

DISTRIBUTION ANALYSIS OF MICRO-EARTHQUAKES IN GEOTHERMAL AREAS BY USING COUPLED VELOCITY-HYPOCENTER AND DOUBLE DIFFERENCE METHODS

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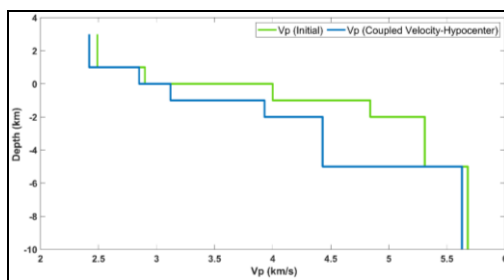
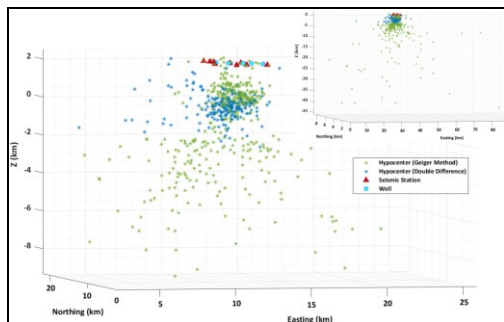
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Graphical abstract



Abstract

Accurate and unbiased determination of hypocenter locations remains a major challenge in microearthquake (MEQ) studies. The reliability of the determination process greatly contributes to the real-time monitoring of geothermal reservoir activities. Therefore, this study focuses on monitoring geothermal reservoirs through the spatial distribution of MEQ locations using inversion methods. The inversion methods applied in this research include the single hypocenter determination method, which is used to identify the initial locations of MEQs, and the Coupled Velocity-Hypocenter simultaneous inversion method, which is employed to derive a new 1D velocity model for the MEQ cluster area. In addition, the Double-Difference method is utilized to enhance the accuracy of MEQ hypocenter locations by considering coherence factors between hypocenters. The data processing workflow demonstrates that the 1D velocity model refined by incorporating station corrections within the MEQ cluster area significantly improves the accuracy of hypocenter locations in the geothermal zone. This improvement is evidenced by a reduction of the average RMS residual to 0.07 s, indicating that the relocated hypocenters are reliable and that the derived velocity structure effectively represents the subsurface and approximates the true seismic velocity distribution. Most relocated MEQs are concentrated around the injection and production well zones, with approximately 94.3% of events occurring between 0.5 and -1.5 km depth, corresponding to the active geothermal reservoir. These clusters delineate fracture zones characterized by intense fluid flow activity.

Keywords: Coupled velocity-hypocenter, double-difference relocation, geothermal, hypocenter determination, micro-earthquake (MEQ)

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1.0 INTRODUCTION

The development of geothermal systems involves continuous fluid extraction and injection that progressively modify the physical and mechanical properties of reservoir rocks [1]–[3]. Previous studies (e.g., Maurer *et al.*, 2020) have shown that fluid circulation and pressure variations alter the stress field, enhance anisotropy, and trigger fracture activation [4]. These processes affect subsurface conditions by changing pore pressure, stress distribution, and fluid volume, leading to fracture formation or reactivation expressed as clusters of microearthquakes (MEQs) [5], [6]. Consequently, the spatial and temporal distribution of MEQs provides essential insights into reservoir dynamics and fluid migration within geothermal fields [7]–[9].

Accurate MEQ hypocenter determination is crucial for evaluating reservoir performance and operational safety [10]–[13]. However, precision is often constrained by phase-picking errors, limited station geometry, and velocity-model bias. Previous studies highlight that refining network configuration and velocity models significantly improves depth accuracy and reduces hypocentral errors [14], [15].

This study was conducted in a vapor-dominated geothermal field located in East Java, Indonesia, characterized by active fault zones and high subsurface fluid circulation. Seismic monitoring was carried out using eleven broadband seismometers deployed around the production and injection well clusters. Each station operated at a sampling rate of 100 samples per second, with a frequency response of 0.1–50 Hz [16]–[20]. These settings ensured sufficient temporal resolution and signal fidelity for detecting low-magnitude events typical of geothermal environments [21], [22]. After signal-to-noise (S/N) evaluation and precise P-S phase picking, a sequential inversion workflow was implemented to improve spatial accuracy. The Geiger method was first applied to determine initial hypocenters, followed by Coupled Velocity-Hypocenter inversion to refine the 1D P-wave velocity model [15], [23], [24]. Final relocation using the Double-Difference method minimized differential travel-time residuals between closely spaced events. Simultaneously, the V_p/V_s ratio derived from the Wadati diagram was used to assess lithological variations and fluid saturation within the reservoir [14], [25]. This integrated approach effectively reduces velocity-model bias and enhances MEQ location accuracy, providing a robust basis for interpreting fracture connectivity and reservoir dynamics in geothermal systems [9], [20].

2.0 METHODOLOGY

The distribution of micro-earthquake (MEQ) hypocenters was determined through an inversion

workflow designed to maximize spatial accuracy and minimize residual errors [26]. The precision of these results primarily depends on the quality of seismic recordings, the accuracy of P and S wave picking, the network geometry, and the appropriateness of the velocity model [27]. As highlighted by Leong and Zhu (2024), iterative location algorithms achieve optimal solutions by minimizing the misfit between observed and calculated travel times, while Blanck *et al.* (2022) demonstrated that double-difference relocation combined with waveform cross-correlation can substantially enhance event precision [28], [29]. Building on these principles, this study applies an integrated stepwise procedure to ensure reliable and high-resolution mapping of MEQ activity within the geothermal field [30].

