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Constraining families of dynamic models using geological, geodetic and strong ground motion data: the Mw 6.5, October 30th, 2016, Norcia earthquake, Italy

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Abstract

The 2016 Central Italy earthquake sequence is characterized by remarkable rupture complexity, including highly heterogeneous slip across multiple faults in an extensional tectonic regime. The dense coverage and high quality of geodetic and seismic data allow to image intriguing details of the rupture kinematics of the largest earthquake of the sequence, the M_w 6.5 October 30th, 2016 Norcia earthquake, such as an energetically weak nucleation phase. Several kinematic models suggest multiple fault planes rupturing simultaneously, however, the mechanical viability of such models is not guaranteed.

Using 3D dynamic rupture and seismic wave propagation simulations accounting for two fault planes, we constrain “families” of spontaneous dynamic models informed by a high-resolution kinematic rupture model of the earthquake. These families differ in their parameterization of initial heterogeneous shear stress and strength in the framework of linear slip weakening friction.

First, we dynamically validate the kinematically inferred two-fault geometry and rake inferences with models based on only depth-dependent stress and constant friction coefficients. Then, more complex models with spatially heterogeneous dynamic parameters

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allow us to retrieve slip distributions similar to the target kinematic model and yield good agreement with seismic and geodetic observations. We discuss the consistency of the assumed constant or heterogeneous static and dynamic friction coefficients with mechanical properties of rocks at 3-10 km depth characterizing the Italian Central Apennines and their local geological and lithological implications. We suggest that suites of well-fitting dynamic rupture models belonging to the same family generally exist and can be derived by exploiting the trade-offs between dynamic parameters. Our approach will be applicable to validate the viability of kinematic models and classify spontaneous dynamic rupture scenarios that match seismic and geodetic observations at the same time as geological constraints.

Keywords:

earthquake source, data-integrated dynamic modeling, frictional heterogeneity, dynamic rupture, high-performance computing

1. Introduction

Kinematic modeling is a standard tool to image the slip behavior of finite faults during earthquakes of moderate-to-large magnitude. Kinematic models ([Haskell, 1964](#)) prescribe the spatio-temporal evolution of slip on a fault as a result of solving data-driven inverse problems. Automated procedures deriving kinematic models within few hours after significant events are an established part of rapid earthquake response information. Refined kinematic models are often emerging during the months and years after an event using seismic and geodetic data and more advanced numerical methods to closely fit observations with a large number of free parameters (e.g., [Wang et al., 2020](#)). Therefore, most significant earthquakes are characterized by several kinematic models that describe the complexity of the seismic process in terms of slip distribution, activated fault planes, fault geometry, and rupture time evolution. As of recently, kinematic modelers aim to take uncertainties into account, using, for example, a Bayesian approach ([Ragon et al., 2018](#), and references therein) to mitigate errors and assumptions in the forward modeling,

15 in the adopted Greens' function (Yagi and Fukahata, 2011), in data coverage, and in data
16 resolution.

17 Despite recent advances, kinematic models are characterized by an inherent non-
18 uniqueness of the problem (strong trade-offs among kinematic parameters) in addition
19 to aforementioned significant uncertainties and the often required predefinition of fault
20 geometries (with notable exceptions, e.g., Ragon et al., 2018; Shimizu et al., 2020).

21 The scaling and distribution of dynamic source properties can be evaluated from kine-
22 matic source models as a solution of the elastodynamic equation when the rupture history
23 is prescribed a-priori on a fault plane. This approach permits retrieving distributions of
24 the corresponding dynamic parameters without using any constitutive law and without as-
25 sessing if the models would propagate spontaneously (e.g., Tinti et al., 2005; Causse et al.,
26 2014). Fully dynamic modeling of earthquakes provides a physics-based understanding
27 of how earthquakes start, propagate, and stop. Earthquake dynamic rupture simula-
28 tions couple the non-linear interaction of fault yielding and sliding behavior to seismic
29 wave propagation (Harris et al., 2018). Using modern numerical methods and computing
30 infrastructure allows for realistic 3D dynamic rupture scenarios of complex, multi-fault
31 earthquakes (Ando and Kaneko, 2018; Wollherr et al., 2019). Initial conditions, such
32 as geometry, frictional fault strength, tectonic stress state and regional lithology, con-
33 trol rupture propagation style (e.g., pulse vs. crack-like dynamics and sub-Rayleigh vs.
34 super-shear speeds), stress transfers (dynamic triggering, branching), and earthquake ar-
35 rest (e.g., Gabriel et al., 2012; Bai and Ampuero, 2017; Lambert et al., 2021).

36 Since it is challenging to constrain fault stresses and strengths from direct observation,
37 it is common to prescribe fault normal and shear stress as constant or linearly increasing
38 with depth (e.g., Galvez et al., 2014). While matching strong motion records with dynamic
39 rupture simulations can be formulated as an inverse problem with stress and friction as
40 model parameters (e.g., Gallovič et al., 2019), to date, only simplified dynamic rupture
41 simulations are computationally tractable for dynamic source inversion.

42 Few dynamic rupture models have been proposed of moderate size normal faulting
43 events (Gallovič et al., 2019; Aochi and Twardzik, 2020). Surface breaching reverse and
44 normal faulting dynamic models are challenged by free-surface induced normal stress,
45 strength, loss of ground motion symmetry, trapped waves in the hanging wall, and other
46 dynamic and quasi-static effects (e.g., Oglesby et al., 1998; Aochi, 2018).

47 Dynamic models can be affected by parameter trade-offs (Guatteri and Spudich, 2000;
48 Schmedes et al., 2010) and the choice of constitutive law (Dieterich, 1979; Ohnaka et al.,
49 1987). Nevertheless, by reconciling findings from experiments (Di Toro et al., 2011; Col-
50 lettini et al., 2019) and increasingly dense observations, dynamic models can bridge scales
51 and geophysical disciplines to provide insight into the mechanic viability of competing
52 hypothesis for a specific event (Ulrich et al., 2019) or fault system (Murphy et al., 2018;
53 Harris et al., 2021).

54 The 2016 Mw 6.5 Norcia (Italy) earthquake is an example of a normal faulting earth-
55 quake with a moderate magnitude involving a complex set of intersecting faults. Several
56 models proposed for this event (Chiaraluze et al., 2017; Cheloni et al., 2017; Pizzi et al.,
57 2017; Scognamiglio et al., 2018; Walters et al., 2018; Bonini et al., 2019) generally agree
58 on the location of the main slip release. However, most recent models require two or more
59 connected faults to match all observations available from diverse dataset. The inferred
60 multi-fault geometries are not conflicting; a consensus (Scognamiglio et al., 2018; Michele
61 et al., 2020; Bonini et al., 2019; Walters et al., 2018) is emerging for a multiple-fault
62 model composed of a main normal fault parallel to the Apennines backbone, confined to
63 the southeast by an oblique fault, unfavorably oriented with respect to the current tec-
64 tonic regime (Mariucci and Montone, 2020). The proposed composite models suggest that
65 these fault planes slipped simultaneously, posing questions on the dynamic plausibility of
66 co-seismic fault interaction.

67 Here, we focus on the complex kinematic model proposed by Scognamiglio et al. (2018),
68 “S18” hereinafter. We develop a systematic approach to constrain spontaneous dynamic

69 models based on a given kinematic model, allowing us to evaluate its dynamic consistency.
70 Such data-driven physics-based models can complement rapid earthquake response and
71 further the fundamental understanding of complex earthquake rupture processes. Specif-
72 ically, we design and analyze “families” of complex multi-fault dynamic models, each
73 recovering main kinematic characteristics but varying in terms of their initial dynamic
74 parameters which determine frictional strength and stress drop.

75 For each of these families, we conduct dynamic rupture scenarios of the Norcia earth-
76 quake. We detail dynamic models having the same kinematic features as the target “S18”
77 model and we validate them with seismic and geodetic observations, overall slip distri-
78 bution, rake direction, and moment magnitude. We derive a parametrization leading
79 to friction coefficients (static and dynamic) consistent with the mechanical properties of
80 rocks in the Italian Central Apennines. Our approach helps to overcome the difficulties in
81 assigning initial modeling conditions for dynamic rupture models in absolute terms and
82 to discuss the lithological meaning of the derived friction parameters.

83 **2. The M_w 6.5 October 30th 2016 Norcia earthquake**

84 The Amatrice-Visso-Norcia (AVN) seismic sequence ([Chiaraluce et al., 2017](#); [Michele](#)
85 [et al., 2020](#); [Scognamiglio et al., 2018](#)) started on August 24th, 2016, with the Mw 6.0
86 Amatrice earthquake ([Tinti et al., 2016](#)). The largest event, which occurred on October
87 30th, struck the region close to Norcia village with magnitude M_w 6.5 and was preceded
88 only four days earlier, on October 26th, by the Mw 5.9 Visso earthquake. For the Mw
89 6.0 Amatrice event, simplified dynamic rupture inferences from strong ground motion
90 data ([Galović et al., 2019](#); [Aochi and Twardzik, 2020](#)) reveal complex dynamics (e.g.,
91 two asperities and a slow nucleation phase) and imply that rupture arrested south of the
92 secondary fault activated during the Norcia earthquake.

93 Similarly, the Norcia earthquake exhibited a large degree of complexity. Our starting
94 point here is the kinematic model “S18” that involves, in addition to the main normal fault

95 parallel to the Monte Vettore-Monte Bove fault systems, a second fault. This secondary
96 fault is ascribed to the inherited Olevano-Antrdoco-Sibillini Thrust and dislocates as a
97 NNE trending normal fault with a significant strike-slip component (Figure 1). The “S18”
98 model is obtained from jointly inverting strong motion and GPS data, and is validated
99 using InSAR data (Scognamiglio et al., 2018) and relocated aftershocks (Michele et al.,
100 2020).

101 The main kinematic characteristics of the “S18” model are the following: i) both faults
102 dislocate almost simultaneously, reaching a maximum slip of 3 m; ii) the location of the
103 high-slip patches is about 5 km shallower than the hypocenter while less than 20 cm
104 of slip is inferred in the nucleation region; iii) the secondary fault is characterized by
105 a predominantly left-lateral strike-slip mechanism within its largest slip patch, but also
106 features local rake variations; iv) the southern part of the main fault, located behind
107 the secondary fault and activated during the first event of the AVN sequence (the Mw
108 6.0 Amatrice earthquake), is partially reactivated during the Norcia event, with a locally
109 significant amount of slip (≈ 1 m).

110 3. Model setup

111 We use the open-source software package *SeisSol* (www.seissol.org) to model sponta-
112 neous dynamic earthquake rupture across intersecting faults and seismic wave propagation
113 with high-order accuracy in space and time (Figure 2, for details see Appendix A).

114 Modeling complex fault interaction during dynamic rupture propagation is challenging,
115 specifically across fault junctions and interpenetrating fault surfaces (e.g., Douilly et al.,
116 2020). *SeisSol*, which is based on the Arbitrary high-order Derivatives Discontinuous
117 Galerkin method (Dumbser and Käser, 2006), naturally allows for discontinuities and
118 fault branching geometries (Pelties et al., 2014).

119 3.1. Constitutive law

120 We adopt a simple constitutive relationship (Figure 3) to focus on the effects of het-
121 erogeneities in fault strength and stress. The linear slip-weakening (LSW) friction law
122 (Barenblatt, 1959) is a simple and widely used constitutive equation derived from theoret-
123 ical and numerical models (Andrews, 1976) of shear crack propagation from a macroscopic
124 perspective (Cocco and Tinti, 2008). This constitutive relation is completely character-
125 ized by the yield strength $\tau_y = \mu_s \sigma_n$, the dynamic frictional resistance $\tau_f = \mu_d \sigma_n$, and the
126 critical slip distance D_c , where μ_s and μ_d are the static and dynamic friction coefficients,
127 respectively, and σ_n is the effective normal stress. The fault begins to rupture when shear
128 stress locally exceeds τ_y and frictional fault strength decreases linearly from a static to a
129 dynamic level over a critical slip distance D_c . For a slip greater than D_c , fault strength
130 remains constant equal to τ_f (i.e., no healing). The distribution across the fault plane of
131 the strength excess $(\tau_y - \tau_0)$, with the initial shear stress τ_0 , and the dynamic stress drop
132 $\Delta\sigma = \tau_0 - \tau_f$, influences the ratio of strain energy and fracture energy, and determines
133 local acceleration or deceleration of the rupture front.

134 Inference of the magnitude and direction of initial stresses is only possible from kine-
135 matic slip models in which the temporal rake rotation is well defined (Spudich et al., 1998)
136 otherwise, additional assumptions are required. Here we assume that the initial traction
137 is co-linear with the accumulated slip in kinematic models to ensure physical plausibility
138 (Tinti et al., 2005).

139 Spontaneous dynamic rupture model is fully defined by the spatial distributions of
140 initial on-fault shear stress, normal stress, static and dynamic friction coefficients, and D_c
141 in addition to the prescribed fault geometry and subsurface structural model. Using the
142 LSW law permits to potentially relate co-seismic fault-constitutive properties directly to
143 observations, e.g., associating friction coefficients of different rocks with inferred values
144 from laboratory experiments. However, scale-invariances and trade-offs between LSW
145 dynamic parameters are well known (Tinti et al., 2009; Goto and Sawada, 2010): dynamic

146 rupture models based on various dynamic parameter choices can fit seismological data
147 equally well (Guatteri and Spudich, 2000). Dynamic parameters cannot be measured in-
148 situ and often lack physical constraints rendering it difficult to determine them prior to
149 (or after) an earthquake. This yields a wide and high-dimensional parameter space which
150 is challenging to fully explore and constrain in a data-driven manner.

151 Therefore, assumptions have to be made when pre-assigning frictional parameters as
152 well as the absolute amplitudes of initial stresses, which both may be heterogeneously
153 distributed acting across the fault planes (e.g., Ripperger et al., 2007; Causse et al.,
154 2014). This motivates our classification of “families” of dynamic models (section 3.4).

155 3.2. Fault geometry

156 We use a two-planar-fault geometry (Figures 1 and 2) derived from Scognamiglio et al.
157 (2018). It consists of a main fault branch N155° trending along the Apennines (hereinafter
158 F155), and a second fault plane striking N210° oblique to the Apennines (hereinafter
159 F210). The main fault geometry aligns well with the SAR interferograms, the TDMT
160 moment tensor solution, and the observed surface rupture (Scognamiglio et al., 2018).
161 The secondary fault plane geometry is supported by geodetic observations, aftershocks’
162 distribution, the inferred non-double-couple component of the mainshock moment tensor,
163 and by moderate earthquakes of NE-SW trending focal mechanisms in the main fault
164 hanging wall (Michele et al., 2020). The dynamic activation of F210, which is shallowly
165 dipping is a major challenge for this model.

166 3.3. Weak dynamic rupture nucleation

167 We use the INGV inferred hypocenter located at 42.84°N, 13.11°E at a depth of
168 9.52 km (<http://cnt.rm.ingv.it/event/8863681>) to prescribe the onset of rupture in all
169 our models. The nucleation region is located on fault F155 and intersects the bottom left
170 corner of the F210 fault (see Figure 1). For the Norcia earthquake, similar to the Amatrice
171 event, only small amounts of slip have been inferred in the hypocentral regions implying

172 a transient, weak nucleation process (Tinti et al., 2016; Gallovič et al., 2019). Weak
173 nucleation in dynamic rupture models is controlled by spatial heterogeneities and the
174 local closeness to failure of the hypocentral region. We find that locally over-stressing the
175 fault (i.e., assuming the initial stress just above yield stress as, e.g. Palgunadi et al., 2020)
176 tends to create artificially large fault slip in the hypocentral area and unrealistic strong
177 pulses in the synthetic seismograms. Instead, we gradually reduce the yield strength in a
178 circular area centered at the hypocenter expanding at time-decreasing speed (Harris et al.,
179 2018) which allows a smooth transition to fully spontaneous dynamic rupture propagation.
180 In conjunction with assuming locally initial shear stresses very close to frictional strength,
181 fault slip in the nucleation area remains limited matching observations.

182 For the models proposed, we choose a nucleation initial forcing speed of 2.8 km/s
183 ($0.7V_s$) and a nucleation radius of 3-6 km, which is of similar size to inferences for the
184 Amatrice event (e.g., Pizzi et al., 2017). To dynamically capture the low energy release
185 and small slip during the weak nucleation phase requires to carefully balance the delicate
186 rupture initiation with spontaneous rupture across both activated fault planes.

187 *3.4. Families of initial dynamic parameters*

188 Dynamic models can be initialized assuming homogeneous or heterogeneous spatial
189 distributions of one or more dynamic parameters governing frictional fault-weakening
190 behavior and initial stresses on the fault plane (e.g., Savran and Olsen, 2020).

191 To limit the complexity of the dynamic parameterization, it is common to attribute all
192 heterogeneities either only to the initial shear stress distribution or to the yield strength
193 (e.g., Gallovič et al., 2019) while considering the other dynamic parameters constant or
194 homogeneously depth-dependent. In fully elastic dynamic models, the radiated waves are
195 only sensitive to the dynamic stress drop but not to the absolute initial stress.

196 The main characteristics of the rocks that belong to a specific seismic zone can add
197 lithology-controlled constraints (e.g., Harris et al., 2021). Laboratory experiments on
198 friction coefficients conducted on different types of rocks (Di Toro et al., 2011; Scuderi

199 [et al., 2013](#); [De Paola et al., 2015](#)) provide possible ranges of frictional parameters for
200 weak and strong faults ([Collettini et al., 2019](#)). Taking laboratory results into account
201 can limit the parameter space to be explored in dynamic models.

202 Based on these considerations, we identify “families” of dynamic models, consistent
203 with field and laboratory observations but differing in their parameterization of hetero-
204 geneous fault stress and strength in the framework of a LSW friction law (Figure 3):

- 205 • Family (Hom) are models based on uniformly depth-dependent stress and strength
206 conditions with constant static (μ_s) and dynamic (μ_d) friction coefficient.
- 207 • Family (A), the “family of heterogeneous stress”, includes all models with constant
208 static and dynamic friction coefficient, linearly depth-dependent initial normal stress
209 but variable initial shear stress τ_0 .
- 210 • Family (B), the “family of heterogeneous strength and stress”, includes all models
211 with constant dynamic friction coefficient, depth-dependent initial normal stress but
212 heterogeneous static friction and initial shear stress.
- 213 • Family (C), the “family of heterogeneous dynamic friction”, includes all models with
214 uniform depth-dependent static friction and initial shear stress but heterogeneous
215 dynamic friction. Family (C) ensures also depth-dependent strength excess.

