

1                    **A High-Resolution Earthquake Catalog Reveals**  
2                    **Coupled Fault Interactions and Multiple Controls on**  
3                    **Seismicity Migration in an Intraplate Earthquake**  
4                    **Swarm**

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15                    **Key Points:**

- 16                    • New machine-learned and relocated catalog reveals reveals two interacting fault  
17                    zones within the 2019–2020 Palghar intraplate swarm.  
18                    • Seismicity migrated as both diffusive and faster-than-diffusive propagating fronts,  
19                    indicating multiple triggering processes.  
20                    • Fault interactions, elastic stress transfer, and possible fluid-assisted deformation  
21                    jointly govern swarm evolution.

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## Abstract

Earthquake swarms in intraplate regions provide a unique opportunity to study fault responses to transient perturbations under low background strain rates. We investigate the 2019–2020 Palghar swarm in western India using data from two local seismic networks to construct a machine-learning-enhanced catalog comprising  $\sim 50,000$  events. High-resolution event relocations and moment tensor solutions reveal reactivation of two shallow normal faults with associated damage zones extending to  $\sim 8$  km depth. Seismicity initiates on the western fault, migrates to the eastern fault, and expands into the damage zones, indicating triggering through fault interactions. Broadly diffusive seismicity migration interspersed with persistent seismicity clusters and faster-than-diffusive migration transients suggests the involvement of both crustal fluids and aseismic slip. Together, these observations show that the evolution of intraplate earthquake swarms may be governed by seismic and aseismic elastic stress transfer coupled with fluid-driven processes, demonstrating how high-resolution earthquake catalogs help disentangle the processes controlling swarm evolution.

### Plain-language summary:

Earthquake swarms represent clusters of seismicity that occur over days to years without a single large mainshock. In stable continental regions, where earthquakes do not originate due to tectonic forcing, these swarms provide valuable clues about what other processes within the earth’s crust causes faults to become active. We investigated the 2019–2020 Palghar earthquake swarm in western India using recordings from two local seismic networks and a machine learning method that detected nearly 50,000 earthquakes. We found that the earthquakes occurred on two shallow, preexisting faults. The activity began on one fault, later migrated to the other, and eventually spread into the area between them. The observed migration of these earthquakes indicates that underground fluids and slow, silent movement along faults might have acted together to control the swarm’s evolution. These findings improve our understanding of how earthquake swarms develop in stable continental regions and provide new insights into the processes that control seismic activity far away from plate boundaries.

## 1 Introduction

Earthquake swarms are clusters of seismicity that occur close together in space and time without a clearly identifiable initiating event (Mogi, 1962; Hill, 1977; Vidale & Shearer, 2006). Swarm activity commonly exhibits complex spatiotemporal migration of seismicity, reflecting the combined influence of multiple physical processes rather than a single triggering mechanism (Hainzl, 2004; Roland & McGuire, 2009; Ruhl et al., 2016; Danré et al., 2022; Fischer et al., 2023; Danré et al., 2024). Characterizing these migration patterns, therefore, provides an important means of distinguishing among the processes responsible for swarm evolution. For example, slow, diffusive migration is commonly interpreted as reflecting fluid-mediated processes wherein the seismicity front is phenomenologically linked to the migration of a pore-pressure diffusion front (Shapiro & Dinske, 2009; Fischer & Hainzl, 2021; Danré et al., 2022). In contrast, faster-than-diffusive migration may reflect elastic stress transfer arising from a wide range of processes, including poroelastic responses to fluid flow within deformable porous media, the rapid propagation of aseismic slip fronts, and stress redistribution through elastic interactions within actively deforming fault networks (Hill, 1977; Rao et al., 1991; Hainzl, 2004; Goebel et al., 2017; Bhattacharya & Viesca, 2019; Hatch et al., 2020; Danré et al., 2024; Almakari et al., 2026). Determining the processes that govern swarm evolution is important for understanding earthquake triggering, the physical controls on the spatiotemporal evolution of earthquake clusters, and the partitioning of strain-energy release across event sizes (Vidale & Shearer, 2006; Roland & McGuire, 2009; Hatch et al., 2020; Ross et al.,

2020; Danré et al., 2024). Intraplate swarms are particularly valuable in this context because they provide a rare window into fault responses to transient stress perturbations under low long-term strain rates, where transient perturbations to fault stability can exceed the secular strain accumulation in amplitude (Hill, 1977; Ellsworth, 2013).

Many intraplate swarms are triggered by transient stress or strength perturbations and are often associated with pre-existing zones of crustal weakness (Rastogi et al., 1997; Stevenson et al., 2006; Srinagesh et al., 2020). Within the stable continental crust of India, several such swarms have been reported, but most are short-lived and poorly monitored, limiting insights into their driving mechanisms (Srivastava & Dube, 1996; Sateesh et al., 2019; Wadhawan et al., 2021). Of these, the Palghar swarm is an exception in that it continued for over two years and produced, as we show later, many tens of thousands of events and was well monitored by multiple seismic networks deployed by government agencies (Srinagesh et al., 2020; Gahalaut et al., 2022; Sharma et al., 2023). Most of the seismic activity occurred within the granitoid basement of the Deccan Volcanic Province within a relatively simple fault system that enables clear interpretation of fault–seismicity relations (Sharma et al., 2020; Pavankumar et al., 2020; Nath et al., 2021). Previous studies, based on smaller, local seismic networks and multiple geophysical observations, have suggested the presence of fluids within the fractured upper crust in the Palghar swarm region (Pavankumar et al., 2020; Nath et al., 2021; Subhadra & Shekar, 2024; Kanaujia et al., 2025). However, substantial discrepancies in hypocentral locations between the seismicity catalogs derived from two major local networks limit our ability to resolve the swarm’s migration patterns and, consequently, to identify the physical processes controlling its initiation and evolution (Sharma et al., 2020; Srinagesh et al., 2020). As a result, the relative roles of pore-pressure diffusion, aseismic slip (if any), and fault–seismicity interactions, as well as the possible source of the fluids, all remain poorly constrained.

Resolving the contribution of these processes requires a dense catalog with accurate earthquake locations, as precise hypocenters and source-property estimates are essential for diagnosing swarm driving mechanisms. To achieve this, we merge waveform data from two local seismic networks to construct a region-specific velocity model and a unified, high-resolution earthquake catalog for the Palghar swarm. We use a machine-learning-based detection workflow to substantially densify the earthquake catalog and refine earthquake locations using relocations based on waveform cross-correlation derived differential travel times. The resulting catalog resolves the three-dimensional geometry and spatiotemporal evolution of the Palghar swarm, revealing two closely spaced fault zones and deformation within an intervening damage zone. We show that seismicity migrated between the two faults through a combination of diffusion-like expansion and transient episodes of rapid propagation, indicating the combined influence of elastic fault interactions, crustal fluids, and transient aseismic slip. These observations demonstrate how fault interactions and transient deformation govern the evolution of intraplate earthquake swarms and provide new insight into earthquake triggering under low background strain-rate regimes.

## 2 Geological setting, data and methods

The Palghar region lies on the western margin of the Deccan Volcanic Province (DVP), where  $\sim 1.5$  km thick basalt flows overlie Archaean–Proterozoic basement rocks and are dissected by inherited rift-related structures, including the Panvel flexure, the West Coast Fault, and several N–S trending basement-rooted lineaments (Auden, 1949; Courtillot et al., 1986, 2003; Chandrasekharam, 1985; Gunnell & Fleitout, 1998; Sheth, 1998; Nath et al., 2021; Sharma et al., 2020, 2023). These inherited structures have hosted diverse modes of seismicity in the past, from the 1993 Latur earthquake and reservoir-triggered seismicity at Koyna to recurrent earthquake swarms (Rao et al., 1991; Srivastava & Dube, 1996; Raju et al., 2000; Hainzl et al., 2014; Sateesh et al., 2019; Wadhawan et al., 2021).

123 Among these swarms, the Palghar sequence stands out for its unusually long du-  
 124 ration (over two years), high event count (tens of thousands of earthquakes), and dense  
 125 instrumental coverage, providing a rare opportunity to constrain—if not the triggering  
 126 mechanisms—at least the processes governing its evolution (Sharma et al., 2020, 2023;  
 127 Gahalaut et al., 2022). Previous studies have linked the Palghar swarm to both meteoric-  
 128 water infiltration and deeper crustal fluids (Pavankumar et al., 2020; Srinagesh et al.,  
 129 2020; Sharma et al., 2020; Nath et al., 2021; Gahalaut et al., 2022; Kanaujia et al., 2025).  
 130 Its exceptional duration and productivity provide a unique opportunity to investigate  
 131 the processes governing earthquake clustering and fault reactivation within the Deccan  
 132 Volcanic Province, with broader implications for intraplate seismicity worldwide.

## 133 2.1 Seismic Data

134 This study uses waveform data from 11 broadband stations operated independently  
 135 by the National Geophysical Research Institute (NGRI) and the National Center for Seis-  
 136 mology (NCS) following the onset of the Palghar swarm, providing dense coverage over  
 137 an area of approximately 30 km  $\times$  30 km (Sharma et al., 2020; Gahalaut et al., 2022)  
 138 (SI Figure S1). We merge the independently compiled NGRI and NCS catalogs by match-  
 139 ing events within 0.5 s of their origin times, yielding a unified dataset with an improved  
 140 mean azimuthal gap ( $\sim 80^\circ$ ) compared with the individual NCS ( $\sim 158^\circ$ ) and NGRI ( $\sim 94^\circ$ )  
 141 networks (SI Figure S5). Earthquakes recorded by at least ten stations are used to de-  
 142 rive a one-dimensional velocity model (Section 2.2), which is subsequently used for earth-  
 143 quake detection, absolute locations, relative relocations, and moment tensor inversions.  
 144 All remaining catalog construction steps rely on automated detection and relative loca-  
 145 tion procedures. Station coordinates and elevations are listed in SI Table 1.

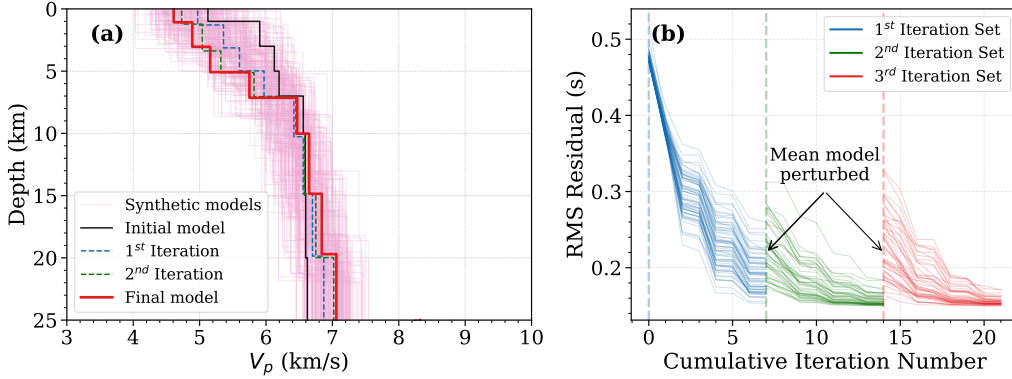
## 146 2.2 One-Dimensional Velocity Modeling

147 Previous studies adopted different velocity models for the NGRI and NCS datasets,  
 148 resulting in substantial discrepancies in earthquake depths (6-7 km Sharma et al. (2020)  
 149 to 4-16 km Srinagesh et al. (2020)). To obtain a consistent velocity model for the merged  
 150 dataset, we applied joint hypocenter determination (JHD) using VELEST (Kissling et al.,  
 151 1995). The inversion used 2,514 well-recorded human-picked earthquakes observed at ten  
 152 or more stations, yielding 25,097 P- and 24,361 S-wave arrivals. A constant  $V_p/V_s$  ra-  
 153 tio of 1.71, estimated from the linear relationship between P- and S-wave arrival times  
 154 (SI Figure S2), was assumed throughout the inversions.

155 The one-dimensional model of Srinagesh et al. (2020) was used as the initial ref-  
 156 erence. To assess model stability and minimize dependence on the starting model, we  
 157 generated ensembles of randomly perturbed velocity models by varying layer velocities  
 158 ( $\pm 0.7$  km  $s^{-1}$ ) and interface depths ( $\pm 2$  km) within prescribed ranges. Three successive  
 159 inversion stages (see SI materials for details) comprising 210 perturbed models in total  
 160 converged to an ensemble of solutions at the end of the third stage (Figure 1). The mean  
 161 velocity model from this final ensemble was adopted for all subsequent earthquake de-  
 162 tection, location, and moment tensor analyses. Station corrections estimated during the  
 163 inversion account for near-surface heterogeneity and were applied in subsequent reloca-  
 164 tions.

## 165 2.3 ML-based earthquake detection and relocation

166 Further earthquake detection and hypocentral determination were performed using  
 167 a multi-stage, fully automated workflow designed to construct a dense and internally  
 168 consistent earthquake catalog for the Palghar swarm. The workflow combines network-  
 169 based detection using the BPF (Back Projection and Matched Filtering) framework  
 170 (Beaucé et al., 2024), probabilistic absolute location, and high-precision relative reloca-  
 171 tion, without reliance on manual phase picking for catalog construction.



**Figure 1.** Joint hypocenter–velocity inversion results for the Palghar swarm region using VELEST. (a) Velocity inversion results after three inversion stages. The initial velocity model of Srinagesh et al. (2020) is shown in black. At each stage, an ensemble of velocity models was generated by randomly perturbing the final mean model from the previous stage by  $\pm 0.7 \text{ km s}^{-1}$  in layer velocity and  $\pm 2 \text{ km}$  in interface depth to assess inversion stability. The resulting ensemble of final velocity models across all 3 stages is shown in pink, and the mean at the end of the stage 3 (thick red line) was adopted in this study. Blue and green dashed lines denote the mean velocity models obtained after the first and second inversion stages, respectively. (b) Evolution of RMS travel-time residuals during the inversion. Vertical dashed lines indicate the boundaries between the three inversion stages.

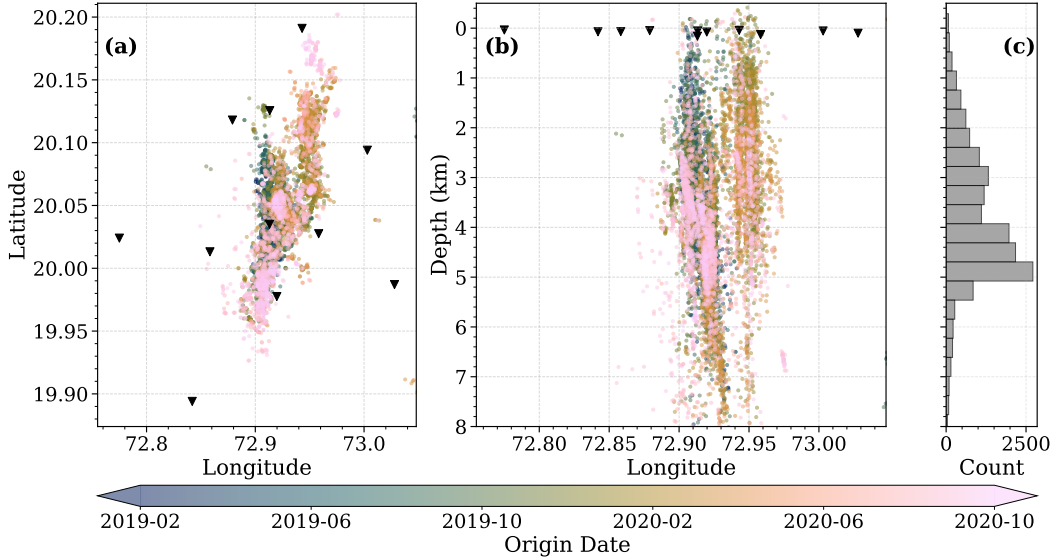
172 Continuous three-component waveform data were first processed using the deep-  
 173 learning phase picker PhaseNet (Zhu & Beroza, 2019) to estimate P- and S-wave arrival  
 174 probabilities. These probabilities were used in the back projection workflow developed  
 175 by (Beaucé et al., 2024), which coherently stacked phase-consistent arrival probabilities  
 176 using theoretical moveout corrections computed from the one-dimensional velocity model  
 177 (Section 2.2). Events exceeding a predefined beam-power threshold (SI Figure S6) yielded  
 178 an initial catalog of 57,826 earthquakes.

