (GOF 35-55) and the broadband model without topography (GOF 40-60). 240 Figure S16 details the model bias and standard deviation over the 0.5-10 s period range. averaged over 20 stations used in this study. A near-zero model bias over a specific period 241 suggests that our simulated ground motions match observations reasonably well. The reference 242 243 model fits the observations only at periods longer than 2 s. Compared to the reference model, the fit of the broadband model without topography (Figure S16b) is improved (30-40% lower bias) 244 at periods shorter than 2 s. The broadband model with topography shows an even better fit (40-245 50% lower bias than the reference model, Figure S16c), while both models preserve a perfect fit 246 at periods longer than 2 s. 247

248 4 Discussion

249 Recorded broadband ground motions are widely used in earthquake engineering to 250 inform the performance-based design of structures. Typically, generic strong-motion waveforms that fit specific ground motion metrics are selected from a strong-motion database for that 251 252 purpose (Iervolino et al., 2010). Also, probabilistic seismic hazard analyses often rely on such 253 so-called ergodic ground-motion models (GMMs, e.g., Petersen et al., 2019). Yet, these may not 254 reflect the conditions of a specific region of interest. Regional synthetic broadband ground motions from 3D dynamic rupture inversions, which offer a physically consistent representation 255 of earthquakes, can sample conditions that are not sufficiently constrained in empirical models 256 257 towards the development of non-ergodic, physics-based GMMs (Graves et al., 2011; Frankel et al., 2018; Moschetti et al., 2017; Wirth et al., 2018; Withers et al., 2020). Our proposed 258 broadband dynamic rupture models can be extended to account for other distinctive regional 259 260 characteristics, such as a listric or segmented fault geometry, 3D velocity models including low velocity layers and basins, and fault zone plasticity (Roten et al., 2014). They may also inform 261 PSHA-targeted kinematic rupture generators while inherently ensuring realistic scaling of 262

earthquake characteristics (e.g., Savran & Olsen, 2020). Our models emphasize the need to
include i) small-scale source characteristics to enhance the high-frequency source radiation
during the rupture propagation, and ii) topography to increase the duration due to scattering. The
duration of the latter effect is controlled by viscoelastic attenuation.

We carefully analyze the effects of adding roughness to a flat fault model. We 267 counterbalance the consequent 3D roughness drag by increasing initial shear traction by τ_{drag}^{3D} 268 (equation 2), calculated using the spatially variable slip amplitude Δ of the reference model. We 269 explored an alternative model (not presented here), with constant Δ equal to the peak slip of the 270 reference model (1.14 m). It generates a higher average τ_{drag}^{3D} of about 3.3 MPa (cf. 1.4 MPa, 271 Section 2.2). It may be possible to identify alternative satisfying models based on constant Δ . 272 Nevertheless, the here presented approach of constraining Δ by the spatially variable reference 273 fault slip appears superior due to its simpler and better constrained parametrization. 274

Although our rough fault model with topography improves the waveform fit at high 275 frequencies, some synthetics still underestimate the observations. More complete matching of the 276 observed records may in future be enabled by: (i) considering smaller length scales (λ_{min}) of fault 277 surface roughness, potentially further increasing high-frequency radiation at the cost of increased 278 computational demands; (ii) incorporating larger-scale non-planar fault geometry such as a listric 279 fault geometry which has been, for instance, suggested from satellite data (Tung and Masterlark, 280 2018) and which may modulate peak ground velocities as a consequence of curvature focusing 281 effect (Passone and Mai, 2017); (iii) probing and quantifying the variability of the predicted 282 shaking using alternative models from the Bayesian ensemble of the dynamic rupture inversion, 283 and (iv) incorporating a more realistic Earth model to capture path and local site conditions, i.e., 284 285 3D velocity models, small-scale scattering media (Imperatori & Mai 2013, Bydlon & Dunham,

286 2015), site corrections (Rodgers et al., 2020) or non-linear soil effects (Roten et al., 2012). In 287 particular, low-velocity sedimentary basins (Lee et al., 2009, Pischiutta et al., 2021) can 288 significantly amplify the amplitude and duration of ground motions, which may lead to improved 289 synthetics for stations with strong site-effects, e.g., CLF with site-class D (Table 1).

290 **5 Conclusions**

We present a novel approach for broadband dynamic rupture modeling constrained from 291 292 low-frequency data towards generating physics-based, non-ergodic ground motion synthetics validated by observations. We generate broadband dynamic rupture models of the 2016 Mw 6.2 293 Amatrice earthquake by combining large-scale heterogeneous stress and frictional parameters, 294 295 inferred from the best-fitting model of a Bayesian dynamic rupture inversion, with self-similar fault roughness and frictional (slip weakening distance) heterogeneity, topography, and 296 viscoelastic seismic attenuation. We empirically quantify the 3D roughness drag by 297 counterbalancing its effective dynamic stress perturbations. We obtain dynamic rupture scenarios 298 that successfully reproduce the low-frequency (<1 Hz) source characteristics of the inverted 299 dynamic model while simultaneously producing a realistic amount of high-frequency (up to ~ 5 300 Hz) radiation. The combined small-scale heterogeneities of the fault geometry, frictional 301 strength, loading, and topography's prolonging coda effect yield comparable with observed 302 303 strong motion records. Our work demonstrates 3D physics-based, broadband earthquake groundmotion simulations that are tightly constrained by data-driven dynamic earthquake source 304 inversion and allows us to quantify the first-order role of dynamic source heterogeneities in the 305 306 broadband seismic wavefield.

307 6 Data and Resources

We use the open-source software package SeisSol, available at https://github.com/SeisSol 308 /SeisSol, branch 'Norcia sequence', commit 181fc85d5c405a8c44fe21869fe736ab1f0206d5. 309 Input files required to run broadband dynamic rupture simulations can be downloaded from 310 https://zenodo.org/record/6386938. The reference dynamic rupture model parameters from the 311 Bayesian inversion are available at <u>https://github.com/fgallovic/fd3d tsn pt/tree/master/</u> 312 example/20160824-Amatrice. The topography data from the Shuttle Radar Topography Mission 313 (SRTM) is retrieved using the SRTM.py python package (https://github.com/tkrajina/srtm.py). 314 Observed strong ground motion waveforms recorded by the Rete Accelerometrica Nazionale 315 (RAN) and the Rete Sismometrica Nazionale, operated by the Italian Department of Civil 316 Protection (DPC) and the Istituto Nazionale di Geofisica e Vulcanologia (INGV) were 317 downloaded from the Engineering Strong-Motion database (https://esm.mi.ingv.it/, Luzi et al., 318 319 2016; Lanzano et al., 2021).

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329 (KAUST) Supercomputing Laboratory (<u>www.hpc.kaust.edu.sa</u>) for computing time on Shaheen
330 II (project k1343).