216 A fully heterogeneous Family (D), the “family of heterogeneous strength, stress and fric-
217 tion”, is here omitted given the high risk of severe data over-fitting.

218 We assume that the effective normal stress increases linearly with depth according to
219 a fixed gradient based on an assumed fluid pressure ratio λ (defined as the fluid pressure
220 over the lithostatic stress, e.g. [Ulrich et al., 2019](#)). The adopted near-hydrostatic fluid
221 pressure ratio λ is 0.4, corresponding to an average gradient around 15 MPa/km. The
222 associated stress and strength parameters (τ_0 , τ_y , and τ_f) also vary linearly as a function
223 of depth (see Figure 3).

224 Family (A) represents the simplest heterogeneous model: static and dynamic friction
225 coefficients are homogeneous while the initial shear stress is heterogeneous. In this group
226 of models, regions with kinematically constrained low fault slip have very high strength
227 excess and small dynamic stress drop. Such areas, if large enough, do not favor sus-
228 tained spontaneous rupture since they require more energy than available to overcome
229 the strength excess. This family potentially allows using laboratory-consistent values for
230 both static and dynamic frictions ($\mu_d \approx 0.2$ and $\mu_s \approx 0.6$, e.g., [Collettini et al., 2019](#)), but
231 is not suited to all kinematic models. Specifically, the “S18” model cannot be reproduced
232 using models belonging to Family (A) due to its low fault slip in the nucleation region:
233 the resulting high strength excess prevents spontaneous rupture propagation.

234 Moreover, models of this family have a very small strength excess in regions of kine-
235 matically inferred high fault slip, such as at the center of the main slip patches (illustrated
236 by the local closeness of τ_0 to τ_y in [Figure 3](#)). Thus, Family (A) dynamic models are also
237 prone to a-causal ruptures, that is, failure may happen at many patches instantaneously.
238 Thus, we refrain from further analysis of Family (A) in the remainder of this paper.

239 Family (B) represents a group of heterogeneous models which are frequently proposed
240 for dynamic source inversions ([Gallovic̆ et al., 2019](#)). Heterogeneities are attributed to the
241 initial shear stress and yield strength, assuming a constant dynamic friction value. The
242 heterogeneity of stress drop is then completely associated with the initial shear stress.
243 Stress drop corresponding to a prescribed distribution of slip can be retrieved in different
244 ways: for example, by relating stress drop and slip in the wavenumber domain (originally
245 proposed by [Andrews \(1980\)](#) and updated by [Ripperger et al. \(2007\)](#)) or by solving the
246 elastodynamic equation using the entire slip-time history at each point of the fault (e.g.,
247 [Tinti et al., 2005](#); [Causse et al., 2014](#)).

248 In this work, we explore two simple approaches. First, we estimate stress drop by
249 assuming direct proportionality with fault slip. In the second approach, we infer the
250 stress drop distribution from the stress change, by imposing the “S18” slip distribution

251 everywhere on the fault (using an arbitrary smooth-step slip-rate function during 1 s),
252 and measuring the final shear stress distribution (simplifying the approach of [Tinti et al.,](#)
253 [2005](#); [Causse et al., 2014](#)).

254 We constrain Family (B)’s yield strength, by assuming a strength excess radially in-
255 creasing from the hypocenter, with a minimum value of 0.1 MPa at the hypocenter. In
256 addition to the smooth nucleation procedure (Sec.3.3), this parametrization facilitates
257 nucleation and yields realistic rupture growth.

258 Family (C) includes models with constant static friction, linearly depth-dependent
259 initial shear stress, and heterogeneous dynamic friction. The resulting yield strength τ_y
260 is only depth-dependent because the effective normal stress is depth-dependent. Hetero-
261 geneities in the dynamic friction coefficient stem from the target stress drop distribution,
262 which is retrieved following two different procedures, as in Family (B). While Family (B)
263 has variable μ_s and constant μ_d , Family (C) has variable μ_d and constant μ_s .

264 We adopt values typical of many lithologies (i.e. 0.5-0.6 for μ_s and 0.1-0.2 for μ_d) for
265 the constant friction in Family (B) and (C). In distinction, variable μ_s and μ_d are obtained
266 respectively from the assumed heterogeneous stress drop, derived from the “S18” fault
267 slip. We further validate the dynamic models belonging to these two families by assuring
268 that the variable friction values are compatible with the expected rocks in the modeled
269 region and depths. Introducing Family (C), which is often disregarded among kinemati-
270 cally constrained dynamic models, is motivated by the fact that most rocks favoring the
271 occurrence of seismic events may share similar μ_s around 0.5-0.6 ([Byerlee, 1978](#)).

272 4. Results

273 The complex spatio-temporal evolution of the AVN sequence, and in particular the
274 distribution and location of the main slip patches of the three main events may suggest
275 strongly heterogeneous initial stress and/or frictional strength and weakening conditions.
276 We first analyze simple models of Family (Hom), based on homogeneous friction, only

277 depth-dependent stress assumptions, to understand which overall conditions favor a spon-
278 taneous multi-fault rupture across the assumed fault geometry. Next, we introduce more
279 realistic heterogeneous dynamic models of Family (B) and (C).

280 4.1. Homogeneous initial conditions

281 Figure 4 (top panels) shows the on-fault distribution of the most important initial
282 conditions and resulting dynamic parameters for three illustrative dynamic scenarios of
283 Family (Hom). We assume constant static and dynamic friction coefficients of 0.6 and 0.2,
284 respectively. Furthermore, we set the initial shear stress τ_0 as 65% of the yield strength τ_y ,
285 which allows dynamic rupture to spontaneously propagate while limiting rupture speed
286 to sub-Rayleigh velocities for most of the fault area. For simplicity, we use on each fault
287 plane a constant shear stress orientation, informed by the average faulting mechanism in
288 the “S18” model: pure normal faulting for the F155 (-90°) and almost pure left-lateral
289 strike faulting (-10°) for the F210 fault.

290 The magnitude of the initial shear stress τ_0 varies on the two fault planes only as
291 a function of depth (Figure 4) following the normal stress gradient. Figure 4 (second
292 row) shows the depth-variations of τ_y , τ_0 and τ_f as cross-sections. Small offsets are
293 the result of the layered density profile. The nucleation is imposed inside a sphere of
294 radius 3 km. Fixing all other parameters, we here explore how dynamic rupture viability
295 on the main and secondary fault is depending on the choice of D_c . We confirm that
296 smaller D_c , i.e. smaller fracture energy with other dynamic parameters kept unchanged,
297 favors dynamic rupture propagation while larger D_c inhibits it. We also find that fault
298 interaction (branching, dynamic triggering, shadowing, and co-seismic static slip effects,
299 e.g. [Kyriakopoulos et al., 2019](#)) is highly sensitive to choices of D_c .

300 For the assumed initial conditions and fault geometries, we find that values of $D_c \approx 1$ -
301 2 m on the F155 main fault allow rupture propagation at sub-Rayleigh velocity (< 3
302 km/s) across most of the slipping area. However, due to the linear depth-dependence of
303 the initial stress, the rupture velocity tends to reach super-shear speeds at shallow depths

304 (e.g., [Tang et al., 2021](#)). At the same time, lower values of D_c on F210 (<0.8 m), are
305 needed to allow dynamic rupture propagation there.

306 In Figure 4 (top row) we show three D_c combinations to illustrate the model sensitiv-
307 ity to this parameter. In the bottom panel, we compare snapshots of slip distributions
308 after a rupture time of 6.75 s for these three models. Their elliptical slip distributions
309 (generated by crack-like dynamics, e.g., [Gabriel et al., 2012](#)) are aided by LSW friction
310 and homogeneous initial conditions. These models are characterized by high slip (> 10 m)
311 in the hypocentral region and by magnitudes much larger than M_w 6.5 (between M_w 7.14
312 and M_w 7.35, see the right-most panel in the second row of Figure 4).

313 Assuming $D_c = 1.2$ m and $D_c = 1.0$ m for F155 and F210, respectively, rupture is not
314 simultaneously propagating along both faults (see snapshot at $t=6.75$ s in panel a) but
315 breaking only the main fault including the area beyond the fault intersection. At a later
316 simulation time (> 8 s, not shown in the figure) slip is observed also on F210, which is
317 dynamically initiated by reflections at the free surface and at the interfaces of the layered
318 velocity structure. Assuming $D_c = 1.2$ m and $D_c = 0.8$ m (panel b) for F155 and F210,
319 respectively, both faults rupture simultaneously. Interestingly, rupture of F155 behind the
320 intersection is initially prevented due to stress shadowing (e.g., [Bhat et al., 2007](#)) from
321 the F210 rupture. Finally, assuming $D_c = 1.8$ m and $D_c = 0.8$ m for F155 and F210
322 (panel c), respectively, shows again simultaneous rupture on both fault planes. Initially,
323 this model features a slower rupture (the rupture front is closer to the hypocenter at 6.75
324 s compared with panel b). Again, rupture propagation beyond the intersection with F210
325 is hindered. In the models of panels b and c, the rupture is able to propagate beyond the
326 fault intersection with a delay of several seconds, which makes this secondary propagation
327 more akin to a triggered event rather than a slow rupture.

328 We find that $D_c^{F210} > 0.8$ m prevents dynamic rupture on F210 (for the here assumed
329 stress conditions and nucleation). As an additional constraint, if dynamic rupture on F210
330 is prevented, F155 can host spontaneous rupture propagation only if $D_c^{F155} < 1.2$ m.

331 Therefore, high values of D_c on the main fault need to be combined with low values
332 of D_c on the secondary fault to allow rupture across both fault planes in the dynamic
333 rupture Family (Hom). Assuming pure normal faulting for both faults results in even less
334 favorable conditions for sustained rupture on F210 and very small D_c values are required
335 to dislocate both fault planes (models not presented).

336 The presented models have the same ratio of initial shear stress τ_0 over yield strength
337 τ_y . Exploring alternative ratios, as well as different ratios on each fault, will likely influence
338 the critical D_c values that allow rupture on one or both faults. A full analysis of this
339 variability, as well as variations in nucleation, is possible but beyond the scope of this
340 study.

341 Fracture energy, defined as $G_c = 1/2(\tau_y - \tau_f) D_c$ (e.g., [Palmer and Rice, 1973](#)), in-
342 creases with depth in Family (Hom) and varies linearly with D_c between models. The
343 average fracture energy in the examples is ≈ 20.6 - 29.3 MJ/m² (see right-most panel in the
344 second row of [Figure 4](#)) which is comparable to estimates inferred for past earthquakes of
345 similar magnitude ([Viesca and Garagash, 2015](#)). We note that fracture energy on both
346 fault planes is roughly equivalent for both models in panels a and b, despite their distinct
347 rupture dynamics.

348 Using simple forward dynamic rupture models of Family (Hom) we show that a multi-
349 fault rupture is plausible. Assuming homogeneous, depth-dependent stress and strength
350 conditions can lead to left-lateral strike-slip faulting on the secondary fault (F210) and
351 normal faulting on the main fault (F155). The synthetic waveforms resulting from dy-
352 namic rupture models of Family (Hom) are very different from observations. While we do
353 not systematically explore the parameter space of all possible constant values of μ_s , μ_d ,
354 and D_c , this nevertheless suggests that the real dynamic initial conditions may have been
355 strongly heterogeneous. We next explore the space of the dynamic parameters with het-
356 erogeneous stress and/or strength conditions to propose dynamic models that reproduce
357 the main features of the “S18” model.

358 *4.2. Heterogeneous initial conditions*

359 We here investigate models of Families (B) and (C) having heterogeneous stress and
360 strength as defined in Section 3.4. We identify plausible rupture models, representative
361 of their respective family. These models are consistent with the “S18” inverted kinematic
362 characteristics and with observations. We do not claim that these models are the dynamic
363 models that best fit the data, due to trade-offs between the dynamic parameters. Instead,
364 we suggest that suites of well-fitting models belonging to the same family exist and can
365 be derived by exploiting the trade-offs between their dynamic parameters. We first show
366 models that assume a direct proportionality between fault slip and stress drop (Sec. 4.2.1
367 and, secondly, models in which the stress drop is kinematically inferred as the stress
368 change associated with the “S18” model (Sec. 4.2.2).

369 *4.2.1. Stress drop proportional to fault slip*

370 In Figure 5 we show the dynamic parameter distributions of two representative models
371 belonging to Family (B) and (C), respectively. Family (B) (panel a) has heterogeneous
372 distributions of initial shear stress and yield strength. The latter is parameterized as a
373 heterogeneous distribution of μ_s in the range of [0.2, 0.7] while μ_d is kept constant at 0.2.
374 Family (C) (panel b) has heterogeneous distribution of dynamic friction μ_d with values
375 between 0.1 and 0.45 and constant $\mu_s = 0.5$.

376 In computing the stress drop for both families (B) and (C), we slightly adapt the “S18”
377 slip distribution at shallow depths (< 2 km) to prevent fault reactivation due to rupture-
378 free-surface interaction mediated by small normal stress. To further prevent near-surface
379 supershear rupture in the uppermost 2 km we use higher values of μ_s (0.7) in Family (B)
380 and we add frictional cohesion $c = 2$ MPa (e.g., [Blanpied et al., 1991](#)) to the yield strength
381 ($\tau_y = \mu_s \sigma_n + c$) in Family (C). The resulting range of the dynamic parameters τ_y , τ_f and
382 τ_0 for the representative models of the two families is very different (see Figure 5).

383 As we have seen in Sec. 4.1 the choice of D_c is fundamental. Yet, D_c is one of the
384 most difficult dynamic parameters to constrain ([Tinti et al., 2009](#)). We find in numerical

385 experiments conducted for both heterogeneous Families (B) and (C) that a constant D_c
 386 value on each fault plane does not permit to recover realistic rupture dynamics. In fact,
 387 imposing a smaller D_c ($D_c < 50$ cm) on both fault planes leads to supershear rupture
 388 velocities. On the other hand, imposing larger D_c values ($D_c > 50$ cm) tends to prevent
 389 the rupture from propagating spontaneously. These strong dynamic trade-offs are also due
 390 to the very small slip in and around the nucleation area (Galović et al., 2019). Thus, we
 391 here decide to assume D_c proportional to slip (Tinti et al., 2009; Brodsky et al., 2020)
 392 which is a common assumption to ensure spontaneous rupture propagation.

393 Based on few trial simulations, we set $D_c = 0.3S_{final}$ (S_{final} is the slip distribution
 394 of the “S18” model) in the shallow part of the fault (down to 4.5 km depth) where the
 395 main patch of slip is located. Below 4.5 km depth, we set $D_c = 0.1S_{final}$, which aids
 396 spontaneous rupture to migrate to the shallow region of larger fault slip (see Figure 5).
 397 The choice of D_c affects the width of the cohesive zone, which has to be numerically well
 398 resolved (Wollherr et al., 2019). We limit D_c to values larger than 0.02-0.06 m (depending
 399 on the family) which ensures that the median of the cohesive zone distribution remains
 400 numerically well resolved (Appendix A).

401 In Figure 6 we show snapshots of fault slip (top) and slip rate (bottom) for one
 402 model of Family (B). Rupture propagates simultaneously on both fault planes. Moreover,
 403 rupture is also able to propagate beyond F210. The interaction of the main rupture
 404 front with the free surface produces back-propagating rupture fronts (interface waves,
 405 Dunham, 2005) of small amplitudes. The nucleation area (Sec. 3.3) results in a weak
 406 nucleation, as desired. The slip distribution features a large patch of slip of up to 3 m
 407 located just above the hypocenter on F155 with a dominant normal component, as well
 408 as a smaller patch of slip with similar maximum amplitude on the F210 fault with a
 409 dominant strike-slip component. The final slip distribution resembles the “S18” model,
 410 but is less heterogeneous. This arises mainly from the assumed proportionality between
 411 slip and stress drop as will become apparent in comparison to models initialized with the

412 stress change computed from the “S18” model (see Sec. 4.2.2).

413 The rupture evolution of a representative dynamic rupture model belonging to Fam-
414 ily (C), characterized by heterogeneous dynamic friction, is shown in Figure 7 using fault
415 slip and slip rates snapshots. The final slip distribution is very similar to the presented
416 Family (B) model, despite the different dynamic conditions. Rupture speed, as well as the
417 peak slip velocity, are also similar in the main area of slip. The Family (C) model features
418 a slightly higher rupture velocity than the Family (B) model towards the northern end of
419 the main fault. Again, this model allows the rupture to propagate behind the secondary
420 fault.

421 The total inferred seismic moments are 1.05×10^{19} N m and 1.3×10^{19} N m for the rep-
422 resentative models of Family (B) and (C), respectively. These values agree with the seismic
423 moment inferred from kinematic inversion in (Scognamiglio et al., 2018) (0.88×10^{19} N m).
424 The average fracture energy computed accounting only for fault cells with slip larger than
425 20% of average slip is $0.7 MJ/m^2$ for Family (B) and $0.61 MJ/m^2$ for Family (C). These
426 averages are smaller than those obtained for models of Family (Hom) (Section 4.1) and
427 consistent with proposed scaling laws between fracture energy and seismic moment (Vi-
428 esca and Garagash, 2015; Tinti et al., 2005).

429 Figure 8 compares synthetic velocity waveforms, with selected observed data in the
430 near-source region. We obtain a surprisingly good fit in both amplitude and phase for
431 both families, given our synthetics are not resulting from a dynamic source inversion. We
432 underline that no static correction has been applied. Moreover, the synthetics of the two
433 families are very similar to each other. Synthetic waveforms at the CNE station, located
434 northwest of the main patch, have similar pulses and amplitudes to the recorded data in
435 both models but are slightly delayed indicating directivity effects not fully captured in
436 either scenario.