179 These initial grid-based hypocenters obtained from the back projection procedure  
 180 were subsequently refined using the probabilistic location algorithm NonLinLoc (Lomax  
 181 et al., 2000). Retaining only events recorded at six or more stations produced 20,638 earth-  
 182 quakes with median horizontal and depth uncertainties of 0.5 km and 1.5 km, respec-  
 183 tively (Figure S7). High-precision relative locations were then obtained using GrowClust  
 184 (Trugman & Shearer, 2017) and waveform cross-correlation of 5–10 Hz filtered seismo-  
 185 grams. Event pairs sharing at least six observations with cross-correlation coefficients  
 186  $\geq 0.8$  were retained, resulting in a final catalog of 16,008 earthquakes. In comparison,  
 187 the manually picked catalog contains only 5,765 well-recorded events.

188 The relocated seismicity delineates two narrow, subparallel north–south–striking  
 189 fault strands confined to depths of 2–6 km (Figure 2). Bootstrap analysis indicates me-  
 190 dian relative location uncertainties of 250 m horizontally and 320 m vertically (SI Fig-  
 191 ure S7), sufficient to robustly resolve fault geometry and swarm evolution.

## 192 2.4 Moment Tensor Inversions

193 To constrain the faulting style of the larger earthquakes within the swarm, we per-  
 194 formed moment tensor (MT) inversions for 37 events ( $M_l \geq 3.0$ ) using MTTime (Ichinose



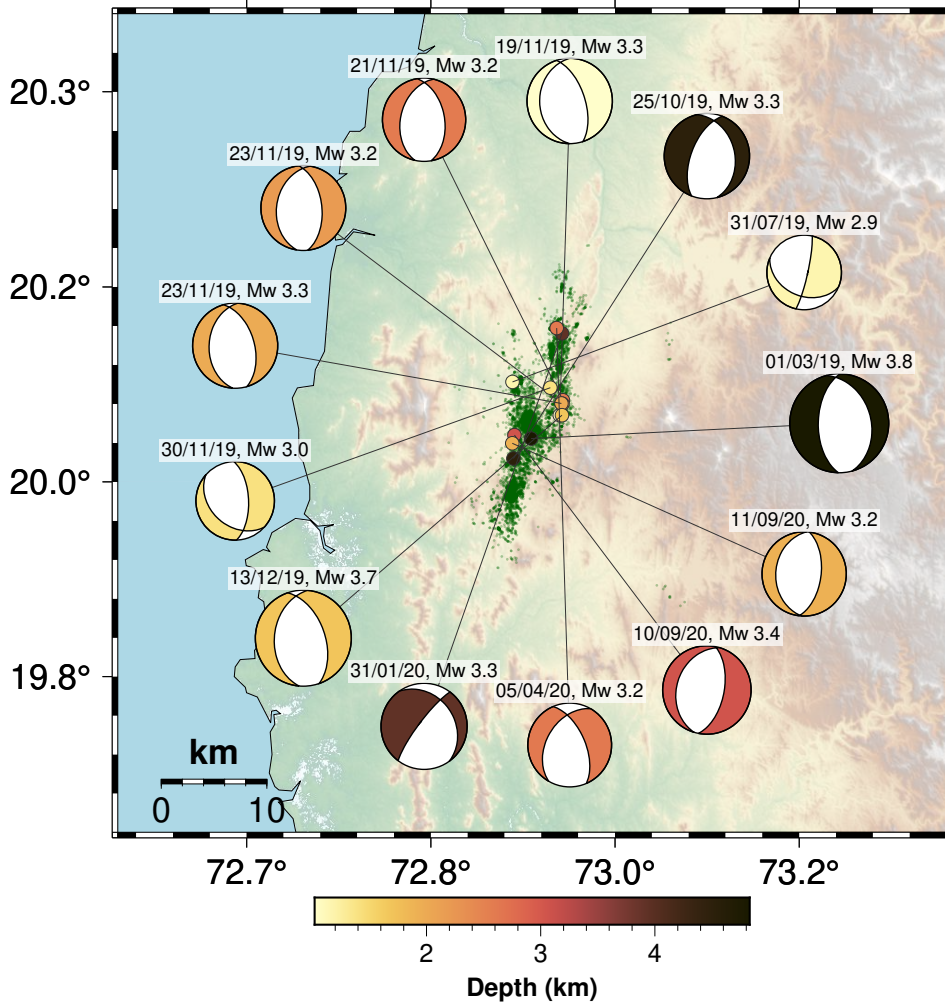
**Figure 2.** Spatial distribution of the *GrowClust*-relocated Palghar earthquake swarm. (a) Map view, and (b) longitude–depth cross-section of relocated seismicity color-coded by origin time, illustrating the spatiotemporal evolution of the swarm. Black inverted triangles denote seismic station locations projected onto the section. (c) Histogram of hypocenter depths showing that seismicity is predominantly concentrated between about 2 and 6 km. The flood basalt is expected to be not more than 1.5 km thick in this region. Therefore, the seismicity is almost entirely concentrated within the granitoid basement.

195 et al., 2014) (see SI materials for details, Text S3). Green’s functions were computed with  
 196 the CPS package (Herrmann, 2013) using the one-dimensional velocity model derived  
 197 in this study. Relative locations from *GrowClust* were used where available, and *NonLinLoc*  
 198 hypocenters were adopted for the remaining nine events. Waveforms were filtered be-  
 199 tween 0.05 and 1 Hz, and all selected solutions achieved variance reductions exceeding  
 200 50% (SI Figure S10; SI Table 2).

201 The MT solutions indicate predominantly normal faulting with minor strike-slip  
 202 components (Figures 3 and S10). Fault strikes closely follow the north–south trends of  
 203 the linear structures outlined by the relocated seismicity, whereas dip angles cluster around  
 204  $\sim 30^\circ$  and  $\sim 60^\circ$ , consistent with conjugate normal faults (SI Figure S9). The rake an-  
 205 gles are likewise consistent with dominantly extensional slip. Together, these observa-  
 206 tions indicate reactivation of pre-existing extensional fault structures and justify inter-  
 207 preting the two relocated seismicity clusters as outlining two subparallel normal faults  
 208 which we refer to as the western and eastern faults in the subsequent analysis.

### 209 3 Results

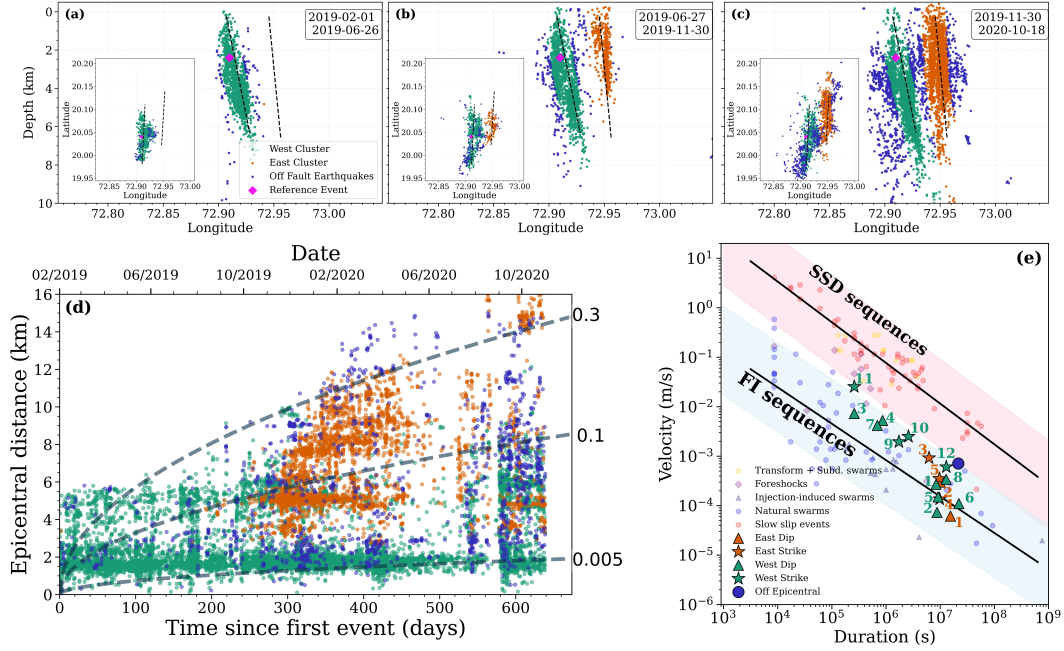
210 The high-resolution ML-enhanced relocated earthquake catalogs developed in this  
 211 study provide substantially improved constraints on the spatiotemporal evolution of the  
 212 Palghar swarm compared to the human-picked catalog that earlier studies have mostly  
 213 relied upon (Sharma et al., 2020, 2023; Srinagesh et al., 2020) (SI Figure S11). During  
 214 the first  $\sim 200$  days, seismicity remains confined to a cluster around the western fault be-  
 215 fore a sudden appearance of seismicity that clusters around the eastern fault (Figures 4a  
 216 and b). We choose to label the seismicity as western, eastern and off-fault clusters de-  
 217 pending on whether the events are within a band of  $\pm 1$  km around each of the visually



**Figure 3.** Moment tensor solutions for earthquakes with  $M_l \geq 3.0$  and variance reduction exceeding 50%. Beachball size scales with magnitude, and colors indicate event depth. Green dots show relocated hypocenters from the full catalog (all events relocated by **GrowClust** plus 9 supplemented from **NonLinLoc**). Only the double-couple component is displayed; full deviatoric solutions are shown in the SI figure S8.

218 interpreted fault structures. The ‘off-fault’ earthquakes between the two faults are pre-  
 219 dominantly activated only after seismicity begins to cluster around both faults likely re-  
 220 flecting ruptures activated by stress perturbations generated by simultaneous activation  
 221 of slip on both faults.

222 We analyze the spatiotemporal evolution of the Palghar earthquake swarm using  
 223 an epicentral-distance vs. time representation of the relocated events, where the former  
 224 is measured relative to the first recorded event from the machine-learned catalog (Origin  
 225 time: 01/02/2019 00:01:53, latitude: 20.04°, longitude: 72.91°, depth: 2.4 km) (Figure  
 226 4). The seismicity migration reveals a broad expansion of earthquake activity over  
 227 time (Figure 4d). The overall envelope is consistent with diffusive growth, with most events  
 228 bounded by diffusivities of  $\sim 0.005\text{--}0.3 \text{ m}^2 \text{ s}^{-1}$ . Superimposed on this long-term expan-  
 229 sion are distinct propagating fronts that define the leading edge of the swarm. These fronts  
 230 are most pronounced during activation of the eastern fault between  $\sim 260$  and 460 days,



**Figure 4.** Spatiotemporal evolution and migration characteristics of the Palghar swarm. (a–c) Relocated earthquakes projected onto longitude–depth sections for the western (green), eastern (orange), and off-fault (blue) clusters. Insets show map views. (d) Epicentral distance versus time with reference diffusion fronts (0.005, 0.1, and 0.3 m<sup>2</sup>/s). (e) Migration velocity versus duration for representative migration episodes in the Palghar swarm plotted within the fluid-induced (FI) and slow-slip–driven (SSD) migration regimes proposed by Danré et al. (2024). Stars and triangles denote along-strike and along-dip migration episodes, respectively (for details, see SI Figures S13, S14 and S15). Green, orange, and blue symbols represent the western fault, eastern fault, and off-fault migration episodes respectively.

231 when earthquakes progressively migrate to larger epicentral distances. In contrast, the  
 232 western fault remains active throughout most of the sequence, with repeated migration  
 233 episodes nucleating from a persistent source region near 1.5 km epicentral distance.

234 The inferred ‘fault’ structures shown in Figures 4a–c clearly show that the epicen-  
 235 tral distance migration is nearly insensitive to any along-dip migration given the steeply  
 236 dipping normal ‘faults’ inferred from the seismicity clustering. Given this, we differen-  
 237 tiate the along-strike and dip migrations of the western and eastern clusters to better  
 238 constrain the rupturing process (SI Figures S13 and S14). The along-dip and along-strike  
 239 distance-time evolution of seismicity along the western fault reveals strongly episodic mi-  
 240 gration behavior. The persistent seismicity band becomes more conspicuous and appears  
 241 both strike- and dip-limited, while multiple seismicity fronts propagate, along-strike and  
 242 dip, both away and towards this seismicity band over time. The along-strike and along-  
 243 dip migration envelopes reveal multiple episodes of outward and back-propagating seis-  
 244 micity fronts on the western fault (SI Figures S13a, c). Migration rates and durations  
 245 for representative episodes are compared with the velocity–duration classification of Danré  
 246 et al. (2024) in Figure 4e.

247 In contrast, the eastern fault exhibits a very different seismicity clustering, with  
 248 much of the early migration occurring prominently northwards along strike, accompa-  
 249 nied by slower back-propagating fronts both along strike and dip (SI Figure S14a and

c). There is a non-persistent seismicity band visible only along strike, but this mostly disappears after about 450 days of seismicity. As for the western cluster, we again estimate the migration rates for some of the visually distinct migration periods and report these within the Danré et al. (2024) classification scheme for reference.

The off-fault seismicity exhibits migration patterns similar to those of the eastern cluster, with nearly identical epicentral migration rates between  $\sim 260$  and  $\sim 360$  days (SI Figure S15). During this interval, much of the off-fault seismicity represents northward migration adjacent to the eastern fault. These observations are robust to a narrower ( $\pm 500$  m) definition of the fault zones (SI Figure S15). Later in the sequence ( $\sim 480$  to  $\sim 640$  days), off-fault seismicity mirrors the southward and updip migration observed along the western fault (SI Figure S16).

## 4 Discussion

The new high-resolution earthquake catalog reveals a substantially more complex evolution of the Palghar swarm than previously recognized (Srinagesh et al., 2020; Sharma et al., 2020, 2023; Nath et al., 2021). Rather than occurring on a single fault zone, the swarm initiated on one of two closely spaced fault segments before migrating to the second one and, finally, progressively expanded into the surrounding fault damage zone. The relocated seismicity further reveals persistent fault-localized activity on the western fault, together with transient propagating fronts, indicating that the observed swarm evolution cannot be explained by a single triggering mechanism. These newly resolved spatiotemporal patterns provide important constraints on the physical processes governing earthquake swarms.