331 References

Ameri, G., Gallovič, F., & Pacor, F. (2012). Complexity of the Mw 6.3 2009 L'Aquila (central
 Italy) earthquake: 2. Broadband strong motion modeling. *Journal of Geophysical*

334 *Research: Solid Earth*, *117*(B4). https://doi.org/10.1029/2011JB008729

- Aochi, H., & Madariaga, R. (2003). The 1999 İzmit, Turkey, Earthquake: Nonplanar Fault
 Structure, Dynamic Rupture Process, and Strong Ground Motion. *Bulletin of the Seismological Society of America*, 93(3), 1249–1266.
 https://doi.org/10.1785/0120020167
- Bruhat, L., Klinger, Y., Vallage, A., & Dunham, E. M. (2020). Influence of fault roughness on
 surface displacement: From numerical simulations to coseismic slip distributions. *Geophysical Journal International*, 220(3), 1857–1877.
 https://doi.org/10.1093/gji/ggz545
- Bydlon, S. A., & Dunham, E. M. (2015). Rupture dynamics and ground motions from
 earthquakes in 2-D heterogeneous media. *Geophysical Research Letters*, 42(6), 1701–
 1709. https://doi.org/10.1002/2014GL062982
- 346 Cheloni, D., De Novellis, V., Albano, M., Antonioli, A., Anzidei, M., Atzori, S., Avallone, A.,
- 347 Bignami, C., Bonano, M., Calcaterra, S., Castaldo, R., Casu, F., Cecere, G., De Luca, C.,
- 348 Devoti, R., Di Bucci, D., Esposito, A., Galvani, A., Gambino, P., ... Doglioni, C. (2017).
- 349 Geodetic model of the 2016 Central Italy earthquake sequence inferred from InSAR and
- 350 GPS data. Geophysical Research Letters, 44(13), 6778–6787.
- 351 https://doi.org/10.1002/2017GL073580

352 Chiaraluce, L., Di Stefano, R., Tinti, E., Scognamiglio, L., Michele, M., Casarotti, E., Cattaneo,

- 353 M., De Gori, P., Chiarabba, C., Monachesi, G., Lombardi, A., Valoroso, L., Latorre, D.,
- 354 & Marzorati, S. (2017). The 2016 Central Italy Seismic Sequence: A First Look at the
- 355 Mainshocks, Aftershocks, and Source Models. Seismological Research Letters, 88(3),
- 356 757–771. https://doi.org/10.1785/0220160221
- Day, S. M., Dalguer, L. A., Lapusta, N., & Liu, Y. (2005). Comparison of finite difference and
 boundary integral solutions to three-dimensional spontaneous rupture. *Journal of Geophysical Research: Solid Earth*, *110*(B12). https://doi.org/10.1029/2005JB003813
- Dumbser, M., & Käser, M. (2006). An arbitrary high-order discontinuous Galerkin method for
 elastic waves on unstructured meshes—II. The three-dimensional isotropic case.
 Geophysical Journal International, 167(1), 319–336. https://doi.org/10.1111/j.1365 246X.2006.03120.x
- Dunham, E. M., Belanger, D., Cong, L., & Kozdon, J. E. (2011). Earthquake Ruptures with 364 Strongly Rate-Weakening Friction and Off-Fault Plasticity, Part 2: Nonplanar Faults. 365 Bulletin of the Seismological Society of America, 101(5), 2308-2322. 366 https://doi.org/10.1785/0120100076 367
- Duru, K., & Dunham, E. M. (2016). Dynamic earthquake rupture simulations on nonplanar faults
 embedded in 3D geometrically complex, heterogeneous elastic solids. *Journal of Computational Physics*, 305, 185–207. https://doi.org/10.1016/j.jcp.2015.10.021
- 371 Eurocode 8, (EC8). (2004). EN 1998-1, Design of structures for earthquake resistance, part 1:
- 372 *General rules, seismic actions and rules for buildings*. European Committee for
 373 Standardization (CEN).
- 374 Fang, Z., & Dunham, E. M. (2013). Additional shear resistance from fault roughness and stress

- levels on geometrically complex faults. *Journal of Geophysical Research: Solid Earth*, *118*(7), 3642–3654. https://doi.org/10.1002/jgrb.50262
- 377 Farr, T. G., Rosen, P. A., Caro, E., Crippen, R., Duren, R., Hensley, S., Kobrick, M., Paller, M.,
- 378 Rodriguez, E., Roth, L., Seal, D., Shaffer, S., Shimada, J., Umland, J., Werner, M.,
- Oskin, M., Burbank, D., & Alsdorf, D. (2007). The Shuttle Radar Topography Mission.
 Reviews of Geophysics, 45(2). https://doi.org/10.1029/2005RG000183
- Frankel, A., Wirth, E., Marafi, N., Vidale, J., & Stephenson, W. (2018). Broadband Synthetic 381 Seismograms for Magnitude 9 Earthquakes on the Cascadia Megathrust Based on 3D 382 Simulations and Stochastic Synthetics, Part 1: Methodology and Overall Results. Bulletin 383 of the Seismological 2347-2369. Societv of America. 108(5A), 384 https://doi.org/10.1785/0120180034 385
- Gallovič, F., & Ampuero, J.-P. (2015). A New Strategy to Compare Inverted Rupture Models
 Exploiting the Eigenstructure of the Inverse Problem. *Seismological Research Letters*,
 86(6), 1679–1689. https://doi.org/10.1785/0220150096
- Gallovič, F., Valentová, Ľ., Ampuero, J.-P., & Gabriel, A.-A. (2019). Bayesian Dynamic Finite Fault Inversion: 1. Method and Synthetic Test. *Journal of Geophysical Research: Solid Earth*, 124(7), 6949–6969. https://doi.org/10.1029/2019JB017510
- Graves, R., Jordan, T. H., Callaghan, S., Deelman, E., Field, E., Juve, G., Kesselman, C.,
 Maechling, P., Mehta, G., Milner, K., Okaya, D., Small, P., & Vahi, K. (2011).
 CyberShake: A Physics-Based Seismic Hazard Model for Southern California. *Pure and*
- 395 *Applied Geophysics*, *168*(3), 367–381. https://doi.org/10.1007/s00024-010-0161-6
- Graves, R. W., & Pitarka, A. (2010). Broadband Ground-Motion Simulation Using a Hybrid
 Approach. Bulletin of the Seismological Society of America, 100(5A), 2095–2123.

398 https://doi.org/10.1785/0120100057

- 399 Harris, R. A., Barall, M., Aagaard, B., Ma, S., Roten, D., Olsen, K., Duan, B., Liu, D., Luo, B.,
- 400 Bai, K., Ampuero, J., Kaneko, Y., Gabriel, A., Duru, K., Ulrich, T., Wollherr, S., Shi, Z.,
- 401 Dunham, E., Bydlon, S., ... Dalguer, L. (2018). A Suite of Exercises for Verifying
- 402 Dynamic Earthquake Rupture Codes. *Seismological Research Letters*, *89*(3), 1146–1162.
- 403 https://doi.org/10.1785/0220170222
- Hartzell, S., Ramírez-Guzmán, L., Meremonte, M., & Leeds, A. (2016). Ground Motion in the
 Presence of Complex Topography II: Earthquake Sources and 3D Simulations. *Bulletin of the Seismological Society of America*, 107(1), 344–358.
 https://doi.org/10.1785/0120160159
- 408 Heinecke, A., Breuer, A., Rettenberger, S., Bader, M., Gabriel, Pelties, C., Bode, A., Barth, W.,
- Liao, X., Vaidyanathan, K., Smelyanskiy, M., & Dubey, P. (2014). Petascale High Order
- 410 Dynamic Rupture Earthquake Simulations on Heterogeneous Supercomputers. SC '14:
- 411 Proceedings of the International Conference for High Performance Computing,