437 *4.2.2. Kinematically inferred stress change*

438 We now present models that belong to Family (B) and (C) in which the stress drop
439 distribution is initialized from the stress change kinematically computed from the “S18”
440 model. We call these models “stress change” models. The stress change models differ from
441 the previously presented models only in their (potential) stress drop distribution. Figure 9
442 shows the imposed heterogeneous distributions of μ_s and μ_d for two models belonging to
443 Families (B) and (C), respectively. Both friction parameters are distributed within the
444 same range (0.2-0.7) but more heterogeneous compared to the models of Section 4.2.1.
445 Note that the large values of μ_d (≈ 0.7) in Figure 9 are fictitious since they are located
446 in areas where rupture does not propagate. Both stress change models show a more
447 heterogeneous distribution also of all other dynamic and kinematic parameters, which
448 is reflected in the complex rupture history shown in Figures S4 and S5. The final slip
449 distribution is more similar to the original model “S18” than the models presented in
450 figures 6 and 7, while the rupture evolution is much more complex than the circular
451 propagation assumed in the kinematic model.

452 In Figure 9 we show the waveform fits for these models. Both stress change models
453 align well with observations. Synthetics of the two models are again similar to each other
454 although differences are more clearly noticeable than in the models shown previously. The
455 more pronounced variability between the models of the two families is expected because
456 they have different and complex slip rate histories.

457 *4.2.3. Geodetic validation*

458 Even if we here do not aim at identifying a best dynamic model for the Norcia earth-
459 quake, we validate all four exemplary heterogeneous dynamic rupture models also with
460 geodetic GPS and InSAR data. We compare in Figure 10 the synthetic deformation along
461 line of sight for the descending and ascending ALOS2 InSAR data and the synthetic co-
462 seismic displacements with GPS observations (Cheloni et al., 2017).

463 The target “S18” model, inverted from strong-motion and GPS data, offers, as ex-

464 pected, the best fit to the GPS data. The dynamic rupture models having the same
465 stress drop assumption yield similar geodetic fits. The “stress change” models, having
466 slip distributions very similar to the original “S18” model, offer the best fit for InSAR
467 data, and reproduce the GPS reasonably well in amplitude and direction, except for a
468 large observed displacement in the footwall region.

469 Models inferred by assuming stress drop proportional to slip (Section 4.2.1), show in
470 general the largest deformation values, still consistent with inversion results, but at worse
471 orientation. This is mainly due to their slip distributions, which reproduce the large-scale
472 features of the target model but not its shallow smaller-scale heterogeneities. Comparison
473 with both ascending and descending InSAR data yields similar conclusions (Figure 10).
474 Note that this dataset can only be discussed qualitatively, as the observed data contain
475 also the deformation produced by the Mw 5.9 Visso earthquake.

476 While our results suggest the existence of dynamic models within both heterogeneous
477 families able to support the dynamic viability of the “S18” kinematic model, model vali-
478 dation with seismological and geodetic data does not identify a preferred family of models.
479 Additional constraints are needed to assign heterogeneities to dynamic parameters, e.g.
480 using friction values consistent with rock properties in the area.

481 5. Discussion

482 We present several dynamic rupture models for the Norcia earthquake to assess if
483 the kinematic model “S18” proposed by [Scognamiglio et al. \(2018\)](#) is dynamically viable
484 (i.e. if the earthquake can propagate spontaneously on both faults). To this end, we
485 design families of dynamic parameters. Family (Hom), the simplest possible distribution
486 of dynamic parameters, allows us to dynamically validate the fault geometry and the
487 average rake values inferred in the “S18” model. Specifically, we find parameter sets
488 that allow for simultaneous spontaneous dynamic rupture of both fault planes (even if
489 the secondary fault is dynamically more challenging to activate). However, homogeneous

490 dynamic conditions lead to earthquake scenarios not agreeing well with observations.

491 The models of Family (B) and (C) with spatially heterogeneous dynamic parameters
492 permit to dynamically retrieve slip distributions similar to model “S18”, yielding a satis-
493 factory fit of the observed waveforms and geodetic observations. We suggest the existence
494 of suites of dynamic models in both families that are able to validate the target kinematic
495 model.

496 However, the dynamic conditions of Family (B) and (C) are very different. In Fam-
497 ily (B), we assume constant dynamic friction ($\mu_d = 0.2$) and heterogeneous static friction,
498 which varies between $\mu_s = 0.2$ and 0.7. In Family (C), we assume constant static friction
499 (in the showed model, we assume $\mu_s = 0.5$) while the dynamic friction is heterogeneous
500 and varies between $\mu_d = 0.1$ and 0.45.

501 Geological data and results from laboratory experiments provide strong evidence for
502 structural and frictional heterogeneities within crustal faults (Collettini et al., 2019). How-
503 ever, the different dynamic parameter assumptions made for Families (B) and (C) have
504 implications for the physical processes occurring on the fault plane during the coseismic
505 stage. In particular, the choice of reliable friction coefficients may be related to the rocks
506 where the event nucleates, propagates, and finally generates the large slip patches.

507 For the Norcia earthquake, the integration of seismic reflection profiles with seismo-
508 logical data shows that the mainshock nucleated within the Triassic Evaporites and prop-
509 agated through the overlaying carbonates (Porreca et al., 2018). The Triassic Evaporites
510 consist of anhydrites and dolostones and laboratory data on these fault rocks show static
511 friction in the range of 0.5-0.6 (Scuderi et al., 2013) with a reduction to 0.4 with increasing
512 temperatures. In addition, the main patch of slip seems to be located within carbonates
513 (Scognamiglio et al., 2018; Porreca et al., 2018), where the static friction is around the
514 Byerlee’s values (0.6) and dynamic friction at high slip rates can be as low as 0.2 (e.g.,
515 De Paola et al., 2015). Experiments conducted at high slip velocities ($> 1m/s$) (Di Toro
516 et al., 2011) show that dynamic friction of different rocks ranges between 0.1 and 0.4.

517 Static friction as low as 0.3-0.2 can be found only in clay-rich rocks (e.g., phyllosilicates).
518 However, friction experiments on carbonates-clay mixtures show that the increase of clay
519 content promotes a clear transition from velocity weakening to velocity strengthening be-
520 havior (Ruggieri et al., 2021). In consideration of these experimental values, the models
521 of Family (B) may be plausible when considering rocks rich in phyllosilicates. Such low
522 static friction values retrieved for Family (B) are located in and around the nucleation
523 zone. Since these conditions may lead to velocity strengthening, this area would be less
524 prone to nucleate (Ruggieri et al., 2021). Finding clay-rich rocks at depths similar to the
525 hypocentral depth is unlikely (Porreca et al., 2018). Since small slip in the nucleation
526 area is a specific earthquake characteristic, we may hypothesize that weak nucleation can
527 result from pre-seismic creep.

528 Following the results of Porreca et al. (2018) and laboratory values, it seems that
529 models belong to Family (C) are promising candidates to represent the friction values
530 of the seismogenic area in the Central Apennines. This family shows the lowest values
531 of dynamic friction (0.1) in the areas of highest slip rate, consistent with laboratory
532 experiments, while the highest dynamic friction values characterize areas of small slip.

533 **6. Conclusions**

534 We propose families of dynamic models for the M_w 6.5 October 30th, 2016 Norcia
535 earthquake that aim to reproduce the main characteristics of the “S18” kinematic model
536 and to assess its mechanical viability. We detail representative models of two families:
537 either with constant dynamic friction coefficient and heterogeneous initial stress and yield
538 strength or with constant static friction coefficient, homogeneous depth-dependent initial
539 stress, and heterogeneous dynamic friction coefficient.

540 In addition to the goodness of fit of seismic waveforms and geodetic deformation
541 (GPS and InSAR) and the ability to reproduce characteristics of the target kinematic
542 model (such as the slip distribution), we propose that geological constraints, e.g. ensuring

543 compatibility of the assumed friction values with experimental values from near-fault
544 rocks, can help to discriminate among plausible dynamic rupture scenarios.

545 Despite the limited resolution of seismological and geodetic data, we believe that
546 future efforts shall be directed towards a new generation of dynamic models of real events
547 including constraints from interdisciplinary geophysical observations. For example, using
548 models of Family (B) or (C), the static and dynamic friction parameters may be chosen
549 based on available geological and lithological constraints, while future high-resolution,
550 near-fault seismic and geodetic data can help to constrain fault characteristics, e.g. D_c ,
551 and relative initial shear loading, in-situ. Reducing the trade-offs among the dynamic
552 parameters by improving the resolution of the seismological data and the knowledge of
553 friction properties of fault rocks are definitely ingredients to combine.

554 The developed approach can be readily applied to various types of earthquakes us-
555 ing kinematic models to constrain dynamic rupture scenarios and enhance data-driven
556 approaches with physics-based implications.

557 *Data and resources*

558 SeisSol is openly available at <https://github.com/SeisSol/SeisSol>. We use commit
559 24b71e4b0b1501782f0369c068dfcc99f57d1bcb. All simulation input files and the jupyter
560 notebooks are accessible at <https://github.com/git-taufiq/NorciaMultiFault>.

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569 *CRedit* authorship contribution statement

570 Elisa Tinti: Conceptualization, Data curation, Formal analysis, Investigation, Writing
571 – original draft, Writing – review & editing. Emanuele Casarotti: Data curation, Formal
572 analysis, Investigation, Writing – original draft, Writing – review & editing. Thomas
573 Ulrich: Methodology, Data curation, Validation, Software, Writing – review & editing
574 Taufiqurrahman: Software, Writing – review & editing Duo Li: Methodology, Writing –
575 review & editing Alice-Agnes Gabriel: Conceptualization, Funding acquisition, Investi-
576 gation, Methodology, Supervision, Resources, Validation, Software, Writing – review &
577 editing. All authors approve on the submitted article.

578 **Appendix A Numerical method and computational mesh**

579 We use SeisSol, a powerful open-source software package ([https://github.com/SeisSol/](https://github.com/SeisSol/SeisSol)
580 [SeisSol](#)), to perform dynamic rupture simulations at the supercomputer SuperMUC-NG
581 at the Leibniz Supercomputing Centre, Germany. *SeisSol* solves the 3-D elastodynamic
582 problem of spontaneous frictional failure across prescribed fault surfaces nonlinearly cou-
583 pled to seismic wave propagation based on an the Arbitrary high-order accurate DE-
584 Rivative Discontinuous Galerkin method (ADER-DG, [Dumbser and Käser, 2006](#); [Heinecke](#)
585 [et al., 2014](#)).

586 *SeisSol* reaches scalable performance up to several thousand nodes on modern super-
587 computers ([Heinecke et al., 2014](#); [Uphoff et al., 2017](#)) and has been applied in large-scale,
588 data-integrated earthquake models, including crustal events ([Wollherr et al., 2019](#); [Ul-](#)
589 [rich et al., 2019](#)), intraplate ([Palgunadi et al., 2020](#)) and megathrust earthquakes ([Uphoff](#)
590 [et al., 2017](#)). SeisSol uses unstructured tetrahedral meshes enabling geometrically com-
591 plex models, such as branching and intersecting faults([Pelties et al., 2014](#)). Aided by
592 a clustered local time-stepping scheme, mesh resolution can be adapted to ensure fine
593 sampling of the faults while satisfying the requirements regarding numerical dispersion
594 of pure wave propagation away from the fault. End-to-end computational optimizations

595 (Uphoff et al., 2017), allows for high efficiency on high-performance computing infrastruc-
596 ture. *SeisSol* is verified in a wide range of community benchmarks (Pelties et al., 2014)
597 by the SCEC/USGS Dynamic Rupture Code Verification project (Harris et al., 2018).

598 Our model domain is discretized into an unstructured computational mesh of four-
599 node linear tetrahedral elements. We use an on-fault spatial discretisation h of 250 m for
600 all models shown in the paper (corresponding to ~ 16 million elements). In the volume, we
601 parametrize the mesh size based on the velocity structure: we allow 3 cells per wavelength
602 of shear waves to ensure resolving a maximum frequency of at least 1 Hz. In most of our
603 simulations, we use basis functions of polynomial order $p = 4$ which leads to fifth-order
604 numerical accuracy in time and space. In *SeisSol*, each triangular fault interface is sub-
605 sampled by $(p + 2)^2$ Gaussian integration points.

606 We ensure all simulation results are sufficiently resolved by following the procedure
607 established in Wollherr et al. (2018), following Day et al. (2005). We measure the cohesive
608 zone size, the region behind the rupture front where the fault strength drops from its
609 static to dynamic level, everywhere on both faults. In a purely elastic setup with depth-
610 dependent heterogeneous initial conditions it is sufficient to resolve the median cohesive
611 zone size Λ by ≈ 1 -2 elements (for $p = 5$) or ≈ 2 -3 elements (for $p = 4$). With $h = 250$ m
612 we ensure that the median cohesive zone size is correctly resolved ($\Lambda > 600$ m) for all our
613 models, except Family (C).

614 Adopting the same mesh for Family (C) models, we increase the resolution by using
615 $p = 5$ (order 6 space-time accuracy). We verify that the fault dynamics of the more
616 heterogeneous Family C models are sufficiently resolved by comparing the on-fault results
617 with results from a finer mesh of fault mesh size $h = 100$ m (corresponding to ~ 33 million
618 elements and median $\Lambda = 253$ m). Rupture arrival time, peak slip-rate, and final slip
619 differ by about 1.5%, 1.4% and 1%, respectively, between these two simulations. Such
620 errors are well within the recommended criteria of (Day et al., 2005).

621 Simulating 30 s of each earthquake scenario using 5th order accuracy in space and

622 time and on fault mesh size $h = 250$ m requires about 600 CPU hours in single precision.

623 **Appendix B Model validation data**

624 The M_w 6.5 October 30th, 2016 Norcia earthquake has been recorded by a dense net-
625 work of strong-motion stations (Figure 1), by Global Positioning System (GPS) stations,
626 and by ALOS-2 satellites. The strong motion stations belong to the National Accelerometric Network (<http://ran.protezionecivile.it>) of the Italian Department of Civil
627 Protection and the National Seismic Network of INGV (Michelini et al., 2016). Strong
628 motion recordings were processed to remove the instrument response, band-pass filtered
629 in the frequency range of 0.02 - 0.5 Hz (Butterworth filter with 2 passes 2 poles), and in-
630 tegrated to obtain ground velocity waveforms. The location of the used stations is shown
631 in Figure 1. The maximum station-epicenter distance is within 45 km. These recorded
632 waveforms are compared with synthetics computed using SeisSol, filtered in the same fre-
633 quency band. The three-components coseismic displacements recorded by campaign GPS
634 stations have been downloaded from the RING website (<http://ring.gm.ingv.it>), and
635 the location of the closest stations is shown in Figure 10.

637 The satellite data (InSAR) acquired by the ascending and descending orbits along
638 the line of sight of ALOS-2 (Cheloni et al., 2017) has a time interval covering both the
639 October 30th Norcia event and the M_w 5.9 Visso earthquake (October 26th). It does
640 not allow discrimination between the surface displacement effects produced by the two
641 earthquakes separately in the northern region (Figure 10). All these data-set have been
642 used in this work to validate the proposed dynamic models.

643 **Appendix C Velocity structure**

644 We adopt the 1D layered model for the Central Apennines of Herrmann et al. (2011)
645 (nnCIA model), constrained by deep crustal profiles, surface-wave dispersion, and tele-
646 seismic P-wave receiver functions. This model consists of five crustal layers above the

647 Moho, including a thin (1.5 km) shallow layer with a relatively low shear wave velocity
648 of 2.14 km/s and a velocity inversion at a depth of 4.5 km (see Figure 2 and Figure S1).
649 The model is routinely adopted for moment tensor inversion for Italian earthquakes and
650 kinematic finite fault inversions in the Apennines Region, including the “S18” model.

651 **Appendix D Supplementary material**

652 **References**

- 653 N. A. Haskell, Total energy and energy spectral density of elastic wave radiation from
654 propagating faults, *Bulletin of the Seismological Society of America* 54 (1964) 1811–
655 1841.
- 656 K. Wang, D. S. Dreger, E. Tinti, R. Bürgmann, T. Taira, Rupture process of the 2019
657 ridgecrest, California mw 6.4 foreshock and mw 7.1 earthquake constrained by seismic
658 and geodetic data, *Bulletin of the Seismological Society of America* 110 (2020) 1603–
659 1626. doi:[10.1785/0120200108](https://doi.org/10.1785/0120200108).
- 660 T. Ragon, A. Sladen, M. Simons, Accounting for uncertain fault geometry in earthquake
661 source inversions – I: theory and simplified application, *Geophysical Journal Interna-*
662 *tional* 214 (2018) 1174–1190. doi:[10.1093/gji/ggy187](https://doi.org/10.1093/gji/ggy187).
- 663 Y. Yagi, Y. Fukahata, Introduction of uncertainty of Green’s function into waveform
664 inversion for seismic source processes, *Geophysical Journal International* 186 (2011)
665 711–720. doi:[10.1111/j.1365-246X.2011.05043.x](https://doi.org/10.1111/j.1365-246X.2011.05043.x).
- 666 K. Shimizu, Y. Yagi, R. Okuwaki, Y. Fukahata, Development of an inversion method
667 to extract information on fault geometry from teleseismic data, *Geophysical Journal*
668 *International* 220 (2020) 1055–1065. doi:[10.1093/gji/ggz496](https://doi.org/10.1093/gji/ggz496).
- 669 E. Tinti, P. Spudich, M. Cocco, Earthquake fracture energy inferred from kinematic
670 rupture models on extended faults, *Journal of Geophysical Research* 110 (2005) B12303.
671 doi:[10.1029/2005JB003644](https://doi.org/10.1029/2005JB003644).
- 672 M. Causse, L. A. Dalguer, P. M. Mai, Variability of dynamic source parameters inferred
673 from kinematic models of past earthquakes, *Geophysical Journal International* 196
674 (2014) 1754–1769. doi:[10.1093/gji/ggt478](https://doi.org/10.1093/gji/ggt478).