The overall migration of seismicity is consistent with a broad diffusive envelope, with apparent hydraulic diffusivities ranging from  $\sim 0.005$  to  $\sim 0.3$   $\text{m}^2 \text{s}^{-1}$ , comparable to values reported for fluid-driven swarms elsewhere (Goebel et al., 2017; Cabrera et al., 2021). Superimposed on this long-term expansion are transient propagation fronts that migrate substantially faster than expected from diffusion alone. In the migration velocity–duration space of Danré et al. (2024), most of the observed migrations fall within the fluid-diffusion regime, whereas a smaller population approaches the transition toward the slow-slip regime identified by Danré et al. (2024) (episodes 3 from the eastern and 3, 4, 7 and 11 from the western clusters in Figure 4e). Although none of the observed faster migration episodes lie unequivocally within the slow-slip regime, the coexistence of distinct migration populations suggests that multiple transient processes contributed to swarm evolution. Additionally, direct comparison with the transitional boundaries proposed by Danré et al. (2024) should be interpreted with caution because their slow-slip–driven migration dataset is derived primarily from epicentral migration in plate-boundary settings, whereas our migration rates are based on along-strike and along-dip propagation within an intraplate fault system.

The inference that multiple processes contributed to swarm evolution is also consistent with contrasting migration styles on the western and eastern fault systems. The western fault hosts a persistent, spatially confined seismicity band from which seismicity repeatedly expands and contracts along strike and dip (SI Figure S13a, c), whereas the eastern fault is dominated by a transient along-strike migration episode that contains several of the  $M_l > 3.0$  earthquakes and lacks a comparable persistent seismicity cluster (SI Figure S14a, c). Additionally, off-fault seismicity between and adjacent to the faults closely mirrors on-fault migration. Following activation of the eastern fault, both on-fault and off-fault seismicity propagated northward at comparable rates during the 260–460 day interval (see SI Figure S15). Similar correspondence is observed west of the western fault, where off-fault seismicity tracks the along-strike and up-dip migration of the main cluster (SI Figure S16). This coordinated spatiotemporal evolution is difficult to reconcile with diffusive processes alone and instead suggests that transient aseismic

301 deformation and elastic stress transfer operated together with diffusive (possibly fluid-  
302 assisted) processes.

303 The Palghar swarm is strongly confined to depths shallower than  $\sim 6$  km (Figure 2c).  
304 Such a sharp lower bound may reflect structural termination of the active fault network,  
305 a permeability barrier restricting deeper fluid migration, or a rheological transition at  
306 depth. Because the swarm is hosted within the granitoid basement of the Deccan Vol-  
307 canic Province, where extensive dyke-related fault networks are expected (Courtillot et  
308 al., 1986; Rubin & Pollard, 1988; Rubin, 1992), structural termination at only  $\sim 6$  km  
309 appears unlikely (Rai & Saha, 2025). Likewise, a rheological transition from velocity-  
310 weakening to velocity-strengthening friction is difficult to reconcile with the shallow cut-  
311 off, regional geothermal gradients (M. L. Gupta & Gaur, 1984; Roy & Rao, 2000; C. Scholz,  
312 1998; C. H. Scholz, 2019), and evidence for deeper crustal low-velocity zones in nearby  
313 swarm regions (Rai & Saha, 2025). Given that the persistent seismicity band within the  
314 western cluster is also not much wider than  $\sim 1$  km along both strike and dip, we favor  
315 a model in which the persistent seismicity band on the western fault marks a localized  
316 fault-valve system connected to a deeper fluid source, with downdip fluid migration lim-  
317 ited by permeability barriers or pore-pressure gradients. Such a persistent fluid reser-  
318 voir may also have provided the initial perturbation that triggered the swarm.

319 Previous studies have attributed the Palghar swarm and similar earthquake swarms  
320 in the Deccan Volcanic Province to monsoon-driven recharge (Hainzl et al., 2014; H. K. Gupta,  
321 2018; Sharma et al., 2020; Nath et al., 2021; Gahalaut et al., 2022; Sharma et al., 2023).  
322 In contrast, the densified catalog does not reveal a systematic relationship between rain-  
323 fall and seismicity during the early, most productive phase of the sequence, although one  
324 increase in seismicity during the later stages follows the end of the monsoon season (SI  
325 Figure S18). While these observations do not exclude a contribution from meteoric recharge,  
326 they provide little evidence that rainfall alone controlled swarm initiation. Instead, our  
327 results favor a deeper fluid source whose interaction with fault geometry and elastic stress  
328 transfer governed the evolution of the Palghar swarm.

## 329 5 Conclusion

330 Previous studies of the Palghar earthquake swarm relied primarily on human-picked,  
331 independently developed catalogs from the NGRI and NCS seismic networks which lacked  
332 the event density and location precision necessary to fully resolve the physical processes  
333 governing the swarm evolution. In this study, we combined waveform data from both  
334 networks and constructed an machine-learning-enhanced high-resolution earthquake cat-  
335 alog to better illuminate the mechanisms controlling the swarm sequence. The refined  
336 relocations reveal two closely spaced structurally-controlled seismicity clusters between  
337 depths of approximately 2-8 km, together with distributed off-fault earthquakes between  
338 them. The spatiotemporal evolution of the sequence exhibits strongly contrasting mi-  
339 gration behaviors between the two fault systems which we interpret as diagnostic of at  
340 least two different physical processes driving the initiation and continued evolution of  
341 the swarm.

342 The overall swarm evolution remains broadly diffusive, suggestive of fluid-driven  
343 migration, whereas several rapid migration episodes approach the transitional regime com-  
344 monly associated with transient aseismic slip. At the same time, the delayed activation  
345 of the eastern cluster and the subsequent emergence of off-fault seismicity provide strong  
346 evidence for elastic stress transfer both between the two principal fault systems and among  
347 earthquakes occurring within the surrounding fault damage zone. While transient aseis-  
348 mic slip may have contributed to at least some of the rapid migration episodes, the Pal-  
349 ghar sequence more broadly illustrates how long-lived earthquake swarms in stable con-  
350 tinental interiors can emerge from coupled interactions among (possible) fluid migration,  
351 transient aseismic deformation, and stress transfer within structurally complex inher-

352 ited fault networks. More generally, our results demonstrate that high-resolution earth-  
 353 quake catalogs, combined with detailed fault geometry and spatiotemporal migration anal-  
 354 yses, can distinguish among these interacting processes and provide new constraints on  
 355 the physical mechanisms governing intraplate earthquake swarms.

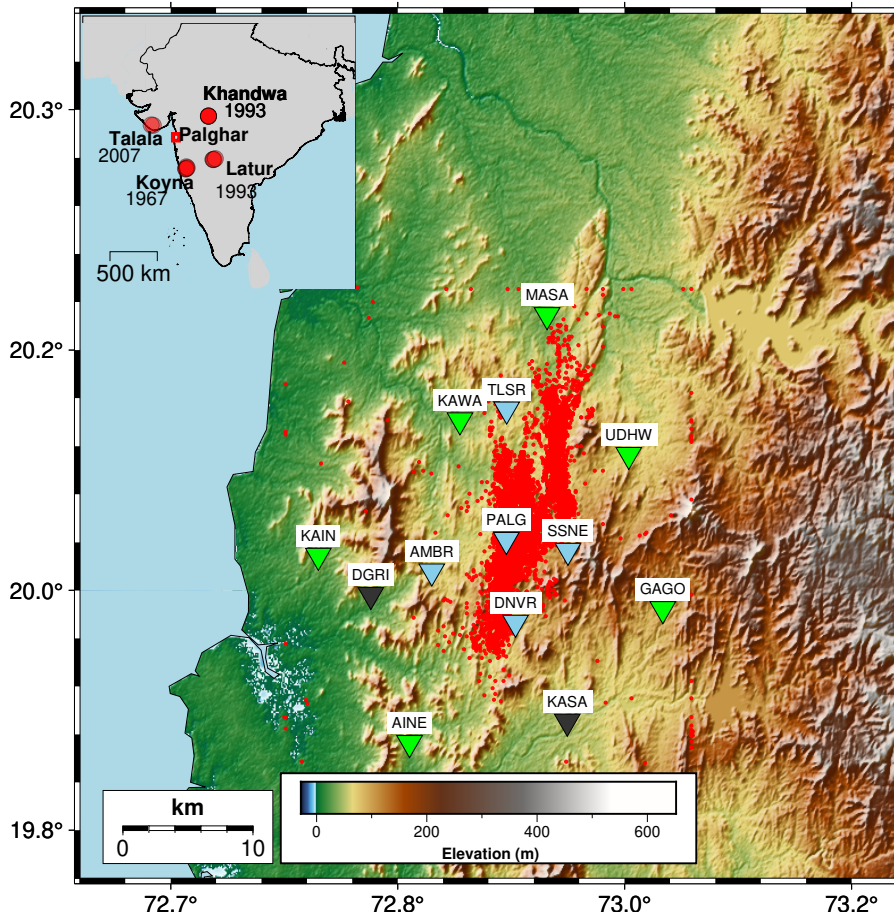
## 356 References

- 357 Almakari, M., Kheirdast, N., Villafuerte, C., Thomas, M., Dubernet, P., Cheng,  
 358 J., ... Bhat, H. (2026). Fault volume digital twin to reproduce the full slip  
 359 spectrum, scaling, and statistical laws. *Journal of Geophysical Research: Solid*  
 360 *Earth*, *131*(5), e2025JB032915.
- 361 Auden, J. (1949). Dykes in western india—a discussion on their relationships with the  
 362 deccan traps. *Trans. Nat. Inst. Sci. India*, *3*, 123–157.
- 363 Beaucé, E., Frank, W. B., Seydoux, L., Poli, P., Groebner, N., van der Hilst, R. D.,  
 364 & Campillo, M. (2024). Bpmf: A backprojection and matched-filtering work-  
 365 flow for automated earthquake detection and location. *Seismological Research*  
 366 *Letters*, *95*(2A), 1030–1042.
- 367 Bhattacharya, P., & Viesca, R. C. (2019). Fluid-induced aseismic fault slip outpaces  
 368 pore-fluid migration. *Science*, *364*(6439), 464–468.
- 369 Cabrera, L., Poli, P., & Frank, W. B. (2021). Tracking the spatio-temporal evolution  
 370 of foreshocks preceding the mw 6.3 2009 l’aquila earthquake 2. *interaction*, *14*,  
 371 15.
- 372 Chandrasekharam, D. (1985). Structure and evolution of the western continental  
 373 margin of india deduced from gravity, seismic, geomagnetic and geochronologi-  
 374 cal studies. *Physics of the Earth and Planetary interiors*, *41*(2-3), 186–198.
- 375 Courtillot, V., Besse, J., Vandamme, D., Montigny, R., Jaeger, J.-J., & Cappetta, H.  
 376 (1986). Deccan flood basalts at the cretaceous/tertiary boundary? *Earth and*  
 377 *Planetary Science Letters*, *80*(3-4), 361–374.
- 378 Courtillot, V., Davaille, A., Besse, J., & Stock, J. (2003). Three distinct types of  
 379 hotspots in the earth’s mantle. *Earth and Planetary Science Letters*, *205*(3-4),  
 380 295–308.
- 381 Danré, P., De Barros, L., Cappa, F., & Ampuero, J.-P. (2022). Prevalence of aseis-  
 382 mic slip linking fluid injection to natural and anthropogenic seismic swarms.  
 383 *Journal of Geophysical Research: Solid Earth*, *127*(12), e2022JB025571.
- 384 Danré, P., De Barros, L., Cappa, F., & Passarelli, L. (2024). Parallel dynamics of  
 385 slow slips and fluid-induced seismic swarms. *Nature Communications*, *15*(1),  
 386 8943.
- 387 Ellsworth, W. L. (2013). Injection-induced earthquakes. *science*, *341*(6142),  
 388 1225942.
- 389 Fischer, T., & Hainzl, S. (2021). The growth of earthquake clusters. *Frontiers in*  
 390 *Earth Science*, *9*, 638336.
- 391 Fischer, T., Hainzl, S., & Vlček, J. (2023). Fast migration episodes within earth-  
 392 quake swarms. *Geophysical Journal International*, *235*(1), 312–325.
- 393 Gahalaut, K., Gahalaut, V., Naresh, B., Shekar, M., Sunilkumar, T., & Srinagesh,  
 394 D. (2022). Long duration non-volcanic and non-tectonic palghar earthquake  
 395 swarm in the stable continental region of india—role of seasonal rainfall and  
 396 earthquake cascading. *Journal of Seismology*, *26*(3), 545–554.
- 397 Goebel, T., Weingarten, M., Chen, X., Haffener, J., & Brodsky, E. (2017). The 2016  
 398 mw5. 1 fairview, oklahoma earthquakes: Evidence for long-range poroelastic  
 399 triggering at  $\approx$  40 km from fluid disposal wells. *Earth and Planetary Science*  
 400 *Letters*, *472*, 50–61.
- 401 Gunnell, Y., & Fleitout, L. (1998). Shoulder uplift of the western ghats passive mar-  
 402 gin, india: a denudational model. *Earth Surface Processes and Landforms: The*  
 403 *Journal of the British Geomorphological Group*, *23*(5), 391–404.

