Detecting long-lasting transients of earthquake activity on a fault system by monitoring apparent stress, ground motion and clustering

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Damaging earthquakes result from the evolution of stress in the brittle upper-crust, but the understanding of the mechanics of faulting cannot be achieved by only studying the large ones, which are rare. Considering a fault as a complex system, microearthquakes allow to set a benchmark in the system evolution. Here, we investigate the possibility to detect when a fault system starts deviating from a predefined benchmark behavior by monitoring the temporal and spatial variability of different micro-and-small magnitude earthquakes properties. We follow the temporal evolution of the apparent stress and of the event-specific residuals of ground shaking. Temporal and spatial clustering properties of microearthquakes are monitored as well. We focus on a fault system located in Southern Italy, where the Mw 6.9 Irpinia earthquake occurred in 1980. Following the temporal evolution of earthquakes parameters and their time-space distribution, we can identify two long-lasting phases in the seismicity patterns that are likely related to high pressure fluids in the shallow crust, which were otherwise impossible to decipher. Monitoring temporal and spatial variability of micro-to-small earthquakes source parameters at near fault observatories can have high potential as tool for providing us with new understanding of how the machine generating large earthquakes works.
quality micro-earthquake recordings, as shown by previous studies where source parameters estimates (e.g., $M_0$, corner frequency $f_c$, stress drop $\Delta \sigma$, radiation efficiency $\eta_{SW}$) for hundreds of earthquakes\(^9\) were computed and a few microearthquake sequences\(^10\) identified.

The wealth of information carried by the nowadays available very large numbers of micro-to-small earthquakes is also pushing the seismological community to look for novel data analysis strategies with the aim to figure out the crustal strength spatial variation and its time evolution\(^11–14\). In this study, we analyze the temporal evolution of micro-to-small magnitude earthquakes at INFO, in southern Italy, in terms of apparent stress ($\tau_a$)\(^15\), between-event residuals ($\delta_{Be}$)\(^16\) for the peak ground velocity (PGV) and applying a cluster analysis\(^17,18\).

\(\tau_a\) was introduced in the 1960s\(^15\) and it is defined as the ratio between radiated energy ($E_S$) and seismic moment ($M_0$) multiplied by the rigidity of the source medium ($\mu$)\(^19,20\) [i.e., $\tau_a = \frac{E_S}{\mu M_0}$]. Therefore, $\tau_a$ measures the amount of seismic energy radiated per unit fault area ($A$) and unit fault slip ($D$)\(^21\) [i.e., $\tau_a = \frac{E_S}{\mu A D}$]. When interpreted also as the difference between the average stress loading ($\tau$) and the average stress that resist fault slip ($\tau_R$), $\tau_a$ represents the stress amount causing seismic energy radiation\(^22\). High $\tau_a$ values are associated with either high average stress level or lower frictional dynamic stress (i.e., conditions which may lead to material failure or facilitate seismic slip, respectively\(^23\)); and, hence, $\tau_a$ can be used as a proxy for the crustal strength. Several recent studies used micro-to-moderate seismicity to compute $\tau_a$ for different purposes such as: relating $\tau_a$ to the stress causing earthquake fault slip\(^22\); for the time-dependent seismic hazard assessment and earthquake forecast in China\(^24\); for the identification of the migrating stress front in rock mass during mining activities\(^25,26\). Moreover,

Figure 1. (a) Locations of the earthquakes considered in this study (yellow stars) and those recorded by ISNet but not considered (gray stars). ISNet seismic stations (white triangles). CO$_2$ degassing site Mefite d’Ansanto (green square). Limits of the central sector (blue lines) and section A-A track (red line). (b) Distribution of $M_w$ for the considered events, completeness magnitude $M_c$ (dashed line). (c) $M_0$ and $M_w$ versus time. (d) Apparent stress ± 1 standard deviations shown as vertical bars. The map was made using MATLAB (R2016b; 9.1.0.441655; https://it.mathworks.com/, last accessed June 2019).
the variability in the source parameters also affects the event-specific variability of ground shaking. In Probabilistic Seismic Hazard Assessment, the ground shaking generated by a specific earthquake is predicted using Ground Motion Prediction Equations (GMPEs)\(^{27}\). Since most GMPEs model the source scaling considering only Mw, the ground shaking variability observed for earthquakes having the same Mw but showing differences in other source parameters (e.g., rupture velocity and stress drop, $\Delta \tau$) is mapped to the between-event residual distribution $\delta\text{Be}$\(^{16}\) (i.e., which is the average discrepancy for a given event of the observations at different stations with respect to the median prediction). Recent studies\(^{28-31}\) highlighted the correlation between $\Delta \tau$ variability and $\delta\text{Be}$ computed at high frequencies. Therefore, along with $\tau_a$, we also monitor the temporal variability of $\delta\text{Be}$.