- 675 R. A. Harris, B. Aagaard, M. Barall, S. Ma, D. Roten, K. Olsen, B. Duan, D. Liu, B. Luo,
676 K. Bai, J. P. Ampuero, Y. Kaneko, A. A. Gabriel, K. Duru, T. Ulrich, S. Wollherr,
677 Z. Shi, E. Dunham, S. Bydlon, Z. Zhang, X. Chen, S. N. Somala, C. Pelties, J. Tago,
678 V. M. Cruz-Atienza, J. Kozdon, E. Daub, K. Aslam, Y. Kase, K. Withers, L. Dalguer,
679 A suite of exercises for verifying dynamic earthquake rupture codes, *Seismological*
680 *Research Letters* 89 (2018) 1146–1162. doi:[10.1785/0220170222](https://doi.org/10.1785/0220170222).
- 681 R. Ando, Y. Kaneko, Dynamic Rupture Simulation Reproduces Spontaneous Multifault
682 Rupture and Arrest During the 2016 M w 7.9 Kaikoura Earthquake, *Geophysical*
683 *Research Letters* 45 (2018) 875–12. doi:[10.1029/2018GL080550](https://doi.org/10.1029/2018GL080550).
- 684 S. Wollherr, A. Gabriel, P. M. Mai, Landers 1992 “Reloaded”: Integrative Dynamic
685 Earthquake Rupture Modeling, *Journal of Geophysical Research: Solid Earth* 124
686 (2019) 6666–6702. doi:[10.1029/2018JB016355](https://doi.org/10.1029/2018JB016355).
- 687 A.-A. Gabriel, J.-P. Ampuero, L. A. Dalguer, P. M. Mai, The transition of dynamic
688 rupture styles in elastic media under velocity-weakening friction, *Journal of Geophysical*
689 *Research: Solid Earth* 117 (2012) 9311. doi:[10.1029/2012JB009468](https://doi.org/10.1029/2012JB009468).
- 690 K. Bai, J.-P. Ampuero, Effect of Seismogenic Depth and Background Stress on Phys-
691 ical Limits of Earthquake Rupture Across Fault Step Overs, *Journal of Geophysical*
692 *Research: Solid Earth* 122 (2017) 280–10. doi:[10.1002/2017JB014848](https://doi.org/10.1002/2017JB014848).
- 693 V. Lambert, N. Lapusta, S. Perry, Propagation of large earthquakes as self-healing pulses
694 or mild cracks, *Nature* 591 (2021) 252–258. doi:[10.1038/s41586-021-03248-1](https://doi.org/10.1038/s41586-021-03248-1).
- 695 P. Galvez, J.-P. Ampuero, L. A. Dalguer, S. N. Somala, T. Nissen-Meyer, Dynamic
696 earthquake rupture modelled with an unstructured 3-D spectral element method applied
697 to the 2011 M9 Tohoku earthquake, *Geophysical Journal International* 198 (2014) 1222–
698 1240. doi:[10.1093/gji/ggu203](https://doi.org/10.1093/gji/ggu203).

- 699 F. Gallovič, Valentová, J. P. Ampuero, A. A. Gabriel, Bayesian Dynamic Finite-Fault
700 Inversion: 2. Application to the 2016 Mw 6.2 Amatrice, Italy, Earthquake, Journal of
701 Geophysical Research: Solid Earth 124 (2019) 6970–6988. doi:[10.1029/2019JB017512](https://doi.org/10.1029/2019JB017512).
- 702 H. Aochi, C. Twardzik, Imaging of Seismogenic Asperities of the 2016 ML 6.0 Amatrice,
703 Central Italy, Earthquake Through Dynamic Rupture Simulations, Pure and Applied
704 Geophysics 177 (2020) 1931–1946. doi:[10.1007/s00024-019-02199-z](https://doi.org/10.1007/s00024-019-02199-z).
- 705 D. D. Oglesby, R. J. Archuleta, S. B. Nielsen, Earthquakes on dipping faults: The effects
706 of broken symmetry, Science 280 (1998) 1055–1059. doi:[10.1126/science.280.5366.](https://doi.org/10.1126/science.280.5366.1055)
707 [1055](https://doi.org/10.1126/science.280.5366.1055).
- 708 H. Aochi, Dynamic asymmetry of normal and reverse faults due to constrained depth-
709 dependent stress accumulation, Geophysical Journal International 215 (2018) 2134–
710 2143. doi:[10.1093/gji/ggy407](https://doi.org/10.1093/gji/ggy407).
- 711 M. Guatteri, P. Spudich, What can strong-motion data tell us about slip-weakening
712 fault-friction laws?, Bulletin of the Seismological Society of America 90 (2000) 98–116.
713 doi:[10.1785/0119990053](https://doi.org/10.1785/0119990053).
- 714 J. Schmedes, R. J. Archuleta, D. Lavallée, Correlation of earthquake source parameters
715 inferred from dynamic rupture simulations, Journal of Geophysical Research 115 (2010)
716 B03304. doi:[10.1029/2009JB006689](https://doi.org/10.1029/2009JB006689).
- 717 J. H. Dieterich, Modeling of rock friction: 1. Experimental results and constitutive equa-
718 tions, Journal of Geophysical Research 84 (1979) 2161. doi:[10.1029/JB084iB05p02161](https://doi.org/10.1029/JB084iB05p02161).
- 719 M. Ohnaka, Y. Kuwahara, K. Yamamoto, Constitutive relations between dynamic phys-
720 ical parameters near a tip of the propagating slip zone during stick-slip shear failure,
721 Tectonophysics 144 (1987) 109–125. doi:[10.1016/0040-1951\(87\)90011-4](https://doi.org/10.1016/0040-1951(87)90011-4).

- 722 G. Di Toro, R. Han, T. Hirose, N. De Paola, S. Nielsen, K. Mizoguchi, F. Ferri, M. Cocco,
723 T. Shimamoto, Fault lubrication during earthquakes., *Nature* 471 (2011) 494–498.
724 doi:[10.1038/nature09838](https://doi.org/10.1038/nature09838).
- 725 C. Collettini, T. Tesei, M. M. Scuderi, B. M. Carpenter, C. Viti, Beyond Byerlee friction
726 , weak faults and implications for slip behavior, *Earth and Planetary Science Letters*
727 519 (2019) 245–263. doi:[10.1016/j.epsl.2019.05.011](https://doi.org/10.1016/j.epsl.2019.05.011).
- 728 T. Ulrich, A. A. Gabriel, J. P. Ampuero, W. Xu, Dynamic viability of the 2016 Mw
729 7.8 Kaikōura earthquake cascade on weak crustal faults, *Nature Communications* 10
730 (2019) 1–16. doi:[10.1038/s41467-019-09125-w](https://doi.org/10.1038/s41467-019-09125-w).
- 731 S. Murphy, G. Di Toro, F. Romano, A. Scala, S. Lorito, E. Spagnuolo, S. Aretusini,
732 G. Festa, A. Piatanesi, S. Nielsen, Tsunamigenic earthquake simulations using experi-
733 mentally derived friction laws, *Earth and Planetary Science Letters* 486 (2018) 155–165.
734 doi:[10.1016/j.epsl.2018.01.011](https://doi.org/10.1016/j.epsl.2018.01.011).
- 735 R. A. Harris, M. Barall, D. A. Lockner, D. E. Moore, D. A. Ponce, R. W. Graymer,
736 G. Funning, C. A. Morrow, C. Kyriakopoulos, D. Eberhart-Phillips, A Geology and
737 Geodesy Based Model of Dynamic Earthquake Rupture on the Rodgers Creek-Hayward-
738 Calaveras Fault System, California, *Journal of Geophysical Research: Solid Earth* 126
739 (2021) e2020JB020577. doi:[10.1029/2020jb020577](https://doi.org/10.1029/2020jb020577).
- 740 L. Chiaraluce, R. Di Stefano, E. Tinti, L. Scognamiglio, M. Michele, E. Casarotti, M. Cat-
741 taneo, P. De Gori, C. Chiarabba, G. Monachesi, A. Lombardi, L. Valoroso, D. Latorre,
742 S. Marzorati, The 2016 central Italy seismic sequence: A first look at the main-
743 shocks, aftershocks, and source models, *Seismological Research Letters* 88 (2017).
744 doi:[10.1785/0220160221](https://doi.org/10.1785/0220160221).
- 745 D. Cheloni, V. De Novellis, M. Albano, A. Antonioli, M. Anzidei, S. Atzori, A. Aval-
746 lone, C. Bignami, M. Bonano, S. Calcaterra, R. Castaldo, F. Casu, G. Cecere,

747 C. De Luca, R. Devoti, D. Di Bucci, A. Esposito, A. Galvani, P. Gambino, R. Giu-
748 liani, R. Lanari, M. Manunta, M. Manzo, M. Mattone, A. Montuori, A. Pepe, S. Pepe,
749 G. Pezzo, G. Pietrantonio, M. Polcari, F. Riguzzi, S. Salvi, V. Sepe, E. Serpelloni,
750 G. Solaro, S. Stramondo, P. Tizzani, C. Tolomei, E. Trasatti, E. Valerio, I. Zinno,
751 C. Doglioni, Geodetic model of the 2016 Central Italy earthquake sequence inferred
752 from InSAR and GPS data, *Geophysical Research Letters* 44 (2017) 6778–6787.
753 doi:[10.1002/2017GL073580](https://doi.org/10.1002/2017GL073580).

754 A. Pizzi, A. Di Domenica, F. Gallovič, L. Luzi, R. Puglia, Fault Segmentation as Con-
755 straint to the Occurrence of the Main Shocks of the 2016 Central Italy Seismic Sequence,
756 *Tectonics* 36 (2017) 2370–2387. doi:[10.1002/2017TC004652](https://doi.org/10.1002/2017TC004652).

757 L. Scognamiglio, E. Tinti, E. Casarotti, S. Pucci, F. Villani, M. Cocco, F. Magnoni,
758 A. Michellini, D. Dreger, Complex Fault Geometry and Rupture Dynamics of the Mw
759 6.5, 30 October 2016, Central Italy Earthquake, *Journal of Geophysical Research: Solid*
760 *Earth* 123 (2018) 2943–2964. doi:[10.1002/2018JB015603](https://doi.org/10.1002/2018JB015603).

761 R. J. Walters, L. C. Gregory, L. N. J. Wedmore, T. J. Craig, K. Mccaffrey, M. Wilkinson,
762 J. Chen, Z. Li, J. R. Elliott, H. Goodall, F. Iezzi, F. Livio, A. M. Michetti, G. Roberts,
763 E. Vittori, Dual control of fault intersections on stop-start rupture in the 2016 Central
764 Italy seismic sequence 500 (2018) 1–14. doi:[10.1016/j.eps1.2018.07.043](https://doi.org/10.1016/j.eps1.2018.07.043).

765 L. Bonini, R. Basili, P. Burrato, V. Cannelli, U. Fracassi, F. E. Maesano, D. Melini,
766 G. Tarabusi, M. M. Tiberti, P. Vannoli, G. Valensise, Testing Different Tectonic Models
767 for the Source of the Mw 6.5, 30 October 2016, Norcia Earthquake (Central Italy): A
768 Youthful Normal Fault, or Negative Inversion of an Old Thrust?, *Tectonics* 38 (2019)
769 990–1017. doi:[10.1029/2018TC005185](https://doi.org/10.1029/2018TC005185).

770 M. Michele, L. Chiaraluce, R. Di Stefano, F. Waldhauser, Fine-scale structure of the
771 2016–2017 central Italy seismic sequence from data recorded at the Italian national

772 network, *Journal of Geophysical Research: Solid Earth* 125 (2020) 1–26. doi:[10.1029/
773 2019JB018440](https://doi.org/10.1029/2019JB018440).

774 M. T. Mariucci, P. Montone, IPSI 1.4, Database of Italian Present-day Stress Indicators,
775 2020. doi:<http://doi.org/10.13127/IPSI.1.4>.

776 E. Tinti, L. Scognamiglio, A. Michelini, M. Cocco, Slip heterogeneity and directivity
777 of the ML 6.0, 2016, Amatrice earthquake estimated with rapid finite-fault inversion,
778 *Geophysical Research Letters* 43 (2016). doi:[10.1002/2016GL071263](https://doi.org/10.1002/2016GL071263).

779 R. Douilly, D. D. Oglesby, M. L. Cooke, J. L. Hatch, Dynamic models of earthquake
780 rupture along branch faults of the eastern San Geronimo pass region in California using
781 complex fault structure, *Geosphere* 16 (2020) 474–489. doi:[10.1130/GES02192.1](https://doi.org/10.1130/GES02192.1).

782 M. Dumbser, M. Käser, An arbitrary high-order discontinuous Galerkin method for elastic
783 waves on unstructured meshes - II. The three-dimensional isotropic case, *Geophysical
784 Journal International* 167 (2006) 319–336. doi:[10.1111/j.1365-246X.2006.03120.x](https://doi.org/10.1111/j.1365-246X.2006.03120.x).

785 C. Pelties, A. A. Gabriel, J. P. Ampuero, Verification of an ADER-DG method for
786 complex dynamic rupture problems, *Geoscientific Model Development* 7 (2014) 847–
787 866. doi:[10.5194/gmd-7-847-2014](https://doi.org/10.5194/gmd-7-847-2014).

788 G. I. Barenblatt, The formation of equilibrium cracks during brittle fracture. General
789 ideas and hypotheses. Axially-symmetric cracks, *Journal of Applied Mathematics and
790 Mechanics* 23 (1959) 622–636. doi:[10.1016/0021-8928\(59\)90157-1](https://doi.org/10.1016/0021-8928(59)90157-1).

791 D. J. Andrews, Rupture Propagation With Finite Stress in Antiplane Strain, *Journal of
792 Geophysical Research* 81 (1976) 3575–3582. doi:[10.1029/JB081i020p03575](https://doi.org/10.1029/JB081i020p03575).

793 M. Cocco, E. Tinti, Scale dependence in the dynamics of earthquake propagation: Ev-
794 idence from seismological and geological observations, *Earth and Planetary Science
795 Letters* 273 (2008) 123–131.

- 796 P. Spudich, M. Guatteri, K. Otsuki, J. Minagawa, Use of Fault striations and Disloca-
797 tion Models to Infer Tectonic Shear Stress during the 1995 Hyogo-ken Nanbu (Kobe)
798 Earthquake, Technical Report 2, 1998. URL: [http://pubs.geoscienceworld.org/
799 ssa/bssa/article-pdf/88/2/413/2709377/BSSA0880020413.pdf](http://pubs.geoscienceworld.org/ssa/bssa/article-pdf/88/2/413/2709377/BSSA0880020413.pdf).
- 800 E. Tinti, M. Cocco, E. Fukuyama, A. Piatanesi, Dependence of slip weakening distance
801 (Dc) on final slip during dynamic rupture of earthquakes, *Geophysical Journal Inter-
802 national* 177 (2009) 1205–1220. doi:[10.1111/j.1365-246X.2009.04143.x](https://doi.org/10.1111/j.1365-246X.2009.04143.x).
- 803 H. Goto, S. Sawada, Trade-offs among dynamic parameters inferred from results of dy-
804 namic source inversion, *Bulletin of the Seismological Society of America* 100 (2010)
805 910–922. doi:[10.1785/0120080250](https://doi.org/10.1785/0120080250).
- 806 J. Ripperger, J.-P. Ampuero, P. M. Mai, D. Giardini, Earthquake source characteristics
807 from dynamic rupture with constrained stochastic fault stress, *Journal of Geophysical
808 Research: Solid Earth* 112 (2007) 4311. doi:[10.1029/2006JB004515](https://doi.org/10.1029/2006JB004515).
- 809 K. H. Palgunadi, A. A. Gabriel, T. Ulrich, J. A. Lopez-Comino, P. M. Mai, Dynamic
810 fault interaction during a fluid-injection-induced earthquake: The 2017 mw 5.5 pohang
811 event, *Bulletin of the Seismological Society of America* 110 (2020) 2328–2349. doi:[10.
812 1785/0120200106](https://doi.org/10.1785/0120200106).
- 813 W. H. Savran, K. B. Olsen, Kinematic Rupture Generator Based on 3-D Spontaneous
814 Rupture Simulations Along Geometrically Rough Faults, *Journal of Geophysical Re-
815 search: Solid Earth* 125 (2020) 1–22. doi:[10.1029/2020JB019464](https://doi.org/10.1029/2020JB019464).
- 816 M. M. Scuderi, A. R. Niemeijer, C. Collettini, C. Marone, Frictional properties and slip
817 stability of active faults within carbonate-evaporite sequences: The role of dolomite
818 and anhydrite, *Earth and Planetary Science Letters* 369-370 (2013) 220–232. doi:[10.
819 1016/j.epsl.2013.03.024](https://doi.org/10.1016/j.epsl.2013.03.024).

- 820 N. De Paola, R. E. Holdsworth, C. Viti, C. Collettini, R. Bullock, Can grain size sensitive
821 flow lubricate faults during the initial stages of earthquake propagation?, *Earth and*
822 *Planetary Science Letters* 431 (2015) 48–58. doi:[10.1016/j.epsl.2015.09.002](https://doi.org/10.1016/j.epsl.2015.09.002).
- 823 D. J. Andrews, A stochastic fault model: 1. Static case, Technical Report B7, 1980.
824 doi:[10.1029/JB085iB07p03867](https://doi.org/10.1029/JB085iB07p03867).
- 825 J. Byerlee, Friction of Rocks, in: *Rock Friction and Earthquake Prediction*, Birkhäuser
826 Basel, 1978, pp. 615–626. doi:[10.1007/978-3-0348-7182-2_{_}4](https://doi.org/10.1007/978-3-0348-7182-2_{_}4).
- 827 C. Kyriakopoulos, D. D. Oglesby, T. K. Rockwell, A. J. Meltzner, M. Barall, J. M.
828 Fletcher, D. Tulanowski, Dynamic Rupture Scenarios in the Brawley Seismic Zone,
829 Salton Trough, Southern California, *Journal of Geophysical Research: Solid Earth* 124
830 (2019) 3680–3707. doi:[10.1029/2018JB016795](https://doi.org/10.1029/2018JB016795).
- 831 R. Tang, J. Yuan, L. Gan, Free-Surface-Induced Supershear Transition in 3-D Simulations
832 of Spontaneous Dynamic Rupture on Oblique Faults, *Geophysical Research Letters* 48
833 (2021) e2020GL091621. doi:[10.1029/2020GL091621](https://doi.org/10.1029/2020GL091621).
- 834 H. S. Bhat, M. Olives, R. Dmowska, J. R. Rice, Role of fault branches in earthquake
835 rupture dynamics, *Journal of Geophysical Research* 112 (2007) B11309. doi:[10.1029/
836 2007JB005027](https://doi.org/10.1029/2007JB005027).
- 837 A. C. Palmer, J. R. Rice, The growth of slip surfaces in the progressive failure of over-
838 consolidated clay, *Proceedings of the Royal Society of London. A. Mathematical and*
839 *Physical Sciences* 332 (1973) 527–548. doi:[10.1098/rspa.1973.0040](https://doi.org/10.1098/rspa.1973.0040).
- 840 R. C. Viesca, D. I. Garagash, Ubiquitous weakening of faults due to thermal pressuriza-
841 tion, *Nature Geoscience* 8 (2015) 875–879. doi:[10.1038/ngeo2554](https://doi.org/10.1038/ngeo2554).
- 842 M. L. Blanpied, D. A. Lockner, J. D. Byerlee, Fault stability inferred from granite sliding

843 experiments at hydrothermal conditions, *Geophysical Research Letters* 18 (1991) 609–
844 612. doi:[10.1029/91GL00469](https://doi.org/10.1029/91GL00469).