412 *Networking, Storage and Analysis*, 3–14. https://doi.org/10.1109/SC.2014.6

- Ida, Y. (1972). Cohesive force across the tip of a longitudinal-shear crack and Griffith's specific
 surface energy. *Journal of Geophysical Research (1896-1977)*, 77(20), 3796–3805.
 https://doi.org/10.1029/JB077i020p03796
- 416 Iervolino, I., Galasso, C., & Cosenza, E. (2010). REXEL: computer aided record selection for
- 417 code-based seismic structural analysis. *Bulletin of Earthquake Engineering*, 8(2), 339–
 418 362. https://doi.org/10.1007/s10518-009-9146-1
- 419 Imperatori, W., & Mai, P. M. (2013). Broad-band near-field ground motion simulations in 3-
- Imperatori, W., & Mai, P. M. (2013). Broad-band near-field ground motion simulations in 3dimensional scattering media. *Geophysical Journal International*, 192(2), 725–744.

421 https://doi.org/10.1093/gji/ggs041

- Käser, M., Hermann, V., & Puente, J. de la. (2008). Quantitative accuracy analysis of the
 discontinuous Galerkin method for seismic wave propagation. *Geophysical Journal International*, *173*(3), 990–999. https://doi.org/10.1111/j.1365-246X.2008.03781.x
- 425 Kheirdast, N., Ansari, A., & Custódio, S. (2021). Neuro-Fuzzy Kinematic Finite-Fault Inversion:
- 426 2. Application to the Mw6.2, August/24/2016, Amatrice Earthquake. *Journal of*427 *Geophysical Research: Solid Earth*, *126*(8), e2020JB020773.
 428 https://doi.org/10.1029/2020JB020773
- Konno, K., & Ohmachi, T. (1998). Ground-motion characteristics estimated from spectral ratio
 between horizontal and vertical components of microtremor. *Bulletin of the Seismological Society of America*, 88(1), 228–241.
- Langer, L., Gharti, H. N., & Tromp, J. (2019). Impact of topography and three-dimensional
 heterogeneity on coseismic deformation. *Geophysical Journal International*, 217(2),
 866–878. https://doi.org/10.1093/gji/ggz060
- 435 Lanzano, G., Luzi, L., Cauzzi, C., Bienkowski, J., Bindi, D., Clinton, J., Cocco, M., D'Amico,
- 436 M., Douglas, J., Faenza, L., Felicetta, C., Gallovic, F., Giardini, D., Ktenidou, O.,
- 437 Lauciani, V., Manakou, M., Marmureanu, A., Maufroy, E., Michelini, A., ...
- Theodoulidis, N. (2021). Accessing European Strong-Motion Data: An Update on ORFEUS Coordinated Services. *Seismological Research Letters*, *92*(3), 1642–1658.
- 440 https://doi.org/10.1785/0220200398
- 441 Lee, S.-J., Komatitsch, D., Huang, B.-S., & Tromp, J. (2009). Effects of Topography on Seismic-
- Wave Propagation: An Example from Northern Taiwan. *Bulletin of the Seismological Society of America*, 99(1), 314–325. https://doi.org/10.1785/0120080020

444	Luzi, L., Puglia, R., Russo, E., D'Amico, M., Felicetta, C., Pacor, F., Lanzano, G., Çeken, U.,						
445	Clinton, J., Costa, G., Duni, L., Farzanegan, E., Gueguen, P., Ionescu, C., Kalogeras, I.,						
446	Özener, H., Pesaresi, D., Sleeman, R., Strollo, A., & Zare, M. (2016). The Engineering						
447	Strong-Motion Database: A Platform to Access Pan-European Accelerometric Data.						
448	Seismological Research Letters, 87(4), 987–997. https://doi.org/10.1785/0220150278						
449	Madariaga, R. (1977). High-frequency radiation from crack (stress drop) models of earthquake						
450	faulting. Geophysical Journal International, 51(3), 625–651.						
451	https://doi.org/10.1111/j.1365-246X.1977.tb04211.x						
452	Mai, P. M., Imperatori, W., & Olsen, K. B. (2010). Hybrid Broadband Ground-Motion						
453	Simulations: Combining Long-Period Deterministic Synthetics with High-Frequency						
454	Multiple S-to-S BackscatteringHybrid Broadband Ground-Motion Simulations:						
455	Combining Deterministic Synthetics with Backscattering. Bulletin of the Seismological						
456	Society of America, 100(5A), 2124–2142. https://doi.org/10.1785/0120080194						
457	Mai, P. M., Schorlemmer, D., Page, M., Ampuero, J., Asano, K., Causse, M., Custodio, S., Fan,						
458	W., Festa, G., Galis, M., Gallovic, F., Imperatori, W., Käser, M., Malytskyy, D.,						
459	Okuwaki, R., Pollitz, F., Passone, L., Razafindrakoto, H. N. T., Sekiguchi, H., Zielke,						
460	O. (2016). The Earthquake-Source Inversion Validation (SIV) Project. Seismological						
461	Research Letters, 87(3), 690-708. https://doi.org/10.1785/0220150231						
462	Michele, M., Chiaraluce, L., Di Stefano, R., & Waldhauser, F. (2020). Fine-Scale Structure of						
463	the 2016–2017 Central Italy Seismic Sequence From Data Recorded at the Italian						
464	National Network. Journal of Geophysical Research: Solid Earth, 125(4),						
465	e2019JB018440. https://doi.org/10.1029/2019JB018440						

466 Michele, M., Di Stefano, R., Chiaraluce, L., Cattaneo, M., De Gori, P., Monachesi, G., Latorre,

- D., Marzorati, S., Valoroso, L., Ladina, C., Chiarabba, C., Lauciani, V., & Fares, M.
 (2016). The Amatrice 2016 seismic sequence: A preliminary look at the mainshock and
 aftershocks distribution. *Annals of Geophysics; Vol 59, Fast Track (2016), 59.*
- 470 Moschetti, M. P., Hartzell, S., Ramírez-Guzmán, L., Frankel, A. D., Angster, S. J., &
- 471 Stephenson, W. J. (2017). 3D Ground Motion Simulations of Mw 7 Earthquakes on the
- 472 Salt Lake City Segment of the Wasatch Fault Zone: Variability of Long-Period ($T \ge 1$ s)
- 473 Ground Motions and Sensitivity to Kinematic Rupture Parameters. *Bulletin of the* 474 *Seismological Society of America*, 107(4), 1704–1723.
- 475 https://doi.org/10.1785/0120160307
- 476 Okubo, K., Bhat, H. S., Rougier, E., Marty, S., Schubnel, A., Lei, Z., Knight, E. E., & Klinger,
- Y. (2019). Dynamics, Radiation, and Overall Energy Budget of Earthquake Rupture With
 Coseismic Off-Fault Damage. *Journal of Geophysical Research: Solid Earth*, *124*(11),
- 479 11771–11801. https://doi.org/10.1029/2019JB017304
- Olsen, K. B., & Mayhew, J. E. (2010). Goodness-of-fit Criteria for Broadband Synthetic
 Seismograms, with Application to the 2008 Mw 5.4 Chino Hills, California, Earthquake.
 Seismological Research Letters, *81*(5), 715–723. https://doi.org/10.1785/gssrl.81.5.715
- Palmer, A. C., Rice, J. R., & Hill, R. (1973). The growth of slip surfaces in the progressive
 failure of over-consolidated clay. *Proceedings of the Royal Society of London. A. Mathematical and Physical Sciences*, 332(1591), 527–548.
 https://doi.org/10.1098/rspa.1973.0040
- Passone, L., & Mai, P. M. (2017). Kinematic Earthquake Ground-Motion Simulations on Listric
 Normal Faults. *Bulletin of the Seismological Society of America*, 107(6), 2980–2993.
- 489 https://doi.org/10.1785/0120170111