Beside monitoring $\tau_a$ and $\delta\text{Be}$, we studied the observed seismicity applying a cluster analysis. Clustering analyses\(^{32,33}\) are considered essential elements for identifying the existence of different event populations (e.g., foreshocks, aftershocks, swarms, etc.), which can allow to better understand the seismic stress redistribution into the crust and the earthquake physics. The nearest-neighbor approach\(^{37,38}\) exploits the earthquake time-space-magnitude information to derive a generalized distance $\eta$ between pairs of earthquakes, which can be used to isolate clusters from the background seismicity.

We considered about 2300 local earthquakes recorded by INFO over the last ten years. Inspired by previous works\(^{39,40}\), we mapped the heterogeneity distribution of $\tau_a$ over the Irpinia fault system. The source parameters $M_0$ and $E_0$ used to derive $\tau_a$ are obtained by applying the same approach used for the 2016–2017 Central Italy seismic\(^{41,42}\). Then, focusing on the more densely instrumented central sector of the Irpinia fault (i.e., a segment 60 km long), where the Mw 6.9, 1980 earthquake occurred, we studied the temporal evolution of $\tau_a$, $\delta\text{Be}$ and $\eta$ considering small and moderate magnitude earthquakes and looking for trends which are coherent among the parameters. To achieve this goal, we compared the cumulative of the considered parameters with the cumulative of a synthetic time series constructed assuming that the monitored parameters are constant for all events and equal to the median of their distributions. It is worth noting that, in order to define our monitoring strategy, we applied a retrospective approach; the prospective validation of the proposed methodology is ongoing considering the real-time data acquired at INFO. In the following, we refer to the synthetic time series as ‘reference model’, and they represent our benchmarks for the background fault activity. We therefore studied the difference between the cumulative experimental and synthetic series over time, which we define as residual apparent stress, $\Delta \tau_a$, and residual nearest-neighbor distance, $\Delta \eta$. Conversely, being $\delta\text{Be}$ a parameter with a zero-average distribution, we simply considered its cumulative in time, $\Sigma \delta\text{Be}$. This approach allowed us to detect variations in all three studied parameters lasting few months and culminating with two Mw 3.5 earthquakes (i.e., occurred in the proximity of the Mw 6.9, 1980 earthquake’s hypocenter).

Results

$\tau_a$ spatial distribution over the Irpinia fault system. We examined 2336 earthquakes with Mw between 0.3 and 3.8 occurred within a buffer of +/- 30 km along the Irpinia fault system using hypocenter locations of the network bulletin from 2008 to 2018 (Fig. 1a–c). Uncertainties in the events locations are mostly within 1 km both horizontally and vertically (i.e., the median error in location is ~0.5 km, Fig. S1). We assessed $E_a$ and $M_0$ by extracting from the S-wave time window proxies for these two source parameters and correcting them for attenuation along the path\(^{43,53}\) (see "Method" section). So far, applications where $E_a$ is estimated for small magnitude earthquakes has been rather limited, mainly due to difficulties in its estimation from band-limited recordings\(^{36}\). $E_a$ can be computed either in the spectral or time domain\(^{33,37}\). Robust $E_a$ estimates can be achieved only when recordings at short distances from dense networks of high quality instruments deployed with good azimuthal coverage around the source are available\(^{4}\).