845 E. E. Brodsky, J. J. Mori, L. Anderson, F. M. Chester, M. Conin, E. M. Dunham,
846 N. Eguchi, P. M. Fulton, R. Hino, T. Hirose, M. J. Ikari, S. Kodaira, W. Lin, Y. Naka-
847 mura, H. S. Rabinowitz, C. Regalla, F. Remitti, C. Rowe, D. M. Saffer, S. Saito,
848 J. Sample, Y. Sanada, H. M. Savage, T. Sun, S. Toczko, K. Ujiie, The State of Stress
849 on the Fault Before , During , and After a Major Earthquake (2020).

850 E. M. Dunham, Dissipative interface waves and the transient response of a three-
851 dimensional sliding interface with Coulomb friction, *Journal of the Mechanics and*
852 *Physics of Solids* 53 (2005) 327–357. doi:[10.1016/j.jmps.2004.07.003](https://doi.org/10.1016/j.jmps.2004.07.003).

853 M. Porreca, G. Minelli, M. Ercoli, A. Brobia, P. Mancinelli, F. Cruciani, C. Giorgetti,
854 F. Carboni, F. Mirabella, G. Cavinato, A. Cannata, C. Pauselli, M. R. Barchi, Seismic
855 Reflection Profiles and Subsurface Geology of the Area Interested by the 2016-2017
856 Earthquake Sequence (Central Italy), *Tectonics* 37 (2018) 1116–1137. doi:[10.1002/
857 2017TC004915](https://doi.org/10.1002/2017TC004915).

858 R. Ruggieri, M. M. Scuderi, F. Trippetta, E. Tinti, M. Brignoli, S. Mantica, S. Petroselli,
859 L. Osculati, G. Volontè, C. Collettini, The role of shale content and pore-water satura-
860 tion on frictional properties of simulated carbonate faults, *Tectonophysics* 807 (2021)
861 228811. doi:[10.1016/j.tecto.2021.228811](https://doi.org/10.1016/j.tecto.2021.228811).

862 A. Heinecke, A. Breuer, S. Rettenberger, M. Bader, A. A. Gabriel, C. Pelties, A. Bode,
863 W. Barth, X. K. Liao, K. Vaidyanathan, M. Smelyanskiy, P. Dubey, Petascale High
864 Order Dynamic Rupture Earthquake Simulations on Heterogeneous Supercomputers,
865 in: *International Conference for High Performance Computing, Networking, Storage*
866 *and Analysis, SC*, volume 2015-January, IEEE Computer Society, 2014, pp. 3–14.
867 doi:[10.1109/SC.2014.6](https://doi.org/10.1109/SC.2014.6).

- 868 C. Uphoff, S. Rettenberger, M. Bader, E. H. Madden, T. Ulrich, S. Wollherr, A. A.
869 Gabriel, Extreme scale multi-physics simulations of the tsunamigenic 2004 sumatra
870 megathrust earthquake, in: Proceedings of the International Conference for High
871 Performance Computing, Networking, Storage and Analysis, SC 2017, volume 16,
872 Association for Computing Machinery, Inc, New York, NY, USA, 2017, pp. 1–16.
873 doi:[10.1145/3126908.3126948](https://doi.org/10.1145/3126908.3126948).
- 874 S. Wollherr, A. A. Gabriel, C. Uphoff, Off-fault plasticity in three-dimensional dynamic
875 rupture simulations using a modal Discontinuous Galerkin method on unstructured
876 meshes: Implementation, verification and application, *Geophysical Journal Interna-*
877 *tional* 214 (2018) 1556–1584. doi:[10.1093/GJI/GGY213](https://doi.org/10.1093/GJI/GGY213).
- 878 S. M. Day, L. A. Dalguer, N. Lapusta, Y. Liu, Comparison of finite difference and bound-
879 ary integral solutions to three-dimensional spontaneous rupture, *Journal of Geophysical*
880 *Research: Solid Earth* 110 (2005) 1–23. doi:[10.1029/2005JB003813](https://doi.org/10.1029/2005JB003813).
- 881 A. Michelini, L. Margheriti, M. Cattaneo, G. Cecere, G. D’Anna, A. Delladio, M. Moretti,
882 S. Pintore, A. Amato, A. Basili, A. Bono, P. Casale, P. Danecek, M. Demartin,
883 L. Faenza, V. Lauciani, A. Giovanni Mandiello, A. Marchetti, C. Marcocci, S. Mazza,
884 F. Mariano Mele, A. Nardi, C. Nostro, M. Pignone, M. Quintiliani, S. Rao, L. Scog-
885 namiglio, G. Selvaggi, The Italian National Seismic Network and the earthquake and
886 tsunami monitoring and surveillance systems, *Advances in Geosciences* 43 (2016) 31–38.
887 doi:[10.5194/adgeo-43-31-2016](https://doi.org/10.5194/adgeo-43-31-2016).
- 888 R. B. Herrmann, L. Malagnini, I. Munafò, Regional moment tensors of the 2009 L’Aquila
889 earthquake sequence, *Bulletin of the Seismological Society of America* 101 (2011) 975–
890 993. doi:[10.1785/0120100184](https://doi.org/10.1785/0120100184).
- 891 M. Barall, R. A. Harris, Metrics for comparing dynamic earthquake rupture simulations,
892 *Seismological Research Letters* 86 (2015) 223–235. doi:[10.1785/0220140122](https://doi.org/10.1785/0220140122).

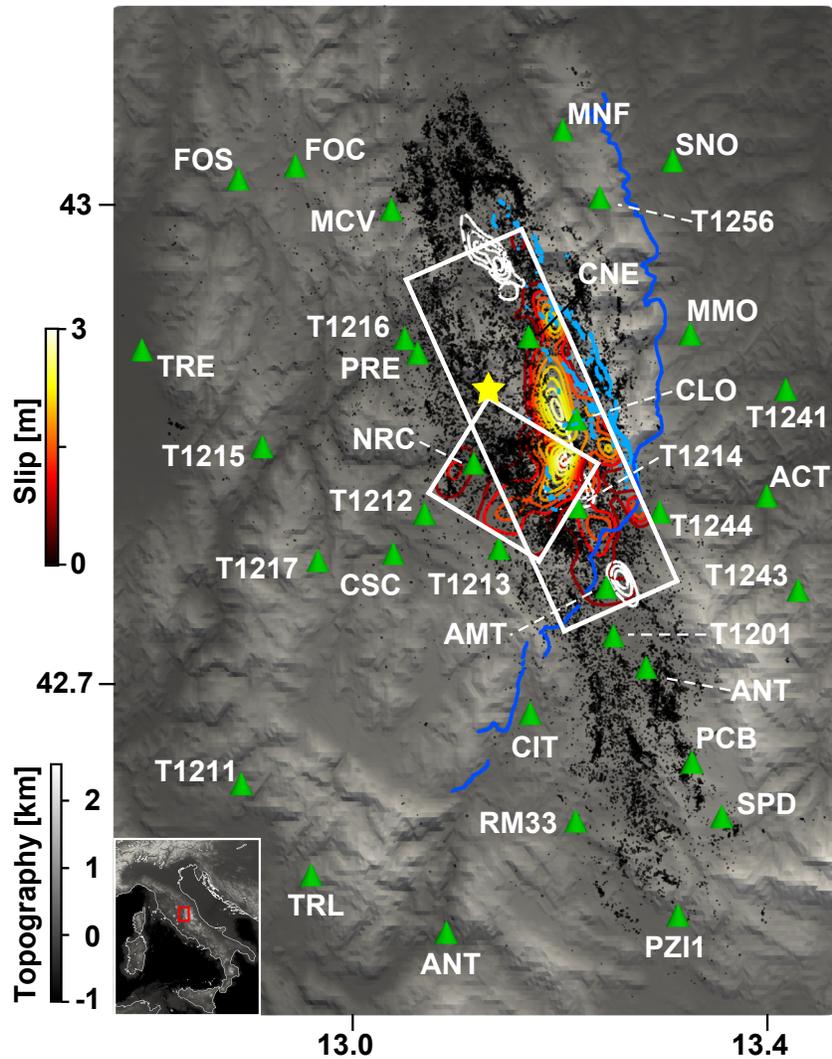


Figure 1: Map of the study area. Black dots: Amatrice–Visso–Norcia seismic sequence relocated earthquakes from [Michele et al. \(2020\)](#); darker blue lines: fault traces of OAS (Olevano-Antrodoco-Sibillini) thrust fronts; light blue lines: observed surface offsets. Green triangles denote the strong motion stations. Yellow star shows the epicenter of the 2016 Norcia event adopted in this study. White contours are the slip distribution for Visso and Amatrice events, from [Tinti et al. \(2016\)](#); [Chiaraluce et al. \(2017\)](#). The slip distribution of the Norcia event inferred by [Scognamiglio et al. \(2018\)](#) is shown by coloured contours.

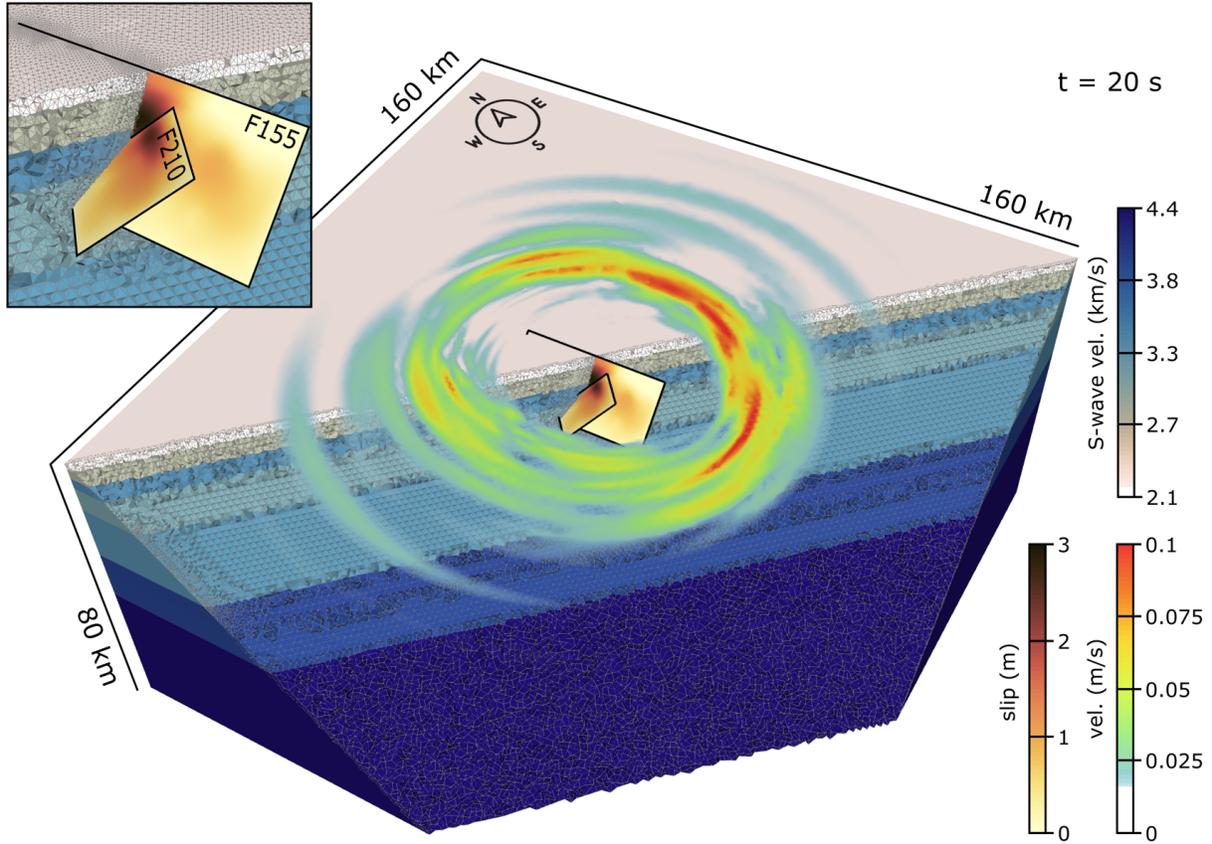


Figure 2: Snapshot of the ground surface wavefield (absolute particle velocity in m/s) at a simulation time of 20 s. The two-faults model, as well as the unstructured mesh incorporating the interface layers of the 1D layered velocity model (nnCIA model, [Herrmann et al. \(2011\)](#)) and featuring refined resolution in the vicinity of the faults, are also shown. The inset provides a zoomed view on the two fault planes, colored by the slip distribution of the exemplary model of Family (B) in which stress drop is assumed proportional to slip. The two-planar-fault geometry [Scognamiglio et al. \(2018\)](#) consists of a main fault branch N155° trending along the Apennines and dipping 47° to the SW (hereinafter F155), and a second fault plane striking N210° oblique to the Apennines and dipping 36° to the NW (hereinafter F210). The main fault is 34 km long and 16 km wide (downdip), while the secondary fault is 10 km long and 14 km wide. F155 reaches the modeled free surface, while the top border of F210 is 1.8 km below the modeled ground surface.

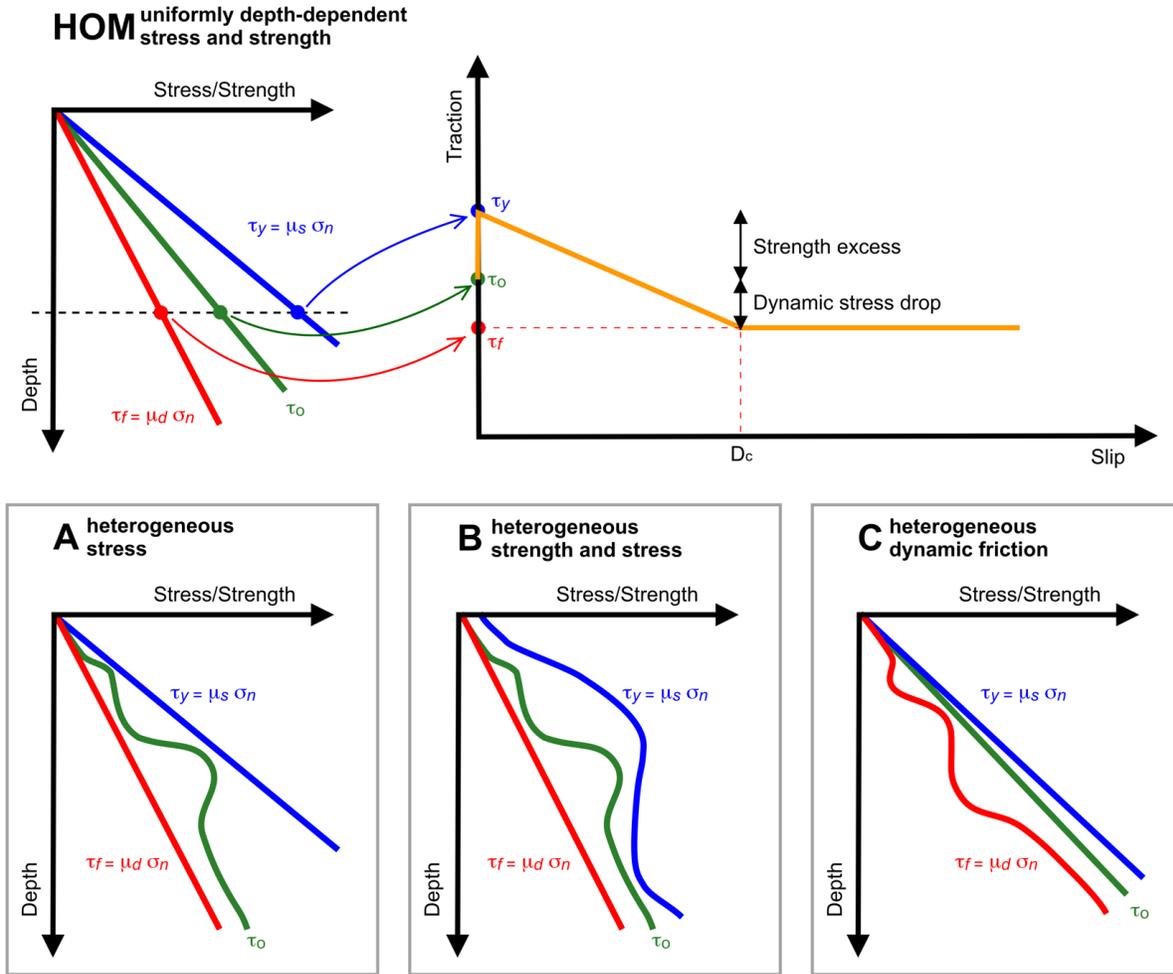


Figure 3: Variation with depth of dynamic parameters describing the LSW law, classified in four families of dynamic rupture models proposed in this work. Family (Hom) encompasses models based on laterally-invariant and linearly depth-dependent stress and strength conditions with constant static and dynamic friction coefficients. Family (A), called “family of heterogeneous stress”, includes models with constant static and dynamic friction, linearly depth-dependent normal stress, and variable initial shear stress τ_0 . Family (B), called “family of heterogeneous strength and stress”, includes all models with constant dynamic friction, linearly depth-dependent normal stress, and heterogeneous static friction and initial shear stress. Family (C), called “family of heterogeneous dynamic friction”, includes all models with linearly depth-dependent normal stress and initial shear stress, constant static friction, and heterogeneous dynamic friction.