- 404 Gupta, H. K. (2018). Reservoir triggered seismicity (rts) at koyna, india, over the  
 405 past 50 yrs. *Bulletin of the Seismological Society of America*, 108(5B), 2907–  
 406 2918.
- 407 Gupta, M. L., & Gaur, V. K. (1984). Surface heat flow and probable evolution of  
 408 deccan volcanism. *Tectonophysics*, 105(1–4), 309–318.
- 409 Hainzl, S. (2004). Seismicity patterns of earthquake swarms due to fluid intrusion  
 410 and stress triggering. *Geophysical Journal International*, 159(3), 1090–1096.
- 411 Hainzl, S., Aggarwal, S., Khan, P., & Rastogi, B. (2014). Monsoon-induced earth-  
 412 quake activity in talala, gujarat, india. *Geophysical Journal International*,  
 413 200(1), 627–637.
- 414 Hatch, R. L., Abercrombie, R. E., Ruhl, C. J., & Smith, K. D. (2020). Evidence  
 415 of aseismic and fluid-driven processes in a small complex seismic swarm near  
 416 virginia city, nevada. *Geophysical Research Letters*, 47(4), e2019GL085477.
- 417 Herrmann, R. B. (2013). Computer programs in seismology: An evolving tool for in-  
 418 struction and research. *Seismological Research Letters*, 84(6), 1081–1088.
- 419 Hill, D. P. (1977). A model for earthquake swarms. *Journal of Geophysical Research*,  
 420 82(8), 1347–1352.
- 421 Ichinose, G., Roman-Nieves, J., & Kraft, G. (2014). Moment tensor inversion toolkit  
 422 (mtinv) documentation, manual and tutorial. *Seismol. Res. Lett.*(2), 37–57.
- 423 Kanaujia, J., Kumar, M. R., Vengala, P. K., & Shekar, M. (2025). Indication of  
 424 fluid-driven seismicity in the palghar region, india, from local earthquake to-  
 425 mography. *Bulletin of the Seismological Society of America*, 115(2), 489–504.
- 426 Kissling, E., Kradolfer, U., & Maurer, H. (1995). Program velest user’s guide-short  
 427 introduction. *Institute of Geophysics, ETH Zurich*, 22.
- 428 Lomax, A., Virieux, J., Volant, P., & Berge-Thierry, C. (2000). Probabilistic earth-  
 429 quake location in 3d and layered models: Introduction of a metropolis-gibbs  
 430 method and comparison with linear locations. In *Advances in seismic event*  
 431 *location* (pp. 101–134). Springer.
- 432 Mogi, K. (1962). Magnitude-frequency relation for elastic shocks accompanying  
 433 fractures of various materials and some related problems in earthquakes. *Bull.*  
 434 *Earthq. Res. Inst., Univ. Tokyo*, 40, 831–853.
- 435 Nath, B., Singh, R. P., Gahalaut, V. K., & Singh, A. P. (2021). Dynamic rela-  
 436 tionship study between the observed seismicity and spatiotemporal pattern of  
 437 lineament changes in palghar, north maharashtra (india). *Remote Sensing*,  
 438 14(1), 135.
- 439 Pavankumar, G., Chakravarthi, N., Demudu Babu, M., & Manglik, A. (2020). Mag-  
 440 netotelluric study of an intraplate seismic zone in the palghar region of the  
 441 deccan volcanic province, india. *Journal of Earth System Science*, 129(1),  
 442 133.
- 443 Rai, S. S., & Saha, G. (2025). Deccan volcanism and related seismic unrest in the  
 444 koyna–warna region, india. *Seismological Research Letters*, 96(3), 1645–1653.
- 445 Raju, P. S., Srinivasan, A., Raghavan, R., & Kousalya, M. (2000). Micro-tremor ac-  
 446 tivity in jubilee hills area of hyderabad, andhra pradesh. *Geological Society of*  
 447 *India*, 443–445.
- 448 Rao, D., Jambusaria, B., Srivastava, S., Srivastava, N., Hamid, A., Desai, B., & Sri-  
 449 vastava, H. (1991). Earthquake swarm activity in south gujarat. *Mausam*,  
 450 42(1), 89–98.
- 451 Rastogi, B., Mandal, P., & Kumar, N. (1997). Seismicity around dhamni dam, ma-  
 452 harashtra, india. *pure and applied geophysics*, 150(3), 493–509.
- 453 Roland, E., & McGuire, J. J. (2009). Earthquake swarms on transform faults. *Geo-*  
 454 *physical Journal International*, 178(3), 1677–1690.
- 455 Ross, Z. E., Cochran, E. S., Trugman, D. T., & Smith, J. D. (2020). 3d fault ar-  
 456 chitecture controls the dynamism of earthquake swarms. *Science*, 368(6497),  
 457 1357–1361.

- 458 Roy, S., & Rao, R. (2000). Heat flow in the indian shield. *Journal of Geophysical*  
 459 *Research: Solid Earth*, 105(B11), 25587–25604.
- 460 Rubin, A. M. (1992). Dike-induced faulting and graben subsidence in volcanic rift  
 461 zones. *Journal of Geophysical Research: Solid Earth*, 97(B2), 1839–1858.
- 462 Rubin, A. M., & Pollard, D. D. (1988). Dike-induced faulting in rift zones of iceland  
 463 and afar. *Geology*, 16(5), 413–417.
- 464 Ruhl, C., Abercrombie, R., Smith, K., & Zaliapin, I. (2016). Complex spatiotem-  
 465 poral evolution of the 2008 mw 4.9 mogul earthquake swarm (reno, nevada):  
 466 Interplay of fluid and faulting. *Journal of Geophysical Research: Solid Earth*,  
 467 121(11), 8196–8216.
- 468 Sateesh, A., Mahesh, P., Singh, A., Kumar, S., Chopra, S., & Kumar, M. R. (2019).  
 469 Are earthquake swarms in south gujarat, northwestern deccan volcanic  
 470 province of india monsoon induced? *Environmental Earth Sciences*, 78(13),  
 471 381.
- 472 Scholz, C. (1998). Earthquakes and friction laws. *Nature*, 391, 37–42. doi: 10.1038/  
 473 34097
- 474 Scholz, C. H. (2019). *The mechanics of earthquakes and faulting*. Cambridge univer-  
 475 sity press.
- 476 Shapiro, S. A., & Dinske, C. (2009). Fluid-induced seismicity: Pressure diffusion and  
 477 hydraulic fracturing. *Geophysical Prospecting*, 57(2), 301–310.
- 478 Sharma, V., Kumar, D., Sharma, B., & Chingtham, P. (2023). Characteristics of  
 479 earthquake swarm activity observed in the palghar region of indian penin-  
 480 sula from january 2019 to october 2020. *Quaternary Science Advances*, 12,  
 481 100099.
- 482 Sharma, V., Wadhawan, M., Rana, N., Sreejith, K., Agrawal, R., Kamra, C., ...  
 483 Gahalaut, V. K. (2020). A long duration non-volcanic earthquake sequence  
 484 in the stable continental region of india: The palghar swarm. *Tectonophysics*,  
 485 779, 228376.
- 486 Sheth, H. (1998). A reappraisal of the coastal panel flexure, deccan traps, as a  
 487 listric-fault-controlled reverse drag structure. *Tectonophysics*, 294(1-2), 143–  
 488 149.
- 489 Srinagesh, D., Singh, D. K., Vikas, G., Naresh, B., Roy, S., Murthy, Y., ... others  
 490 (2020). An appraisal of recent earthquake activity in palghar region, maha-  
 491 rashtra, india. *Curr Sci*, 118, 1592–1598.
- 492 Srivastava, H., & Dube, R. (1996). Comparison of precursory and non-precursory  
 493 swarm activity in peninsular india. *Tectonophysics*, 265(3-4), 327–339.
- 494 Stevenson, D., Gangopadhyay, A., & Talwani, P. (2006). Booming plutons: Source of  
 495 microearthquakes in south carolina. *Geophysical Research Letters*, 33(3).
- 496 Subhadra, N., & Shekar, M. (2024). The attenuation mechanism of high-frequency  
 497 seismic waves in the palghar swarm earthquake source area in western india.  
 498 *Journal of Asian Earth Sciences*, 276, 106306.
- 499 Trugman, D. T., & Shearer, P. M. (2017). Growclust: A hierarchical clustering  
 500 algorithm for relative earthquake relocation, with application to the spanish  
 501 springs and sheldon, nevada, earthquake sequences. *Seismological Research*  
 502 *Letters*, 88(2A), 379–391.
- 503 Vidale, J. E., & Shearer, P. M. (2006). A survey of 71 earthquake bursts across  
 504 southern california: Exploring the role of pore fluid pressure fluctuations  
 505 and aseismic slip as drivers. *Journal of Geophysical Research: Solid Earth*,  
 506 111(B5).
- 507 Wadhawan, M., Rana, N., Gahalaut, V., Singh, M., Singh, K., Suresh, G., ... others  
 508 (2021). Monsoonal rainfall induced shallow earthquake swarm in the amravati  
 509 district of the central india. *Journal of Earth System Science*, 130(1), 29.
- 510 Zhu, W., & Beroza, G. C. (2019). Phasenet: a deep-neural-network-based seismic  
 511 arrival-time picking method. *Geophysical Journal International*, 216(1), 261–  
 512 273.

## Supporting Information



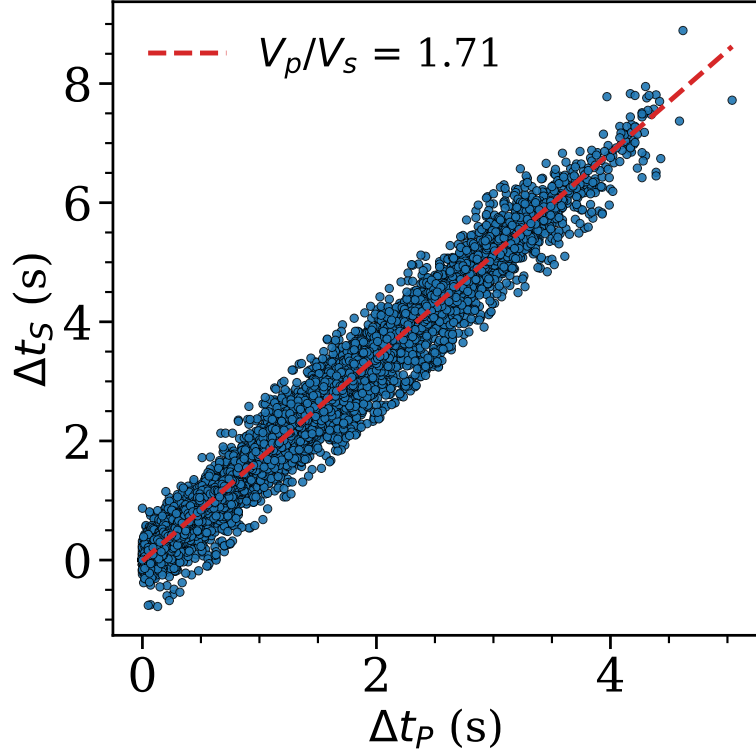
**Figure S1.** Map of the Palghar region showing the distribution of the seismic stations (triangles) used in this study. Green triangles denote stations operated by NGRI, while sky-blue triangles denote those operated by NCS. Note that the KASA and DGRI stations were later relocated to TLSR and DNVR, respectively, which were also managed by NCS. Red dots represent earthquake epicenters derived from machine learning-based detections and located using Non-LinLoc. The inset map of India highlights the Palghar region (red box) and marks other major intraplate earthquakes and swarm occurrences across the Indian peninsular region.

### Text S1. Velocity Model Constraints and Robustness Tests

To evaluate the robustness of the final 1-D velocity model derived for the Palghar swarm, we examine the stability of the inversion results, the sensitivity of the recovered velocity structure, and the consistency of the adopted  $V_p/V_s$  ratio.

#### S1.1 Estimation of the $V_p/V_s$ Ratio

The  $V_p/V_s$  ratio used in the velocity inversion was independently estimated from the manually picked P- and S-wave arrival times of the earthquakes used for velocity modeling. Figure S2 shows the relationship between the differential S- and P-wave travel ( $\Delta t_S$

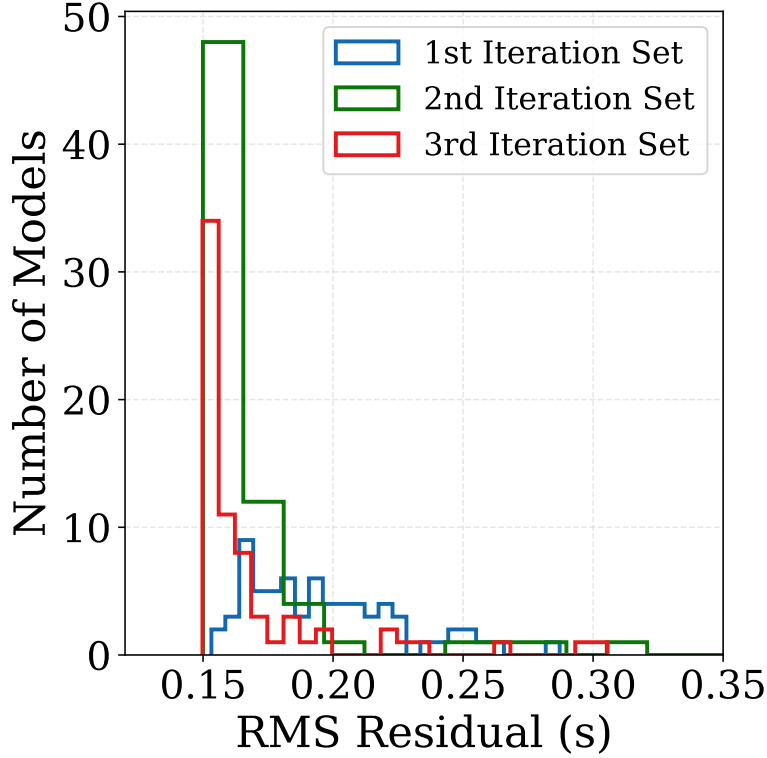


**Figure S2.** Relationship between the differential S- and P-wave travel times ( $\Delta t_S$  and  $\Delta t_P$ ) for the earthquakes used to estimate the  $V_p/V_s$  ratio. The dashed red line represents the best-fitting linear regression, yielding  $V_p/V_s \approx 1.71$ .

522 and  $\Delta t_P$ ) times measured at the same stations for the selected earthquakes. The slope  
 523 of the best-fitting regression line yields a  $V_p/V_s$  ratio of approximately 1.71.

## 524 S1.2 RMS Residual Statistics of Velocity Inversions

525 To reduce dependence on the starting velocity model, synthetic velocity models were  
 526 generated by perturbing the reference model of Srinagesh et al. (2020). Layer velocities  
 527 were randomly perturbed within  $\pm 0.7 \text{ km s}^{-1}$  and interface depths within  $\pm 2 \text{ km}$  to pro-  
 528 duce 70 alternative starting models for each inversion stage. Each model was inverted  
 529 independently using VELEST in simultaneous inversion mode. We adopted INVERTRATIO  
 530 = 2, such that the velocity model was updated only every second iteration, while earth-  
 531 quake hypocenters and station corrections were updated during every iteration. This strat-  
 532 egy allows the hypocenters and station delays to stabilize before subsequent velocity up-  
 533 dates, thereby improving inversion stability and convergence. After the completion of  
 534 each stage, consisting of 7 iterations, the final velocity model was obtained by comput-  
 535 ing the layer-by-layer mean of all successfully inverted velocity models. This averaged  
 536 model was then used as the starting model for the next inversion stage and was again  
 537 perturbed randomly to generate an ensemble of starting models. Three successive inver-  
 538 sion stages were performed, resulting in a total of 210 inverted velocity models, from which  
 539 the final mean velocity model was selected for use in all subsequent analyses.