The linear scaling between the logarithm of $E_a$ and $M_0$ holds over 5 orders of magnitude for seismic moment (Fig. S2), and the scaled energy (i.e., $E_a/M_0$) variations with $M_0$ well compares to previous results obtained for several tectonic areas\(^{44}\), and in particular for Central Italy\(^{45}\) (Fig. S3). We derived $\tau_a$ accounting for the depth variations of the crustal shear modulus $\mu$, according to the 1D velocity model calibrated for the investigated area\(^{39}\). Figure 1d shows that $\tau_a$ varies between $10^{-3}$ MPa and about 2 MPa over the decade of observations without, at a first glance, any clear trend. We analyzed the spatial distribution of $\tau_a$ along a section parallel to the Irpinia fault system dividing the fault into a $14 \times 7$ array of subfaults having a length of 10 km along the strike and a down-dip width of 5 km. Within each subfault, we stack the $\tau_a$ in logarithmic scale considering only events with Mw < 3.0, we compute the average $\tau_a$ value, and we assign it to the center of the subfault patch. Figure 2 shows that spatial interpolation of $\tau_a$ is characterized by large lateral and depth variations. $\tau_a$ value smaller than about 0.06 MPa (i.e. the average $\tau_a$) are confined to depths between 10 km and 15 km in the central fault sector (i.e., please note that the Mw 6.9, 1980 Irpinia earthquake nucleated at ~12 km), and to depths shallower than about 5 km in both the northern and southern fault sectors.

The distribution of hypocenters of events with Mw larger than 3.0 is also peculiar (i.e., please note that they are colored per their $\tau_a$ value in Fig. 2, and we remind they were not used to create the $\tau_a$ section). In the northern and southern portion of the fault, they are mostly located within areas with large $\tau_a$, whereas in the central sector, they are mostly distributed along the $\tau_a = 0.06$ MPa boundary.

$\tau_a$ temporal evolution. Several previous studies\(^{46,47}\) showed that the northern and southern Irpinia sectors are characterized by a complex tectonic, as resulting from the overlap of a strike-slip regime (i.e., related to the Benevento and Potenza faults to the north and to the south, respectively) with the extensional regime dominating the central sector of the Irpinia fault. Since such tectonic complexities would require a specific interpretation of the results for each sector, we restrict our interest to the more densely instrumented central sector of the Irpinia fault for studying the time evolution of $\tau_a$. This sector extends for 60 km, and it includes the area where the Mw 6.9, 1980 earthquake occurred. Furthermore, aiming to avoid possible biases due to variation in the number of
stations during the last decade, we considered only earthquakes above the completeness magnitude at INFO (i.e., Mc = 1.6), thus in total, 928 earthquakes are considered.

Figure (3a,b) shows the histogram of the $\tau_a$ values, their cumulative distribution (i.e., $\Sigma \tau_a$obs) and the cumulative distribution of the reference model (i.e., $\Sigma \tau_a$syn, the model for constant apparent stress). We investigated the deviations of $\Sigma \tau_a$obs from $\Sigma \tau_a$syn (Fig. 3b) by computing their residuals $\Delta \tau_a = \Sigma \tau_a$obs - $\Sigma \tau_a$syn. Events are represented in the $\Delta \tau_a$ curve (Fig. 4a) with different symbols: green dots for $M_w < 3.0$; white squares for $M_w > 3.0$. Considering the average magnitude uncertainty for the mainshocks, clustered subpopulations of events could be identified. To this aim, we applied the uncertainty associated to $\Delta \tau_a$ through bootstrap analysis using 2000 realizations. This means that, for each realization, we randomly resample the $\tau_a$ population, we compute the median value and the two cumulative distributions from a reference model characterized by apparent stress constant for all events. Since the variability affecting source parameters leaves an imprint in the variability of the ground shaking generated at the same distance and a half and two years and a half, respectively. After these two cycles, since 2014 we observe a rather slow and regular increase in $\Delta \tau_a$ lasting for about two years. The increase of $\Delta \tau_a$ before the two Mw 3.5 lasted about 8 and half months before the first event, and about 4 and half months before the second one. The decreasing phases lasted one year and a half and two years and half, respectively. As for $\tau_a$, no clear temporal trends in $\delta$Be are observed. Therefore, we studied the temporal evolution of its cumulative, $\Sigma \delta$Be (Fig. 4b), interpreted in comparison with $\Delta \tau_a$. The uncertainty associated to $\Sigma \delta$Be is computed by bootstrapping the dataset considering 2000 replications, as for $\Delta \tau_a$. Interestingly, locally $\Sigma \delta$Be shows a (positive) trend over the two periods culminating with the two Mw 3.5 earthquakes, where also $\Delta \tau_a$ increase, strengthening the hypothesis that a background process influences the dynamical characteristics of microseismicity rupture process in the investigated fault system. Or, in other words, anomalies in the temporal distribution of $\tau_a$ leave an imprint also in the ground shaking variability.