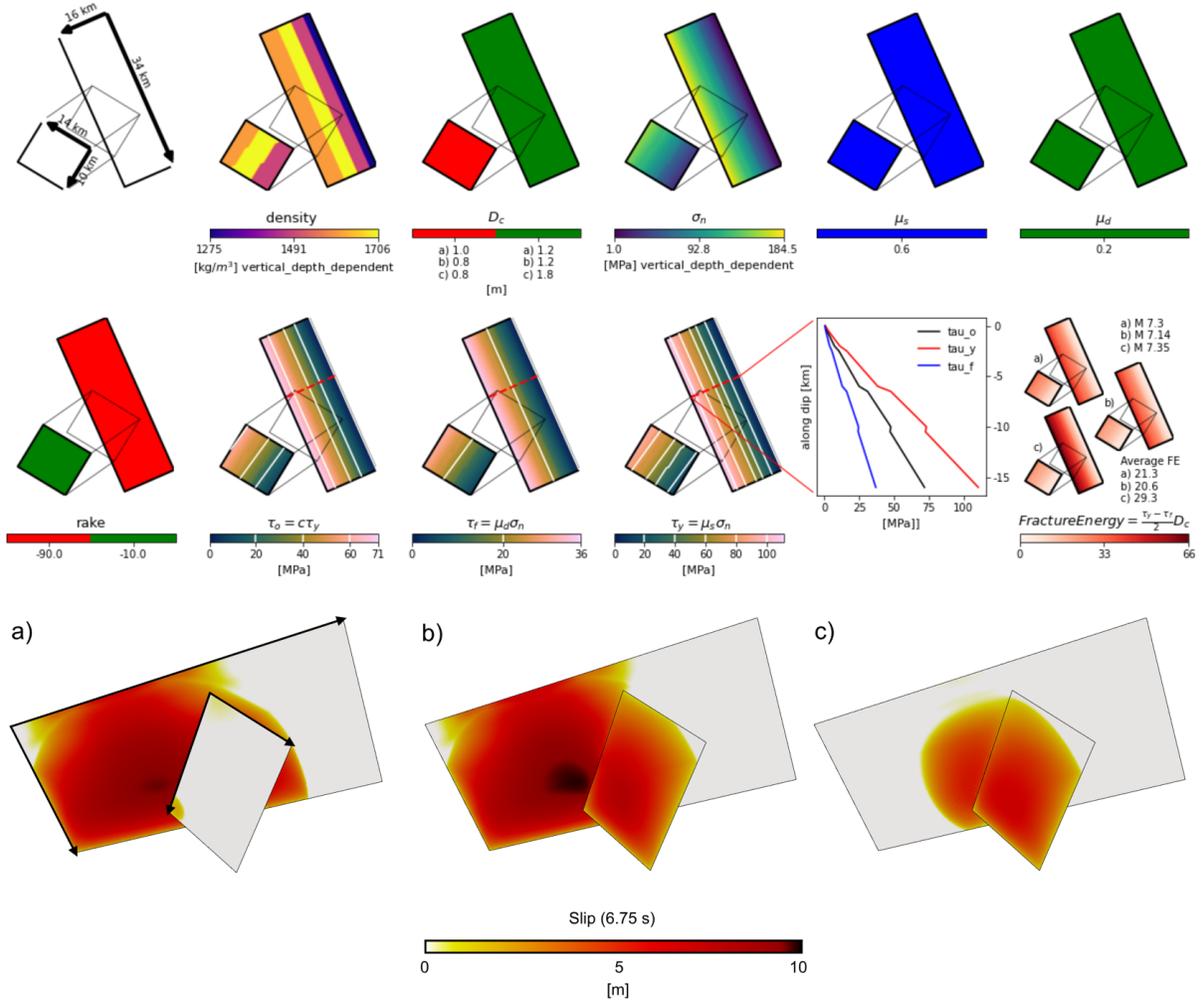


Figure 4: Parametrization and rupture dynamics of representative dynamic rupture models belonging to Family (Hom). Upper panels: example of distribution of dynamic parameters in homogeneous stress conditions on both the fault planes (Family (Hom)). Bottom panels: slip distribution after 6.75 s of rupture initiation for models with: a) $D_c^{F155} = 1.2$ m and $D_c^{F210} = 1.0$ m ; b) $D_c^{F155} = 1.2$ m and $D_c^{F210} = 0.8$ m; c) $D_c^{F155} = 1.8$ m and $D_c^{F210} = 0.8$ m . The fracture energy panel indicates the average values of fracture energy E_g and the moment magnitude M_w values after the ruptures termination.

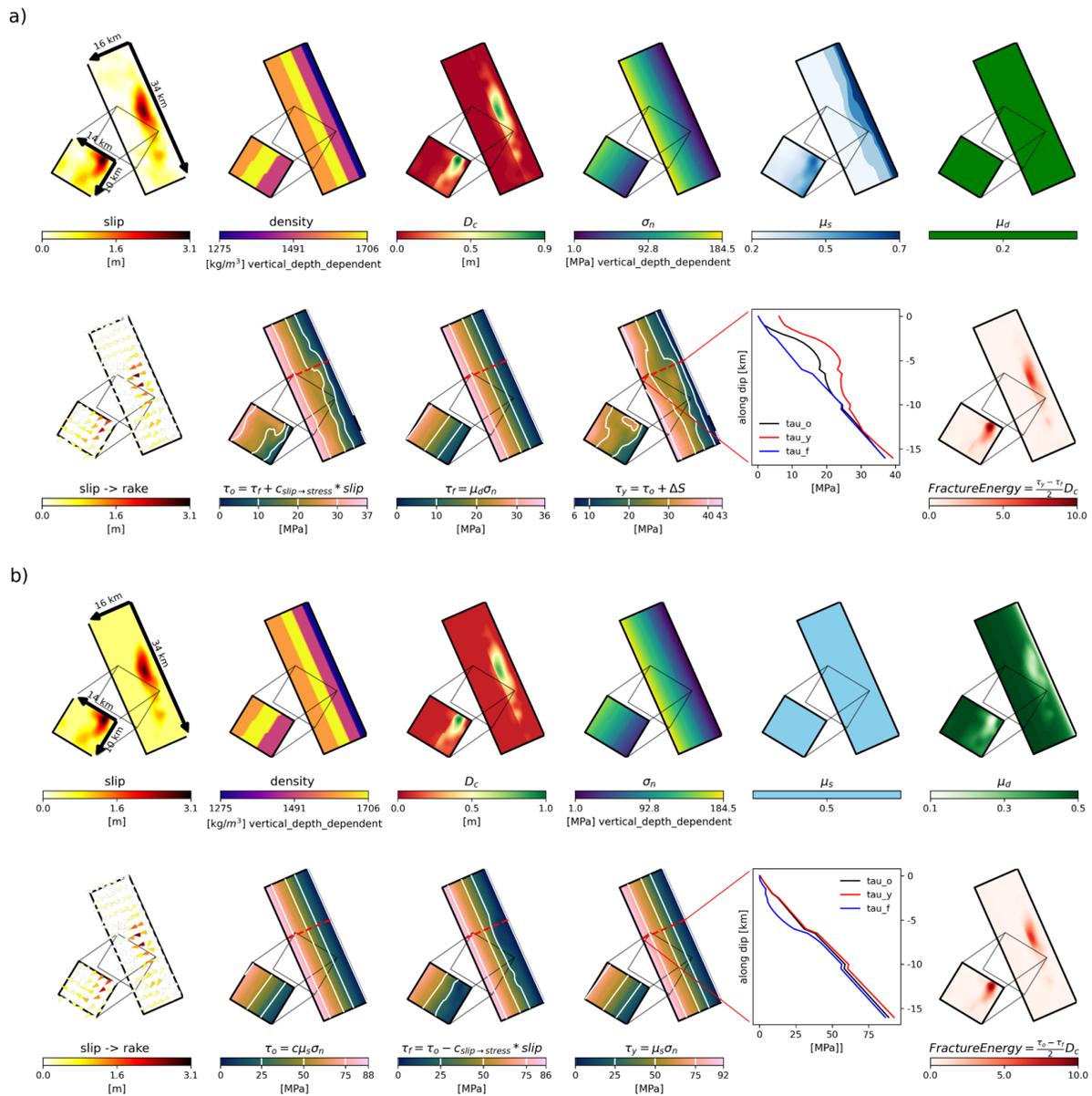


Figure 5: Distribution of the dynamic rupture parameters of the two exemplary models of Family (B) (panel a) and (C) (panel b)

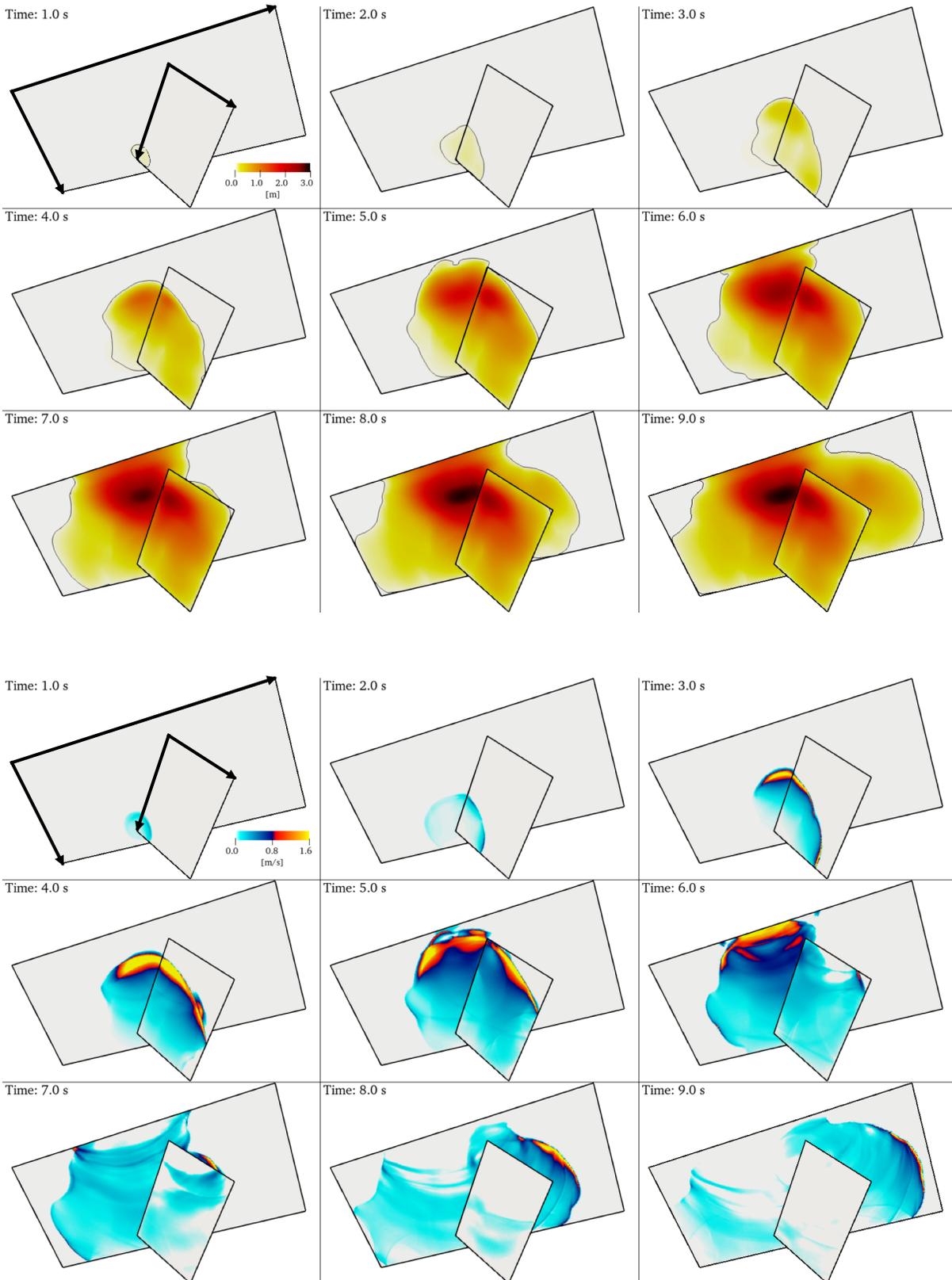


Figure 6: Dynamics of the exemplary model belonging to Family (B) inferred by assuming stress drop proportional to slip. Snapshots, every one second, of slip (m, top) and slip rate (m/s, bottom)

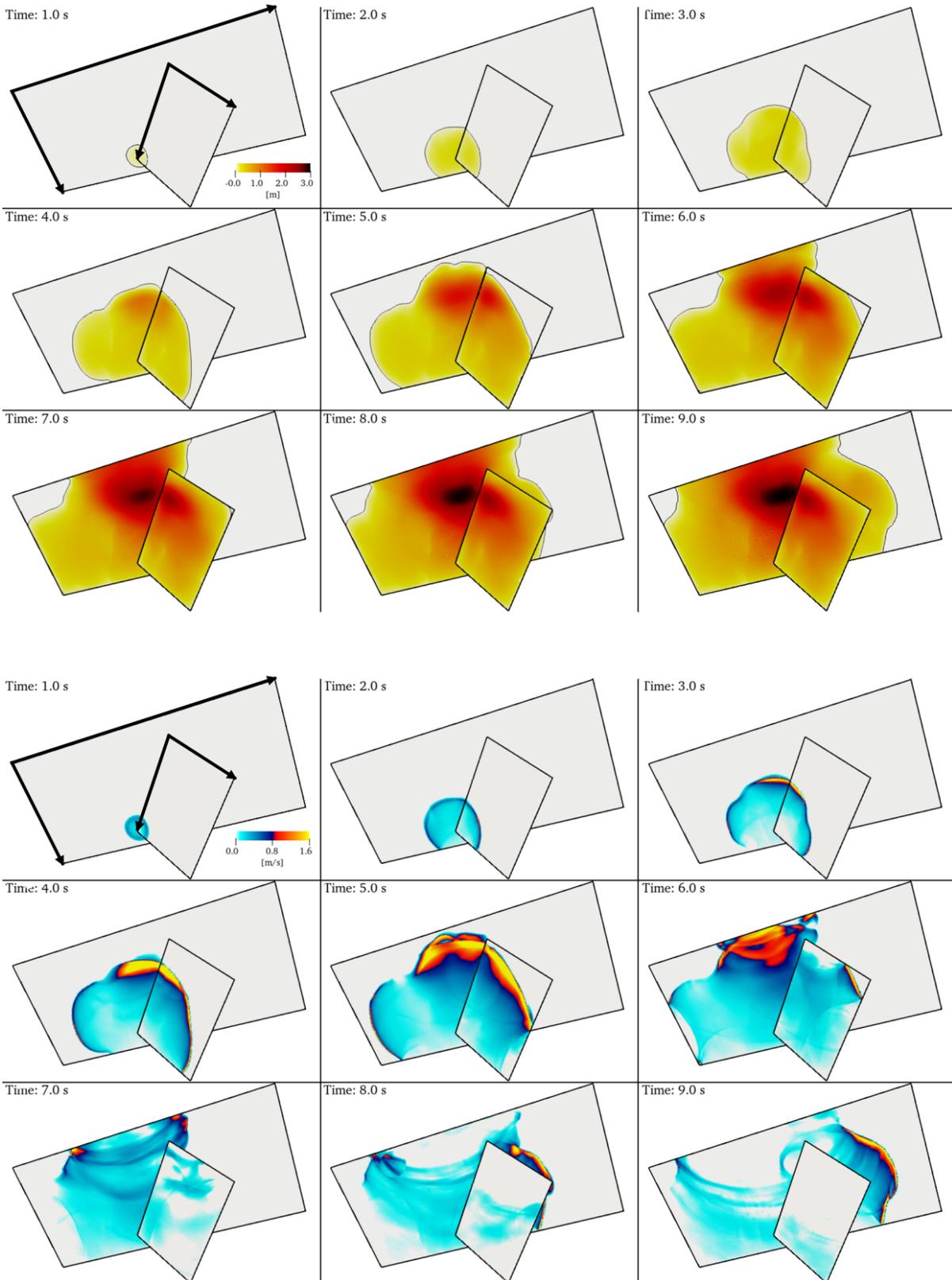


Figure 7: Dynamics of the exemplary model belonging to Family (C) inferred by assuming stress drop proportional to slip. Snapshots, every one second, of slip (m, top) and slip rate (m/s, bottom)

stress proportional to slip models

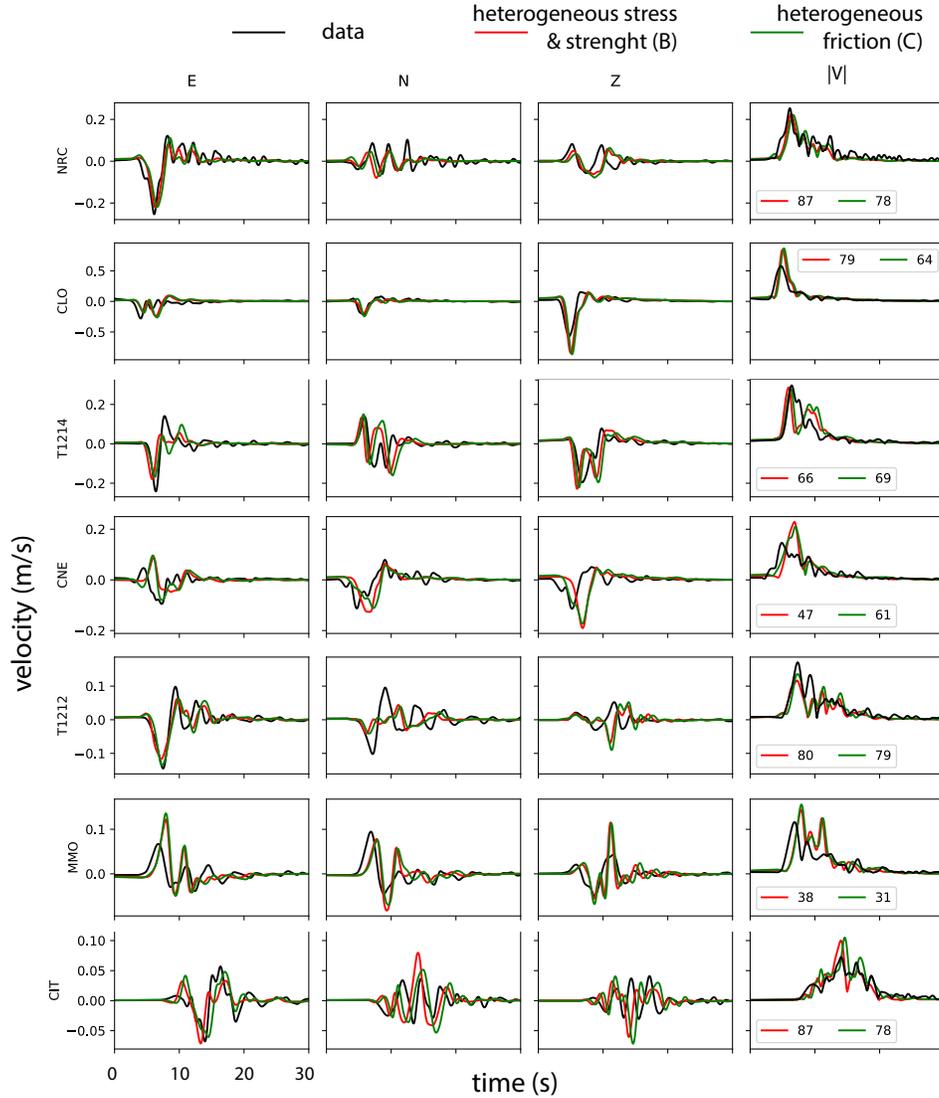


Figure 8: Comparison of synthetic strong-motion velocity waveforms (red and green for Family (B) and (C) models, respectively) inferred by assuming stress drop proportional to slip with observations (black) at selected stations. We quantify the waveform fit using the metric equation suggested by [Barall and Harris \(2015\)](#) on the time-history of the 3D absolute velocity vector. The fit can vary between -100% to $+100\%$ from worst to best, respectively. Both families give similar goodness of fit ($VR_B = 55.5\%$ for Family (B) and $VR_C = 49.7\%$ for Family (C)). Numbers in the fourth column represent goodness of fit for each station and model. Station location is shown in figure 1. Additional waveform comparisons are shown in Figure S2 and S3.