- Pelties, C., de la Puente, J., Ampuero, J.-P., Brietzke, G. B., & Käser, M. (2012). Threedimensional dynamic rupture simulation with a high-order discontinuous Galerkin
 method on unstructured tetrahedral meshes. *Journal of Geophysical Research: Solid Earth*, 117(B2). https://doi.org/10.1029/2011JB008857
- Pelties, C., Gabriel, A.-A., & Ampuero, J.-P. (2014). Verification of an ADER-DG method for
 complex dynamic rupture problems. *Geoscientific Model Development*, 7(3), 847–866.
 https://doi.org/10.5194/gmd-7-847-2014
- 497 Petersen, M. D., Shumway, A. M., Powers, P. M., Mueller, C. S., Moschetti, M. P., Frankel, A.
- 498 D., Rezaeian, S., McNamara, D. E., Luco, N., Boyd, O. S., Rukstales, K. S., Jaiswal, K.
- 499 S., Thompson, E. M., Hoover, S. M., Clayton, B. S., Field, E. H., & Zeng, Y. (2019). The
- 2018 update of the US National Seismic Hazard Model: Overview of model and
 implications. *Earthquake Spectra*, 36(1), 5–41.
 https://doi.org/10.1177/8755293019878199
- Peyrat, S., & Olsen, K. B. (2004). Nonlinear dynamic rupture inversion of the 2000 Western
 Tottori, Japan, earthquake. *Geophysical Research Letters*, 31(5).
 https://doi.org/10.1029/2003GL019058
- Pischiutta, M., Akinci, A., Tinti, E., & Herrero, A. (2021). Broad-band ground-motion
 simulation of 2016 Amatrice earthquake, Central Italy. *Geophysical Journal International*, 224(3), 1753–1779. https://doi.org/10.1093/gji/ggaa412
- 509 Pitarka, A., Akinci, A., De Gori, P., & Buttinelli, M. (2021). Deterministic 3D Ground-Motion
- 510 Simulations (0–5 Hz) and Surface Topography Effects of the 30 October 2016 Mw 6.5
- 511 Norcia, Italy, Earthquake. Bulletin of the Seismological Society of America.
 512 https://doi.org/10.1785/0120210133

Pizzi, A., Di Domenica, A., Gallovič, F., Luzi, L., & Puglia, R. (2017). Fault Segmentation as 513 Constraint to the Occurrence of the Main Shocks of the 2016 Central Italy Seismic 514 Sequence. Tectonics, 36(11), 2370–2387. https://doi.org/10.1002/2017TC004652 515 Premus, J., Gallovič, F., Hanyk, L., & Gabriel, A. (2020). FD3D TSN: A Fast and Simple Code 516 for Dynamic Rupture Simulations with GPU Acceleration. Seismological Research 517 Letters, 91(5), 2881–2889. https://doi.org/10.1785/0220190374 518 Ragon, T., Sladen, A., & Simons, M. (2018). Accounting for uncertain fault geometry in 519 earthquake source inversions – I: theory and simplified application. *Geophysical Journal* 520 International, 214(2), 1174–1190. https://doi.org/10.1093/gji/ggy187 521 Ripperger, J., Mai, P. M., & Ampuero, J.-P. (2008). Variability of Near-Field Ground Motion 522 from Dynamic Earthquake Rupture Simulations. Bulletin of the Seismological Society of 523 America, 98(3), 1207–1228. https://doi.org/10.1785/0120070076 524 Rodgers, A. J., Pitarka, A., Pankajakshan, R., Sjögreen, B., & Petersson, N. A. (2020). Regional-525 Scale 3D Ground-Motion Simulations of Mw 7 Earthquakes on the Hayward Fault, 526 Northern California Resolving Frequencies 0-10 Hz and Including Site-Response 527 Corrections. Seismological Bulletin of the Societv of America. 528 https://doi.org/10.1785/0120200147 529 Roten, D., Olsen, K. B., Day, S. M., Cui, Y., & Fäh, D. (2014). Expected seismic shaking in Los 530 Angeles reduced by San Andreas fault zone plasticity. Geophysical Research Letters, 531 41(8), 2769–2777. https://doi.org/10.1002/2014GL059411 532 Roten, D., Olsen, K. B., & Pechmann, J. C. (2012). 3D Simulations of M 7 Earthquakes on the 533 Wasatch Fault, Utah, Part II: Broadband (0-10 Hz) Ground Motions and Nonlinear Soil 534 535 Behavior. Bulletin of the Seismological Society of America, 102(5), 2008-2030.

- 536 https://doi.org/10.1785/0120110286
- Sambridge, M. (2013). A Parallel Tempering algorithm for probabilistic sampling and
 multimodal optimization. *Geophysical Journal International*, 196(1), 357–374.
 https://doi.org/10.1093/gii/ggt342
- Savran, W. H., & Olsen, K. B. (2020). Kinematic Rupture Generator Based on 3-D Spontaneous
 Rupture Simulations Along Geometrically Rough Faults. *Journal of Geophysical Research: Solid Earth*, *125*(10), e2020JB019464. https://doi.org/10.1029/2020JB019464
- Scognamiglio, L., Tinti, E., Casarotti, E., Pucci, S., Villani, F., Cocco, M., Magnoni, F.,
 Michelini, A., & Dreger, D. (2018). Complex Fault Geometry and Rupture Dynamics of
 the MW 6.5, 30 October 2016, Central Italy Earthquake. *Journal of Geophysical Research: Solid Earth*, *123*(4), 2943–2964. https://doi.org/10.1002/2018JB015603
- Shi, Z., & Day, S. M. (2013). Rupture dynamics and ground motion from 3-D rough-fault
 simulations. *Journal of Geophysical Research: Solid Earth*, *118*(3), 1122–1141.
 https://doi.org/10.1002/jgrb.50094
- Shimizu, K., Yagi, Y., Okuwaki, R., & Fukahata, Y. (2020). Development of an inversion
 method to extract information on fault geometry from teleseismic data. *Geophysical Journal International*, 220(2), 1055–1065. https://doi.org/10.1093/gji/ggz496
- Simmetrix Inc. (2017). SimModeler: Simulation Modeling Suite 11.0 Documentation.
 www.simmetrix.org
- Takemura, S., Furumura, T., & Maeda, T. (2015). Scattering of high-frequency seismic waves
 caused by irregular surface topography and small-scale velocity inhomogeneity.
 Geophysical Journal International, 201(1), 459–474. https://doi.org/10.1093/gji/ggv038
- 558 Tinti, E., Casarotti, E., Ulrich, T., Taufiqurrahman, T., Li, D., & Gabriel, A.-A. (2021).