$\eta$ temporal evolution. We analyzed the statistical features of the recorded earthquakes to verify if, beyond the background seismicity, clustered subpopulations of events could be identified. To this aim, we applied the
nearest-neighbor approach\textsuperscript{17,18} and computed the generalized distance between pairs of earthquakes, $\eta$ (see Method).

Figure 3e shows the rescaled time and space components ($T$, $R$) of $\eta$, whereas each point in the plot represents an event. The cloud of points is distributed mainly between $\eta = -1$ and $-3$. The distribution of $\eta$ is unimodal (with median value $\eta$ equal to $-2.7$; Fig. 3f), and we cannot identify any clear cluster from the background seismicity\textsuperscript{18}. Following the analyses done for $\Delta \tau_a$ and $\Sigma \delta_{Be}$, which suggested the presence of cyclic variations in the earthquakes source properties, we investigated the temporal evolution of $\eta$. Figure 3g shows the cumulative of $\eta$ compared to the reference cumulative constructed assuming $\eta$ to be constant and equal to its median value, $\eta^\circ$, whereas Fig. 4c shows the difference of the two cumulative functions $\Delta \eta = \Sigma \eta - \Sigma \eta^\circ$ versus time. The uncertainty associated to $\Delta \eta$ is again derived by a bootstrap analysis\textsuperscript{43} using 2000 realizations.

It is worth noting the presence of significant (negative) trends in $\Delta \eta$ before the occurrence of the two Mw 3.5 (Fig. 4c), which seems well related in time with variations of $\Delta \tau_a$ and $\Sigma \delta_{Be}$. However, while $\Delta \tau_a$ is related to dynamic properties of the earthquakes, for $\Delta \eta$ we still miss a physical interpretation. By construction, the cumulative series $\Sigma \eta$ and $\Sigma \eta^\circ$ have the same intercurrence time $t$ (See Eqs 5–7 in section 'Method'). Furthermore, changes in the earthquake distance $r$ (such as to justify a decrease in $\Delta \eta$) should appear as multi-modal distribution of $\eta$ (i.e., a mode for each cluster); while, we see for $\eta$ an unimodal distribution. Hence, both $t$ and $r$ can in first approximation be neglected. Looking at Eqs (5) and (6), two parameters that might play a role on the variations of $\Delta \eta$ are the fractal dimension $d$ and the parameter of the Gutenberg-Richter parameter $b$. Since a correlation between $d$ and the $b$ is reported in literature e.g.\textsuperscript{45,46}, we focused only on the relation between $d$ and $\Delta \eta$. To verify if a relation between changes in $d$ and $\Delta \eta$ does exist, we used the approach proposed by Grassberger and Procaccia (1983)\textsuperscript{47} to estimate $d$. Following a previous study\textsuperscript{45}, we used sliding windows of 200 events, shifted by...

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**Figure 3.** (a) Distribution of $\tau_a$ and median value ($\tau^\circ_a$, dashed black line). (b) Cumulative of $\tau_a$ (green) and of $\tau^\circ_a$ (black). (c) Distribution of the between-event residuals ($\delta_{Be}$) computed for PGV and median value (dashed black line). (d) $\delta_{Be}$ versus time. (e) 2D distribution of the rescaled time $T$ and distance $R$. The dotted lines correspond to different values of the nearest-neighbor distance $\eta$. (f) Distribution of $\eta$ and median value $\eta^\circ$ (dashed black line). (g) Cumulative of $\eta$ (yellow) and of $\eta^\circ$ (black).
one event per time. Figure (S5) shows the evolution of the fractal dimension of hypocenters with time. Despite, in our opinion, the analysis for estimating $d$ is less robust than the procedure for computing $\eta$, mainly because the former implies a best-fit procedure, we consider remarkable the good correlation between results from the two analyses (i.e., $d$ and $\Delta\eta$). Indeed, also $d$ shows a decrease starting from the 2009, before the first Mw 3.5 earthquakes, and a second one during the 2011, before the second Mw 3.5 (Fig. S5).