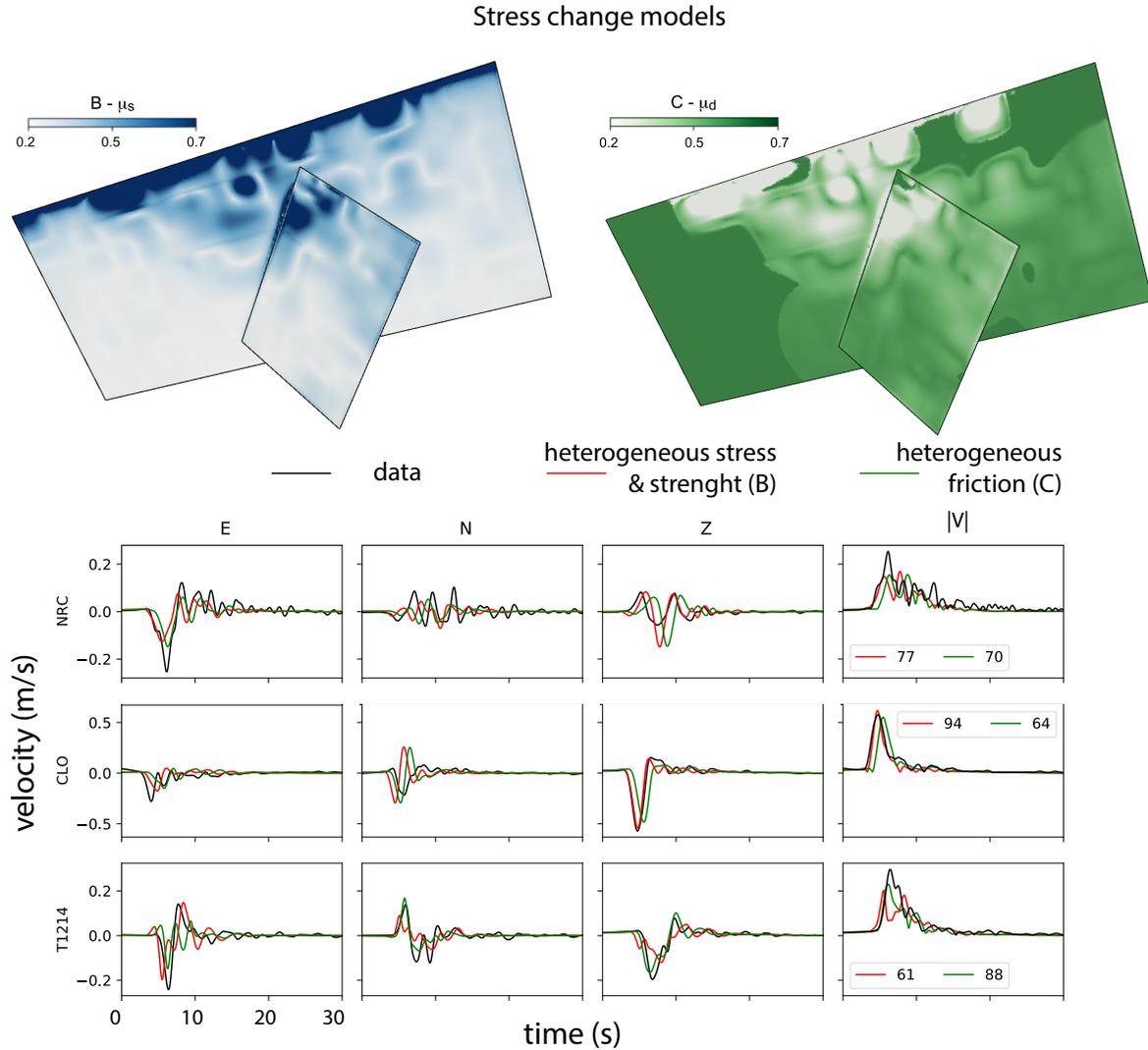


Figure 9: Top: distribution of static (left) and dynamic (right) friction parameters for exemplary models of Family (B) and (C), respectively, obtained with the stress change procedure (Section 4.2.2). Bottom: comparison of synthetic velocity waveforms (red, green for Family (B) and (C) models, respectively) obtained with the stress change procedure with observation (black) at selected stations. Numbers in the fourth column represent goodness of fit for each station and model. Station locations are shown in figure 1. Additional waveform comparisons are shown in Figure S6 and S7.

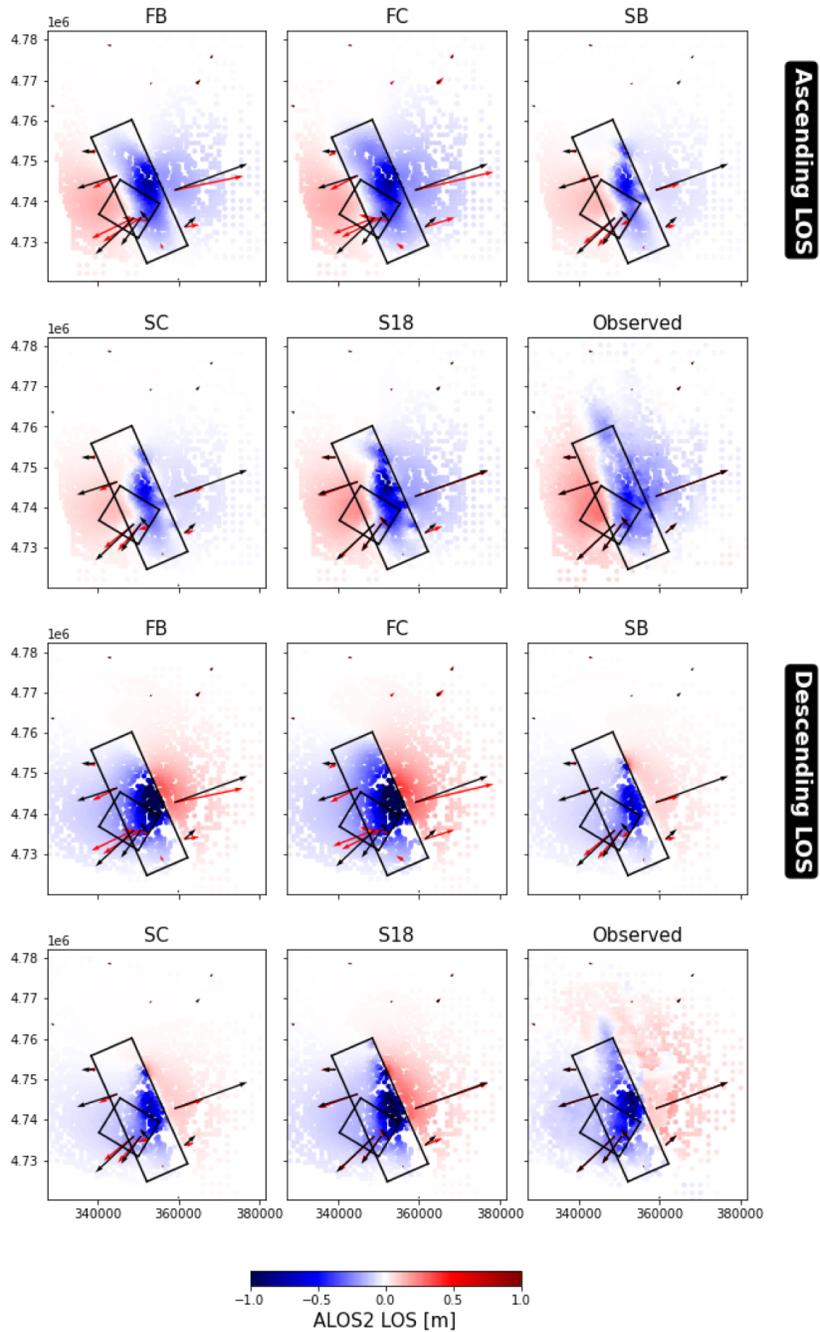


Figure 10: Measured ground displacements along line of sight for the ascending and descending ALOS2 InSAR data (Cheloni et al., 2017) compared with synthetics of all four presented dynamic models and of the original “S18” kinematic model. Each panel reports also the observed ground displacements at GPS stations (black arrows) and the synthetics the corresponding model (colored arrows). Geographical coordinates are expressed in UTM (zone 33)

893 **7. Supplementary material**

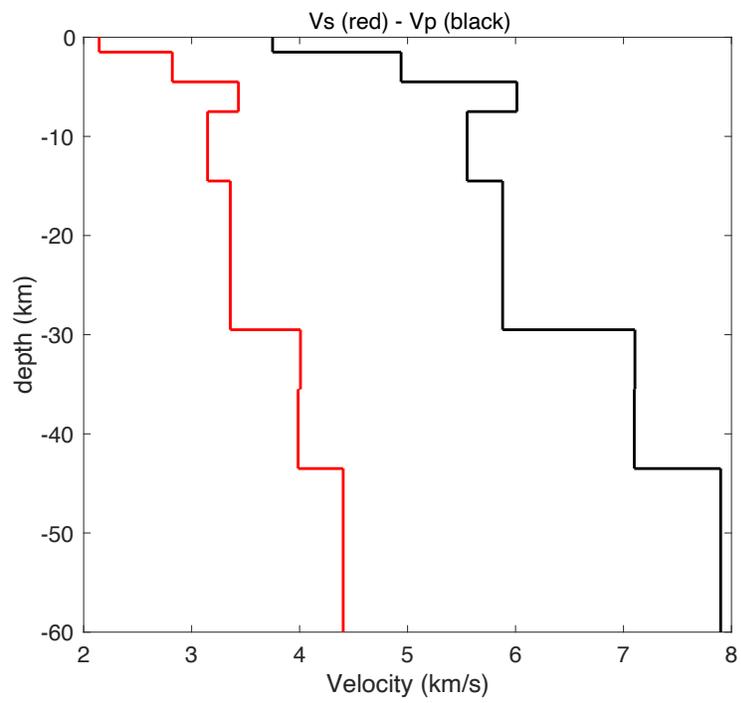


Figure S1: Velocity model by (Herrmann et al., 2011), adopted in this study

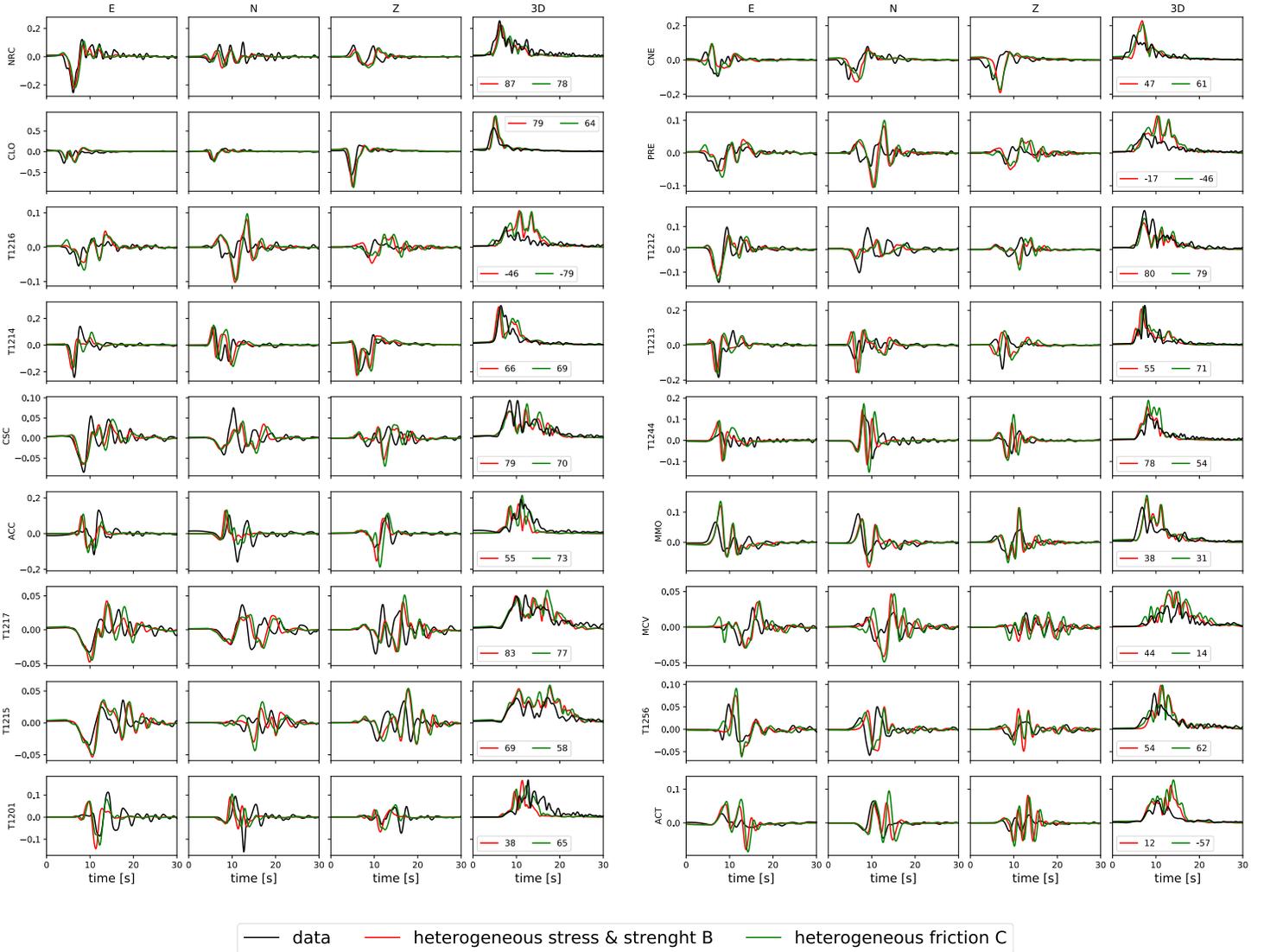


Figure S2: Comparison of synthetic strong-motion velocity waveforms (red and green for Family (B) and (C) models, respectively) at all stations inferred by assuming stress drop proportional to slip with observation (black). The variance reduction (VR) for both model are: $VR_B = 55.5$ and $VR_C = 49.7$ (1/2)