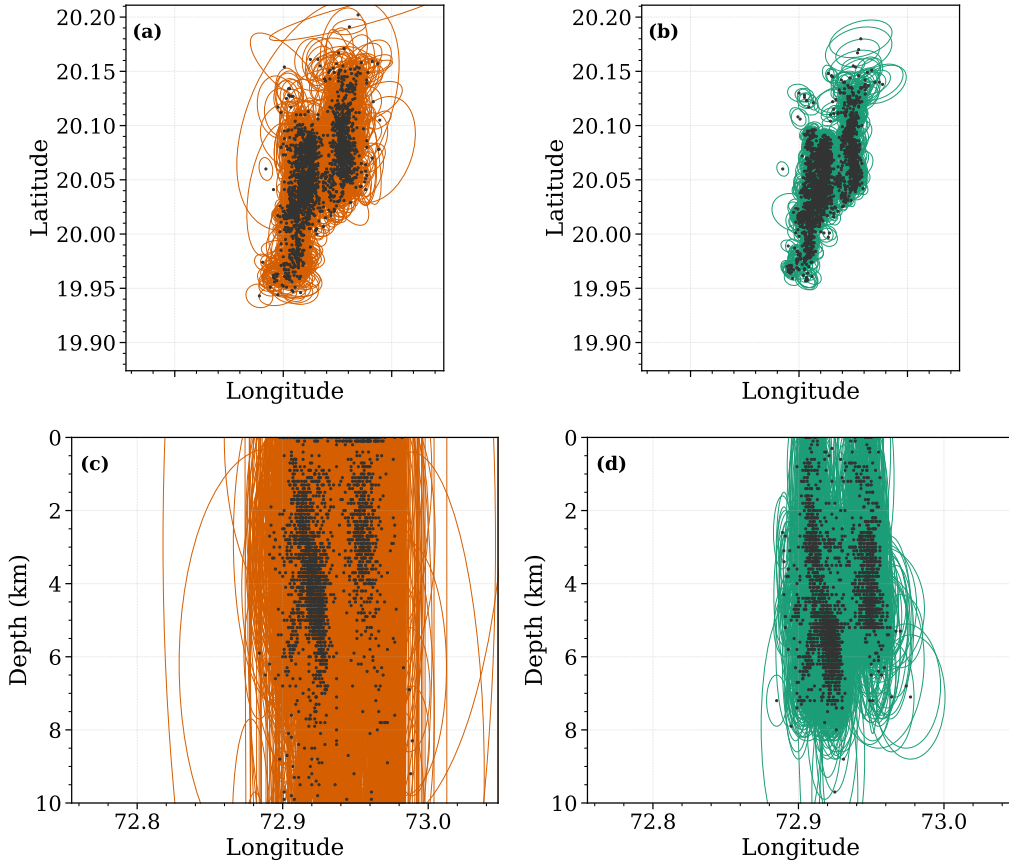
**Figure S3.** Distribution of RMS travel-time residuals obtained from velocity inversions during the three iteration stages of the inversion. The progressive reduction and convergence of RMS values indicate that the inversion approaches a stable velocity model.

540 Figure S3 shows the distribution of RMS travel-time residuals obtained from the  
 541 inversions during the three iteration stages described in the main text. The first iteration  
 542 set displays a broader range of RMS values reflecting the diversity of starting models.  
 543 With subsequent inversion stages, the RMS residuals progressively decrease and converge  
 544 toward a narrow range of values, indicating that the inversion approaches a stable  
 545 minimum-misfit solution.

546 Using the final velocity model, earthquakes were relocated with **HypoInverse**. A  
 547 comparison of location error ellipses before and after velocity inversion is shown in Fig-  
 548 ure S4, illustrating the reduction in location uncertainties achieved with the updated model.

## 549 **Text S2. Earthquake detection, phase picking, and relocation**

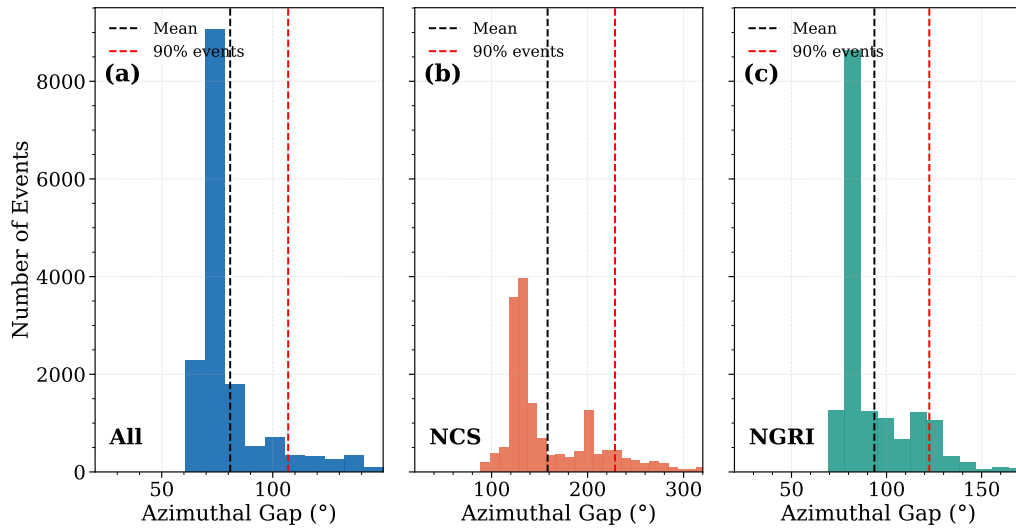
550 Earthquake detection and initial earthquake locations obtained using the BackPro-  
 551 jection and Matched-Filtering (BPMF) workflow (Beaucé et al., 2024). Within the work-  
 552 flow, continuous three-component waveform data from the NGRI and NCS seismic net-  
 553 works were corrected for instrument response and band-pass filtered between 1 and 20  
 554 Hz. The original pretrained **PhaseNet** model (Zhu & Beroza, 2019) was applied to the  
 555 continuous waveforms to generate continuous P- and S-wave probability time series, which  
 556 were subsequently backprojected across the seismic network to identify coherent earth-  
 557 quake signals. Earthquake detections were identified using a backprojection threshold



**Figure S4.** Comparison of earthquake location uncertainty ellipses before and after velocity inversion. Panels (a) and (c) show epicentral and depth–longitude distributions obtained using the initial velocity model, while panels (b) and (d) show results after relocation using the final velocity model derived in this study. Ellipses represent location uncertainties estimated from *HypoInverse*. The comparison illustrates the reduction in location uncertainty achieved using the updated velocity model.

558 of 2 (SI Figure S6), producing an initial machine-learning (ML) catalog of 57,826 detected  
 559 events. The ML catalog recovered approximately 95% of the manually picked high-quality  
 560 earthquakes (recorded by at least six stations) while substantially increasing the total  
 561 number of detected events.

562 For each detected event, P- and S-wave arrival times were refined using *PhaseNet*  
 563 in picking mode. The automatically determined phase arrivals were subsequently used  
 564 as input to the probabilistic earthquake location program *NonLinLoc* (Lomax et al., 2000).  
 565 Earthquake hypocenters were determined using the Equal Differential Time (EDT\_OT\_WT\_ML)  
 566 location algorithm, which combines differential travel-time residuals, weighted phase ob-  
 567 servations, and maximum-likelihood estimation to obtain robust hypocenter locations  
 568 that are less sensitive to outlying phase picks. Only events with picks from at least six  
 569 stations were considered for location, resulting in 20,638 well-located earthquakes. Ap-  
 570 proximately 88% of the manually picked earthquakes were successfully recovered after  
 571 the initial location stage.



**Figure S5.** Distribution of azimuthal gap values for earthquake locations obtained using different seismic network configurations. (a) Azimuthal gap distribution for all relocated events used in this study. (b) Azimuthal gap distribution for locations determined using only the NCS network. (c) Azimuthal gap distribution for locations determined using only the NGRI network. Black dashed lines indicate the mean azimuthal gap for each dataset, while red dashed lines mark the azimuthal gap below which 90% of the events are located. The distributions show that most relocated events have azimuthal gaps below  $\sim 107^\circ$  for the full dataset, whereas the individual NCS and NGRI network solutions exhibit larger azimuthal gaps due to reduced station coverage.

**Table 1.** Seismic station information used in this study. Latitude and longitude are given in degrees, and elevation is given in meters above sea level.

Station	Latitude (°N)	Longitude (°E)	Elevation (m)
AINE	19.8940	72.8420	73
GAGO	19.9870	73.0280	101
KAIN	20.0240	72.7750	37
KAWA	20.1180	72.8790	51
MASA	20.1910	72.9430	44
UDHW	20.0940	73.0030	57
PALG	20.0350	72.9131	162
TLSR	20.1255	72.9133	57
SSNE	20.0275	72.9583	130
AMBR	20.0131	72.8583	70
DNVR	19.9775	72.9198	78

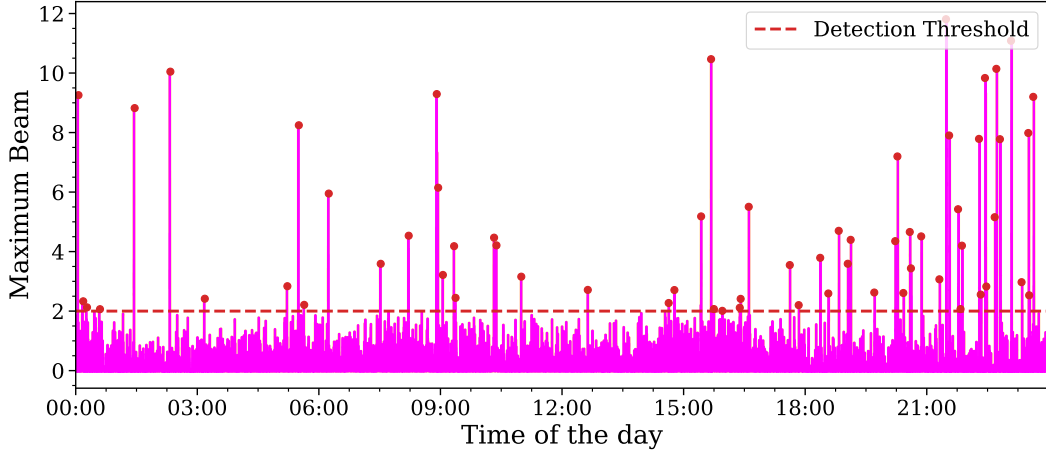
572 The **NonLinLoc** catalog was subsequently relocated using **GrowClust** (Trugman &  
573 Shearer, 2017), which refines relative earthquake locations by inverting differential travel  
574 times derived from waveform cross-correlation. Relative relocations were performed for  
575 event pairs with waveform cross-correlation coefficients of at least 0.8. This procedure  
576 yielded a final relocated catalog of 16,008 earthquakes, corresponding to approximately  
577 81% of the manually picked catalog. The relocated catalog forms the basis for all spa-  
578 tiotemporal analyses presented in this study.

579 Uncertainties in earthquake locations were estimated using the **GrowClust** boot-  
580 strap procedure, in which hypocenter locations are perturbed, and the relocation is re-  
581 peated to evaluate solution stability. Horizontal and vertical location uncertainties for  
582 events relocated with **GrowClust** are shown in Figure S7, which compares these uncer-  
583 tainties with those obtained from **NonLinLoc**, where location errors are derived from con-  
584 fidence regions of the posterior probability density function. The comparison highlights  
585 the improved relative location precision achieved using the cross-correlation-based re-  
586 location approach.

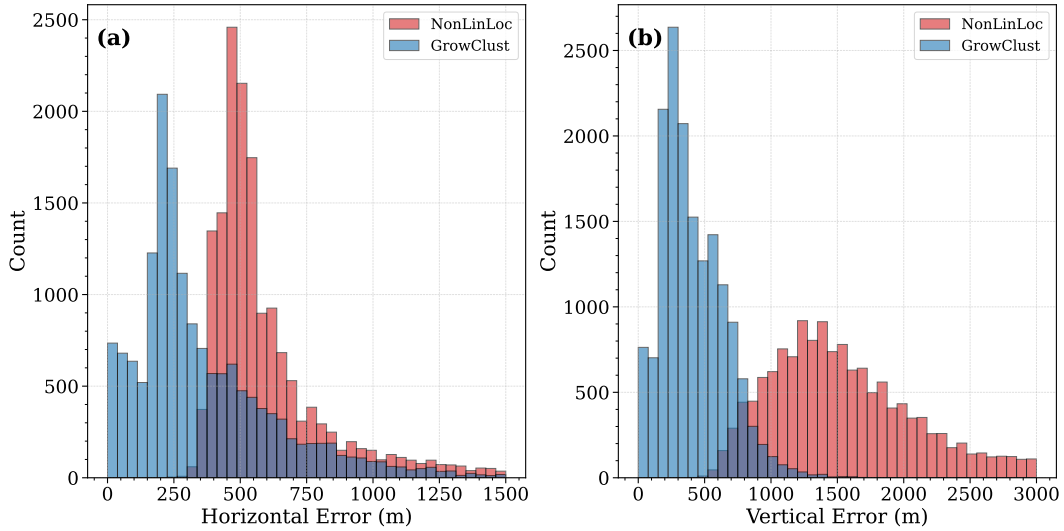
### 587 **Text S3. Moment tensor solutions**

588 This section provides additional details of the moment tensor inversion procedure  
589 summarized in the main text. Moment tensor inversions were performed using **MTTime**  
590 (Ichinose et al., 2014), with displacement Green’s functions computed using the **Com-**  
591 **puter Programs in Seismology (CPS)** package (Herrmann, 2013) for the final one-dimensional  
592 velocity model derived in this study. Instrument responses were removed prior to pro-  
593 cessing, and the three-component broadband waveforms were converted to displacement  
594 and band-pass filtered between 0.05 and 1 Hz using a two-pole Butterworth filter. For  
595 each station, waveform windows were extracted around the P-wave arrival, extending from  
596 2 s before the arrival to 6–8 s after the arrival, with the post-arrival window length ad-  
597 justed according to the source–receiver distance.

598 Inversion quality was evaluated using the variance reduction (VR) between the ob-  
599 served and synthetic waveforms across all available stations. Only solutions with VR ex-  
600 ceeding 50% were retained. Table 2 summarizes the source parameters, nodal-plane ge-  
601 ometry, moment tensor decomposition, and inversion quality metrics for the 37 accepted  
602 events, while Figure S8 shows the spatial distribution of the corresponding deviatoric mo-  
603 ment tensor solutions for the same set of earthquakes from Figure 3.