- Tinti, E., Scognamiglio, L., Michelini, A., & Cocco, M. (2016). Slip heterogeneity and
 directivity of the ML 6.0, 2016, Amatrice earthquake estimated with rapid finite-fault
 inversion. *Geophysical Research Letters*, 43(20), 10,745-10,752.
 https://doi.org/10.1002/2016GL071263
- Tung, S., & Masterlark, T. (2018). Resolving Source Geometry of the 24 August 2016 Amatrice,
 Central Italy, Earthquake from InSAR Data and 3D Finite-Element Modeling. *Bulletin of the Seismological Society of America*, 108(2), 553–572.
 https://doi.org/10.1785/0120170139
- Valentová, Ľ., Gallovič, F., & Hok, S. (2021). Near-Source Ground Motions and Their
 Variability Derived from Dynamic Rupture Simulations Constrained by NGA-West2
 GMPEs. *Bulletin of the Seismological Society of America*, *111*(5), 2559–2573.
 https://doi.org/10.1785/0120210073
- Walters, R. J., Gregory, L. C., Wedmore, L. N. J., Craig, T. J., McCaffrey, K., Wilkinson, M.,
- 575 Chen, J., Li, Z., Elliott, J. R., Goodall, H., Iezzi, F., Livio, F., Michetti, A. M., Roberts,
- G., & Vittori, E. (2018). Dual control of fault intersections on stop-start rupture in the
 2016 Central Italy seismic sequence. *Earth and Planetary Science Letters*, 500, 1–14.
 https://doi.org/10.1016/j.epsl.2018.07.043
- Wirth, E. A., Frankel, A. D., Marafi, N., Vidale, J. E., & Stephenson, W. J. (2018). Broadband
 Synthetic Seismograms for Magnitude 9 Earthquakes on the Cascadia Megathrust Based
 on 3D Simulations and Stochastic Synthetics, Part 2: Rupture Parameters and Variability.

- Bulletin of the Seismological Society of America, 108(5A), 2370–2388.
 https://doi.org/10.1785/0120180029
- Withers, K. B., Moschetti, M. P., & Thompson, E. M. (2020). A Machine Learning Approach to
 Developing Ground Motion Models From Simulated Ground Motions. *Geophysical Research Letters*, 47(6), e2019GL086690. https://doi.org/10.1029/2019GL086690
- 587 Withers, K. B., Olsen, K. B., Day, S. M., & Shi, Z. (2018). Ground Motion and Intraevent
- 588 Variability from 3D Deterministic Broadband (0–7.5 Hz) Simulations along a Nonplanar
- 589 Strike-Slip Fault. Bulletin of the Seismological Society of America, 109(1), 229–250.
- 590 https://doi.org/10.1785/0120180006
- Wollherr, S., Gabriel, A.-A., & Uphoff, C. (2018). Off-fault plasticity in three-dimensional
 dynamic rupture simulations using a modal Discontinuous Galerkin method on
 unstructured meshes: Implementation, verification and application. *Geophysical Journal International*, 214(3), 1556–1584. https://doi.org/10.1093/gji/ggy213
- 595 Yamashita, T. (2000). Generation of microcracks by dynamic shear rupture and its effects on
- ⁵⁹⁶ rupture growth and elastic wave radiation. *Geophysical Journal International*, 143(2),
- 597 395–406. https://doi.org/10.1046/j.1365-246X.2000.01238.x



Figure 1. (a) Three-dimensional dynamic rupture model setup of the 2016 Mw 6.2 Amatrice, Central Italy, earthquake. Snapshot of the absolute surface velocity at a simulation time of 16 seconds. The model is discretized by an unstructured tetrahedral mesh refined in the vicinity of the fault and the high-resolution topography. Twenty strong-motion stations used in this study are marked in black (see Table S1). Mesh elements are colored by shear wave velocity (Vs). (b) and (c) Fault slip for the smooth Bayesian dynamic source inversion reference model (b) and the broadband dynamic rupture model (c). Black curves represent rupture front contours every 1 s.



Figure 2. (a) and (b) Comparison of dynamic parameters used in the reference model (a), and the broadband rough fault model (b). Fractal heterogeneity is also added to the distribution of slip weakening distance (D_c). The black contour marks the nucleating negative strength excess area. (c) and (d) Dynamic rupture propagation in the reference (c) and broadband (d) rough fault models of the Amatrice earthquake. Snapshots of the absolute fault slip rates illustrate the similar space-time evolution in both models.



Figure 3. Comparison of observed (black) and simulated (red) components (NS, EW, and Z) of (a) ground velocity (in cm/s) and (b) acceleration (in cm/s²) band-pass filtered between 0.05 and 5 Hz for all 20 stations (Figure 1). Synthetics are from the broadband dynamic rupture scenario incorporating fault roughness, D_c heterogeneity, and topography. Both observed and synthetic waveforms are scaled by their maximum value, which is indicated on the right-hand side of each plot. Velocity waveforms are scaled by the maximum value of the observed records at each station, while acceleration waveforms are scaled component-wise.



Figure 4. Broadband velocity and acceleration waveforms and Fourier amplitude spectra. (top, 618 middle) Comparison of EW component of synthetic ground-velocity (top rows) and acceleration 619 620 (middle rows) waveforms from the broadband rough fault model with topography (red), the broadband rough fault model without topography (green), and the reference model (grey) 621 compared with observations (black) at five selected stations (see Figure 1). All waveforms are 622 623 scaled by their maximum values, indicated on the left-hand side of each plot. (bottom) Smoothed (using the method of Konno & Ohmachi, 1998) Fourier amplitude spectra (FAS) of the velocity 624 625 waveforms. The observed data is tapered with a 35 s cosine window.

AGU PUBLICATIONS

1	Geophysical Research Letters						
2	Supporting Information for						
3 4 5	Broadband dynamic rupture modeling with fractal fault roughness, frictional heterogeneity, viscoelasticity and topography: the 2016 Mw 6.2 Amatrice, Italy earthquake						
6	T. Taufiqurrahman ¹ , AA. Gabriel ¹ , T. Ulrich ¹ , L. Valentová ² , F. Gallovič ²						
7 8	¹ Department of Earth and Environmental Sciences, Faculty of Geosciences, Ludwig Maximilian University, Munich, Germany						
9 10	² Department of Geophysics, Faculty of Mathematics and Physics, Charles University, Prague, Czech Republic						
11							
12	Contents of this file						
13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29	 Text S1: Numerical discretization and resolution Text S2: Empirical quantification of the 3D roughness drag Text S3: Adaption of the reference initial loading to account for the 3D roughness drag Text S4: Goodness-of-fit (GOF) of broadband and reference ground motions Table S1: Strong motion stations Table S2: 1D velocity model Figure S1: Coefficient <i>C</i> which allows recovering the planar rupture dynamics on the rough fault model and the number of elements per roughness wavelength <i>n</i> used to resolve rough fault models. Figure S2: Comparison of the relative prestress ratio R for the planar reference model and the rough fault broadband model with and without roughness drag correction. Figure S3: Moment rate release, moment spectrum, and 2nd-time derivatives of moment rate Figure S4: Comparison of the observed and synthetic velocity and acceleration waveforms of the planar fault model 						
30 31	11. Figure S5: Comparison of the observed and synthetic velocity and acceleration waveforms of the broadband rough fault model without topography						