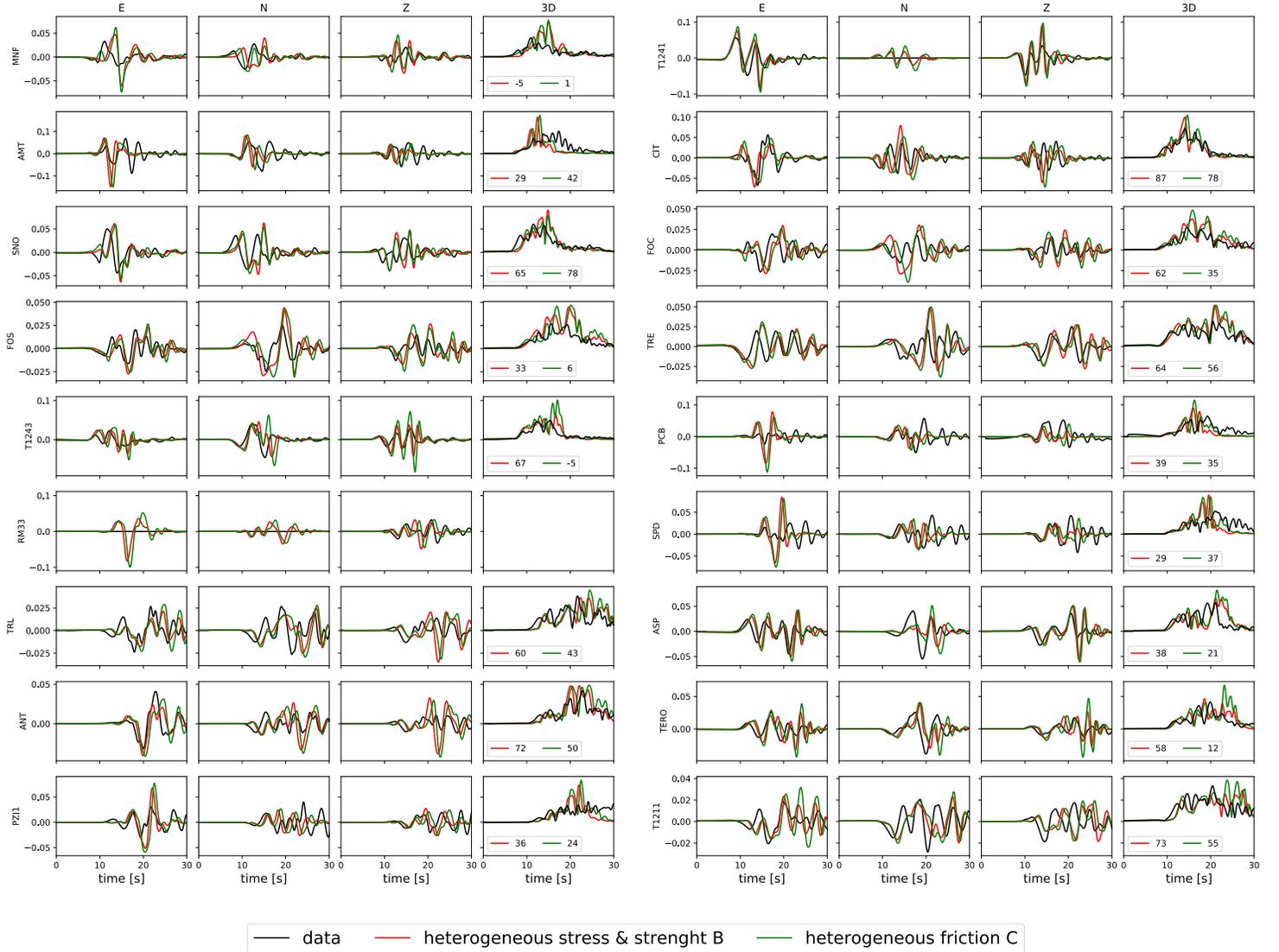


Figure S3: Comparison of synthetic strong-motion velocity waveforms (red and green for Family (B) and (C) models, respectively) at all stations inferred by assuming stress drop proportional to slip with observation (black). The variance reduction (VR) for both models are: $VR_B = 55.5$ and $VR_C = 49.7$ (2/2)

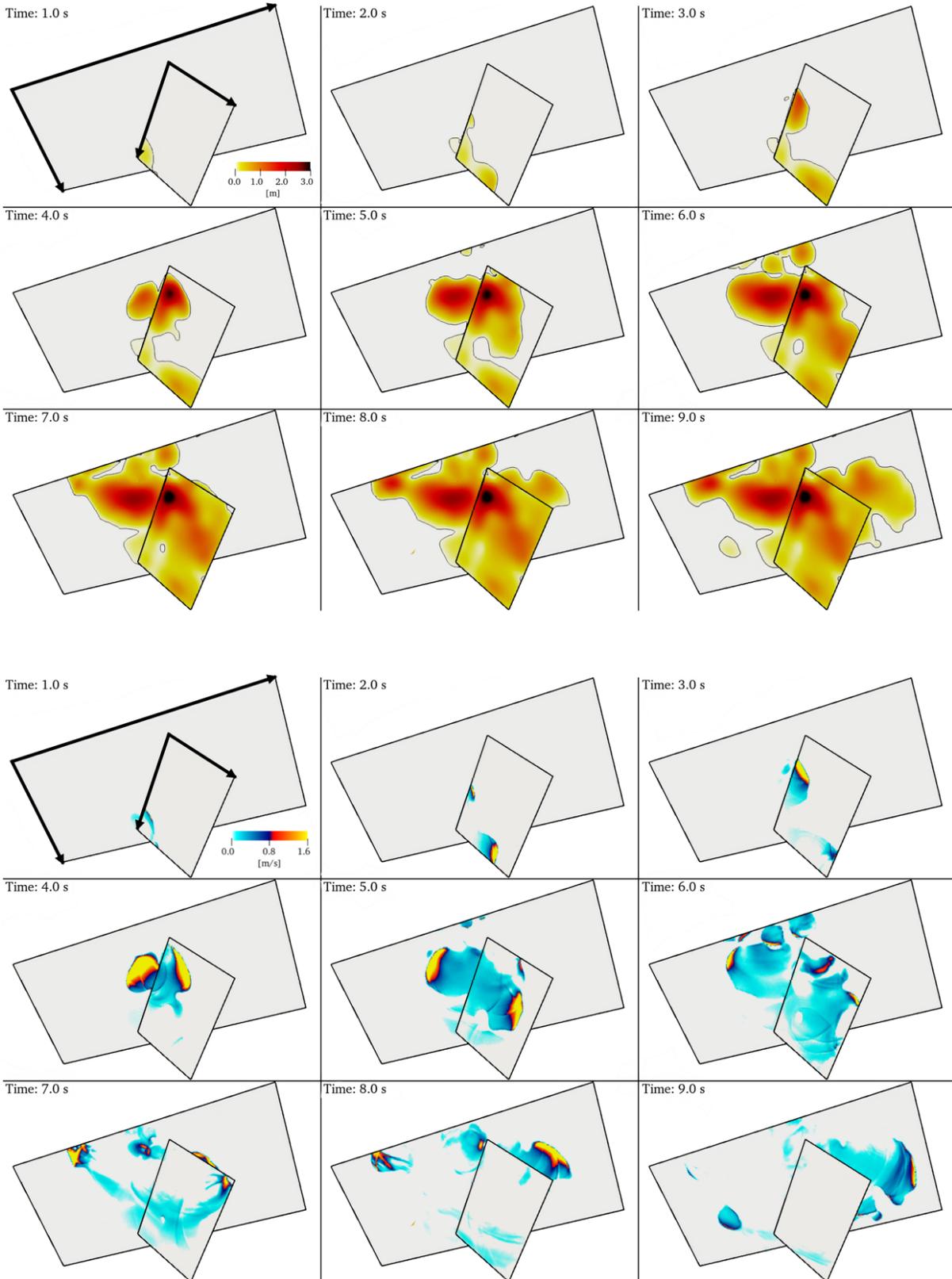


Figure S4: Dynamics of the exemplary model belonging to Family (B) based on the stress change procedure. Snapshots, every one second, of slip (m, top) and slip rate (m/s, bottom).

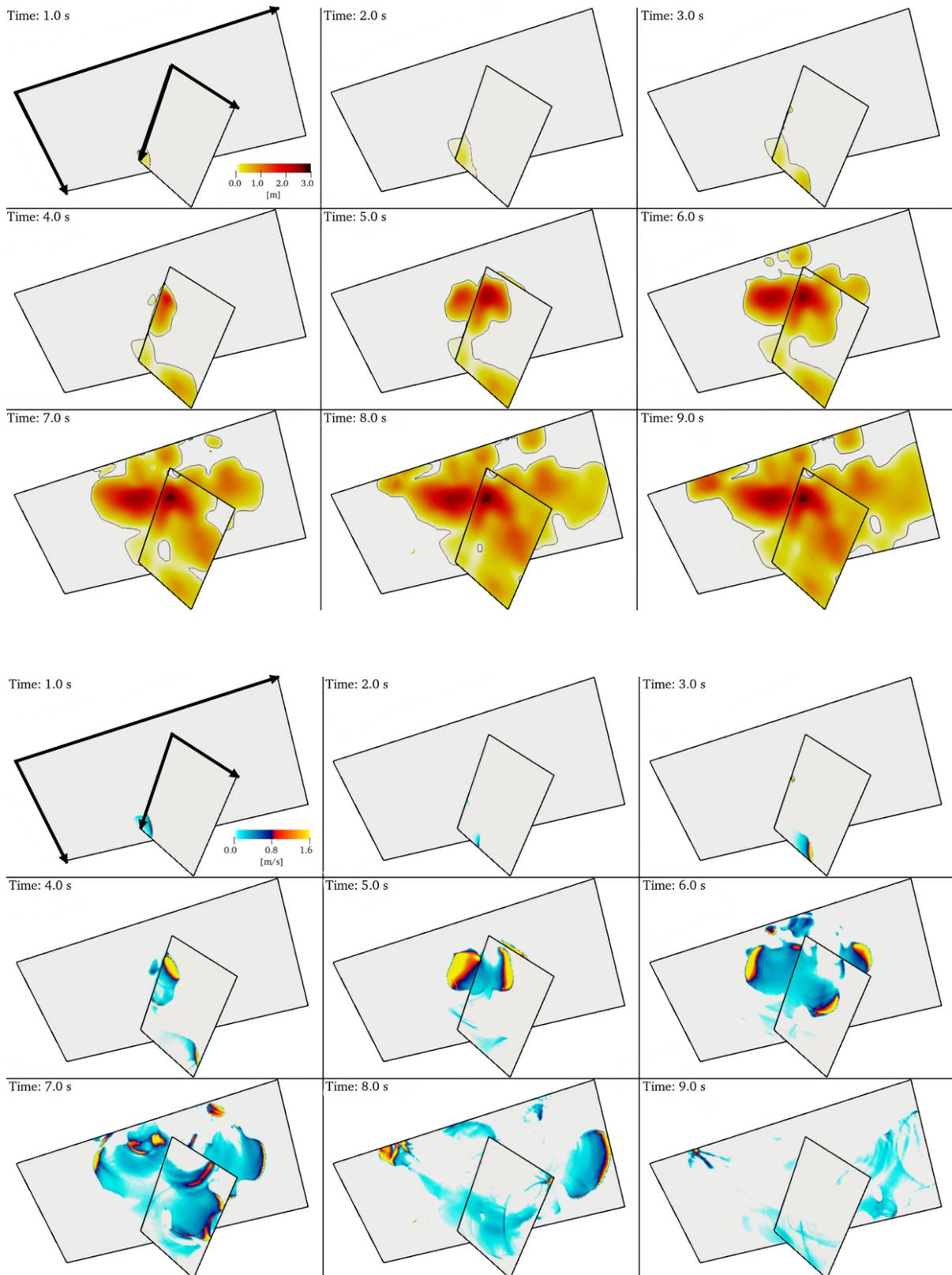


Figure S5: Dynamics of the exemplary model belonging to Family (C) based on the stress change procedure. Snapshots, every one second, of slip (m, top) and slip rate (m/s, bottom).

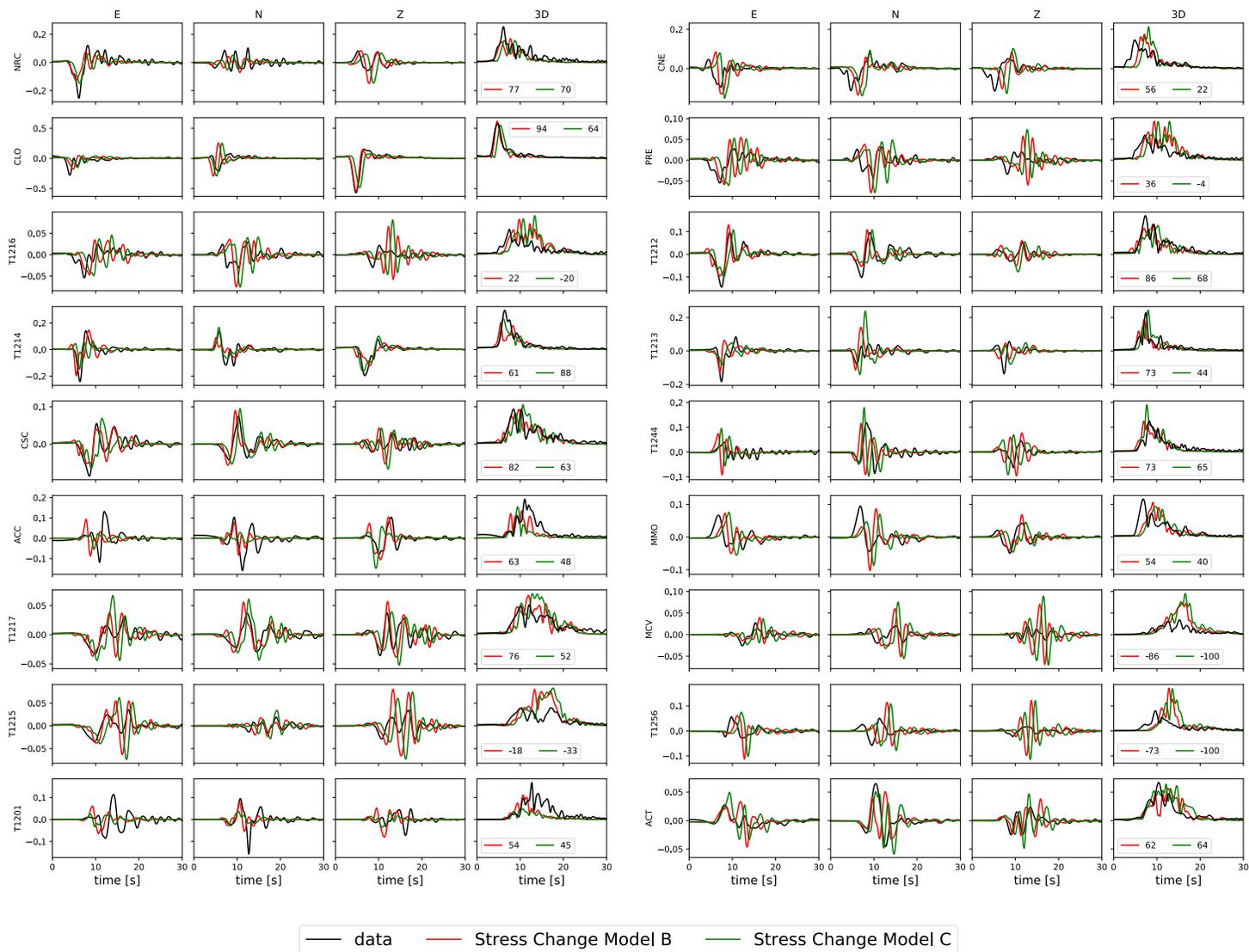


Figure S6: Comparison of synthetic strong-motion velocity waveforms (red and green for Family (B) and (C) models, respectively) at all stations derived from the stress change procedure with observation (black). The variance reduction (VR) for both models are: $VR_B = 58.0$ and $VR_C = 48.6$ (1/2)

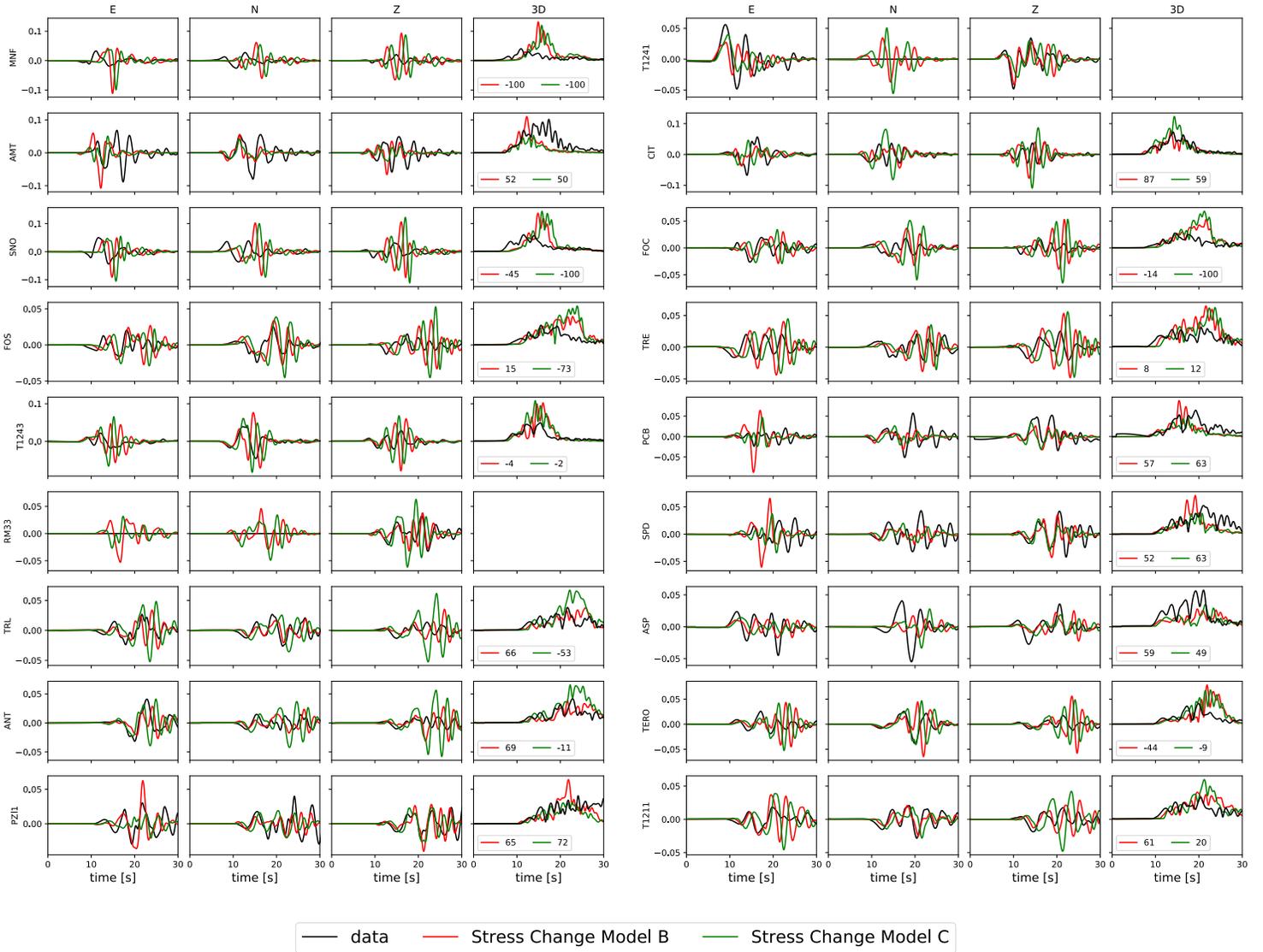


Figure S7: Comparison of synthetic strong-motion velocity waveforms (red and green for Family (B) and (C) models, respectively) at all stations derived from the stress change approach with observations (black). The variance reduction (VR) for both models are: $VR_B = 58.0$ and $VR_C = 48.6$ (2/2)