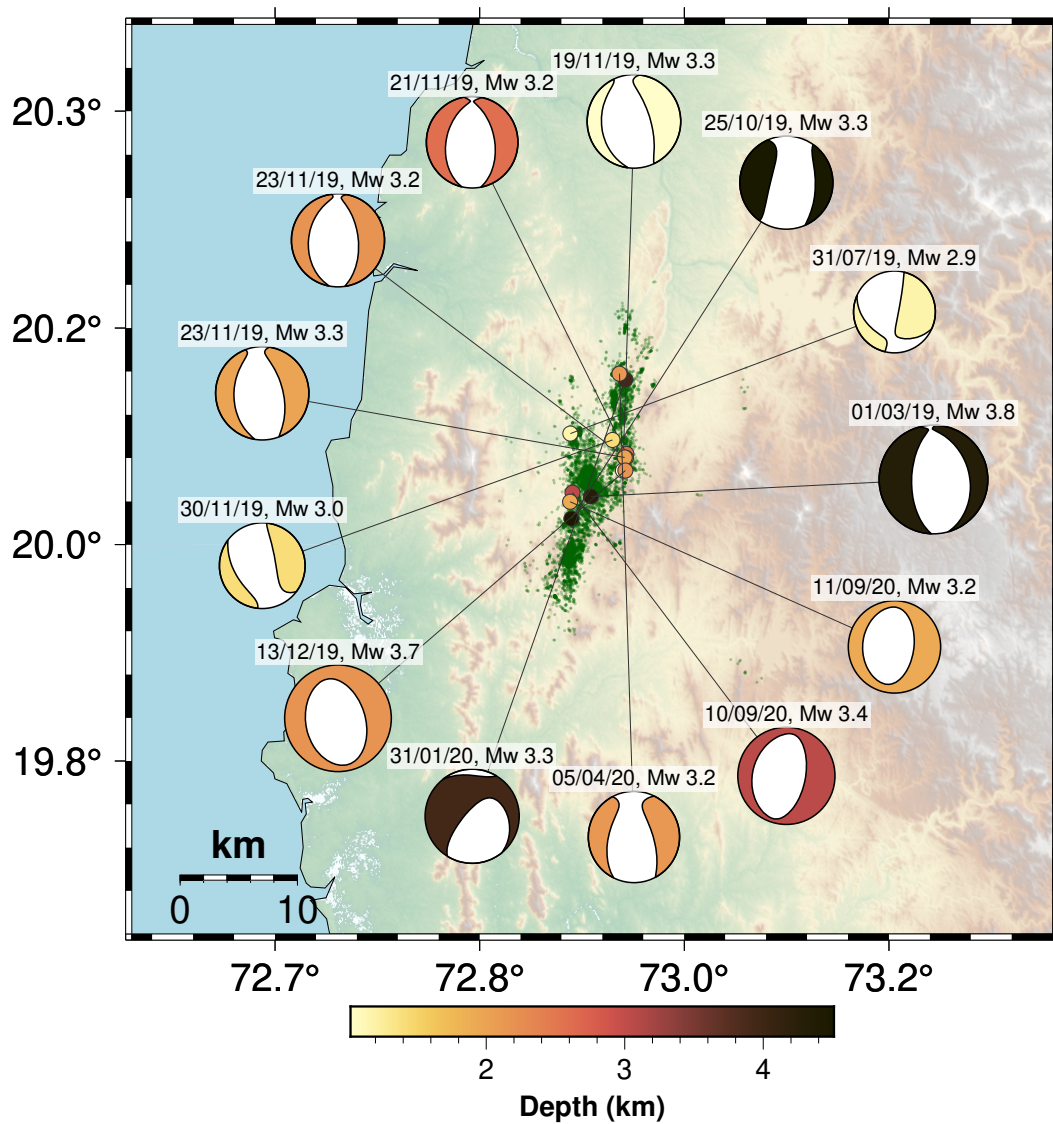
**Figure S6.** Example of the maximum beam power time series computed across the seismic network for a single day during the swarm sequence. The red dashed line marks the detection threshold (beam power = 2), and peaks exceeding this threshold were identified as candidate earthquakes. A total of 63 detections were identified for the day shown.



**Figure S7.** Comparison of earthquake location uncertainties obtained from NonLinLoc and GrowClust, showing (a) horizontal and (b) depth uncertainties.

604 The distributions of strike, dip, and rake angles are summarized in Figure S9. The  
 605 strike orientations define two dominant north–south trending populations that closely  
 606 match the relocated seismicity clusters and support the interpretation of two subparal-  
 607 lel fault strands. Dip angles cluster around  $\sim 30^\circ$  and  $\sim 60^\circ$ , consistent with conju-  
 608 gate normal faults, whereas rake angles are concentrated near  $270^\circ$ , indicating predom-  
 609 inantly normal faulting with only minor strike-slip components.

610 Representative waveform fits are shown in Figure S10. The examples correspond  
 611 to the accepted solutions with the highest (65%) and lowest (50%) variance reductions.  
 612 In both cases, the synthetic waveforms reproduce the observed radial, transverse, and  
 613 vertical components well over the adopted frequency band, demonstrating that the in-  
 614 ferred source mechanisms are robust across the range of accepted inversion qualities.



**Figure S8.** Deviatoric moment tensor solutions for the same set of earthquakes from Figure 3. Beachball symbols represent full deviatoric solutions (double-couple + CLVD) plotted at relocated epicenters and color-coded by depth.

615  
616

#### **Text S4. Influence of Catalog Resolution on Inferred Swarm Migration**

617 The migration behavior inferred from the Palghar swarm depends strongly on the  
618 accuracy and density of the earthquake catalog. SI Figure S11 compares the distance-  
619 time evolution obtained from the initial human-picked catalog containing 10,377 events  
620 with the migration envelopes derived from the relocated catalog used in this study. Al-  
621 though the initial catalog captures the broad temporal expansion of the swarm, the sub-  
622 stantial scatter in earthquake locations obscures the detailed migration fronts and makes  
623 it difficult to distinguish the different phases of swarm evolution. In particular, the de-  
624 layed activation of the eastern fault, the rapid migration episodes, and the temporal evo-  
625 lution of the leading seismicity fronts are not clearly resolved.

**Table 2.** Moment tensor solutions for earthquakes with  $M_l > 3.0$  and variance reduction exceeding 50%. The table lists the event origin time (Event ID), hypocentral location, magnitude, and the strike, dip, and rake parameters of the two nodal planes obtained from the moment tensor inversion. The percentage contributions of the double-couple (DC) and compensated linear vector dipole (CLVD) components of the deviatoric moment tensor solution are also reported.

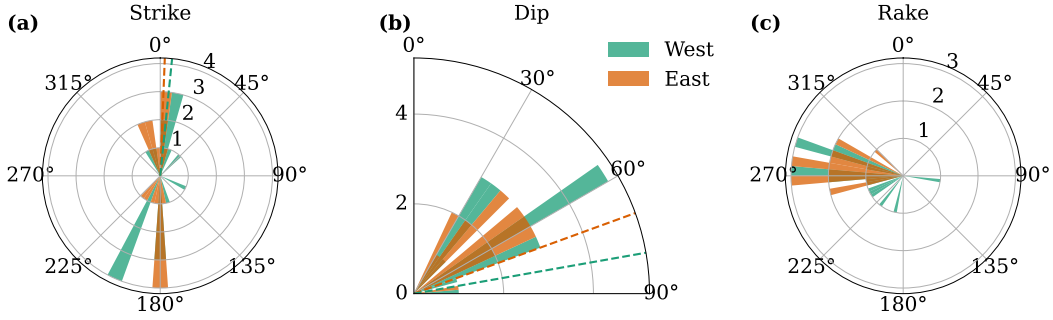
Event ID	Lat (°N)	Lon (°E)	Depth (km)	Fault Plane 1 (S/D/R)	Fault Plane 2 (S/D/R)	DC/CLVD/ISO (%)	VR (%)	$M_w$
2019_03_01_05_44_25	20.037	72.924	4.83	345/36/-103	181/55/-81	97/3/0	58.2	3.79
2019_03_10_08_25_26	20.054	72.908	1.64	1/57/-96	192/33/-81	98/2/0	52.9	2.74
2019_03_10_19_30_10	20.051	72.920	5.12	206/65/-67	341/34/-130	85/15/0	55.7	3.17
2019_03_10_22_21_49	20.055	72.908	1.56	335/58/-119	201/42/-52	68/32/0	51.3	2.76
2019_03_31_12_21_57	20.049	72.923	5.58	40/72/-73	176/25/-132	95/5/0	57.1	3.09
2019_05_13_06_31_26	20.046	72.921	6.32	180/31/-115	28/62/-76	84/16/0	59.0	2.99
2019_07_24_19_48_28	20.056	72.909	1.51	181/35/-90	1/55/-90	85/15/0	67.8	2.75
2019_07_31_14_50_03	20.085	72.907	1.18	113/37/-167	13/82/-54	93/7/0	59.2	2.86
2019_10_24_19_12_15	20.063	72.950	1.85	5/54/-84	175/37/-98	83/17/0	59.5	2.79
2019_10_25_22_35_41	20.020	72.908	4.51	201/60/-71	346/36/-120	50/50/0	59.8	3.27
2019_11_19_20_05_44	20.057	72.950	1.02	342/61/-99	180/30/-74	89/11/0	55.1	3.27
2019_11_19_20_26_47	20.059	72.949	0.82	173/32/-89	351/58/-91	94/6/0	55.1	2.79
2019_11_19_20_57_41	20.052	72.952	0.78	351/64/-95	183/26/-79	80/20/0	50.2	2.66
2019_11_21_01_50_28	20.070	72.953	2.60	193/43/-73	350/49/-106	100/0/0	50.2	3.19
2019_11_23_15_10_48	20.067	72.952	2.20	189/37/-78	354/54/-99	98/2/0	60.9	3.25
2019_11_23_20_31_13	20.067	72.951	1.99	344/53/-102	184/39/-74	91/9/0	60.2	3.26
2019_11_24_12_12_25	20.044	72.940	1.58	178/27/-82	349/63/-94	96/4/0	51.1	2.70
2019_11_25_21_15_55	20.039	72.945	1.58	178/27/-82	349/63/-94	96/4/0	51.1	2.70
2019_11_27_01_25_37	20.047	72.952	1.52	2/58/-95	190/32/-83	89/11/0	61.0	2.71
2019_11_30_11_41_33	20.081	72.941	1.42	358/62/-64	131/37/-130	62/38/0	54.6	3.01
2019_12_03_06_09_23	20.093	72.944	1.57	5/59/-78	164/33/-108	92/8/0	55.4	2.65
2019_12_03_22_17_43	19.964	72.889	1.87	13/58/-93	199/33/-85	87/13/0	50.1	2.66
2019_12_13_23_52_38	20.057	72.952	1.70	341/53/-105	185/40/-71	95/5/0	56.9	3.68
2019_12_14_09_05_05	20.101	72.950	1.24	9/59/-72	157/35/-117	97/3/0	52.1	2.70
2019_12_14_10_47_34	20.063	72.951	1.55	179/38/-94	4/52/-87	79/21/0	63.1	2.77
2019_12_15_00_31_28	20.015	72.920	2.91	202/65/-78	356/27/-114	94/6/0	55.7	3.03
2019_12_15_03_49_17	20.002	72.908	3.02	343/42/-144	225/67/-54	88/12/0	61.4	2.79
2019_12_15_19_27_56	20.049	72.922	4.63	181/89/98	277/8/6	96/4/0	59.2	2.82
2020_01_31_07_54_42	20.127	72.952	3.93	217/80/-61	324/30/-160	76/24/0	53.0	3.30
2020_02_23_04_43_02	20.067	72.952	1.92	177/43/-89	355/47/-91	95/5/0	53.2	2.73
2020_03_19_03_49_08	20.048	72.923	5.87	15/68/-66	144/32/-136	91/9/0	52.1	3.09
2020_04_01_18_09_18	20.044	72.921	6.29	165/31/-125	25/65/-71	92/8/0	55.5	3.01
2020_04_05_18_48_30	20.131	72.947	2.63	210/44/-46	337/60/-124	79/21/0	50.3	3.21
2020_09_10_22_27_48	20.040	72.909	3.06	11/58/-93	196/32/-85	91/9/0	53.6	3.39
2020_09_11_01_36_02	20.033	72.907	1.92	7/57/-88	184/33/-93	80/20/0	65.7	3.23
2020_09_21_21_20_44	20.050	72.922	2.87	201/34/-72	359/57/-102	86/14/0	58.5	2.90
2020_09_22_19_11_15	20.038	72.904	1.79	19/62/-90	200/28/-89	64/36/0	51.0	2.78

626 Application of **NonLinLoc** to the densified ML-enhanced catalog improves the over-  
627 all spatial coherence of the seismicity and captures the first-order expansion of the swarm  
628 (SI Figure S12). The eastern, western, and off-fault seismicity clusters become more read-  
629 ily distinguishable, and the broad diffusive character of the swarm is preserved. How-  
630 ever, the location precision remains insufficient to clearly resolve the fault-scale migra-  
631 tion fronts, persistent seismicity bands, and episodic along-strike and along-dip migra-  
632 tion patterns revealed by the subsequent **GrowClust** relocations.

633 The **GrowClust** relocations further sharpen the seismicity distribution and reveal  
634 the detailed spatiotemporal patterns analyzed in the main text. The improved relative  
635 locations delineate persistent localized seismicity along the western fault, distinct mi-  
636 gration episodes along both fault systems, and the close correspondence between on-fault  
637 and off-fault migration patterns. These comparisons demonstrate that the combination  
638 of catalog densification and high-precision relative relocation is essential for understand-  
639 ing the processes governing the evolution of the Palghar swarm.

## 640 **Text S5. Along-Strike, Along-Dip, and Off-Fault Migration Patterns** 641 **of the High-Resolution Relocated Seismicity**

642 To further investigate the migration behavior of the Palghar swarm, we analyzed  
643 the temporal evolution of seismicity projected along the strike and dip directions of the  
644 western and eastern faults. Earthquake locations were projected onto fault-parallel co-



**Figure S9.** Rose diagrams showing the distribution of focal mechanism parameters derived from moment tensor solutions for 37 earthquakes with local magnitude ( $M_l > 3.0$ ) in the Palghar swarm sequence. (a) Distribution of strike angles for the western (green) and eastern (orange) fault systems. The strike distributions consistent with the mapped geometry of the fault network. (b) Distribution of dip angles for both fault systems. Dashed lines indicate the mean dip orientation of the western and eastern fault planes inferred from relocated seismicity. Most events exhibit moderately to steeply dipping fault planes. (c) Distribution of rake angles for the two fault systems. The rake distributions are strongly clustered near  $\sim 270^\circ$ , indicating predominantly normal faulting with a minor strike-slip component.

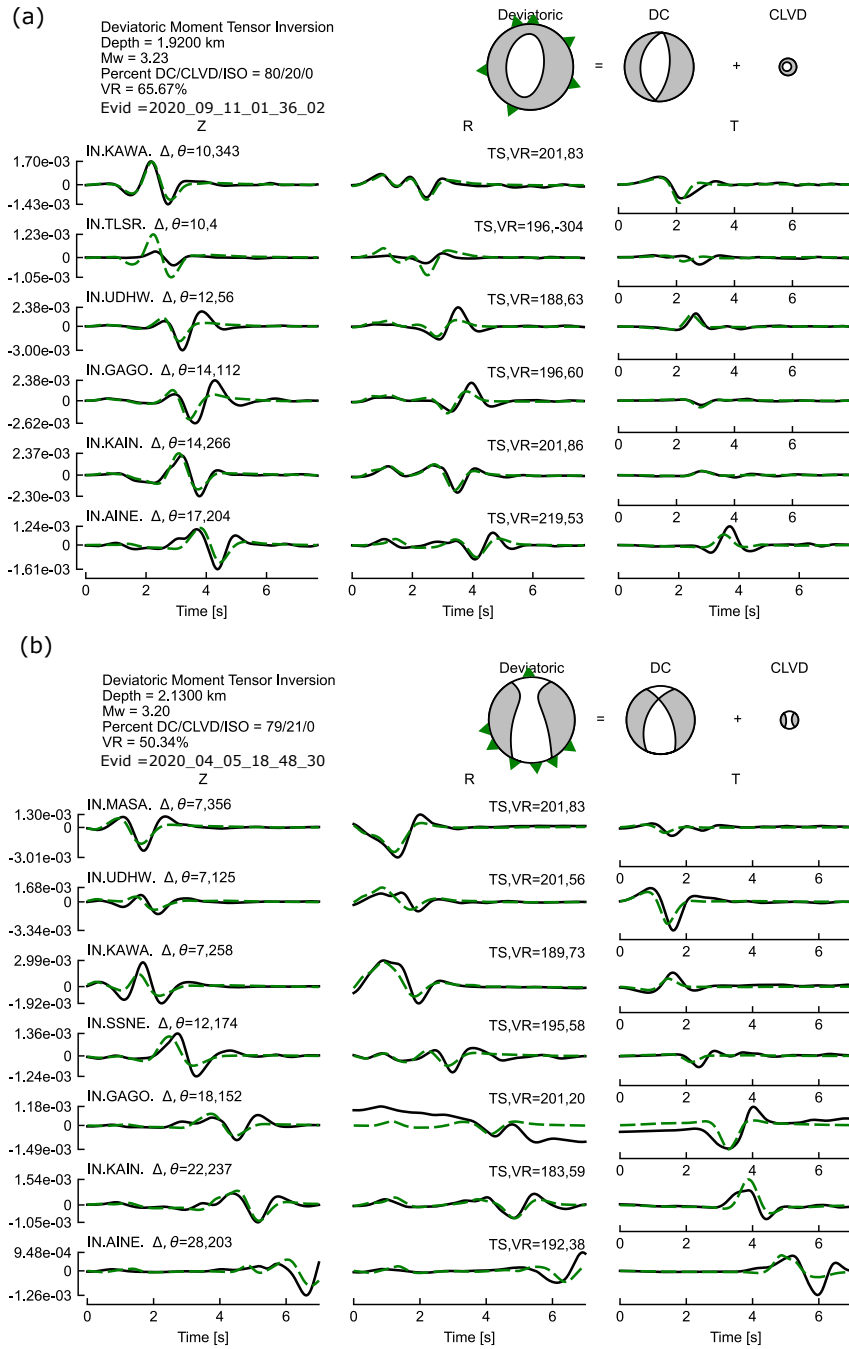
ordinate systems, and the evolution of the migrating seismicity fronts was tracked using the 95th and 5th percentile envelopes of seismicity through time.