To highlight the association between changes in $\Delta\tau_a$, $\Delta\eta$ and $\delta$Be, we performed a correlation analysis. Figure (S6a) shows the comparison of the normalized $\Delta\tau_a$ and $\Delta\eta$ time series together with splines used to guarantee a homogeneous sampling and to emphasize low frequency trend. After the linear trend removal (Fig. S6b), the splines are used to compute the Spearman’s correlation coefficient on moving windows of 50 points and 90% of overlap. The distribution of Spearman’s correlation coefficients shown in Fig. S6c, has a maximum at $-1$, indicating that $\Delta\tau_a$ and $\Delta\eta$ are strongly anti-correlated. A similar analysis was carried out also for $\Delta\tau_a$ and $\delta$Be (Fig. S7). In this case, the Spearman’s correlation coefficients indicate a strong positive correlation (Fig. S7c).

**Discussion**

The seismic velocity and attenuation structure in the Irpinia region is well-known. This region is affected by crustal extension, generating a seismicity characterized by small to moderate magnitude events marked by strong earthquakes, as the Mw 6.9, 1980 Irpinia earthquake. The seismic active rock volume consists of the Apulian Platform carbonates and its basement. The 3-D images of the Irpinia fault system in P and S waves velocity indicate a highly fractured rock volume rich of fluid in the uppermost 15 km, with background seismicity.

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**Figure 4.** (a) Temporal variability of the residual apparent stress ($\Delta\tau_a$). Events with magnitude below Mw 3.0 (circles), $3.0 \leq M_w < 3.5$ (white square with red contour), and those with $M_w = 3.5$ (white stars with red contour). The light green area indicates the median $+/-$ one standard deviation range as resulting from bootstrap analysis. (b) The same as (a), but for $\Sigma\delta$Be. (c) The same as (a), but for $\Delta\eta$s. (d) Distribution of Mw for the considered events versus time.
driven by pore pressure changes in fluid-filled cracks. Further important insights on the underground structure were provided by the 3-D attenuation images\(^5\), showing Q\(_s\) anomalies well correlated to the Mw 6.9, 1980 Irpinia earthquake nucleation area. Moreover, Q\(_s\) images highlight a strong lateral variation associated with the contrast between the thick, slower and high attenuation Miocene basin sediments and the Apulian Platform carbonates\(^5\).

The spatial distribution of \(\tau\) well complement the velocity and attenuation images. Low \(\tau\) values in the shallower portion (i.e. within ~15 km) of the central sector in the Irpinia fault system correspond to the highly fractured carbonate rocks. The increase of \(\tau\) at about 15 km, a depth compatible with the nucleation area of the Mw 6.9, 1980 Irpinia earthquake, likely marks the contrast between carbonates with the underlying basement. In both the northern and southern portions of the analyzed fault system, high \(\tau\) values are observed at shallower depths (i.e., ~5 km), confirming the high heterogeneity and complexity of the Irpinia region. In these external sectors, indeed, focal mechanisms\(^40\–42\) indicate the presence of faults with different kinematic and suggest the intersection between the Irpinia fault system and the Benevento strike-slip zone at north and the Potenza strike-slip zone at south. In support to this hypothesis, the 2012 Benevento and the 1990–1991 Potenza seismic sequences (i.e., respectively, in the northern and southern sectors of the investigated area) present strong similarities: have similar kinematics, are both crustal but have occurred at relatively high depth (~between 15 and 25 km) well below the Apulian carbonate units that represents the upper crust level where the Mw 6.9, 1980 earthquakes hypocenter is located\(^43\–45\).