32 33	12.	Figure S6-S12: Comparison of the observed and synthetic velocity, acceleration waveforms, and Fourier amplitude spectrums (FAS) of both the planar fault and
34		the broadband rough fault model for EW and NS-components
35	13.	Figure S13: Comparison of the observed, synthetic of the planar fault model with
36		topography and without viscoelastic attenuation, and synthetic velocity
37		waveforms of the planar fault model with topography and with viscoelastic
38		attenuation
39	14.	Figure S14: Histograms of Goodness-of-fit (GOF) between observation and
40		velocity waveforms of the planar fault and the broadband rough fault model for
41		all three-components
42	15.	Figure S15: Comparison of Goodness-of-fit (GOF) between the planar fault and
43		the broadband rough fault model for all three components for all stations
44	16.	Figure S16: Model bias plot comparing between observation and velocity
45		waveforms of the planar fault model, and the broadband rough fault model
46	Additi	onal Supporting Information (Files uploaded separately)
17	1	Movie S1: Absolute slip rate (m/s) across the fault for the reference and the
48	1.	broadband rough fault models:
40 49		https://drive.google.com/file/d/1TXFbNAshafduanUGMWZWi413tgzwZNN5/
50		view?usp=sharing
51	2	Movie S2: Top view of fault parallel (u) ground surface velocity for the reference
52	2.	model.
53		https://drive.google.com/file/d/1abwnd5iFs5R8AAA2by1O3NGSrC-SkKcY/
54		view?usp=sharing
55	3	Movie S3: Top view of fault normal (y) ground surface velocity for the reference
56	5.	model:
57		https://drive.google.com/file/d/111vi9le_BusxM6ZGYCtYf4CO0S0HluRO/view?
58		usn=sharing
59	4	Movie S4 [·] Top view of vertical (w) ground surface velocity for the reference
60		model.
61		https://drive.google.com/file/d/1vAvXXEsGGwVROKeOhJXG8J10_ikwOAIf/
62		view?usp=sharing
63	5	Movie S5. Top view of fault parallel (u) ground surface velocity for the
64		broadband rough fault model with topography:
65		https://drive.google.com/file/d/1xdlG5_KPVXe52AogYztJnBFoZcW9czAO/
66		view?usp=sharing
67	6	Movie S6: Top view of fault normal (v) ground surface velocity for the broadband
68	0.	rough fault model with topography.
69		https://drive.google.com/file/d/1PSDfvSM5bS-WO-U_Ls0-wp-oHZC5IGIz/
70		view?usn=sharing
, 0		Terr wep blatting

71 7. Movie S7: Top view of vertical (w) ground surface velocity for the broadband
 72 rough fault model with topography:

https://drive.google.com/file/d/1aIHHXBIBf9VNnDBJkhjZKWBqiavWCG8H/
 view?usp=sharing

75 Text S1. Numerical discretisation and resolution

Our high-resolution 3D dynamic rupture simulations are performed using the open-source software package SeisSol (<u>https://github.com/SeisSol/SeisSol</u>) resolving seismic wave excitation locally up to 10 Hz and ground motions up to at least 5 Hz within 50 km distance of the fault. SeisSol is based on the Arbitrary high-order DERivative Discontinuous Galerkin (ADER-DG) method (Dumbser & Käser, 2006; Pelties et al., 2012, 2014).

82 The dimensions of the model are 300 km x 300 km x 150 km (depth). We 83 gradually increase the element size towards a maximum edge length of 10 km at the 84 domain edges to reduce computational cost without sacrificing accuracy in the region of 85 interest. This region is a highly resolved subdomain spanning 50 km x 50 km horizontally 86 and 10 km in depth, centered at the hypocenter. There the high-order accurate element edge lengths range from 150 to 350 m, adapted to the 1D velocity model and requiring at 87 least two elements resolving the shortest wavelengths (Käser et al., 2008), computed for a 88 target frequency of 5 Hz. 89

90 We adapt the planar fault to adhere to band-limited fractal surface morphology characterized by wavelengths ranging from $\lambda_{\min}=200$ m to $\lambda_{\max}=2$ km. At least n=291 92 elements per wavelength λ_{\min} are required to capture the complexity of the band-limited 93 rough fault model without aliasing with SeisSol, which uses unstructured tetrahedral 94 meshes. In this study, the rough fault is generated using the Fourier transform method. 95 The rough fault can, therefore, be viewed as a weighted sum of sinusoids, which are here 96 approximated by piece-wise bilinear functions due to SeiSol's geometrically linear 97 triangles representing fault surfaces. We note that this leads to an artificial enhancement 98 of the shorter wavelength content of the rough geometry.

99 For example, a 2-node approximation of a sine wave of wavelength λ can be 100 decomposed into the sum of sinusoids of wavelength λ and its multiples λ/p . By using 101 higher n, we decrease the artificial short-wavelengths content of wavelengths shorter than 102 λ_{\min} . The rough fault geometry then better resembles a band-limited fractal distribution. 103 Low n (i.e., 2 to 4) are sufficient to capture the effects of fault roughness on earthquake 104 dynamics and ground motions and also promote well-balanced numerics by limiting the 105 number of elements with dynamic rupture boundaries. Even higher n would allow the 106 discrete fault geometry to approach curvilinear approaches (e.g., Duru & Dunham, 2016) 107 and for a better control of the amplitude spectrum of fault roughness at short 108 wavelengths.

109 For our choice of n=4, the fault is discretized using 50 m sized elements, ensuring 110 that the process zone size and thus rupture dynamics are sufficiently resolved everywhere on the fault (Day et al., 2005; Wollherr et al., 2018). We measure the minimum process 111 112 zone size, the region behind the rupture front where the fault strength drops from its static to dynamic level, as being equal to 200 m. The resulting mesh has more than 80 million 113 114 tetrahedral elements (Figure 1a) and is generated using Simmodeler (Simmetrix Inc., 2017). Simulating 40 s of a broadband Amatrice dynamic rupture earthquake scenario 115 116 using SeisSol with fifth-order accuracy in space and time (i.e., basis functions of 117 polynomial order p=4) and double precision requires 4 hours on 256 nodes of the 118 SuperMUC-NG supercomputer.

119 Text S2. Empirical quantification of the 3D roughness drag

We empirically approximate the roughness drag τ_{drag}^{3D} (Equation 2) through 120 systematic dynamic rupture simulations varying C, the minimum roughness 121 wavelength λ_{\min} , and the number of elements *n* resolving λ_{\min} . We remind the reader that 122 C is a dimensionless coefficient for empirical approximation of the 3D roughness drag. 123 124 We find the preferred scenario for each parameterisation by comparing the space-time 125 evolution of dynamically self-sustained rupture along the fault, seismic moment, peak 126 seismic moment release, and timing of the peak seismic moment release to the reference 127 model. We find that approximating C, as

128
$$C \approx m \lambda_{\min} + b$$
, (S1)

i.e., a linear function with slope m = 0.001 and intercept b = 0.315(1 - 1 / n), allows us to recover broadband models matching the reference model for a range of broadband dynamic rupture models and discretizations (Figure S1a). Figure S1b shows analyzed and preferred values of intercept *b* for varying number of elements *n* per λ_{min} .