For the western fault, seismicity remains strongly concentrated within persistent central bands located near  $\sim 1.2$  km in the along-dip direction and  $\sim 1.6$  km in the along-strike direction. To better resolve the migration of the seismicity fronts, events within these highly populated bands were excluded from the percentile calculations using windows of  $\pm 0.5$  km and  $\pm 0.7$  km in the along-dip and along-strike projections, respectively. The resulting distance-time evolution reveals strongly episodic migration, characterized by multiple propagating fronts, abrupt reversals, and back-propagating seismicity streaks along both strike and dip directions. A prominent down-dip migration episode is observed between approximately 260 and 460 days (migration 6 in SI Figure S13a), followed by a later phase of up-dip and southward migration between approximately 480 and 640 days (migrations 8 and 12 in SI Figure S13a,c). Several additional shorter-duration migration episodes occur throughout the sequence, indicating repeated reorganization of seismicity along the fault.

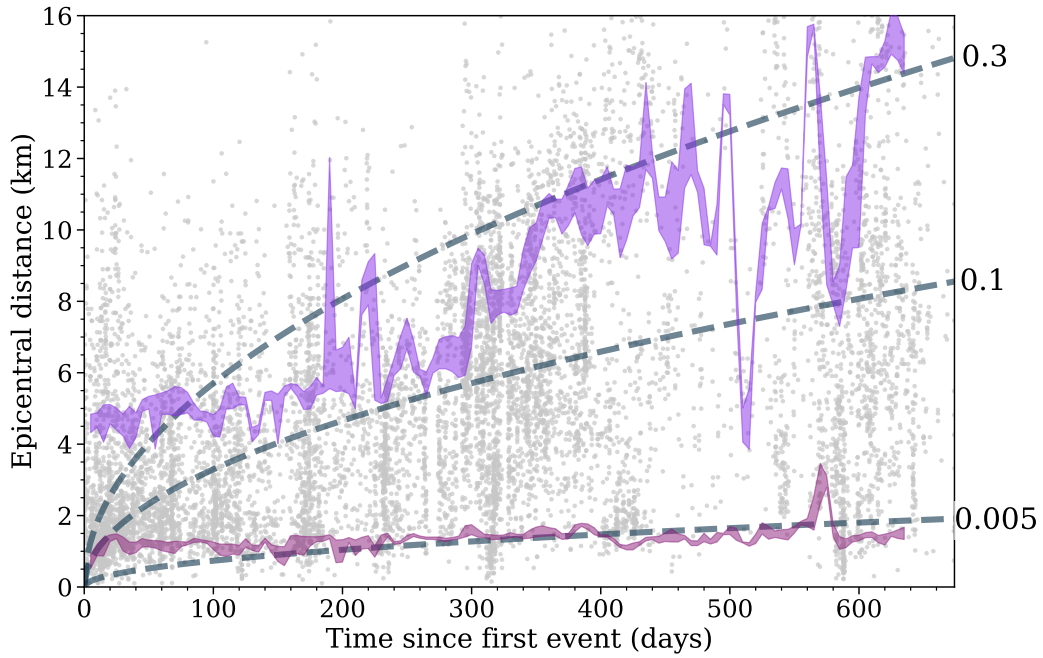
In contrast, the eastern fault exhibits a more spatially localized migration pattern. Following its activation at approximately 260 days, seismicity remains concentrated near the central portion of the fault before progressively expanding along the dip direction. After approximately 400 days, the seismicity envelope broadens, with expansion toward updip and a pronounced down-dip migration of the lower migration front during the later stages of the sequence (migration 2 in SI Figure S14a). In the along-strike projection, the leading migration front propagates northward between approximately 260 and 460 days (migrations 3 and 4 in SI Figure S14c). At the same time, a weaker migration front develops toward negative along-strike distances (migration 5).

To further examine the relationship between on-fault and off-fault seismicity, we compared the epicentral migration of the off-fault events with that of the eastern fault (SI Figure S15 a, d). Between approximately 260 and 460 days, both clusters exhibit nearly identical migration rates. During the initial stage of this interval (approximately 260–360 days), the off-fault seismicity also reproduces the northward migration pattern ob-

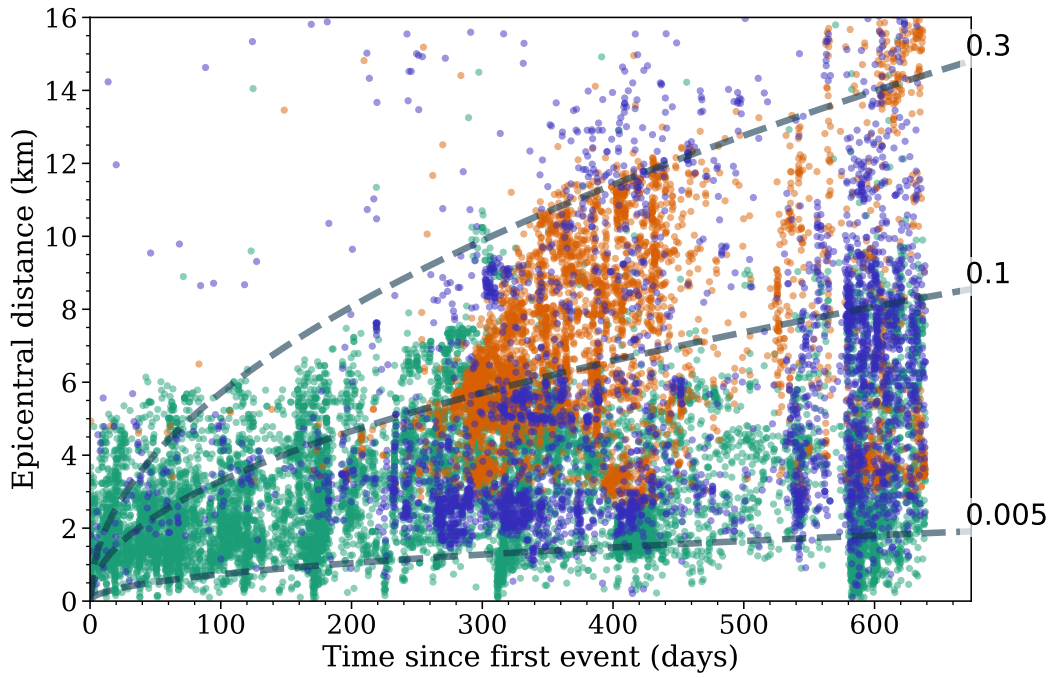
674 served along the eastern fault (SI Figure S15 b, c). This correspondence becomes even  
675 more apparent when the on-fault clusters are restricted to narrower bands of  $\pm 500$  m  
676 around the interpreted fault traces (SI Figure S15 e, f), demonstrating that the observed  
677 migration patterns are robust and insensitive to the choice of fault-zone width. During  
678 the later stages of the sequence (approximately 480–640 days), the southward and up-  
679 dip migration observed along the western fault is similarly mirrored by adjacent off-fault  
680 seismicity immediately west of the fault (SI Figure S16). These observations demonstrate  
681 that the off-fault seismicity closely tracks the migration of the active fault systems through-  
682 out the evolution of the swarm.



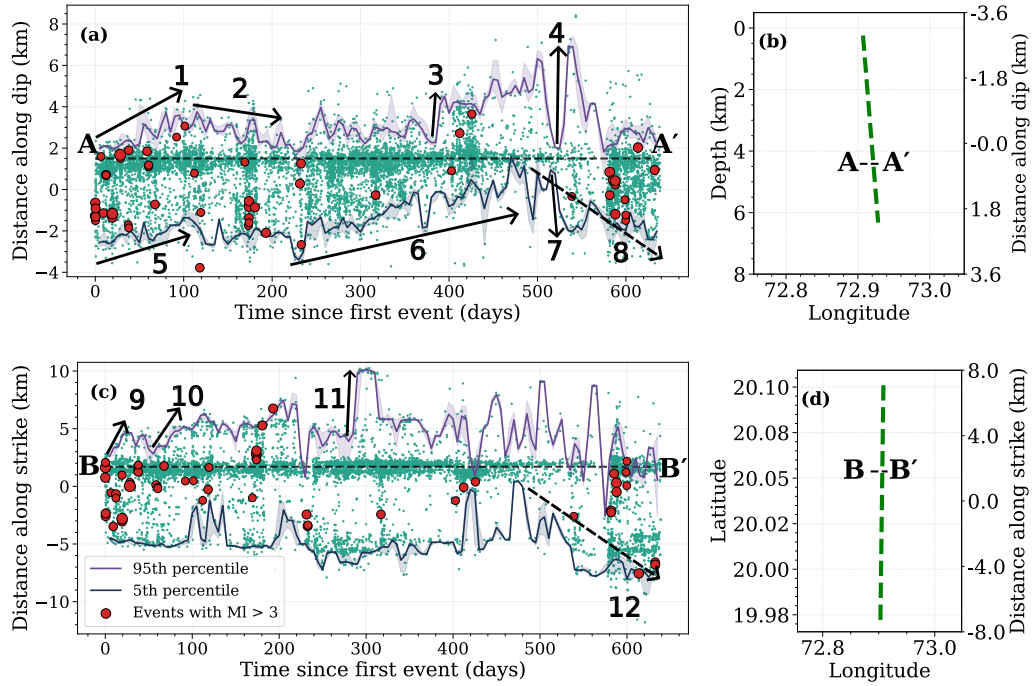
**Figure S10.** Deviatoric moment tensor inversion results for two Palghar earthquakes filtered between 0.05–1 Hz. (a) Event on 2020-09-11 ( $M_w$  3.23) with the highest variance reduction (VR = 65%). (b) Event on 2020-04-05 ( $M_w$  3.21) with the lowest variance reduction (VR = 50%). Black lines are observed waveforms, green lines are synthetics, shown for radial (R), transverse (T), and vertical (Z) components. Beachball diagrams show deviatoric solutions, with percentage decomposition and overall VR listed above each event.



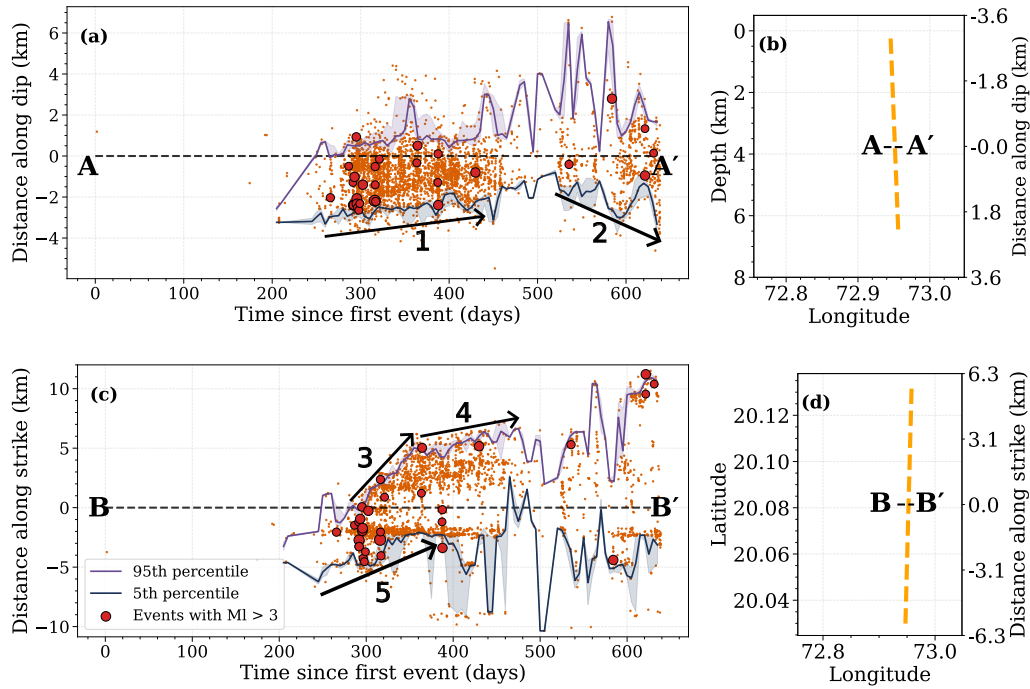
**Figure S11.** Epicentral distance as a function of time since the first detected event for the 10,377 earthquakes in the initial human-picked catalog (gray dots). Dashed curves indicate theoretical diffusion fronts corresponding to apparent diffusivities of  $D = 0.005, 0.1,$  and  $0.3 \text{ m}^2 \text{ s}^{-1}$ . The purple and pink shaded regions show the 95th and 5th percentile migration envelopes obtained from the relocated catalog, with shaded bounds denoting the associated  $\pm 2.5\%$  percentile range. The figure illustrates that the diffuse nature of the initial earthquake locations obscures the detailed migration fronts and spatiotemporal patterns that become evident only after relocation.