The temporal evolution of \(\Delta\tau\) suggested the presence of cyclic variations in the earthquakes source properties and stimulated our curiosity on its physical interpretation. As discussed in previous studies, \(\tau\) can be defined as the product between the seismic radiation efficiency, \(\rho\), and the shear stress causing fault slip, \(\tau_s\) (i.e., \(\tau = \rho \cdot \tau_s\)). In the Irpinia fault region \(\rho\) is low and, in general, corresponds to a small fraction of the energy is spent for creating new fractures\(^4\). Indeed, the Savage and Wood efficiency\(^52\) \(\rho_{SW}\) for Irpinia microearthquakes (i.e., please note that \(\rho_{SW}\) is one-half of \(\rho\) and it represents the shear stress), \(\Delta \tau\), commonly used by seismologists, \(\tau_s = \rho_{SW} \cdot (\Delta \sigma)\), remains almost constant over 4 orders of magnitude\(^4\) (i.e., with median \(\rho_{SW} = 0.06\)). Therefore, we consider reasonable to assume \(\rho\) being constant. Hence, \(\Delta\tau\) can be approximated as follows:

\[
\Delta\tau = \sum_{i}^{n}(\tau_{ij} - \tau^m) = \sum_{i}^{n}(\rho \cdot \tau_s - \rho \cdot \bar{\tau}) \approx \Delta\tau
\]

where the cumulative for the radiation efficiency is approximated with the cumulative of the average radiation efficiency \(\sum_{i} \rho = \sum_{i} \overline{\rho}\). Under this condition, \(\Delta\tau\) can be considered a proxy for \(\Delta\tau\), the fault strength loss. Studying the temporal evolution of \(\Delta\tau\) is akin to study the temporal evolution of the Gutenberg-Richter law b parameter\(^53\). Indeed, while single events provide information of the stress acting in specific location of the fault plane (i.e., for instance by their stress drop), the b parameter captures their collective behavior and is considered well representative of the differential stress acting on the whole fault\(^54\). Similarly, \(\tau\) for single events represents the stress causing seismic energy radiation, but performing the temporal analysis over \(\Delta\tau\), we highlight the collective behavior of the earthquakes, and we provide a first order estimate of the crustal strength for the central sector of the Irpinia fault system. From our analyses, \(\Delta\tau\) increases between about ~1.5 MPa and ~4 MPa observed in the periods preceding the Mw 3.5. Previous studies analyzing microearthquakes in the Irpinia area estimated a median stress drop varying from 1.4 MPa\(^a\) to 3.9 MPa\(^b\), well in agreement with \(\Delta\sigma = 3.5\) MPa estimated for the Mw 6.9, 1980 Irpinia earthquake\(^5\).