133 Text S3. Empirical quantification of the 3D roughness drag

134 We adapt the reference loading, which is prescribed as smoothly varying fault 135 local initial tractions, to an equivalent globally defined Cartesian background stress 136 allowing for geometry-induced small-scale traction heterogeneities. In this way, we can 137 also account for the above quantified 3D roughness drag while preserving the smooth reference initial stress distribution. We build a heterogeneous stress tensor from τ_{dip} , τ_{strike} , 138 and σ_n (see Section 2.2). The reference coordinate system of our model has the x-axis 139 140 aligned with fault strike and the y-axis horizontal, pointing away from the hanging wall. We load the fault by an initial stress tensor (σ_{ii}) defined in a fault coordinate system 141 (x,u,v) aligned with the planar reference model fault. v points up-dip, and u is oriented 142 143 normally to the fault such that (x,u,v) forms a right-handed coordinate system. We set

 $-\sigma_{uv} = \tau_{dip}$, $-\sigma_{xu} = -\sigma_{xv} = \tau_{drag}^{3D}$ to compensate for the 3D roughness drag effects on all 144 initial shear stresses components, and $-\sigma_{xx} = -\sigma_{vv} = -\sigma_{uu} = \sigma_n$ (assuming compressive 145 normal stresses being negative). This stress tensor is finally rotated by 45°, the dip angle 146 147 of the planar reference fault, with respect to the x-axis to the reference coordinate system. 148 The resulting heterogeneous initial loading features roughness-induced small-scale 149 fluctuations of the initial shear and normal tractions (Figure 2b), consisting of both 150 releasing and restraining slopes, in which the fault is brought closer (resp. farther away) 151 from failure (Fig S2a).

Due to the added τ_{drag}^{3D} term, the initial shear traction may exceed the initial fault 152 strength locally. To prevent instantaneous failure across the fault, we limit τ_{dip} and τ_{strike} to 153 154 be at least 0.5 MPa lower than fault strength everywhere on the fault except the 155 nucleation area. Rupture is initiated in an area of negative strength excess of ~1 km 156 radius located 16 km along-strike and 7 km down-dip (highlighted by a black line in 157 Figure 2a,b). We empirically find that in the such modified rough fault model, 30% 158 higher negative strength excess (τ_s - τ_0) of ~1.3 MPa is required to model the dynamically 159 very sensitive weak nucleation from the reference model.

160 Text S4. Goodness of fit of broadband and reference ground motion

We quantify the fit between observations and synthetic ground-motions for all models (reference, broadband without topography, and broadband with topography) using Goodness-of-fit (GOF) metrics (Olsen & Mayhew, 2010). We compute the average GOF of 0.05–5 Hz bandpass filtered signals using the following metrics:

- 165
- Peak ground velocity (PGV)
- Peak ground displacement (PGD)
- Spectral Acceleration (SA) at periods 0.5-10 s
- Fourier amplitude spectra (FS)
- Energy duration (DUR)
- Cumulative energy (ENER)

We do not consider PGA because it is susceptible to site effects, which we do not
account for in this study. Note that in our GOF computations, we exclude station RQT,
for which no NS-component recording is available, and station CLF because of its strong
broadband site-effect (class D).

Figure S14 shows the distribution of average GOF at all station components for the reference and both rough fault models. The histograms in Figure S14 show the prevalence of fairly good values (GOF 45-65) for both reference and rough fault models. The rough fault model with topography fits the observations generally better than the reference model and the rough fault model without topography for all components, with an average GOF of 55, 65 and 55 for EW, NS and Z components, respectively. The best
fit stations (GOF > 55 for all components) are PZI, LSS and TERO, located SE from the
fault. Figure S15 details the average GOF for each station, component, and model. GOF
values near or below 35 are observed at stations NRC and FEMA (EW components),
stations NRC, ASP, and TRE (NS components), and station MNF (Z component).

186 We also calculate residuals r_j of spectral accelerations using the natural logarithm 187 of the ratio of the observation O_i and synthetics S_i for each site *j* as a function of period T_i ,

188
$$r_i(T_i) = \log[O_i(T_i) / S_i(T_i)]$$
 (S3)

189 Here, O_j and S_j are calculated as the geometric mean of the horizontal 190 components. Mean model bias *b* for number of stations N = 20 is defined as

191
$$b(T_i) = 1/N \Sigma r_j(T_i)$$
 (S4)

192 with its standard deviation σ calculated as

193
$$\sigma(T_i) = [1/N \Sigma (r_i(T_i) - b(T_i))^2]^{\frac{1}{2}}$$
 (S5)

194 The model bias for the three tested models are shown in Fig. S15.

195 Table S1. Strong motion stations at which ground motion waveforms are compared in

196	this study. All 20 stations are within a radius of 50 km from the Mw 6.2 Amatrice event
197	epicenter.

Code	Station Name	Longitude (°E)	Latitude (°N)	Site Class* (EC8)
AMT	Amatrice	42.6325	13.2866	В
NRC	Norcia	42.7925	13.0964	В
MNF	Monte Fiegni (Fiastra)	43.0596	13.1844	А
TRL	Terminillo	42.4613	12.9323	В
ANT	Antrodoco	42.4182	13.0786	А
PZI1	Pizzoli	42.4356	13.3262	В
LSS	Leonessa	42.5582	12.9689	А
SPD	Sella Pedicate (Campotosto)	42.5151	13.371	В
FOS	Foligno Seggio	43.0146	12.8351	В
ASP	Ascoli Piceno	42.848	13.6479	В
TRE	Trevi	42.8765	12.7358	С
SPM	Spoleto (Monteluco)	42.7232	12.7512	А
FEMA	Monte Fema	42.9621	13.04976	А
TERO	Teramo	42.62279	13.60393	А
RM33	Pellescritta	42.50898	13.21452	В
RQT	Arquata del Tronto	42.813	13.311	А
SNO	Sarnano	43.0371	13.3041	В
CSC	Cascia	42.719	13.0122	В
MSC	Mascioni (Campotosto)	42.5268	13.3508	В
CLF	Colfiorito	43.03671	12.92043	D

198 *) the Engineering Strong-Motion database (<u>https://esm.mi.ingv.it/</u>, Luzi et al., 2016;

199 Lanzano et al., 2021). Site classification according to EC8 (Eurocode 8, 2004).

Depth (km)	Vp (km/s)	Vs (km/s)	ρ (g/cm ³)	Qp	Qs
0	3.16	1.70	2.50	200	100
1	4.83	2.60	2.84	400	200
2	5.76	3.10	2.94	400	200
5	6.51	3.50	3.15	400	200
27	7.00	3.80	3.26	600	300
42	7.80	4.20	3.50	800	400

Table S2. 1D Velocity model (Ameri et al., 2012) assumed in this study.