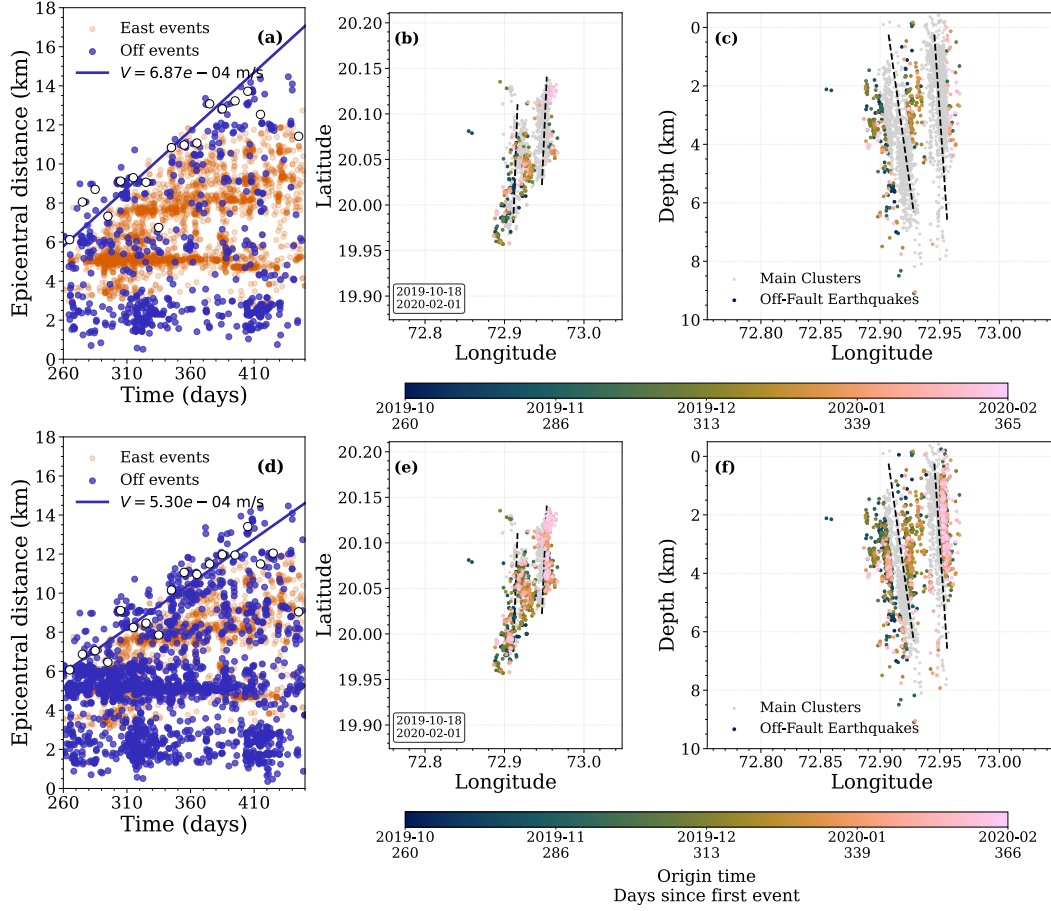
**Figure S12.** Distance–time evolution of the Palghar swarm using 20,638 earthquakes located with `NonLinLoc` and recorded by six or more stations. Green, orange, and blue dots represent earthquakes associated with the western cluster, eastern cluster, and off-fault seismicity, respectively. Epicentral distance from the first detected event is plotted as a function of time since swarm initiation. Dashed curves represent reference diffusion fronts corresponding to apparent diffusivities of 0.005, 0.1, and 0.3  $\text{m}^2 \text{s}^{-1}$ . The figure illustrates the broad spatiotemporal expansion of the swarm resolved by the `NonLinLoc` locations and serves as a reference for comparison with the higher-resolution catalog obtained from `GrowClust` relocations using waveform cross-correlation coefficients  $\geq 0.8$ .



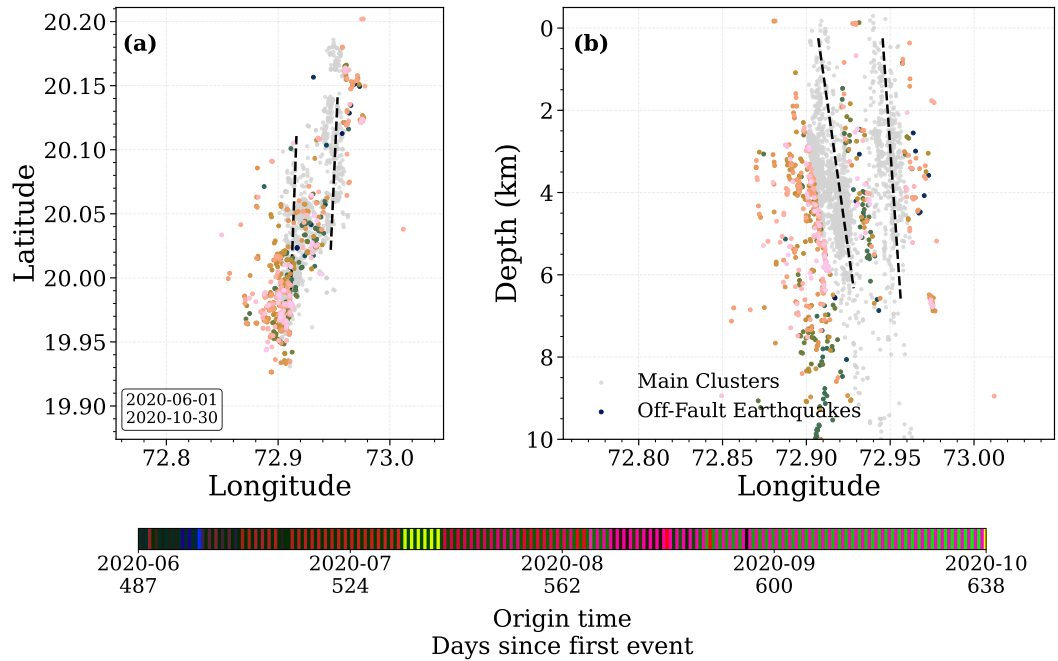
**Figure S13.** Spatiotemporal evolution of seismicity migration along the western fault. (a,c) Earthquake locations are projected onto along-dip and along-strike profiles as a function of time since the first detected event. Green dots represent relocated earthquakes and red circles denote events with  $M_l > 3$ . Purple and dark-blue curves show the 95th and 5th percentile positions of seismicity, respectively. The arrows and numbered labels identify major migration episodes. (b,d) Geometry of the projection profiles used for calculating along-dip and along-strike migration distances.



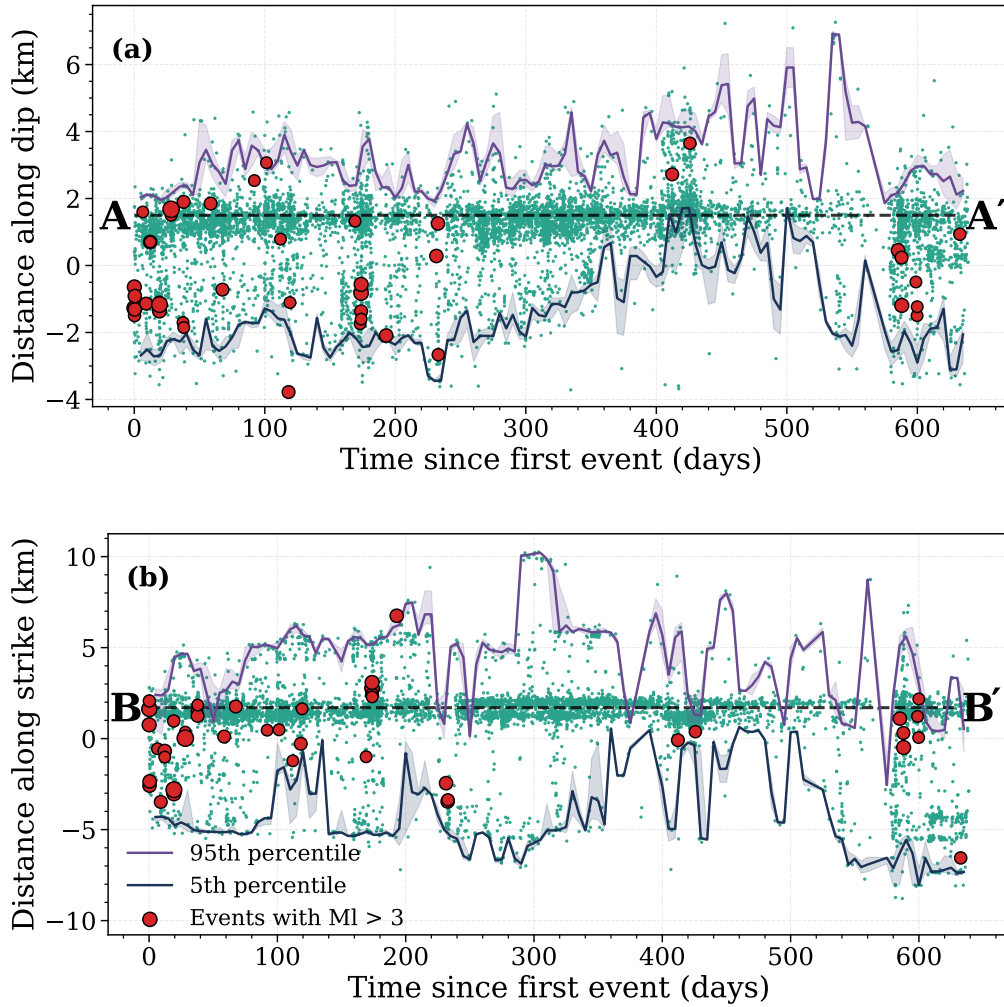
**Figure S14.** Spatiotemporal evolution of seismicity migration along the eastern fault. (a,c) Earthquake locations are projected onto along-dip and along-strike profiles as a function of time since the first detected event. Orange dots represent relocated earthquakes, and red circles denote events with  $M_l > 3$ . Purple and dark-blue curves show the 95th and 5th percentile positions of seismicity, respectively, highlighting the evolution of the migrating seismicity fronts. Arrows indicate the dominant migration direction along strike. (b,d) Geometry of the projection profiles used for calculating along-dip and along-strike migration distances.



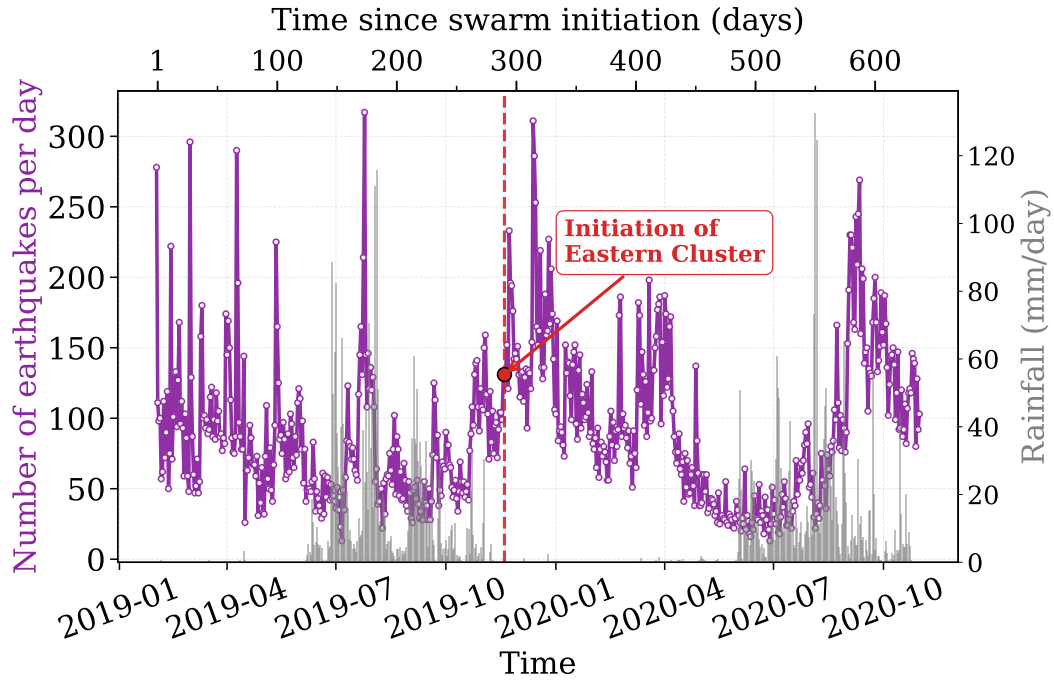
**Figure S15.** Migration of off-fault seismicity associated with the rapid migration episode on the eastern fault (migration 3 in Figure S14). (a) Epicentral distance as a function of time for off-fault earthquakes identified using a  $\pm 1$  km fault-zone definition. Orange and blue dots represent eastern-fault and off-fault earthquakes, respectively. The solid blue line shows a linear fit to the leading migration front of the off-fault seismicity, corresponding to a migration velocity of  $6.87 \times 10^{-4} \text{ m s}^{-1}$ . (b) Map-view and (c) longitude–depth projections of the seismicity during the 260–360 day interval. Gray dots represent on-fault earthquakes associated with the eastern and western fault systems, while colored dots denote off-fault earthquakes color-coded by origin time. The off-fault seismicity exhibits the same northward migration pattern observed along the eastern fault (migration 3 in SI Figure S14). (d–f) Same as (a–c), but using a stricter on-fault definition in which earthquakes within  $\pm 500$  m of the interpreted eastern and western fault planes are classified as on-fault events, with all remaining earthquakes assigned to the off-fault population. The estimated off-fault migration velocity ( $5.30 \times 10^{-4} \text{ m s}^{-1}$ ) and the spatiotemporal migration pattern remain largely unchanged, indicating that the observed off-fault migration is robust and not sensitive to the fault-zone width used to classify earthquakes.



**Figure S16.** Spatiotemporal evolution of off-fault seismicity during the late stages of the Palghar swarm sequence (2020-06-01 to 2020-10-30, 480 to 640 days). Gray dots represent earthquakes associated with the western and eastern faults, while colored dots denote off-fault earthquakes, color-coded by origin time. (a) Map view and (b) longitude–depth projection of the off-fault seismicity. Dashed gray lines indicate the approximate geometry of the western and eastern fault planes inferred from relocated seismicity. The off-fault earthquakes exhibit migration patterns that closely mirror contemporaneous migration along the main fault systems, including the up-dip expansion observed on both the western and eastern faults and the southward migration episode identified on the western fault. These migrations correspond to episodes 8 and 12 on the western fault and episode 2 on the eastern fault (see SI Figures S13 and S14), highlighting the close spatiotemporal relationship between on-fault and off-fault seismicity.



**Figure S17.** Spatiotemporal evolution of seismicity along the western fault after restricting the fault-zone width to  $\pm 500$  m around the interpreted fault plane. Earthquakes are projected onto the along-dip (a) and along-strike (b) directions as a function of time since the first detected event. Green dots represent relocated earthquakes, while red circles denote larger events with local magnitude ( $M_l > 3$ ). The horizontal dashed lines mark the dominant seismicity band. Purple and dark-blue curves show the 95th and 5th percentile positions of seismicity, respectively, with shaded regions indicating the associated  $\pm 2.5\%$  percentile bounds. Despite the substantially narrower fault-zone definition, a persistent and spatially confined seismicity band remains evident in both the along-dip and along-strike projections throughout most of the swarm evolution, demonstrating that the long-lived seismicity concentration along the western fault is a robust feature and not an artifact of the fault-zone width used for event classification.



**Figure S18.** Time evolution of daily earthquake frequency during the Palghar swarm sequence from February 2019 to October 2020, comprising a total of 57,826 earthquakes detected in the ML-based earthquake catalog developed in this study. The purple curve shows the number of earthquakes per day, while gray bars represent daily rainfall. The red dashed line marks the onset of seismicity on the eastern fault. The periods of highest seismicity rates occur during the early stages of the sequence and do not show an obvious correspondence with seasonal rainfall variations. Daily precipitation data were obtained from the NASA Langley Research Center Prediction Of Worldwide Energy Resources (POWER) Project Daily API.