Since the inter-event residuals \(\delta\)Be absorb event-specific component of the ground motion variability not controlled by the seismic moment, the consistency between the trends observed for these two parameters during periods lasting a few months before the two Mw 3.5 earthquakes lead us to hypothesize the existence of background processes influencing the dynamical characteristics of microseismicity rupture process. One further, independent, piece of the puzzle is the decrease of \(\Delta\eta\) (Fig. 4c), which we have shown likely be indicative of a decrease in the fractal dimension \(d\) (Fig. 4b). Empirically, values of \(d\) close to 2, as for the first Mw 3.5 earthquake, or slightly below it (i.e., ~1.5 as for the second Mw 3.5), suggest that the hypocentres of the earthquakes tend to be progressively distributed over a two-dimensional fault plane before the mainshocks\(^5\). Therefore, we can interpret the increase of \(\Delta\tau\) and \(\delta\)Be, combined with the reduction of \(\Delta\eta\), as the result of a progressive activation of seismic asperities in the fault plane that have higher \(\Delta\sigma\) than the surrounding area. The time with which this process take place (i.e., few months) suggest the presence of a slow aseismic force. On the nature of this force, we can provide only a guess. Upscaling procedures based on velocity and attenuation\(^4\) lead to estimate a range of porosity in carbonates (i.e., 4 to 5%) and the fluid composition including brine-CO\(_2\) and/or CH\(_4\)-CO\(_2\). The CO\(_2\) earth degassing in central and southern Italy is largely documented\(^47\–50\). In particular, the Mefite d’Ansanto located in the Irpinia region (Fig. 1a) is considered one of the largest natural emission environments of low temperature CO\(_2\) rich gas ever measured\(^49\). The spatial distribution of the CO\(_2\) degassing areas and seismicity in the Apennines provides an interpretative scheme for the seismicity generation\(^5\). The large upwelling of mantle fluids in the Tyrrhenian hinterland infillate upward through interconnected fractures generated by extensional tectonic regime and generate high CO\(_2\) domains in the shallow crust. Whereas fractures are well interconnected, degassing at the surface is observed. On the contrary, in the Apennines, the arrangements of deeper thrust and low-angle normal faults create verging structure where CO\(_2\) may accumulate and generate overpressurized reservoirs at depths ranging from few kilometres to about 15 km\(^57\). In the case of the 2009 L’Aquila earthquake in Italy (Mw 6.3, central Apennines), strong evidences for the presence of overpressurised fluids near the foreshocks and mainshock were provided\(^51\), suggesting that they contributed to the mainshock rupture. Similar conclusions were proposed also for two earthquakes of magnitudes 5.7 and 6 occurred in 1997 in the northern Apennines\(^5\). In this framework, we can interpret our results as related to CO\(_2\)-brine accumulated and sealed within the rock volume. When the pressure in the deep CO\(_2\)-brine reservoir increases, the seismicity tends to progressively be distributed over two-dimensional fault planes; the process culminates with Mw \(\geq 3.5\) earthquakes, which evidently allows for the creation of way outs large enough for the fluids migration towards
the surface. The sequences are then followed by a slow decrease in the stress level (i.e., $\Delta \tau$ decreases) until a new stress loading cycle starts.

In future, it will be certainly important to focus on areas where large magnitude events have occurred, to investigate if a relation between the rate and amplitude of the $\Delta \tau$ anomalies and the earthquake magnitude does exist. In conclusion, in our vision for NFOs, we think that studying the microseismicity and small-to-moderate earthquakes through a wide set of parameters (e.g., including also focal mechanisms$^{11}$ and medium changes$^{59}$) can have high potential as tool for monitoring the evolution of seismic sequences, identifying the preparatory phase of large earthquakes and providing us with new understanding of how does work the machine generating earthquakes.

**Methods**

**Data processing.** We adopted a procedure proposed for the Central Italy seismic sequence$^{34,35}$, where $M_0$ and $E_0$ are derived from two proxies measured on direct S-waves: the peak displacement (PDS) and the cumulative squared velocity (IV2s), respectively. Processing was carried out as follows: P-waves have been hand-picked on three-component ground velocity recordings sampled at 125 Hz, while S-waves arrival has been estimated considering location and a 1D velocity model optimized for the Irpinia area; Butterworth band pass pre-deconvolution filter with high and low pass corner frequencies at 0.5 Hz and 40 Hz, respectively; Instrumental correction and computation of acceleration, velocity and displacement. The parameters PDS and IV2s are computed considering a pre-event noise window of the same length as the signal window. Finally, the values of the three components of ground motion are averaged for both PDS and IV2s. Furthermore, a subset of data was extracted with the following criteria: hypocentral distance smaller than 60 km; events recorded by a minimum of 3 stations; the sum of SNR for the three components $\geq 200$.

**Seismic moment and radiated energy estimation.** From a database of 2336 earthquakes, source parameters (i.e., $M_0$, $f_c$, $\Delta \sigma$ and $\eta_{(S)}$) were available for 216 events with magnitude ranging between $M_0$, 1.2 and $M_0$, 3.2. The dataset of 216 earthquakes was thus used to link the experimental IV2s and PDS to well-constrained $M_0$, values and theoretical $E_0$ (i.e., obtained from source parameters and the spectral integration$^{37}$ of the squared theoretical Brune’s velocity spectrum for S-waves$^{60}$) and to calibrate empirical attenuation models. Having assumed the Brune’s seismic source model, our $E_0$ and $M_0$ estimates are model dependent. We assumed a linear model, where the attenuation of average values of the proxy (i.e., either IV2s or PDS) as a function of distance is expressed without assuming any a-priori functional form:

$$\log(IV2s(R_{H})) = A + B \log(E_0) + w_i C_j + (1 - w_i)C_{j+1}$$  \hspace{1cm} (2)

and

$$\log(PDS(R_{H})) = D + F \log(M_0) + w_i G_j + (1 - w_i)G_{j+1}$$  \hspace{1cm} (3)

where, the hypocentral distance $R_H$ ranging between 5 km and 60 km is discretized into 12 bins ($N_{\text{bin}}$) with equal width (i.e., 5 km); the index $j = 1, \ldots, N_{\text{bin}}$ indicates the $j$-th node selected such that $R_H$ is between the distances $r_j \leq R_H < r_{j+1}$; the attenuation function is linearized between nodes $r_j$ and $r_{j+1}$ using the weights $w_i$ computed as $w_i = (r_{j+1} - R_H)/(r_{j+1} - r_j)$. The minimum distance is fixed to 5 km given the lack of recordings at shorter distances. The coefficients of Eqs (2) and (3) are determined by solving the over-determined linear system in a least-square sense. The trade-off between $A$ and $C_j$, and between $D$ and $G_j$ is constrained by the condition to be zero at 10 km. It is worth noting that changing the node constrained to zero corresponds to changes of both the $A$ and $C_j$, as well as for $D$ and $G_j$ which compensate each other, so that the coefficient B and F in Eqs (2) and (3) are unaltered. The attenuation models obtained for the Irpinia area agree with those obtained for the Central Italy seismic sequence$^{35}$ (Fig. S5). The attenuation models are reported in Table S1. The event $E_0$ and $M_0$, estimates per event, as well as the derived $\tau$, are obtained by averaging over the recording stations.

**Between event residuals, $\delta E_0$, estimation.** We performed a linear regression analysis to model the PGV scaling with distance and magnitude by searching the best-fit parameters of the following equation:

$$\log(\text{PGV}) = A + B_1 \cdot (M_0 - M_{\text{ref}}) + B_2 \cdot (M_0 - M_{\text{ref}})^2 + C \cdot \log(R),$$  \hspace{1cm} (4)

where $R$ is the hypocentral distance in km and PGV is in cm/s, and the reference magnitude $M_{\text{ref}}$ is set equal to 1. Like previous studies$^{34,35,43}$, we refer to a simple ground motion model, since we do not have the ambition to set up a GMPE for hazard calculation. Our aim is to measure the variation of the between-event variability of ground motion. The result of the regression analysis is provided in Table (S2). The between-event residuals ($\delta E_0$) are computed as the average difference, for any given event, between the PGV measured at different stations and the corresponding values predicted by the GMPE.

**Nearest-neighbor distance, $\eta_{(S)}$ analysis.** The nearest-neighbor approach$^{17,48}$ is based on the computation of the generalized distance between pairs of earthquakes, $\eta$, from the analysis of the time-space distances between pairs of earthquakes. The parameter $\eta$ is derived computing the distances in time (i.e., Rescaled Time) and space (i.e., Rescaled Distance) between an event $i$ and its parent $j$ normalized by the magnitude of the parent event:
where, $m$ is the magnitude ($M_w$), $b$ is the parameter of the Gutenberg-Richter law, $t$ is the earthquake intercur-
tence time, $r$ is the earthquake distance, and $d$ is the fractal dimension. We fixed $b = 1$ and $d = 1.6$, following Zaliapin and Ben-Zion\(^{44}\), and we verified that, as claimed into that study, the cluster analysis is not much sensitive to the selection of precise parameter values.

Finally, $\eta$ is defined as:

\[
\log \eta_{ij} = \log R_{ij} + \log T_{ij}
\]  

(see Zaliapin and Ben-Zion, 2016\(^{48}\) for further details).

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M.P. designed the study, prepared the dataset, and performed the analysis. D.B. and D.S. contributed to the methodological development. All authors contributed to the manuscript redaction and approved the final version of the manuscript.

Competing interests
The authors declare no competing interests.

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