Figure S1. (a) Tested and preferred values of coefficient *C* (Eq. 2), relating τ_{drag}^{3D} to τ_{drag}^{2D} (Eq. 1), for varying minimum roughness wavelength λ_{min} . The preferred values of *C* (shown by full circles) are identified by comparing the moment rate release of the rough fault model to the reference model (b) Tested and preferred values of intercept parameter *b* (Eq. S1) for varying number of elements *n* per λ_{min} . The dashed line corresponds to the fitted function of *b*.



207 Figure S2. Comparison of the relative prestress ratio R for the planar reference model 208 and the rough fault broadband model with and without roughness drag correction. We 209 define the relative prestress ratio R following Aochi & Madariaga (2003) as the ratio of the potential stress drop $\Delta \tau$ to the full breakdown strength drop $\Delta \tau_{\rm b}$, $R = \Delta \tau / \Delta \tau_{\rm b} = (\tau_0 + \tau_0)$ 210 - $\mu_d \sigma_n / ((\mu_s - \mu_d) \sigma_n)$. R=1 indicates a critically stressed fault. (a) R for the reference 211 model, the broadband model (b) without and (e) with τ_{drag}^{3D} ; (c) difference of the R-212 213 parameter between (a) and (b), (f) difference of the R-parameter between (a) and (e); and 214 (d) histogram of (c) and (f).



215 Figure S3. (a) Moment rate release, (b) moment rate spectrum, (c) the 2nd time

- 216 derivative of the moment rate release. Grey curves correspond to the reference model.
- 217 Red and blue curves correspond to the broadband rough fault model with and without D_c
- 218 perturbations, respectively.

a)	N-S	E-W	z	b)	N-S	E-W	z
AMT -			39.79	AMT _	353.79	595.08	238.1
NRC				NRC -	296.16	<u>289.04</u> 43.52	103.73 19.66
MNF			4.84	MNF _	30.43	20.37	40.38
TRL -				TRL -			
ANT				ANT -	14.27		
PZI1 -			4.53	PZI1 -	25.77	31.43	<u>13.32</u>
LSS -	- may man -			LSS –	2.99		
SPD -			7.95	SPD -	<u>84.97</u>	16.4	43.17
FOS				FOS _	57.22 		
ASP	- my hormon -			ASP _	40.88	2.97	18.52
TRE -			7.88	TRE _		52.13 	
SPM				SPM -	31.06		
FEMA			<u>14.83</u>	FEMA -	71.02	211.25	58.25
TERO -			- Muntil	TERO -	36.48	30.1	16.17
RM33 -				RM33 -		72.21 24.37	19.65
RQT	-		15.68	RQT	-	<u>195.9</u> 	95.24
SNO -			4.84	SNO -	29.29 15.68	51.44 	23.77
csc -		- Maphyphin -		csc -	<u>64.85</u> 4.38	89.48 11.12	33.4
MSC -				MSC -	<u>54.89</u> 82.93	86.62 27.26	<u>50.66</u>
CLF			10.22	CLF -	11.91		
ł	35 s			ł	35 s		-* V
	-				-		

219 Figure S4. Comparison of observed (black) and simulated (grey) broadband three-220 component (NS, EW, and Z) of (a) ground-velocity (in cm/s) and (b) acceleration (in 221 cm/s2) waveforms at 20 selected stations (Figure 1). Synthetics are from the reference 222 dynamic rupture scenario based on a planar fault. Both observed and synthetic velocity 223 waveforms are scaled by the maximum value of the observed records at each station, 224 indicated on the right-hand side of each plot, while acceleration waveforms are scaled 225 component-wise. Observed and synthetic waveforms are band-pass filtered between 0.05 226 and 5 Hz.



227 Figure S5. Comparison of observed (black) and simulated (green) broadband threecomponent (NS, EW, and Z) of (a) ground-velocity (in cm/s) and (b) acceleration (in 228 229 cm/s2) waveforms at 20 selected stations (Figure 1). Synthetics are from the rough fault 230 dynamic rupture scenario without topography. Both observed and synthetic velocity 231 waveforms are scaled by the maximum value of the observed records at each station, 232 indicated on the right-hand side of each plot, while acceleration waveforms are scaled 233 component-wise. Observed and synthetic waveforms are band-pass filtered between 0.05 234 and 5 Hz.



235 Figure S6. Effect of rough fault and topography on the velocity and acceleration 236 waveforms and on the Fourier amplitude spectra. (top, middle) Comparison of synthetic ground-velocity (top rows) and acceleration (middle rows) waveforms from the 237 238 broadband rough fault model with topography (red), the broadband rough fault model 239 without topography (green), and the reference model (grey) compared with observations 240 (black). NS component at five selected stations (see Figure 1). All waveforms are scaled by their maximum values, indicated on the left-hand side of each plot. (bottom) 241 242 Smoothed (using the Konno & Ohmachi (1998) method) Fourier amplitude spectra (FAS) 243 of the velocity waveforms.



Figure S7. Same as Figure S4 for stations PZI, LSS, SPD, FOS, and ASP.



Figure S8. Same as Figure S4 for stations TRE, SPM, FEMA, TERO, and RM33.



Figure S9. Same as Figure S4 for stations RQT, SNO, CSC, MSC, and CLF.



Figure S10. Effect of rough fault and topography on the velocity and acceleration 247 248 waveforms and on the Fourier amplitude spectra. (top, middle) Comparison of synthetic 249 ground-velocity (top rows) and acceleration (middle rows) waveforms from the 250 broadband rough fault model with topography (red), the broadband rough fault model 251 without topography (green), and the reference model (grey) compared with observations 252 (black). EW component at five selected stations (see Figure 1). All waveforms are scaled 253 by their maximum values, indicated on the left-hand side of each plot. (bottom) 254 Smoothed (using the Konno & Ohmachi (1998) method) Fourier amplitude spectra (FAS) 255 of the velocity waveforms.



Figure S11. Same as Figure S8 for stations TRE, SPM, FEMA, TERO, and RM33.



Figure S12. Same as Figure S8 for stations RQT, SNO, CSC, MSC, and CLF.



Figure S13. Comparison of observed (black) and simulated ground velocities for a planar fault model with topography with (cyan) and without (orange) viscoelastic attenuation at 20 selected stations (Figure 1). All observed and synthetic velocity waveforms are scaled by the maximum value of the observed records at each station, indicated in cm/s on the right-hand side of each plot. Observed and synthetic waveforms are band-pass filtered between 0.05 and 5 Hz.



Figure S14. Distribution of average Goodness-of-Fit (GOF) for each component (color coded) for (a) the reference model, (b) the broadband rough fault model without topography, and (c) the broadband rough fault model with topography.



Figure S15. Average Goodness-of-Fit (GOF) for each station and component, for the reference (grey circles), the broadband rough fault model without topography (green

triangles), and the broadband rough fault model with topography (red squares).



Figure S16. The model bias (dashed lines) and standard deviation (solid lines) of residuals between observed SA values in the 0.5-10 s period range, averaged over 20 stations and synthetics of (a) reference model, (b) broadband rough fault model without topography, and (c) broadband rough fault model with topography. The bold black line is the median value, the filled area is the 90% confidence interval and the pale filled area is the one-sigma range.