2.1 Waveform Processing and Phase Picking

Waveform data were recorded by eleven broadband seismometers operating at 100 samples per second. The P and S wave arrivals were manually identified using Seisgram2K, which enables precise visualization of amplitude and frequency variations. After applying a strict quality-control threshold, 1,092 reliable phase arrivals were obtained [16]–[19]. Accurate phase picking is critical for hypocenter determination, as emphasized by Pei *et al.* (2025), who demonstrated that consistent manual identification achieves up to 98 % agreement with reference datasets [31]. Li *et al.* (2020) noted that waveform fidelity particularly affects inversion stability, while Zheng *et al.* (2025) and Dahal *et al.* (2024) highlighted that noise suppression and full-waveform analysis substantially reduce timing uncertainty [32]–[34]. Accordingly, this study applies rigorous waveform processing and manual phase picking to ensure the precision and reproducibility of seismic inversion results. The determination of P and S wave arrivals across all stations is shown in Figure 1.

2.2 Wadati Diagram and Geological Consistency

A Wadati diagram constructed from manually picked P and S wave arrivals yielded a V_p/V_s ratio of 1.12, indicating a stable velocity structure characteristic of steam saturated fractures in vapor dominated geothermal systems [35]. Similar findings were reported by Muttaqy *et al.* (2023) in Central and East Java and by Afnimar *et al.* (2015) beneath the Lembang Fault, both identifying low V_p/V_s values in fluid-rich volcanic deposits. The agreement with these studies supports the interpretation that the low V_p/V_s ratio observed here accurately represents steam-bearing fractured zones within the geothermal reservoir [25], [36]. The resulting Wadati diagram demonstrating this consistency is shown in Figure 2.

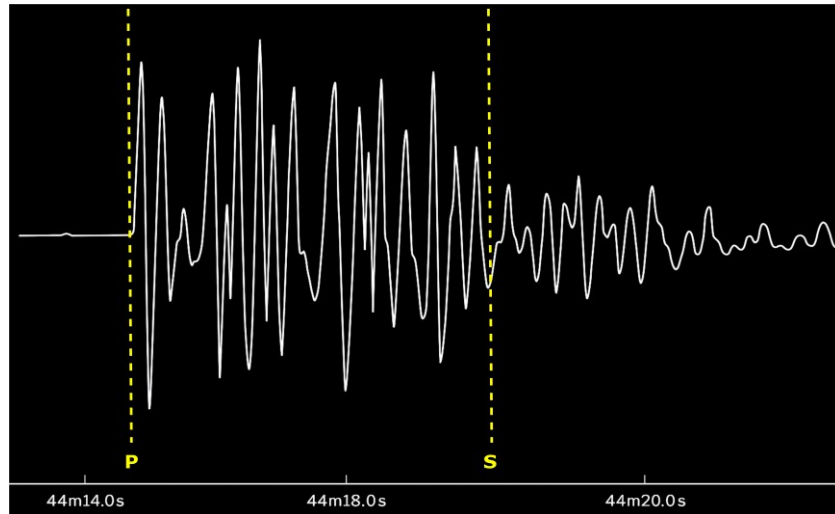


Figure 1 Determination of P and S wave arrival times

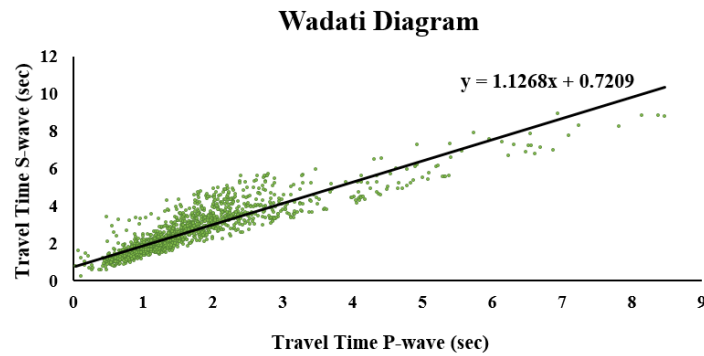


Figure 2 Wadati Diagram for the the selection of events to be consistent with geological information [25], [36]

2.3 Initial Hypocenter Estimation Using the Geiger Method

The initial MEQ hypocenters were estimated using the Geiger method, an iterative least-squares inversion that minimizes the residuals between observed and calculated P- and S-wave travel times. This classical single event approach employs station coordinates and a one dimensional velocity model to derive source positions and origin times [37]. As highlighted by Cheng *et al.* (2018) and Liao *et al.* (2022), the method remains a fundamental tool for earthquake and microseismic localization, efficiently linearizing the nonlinear source-location problem through successive iterations. Despite its sensitivity to phase-picking errors and simplified velocity assumptions, it provides a robust first-order approximation of hypocenter geometry [38], [39]. As noted by Wu *et al.* (2018), such initial estimates form the essential foundation for subsequent velocity-model refinement and event relocation [40].

2.4 Coupled Velocity–Hypocenter Inversion for 1-D Velocity Model Refinement

To improve event location accuracy, a coupled velocity-hypocenter inversion was performed using

MATLAB based routines. This method simultaneously refines the one-dimensional velocity structure, hypocenter coordinates, and station corrections by minimizing travel-time residuals. The inversion utilized initial parameters from the Geiger results, P-S travel times, station coordinates, and the V_p/V_s ratio derived from the Wadati diagram, iterating until the RMS residual fell below 0.10 s with at least eight high-quality phases per event [9]. Previous studies Huang *et al.*, (2019) have shown that joint inversion effectively reduces location bias caused by velocity heterogeneity, enhances the representation of layered structures, and improves microseismic mapping precision in geothermal reservoirs [41]. Consistent with these findings, the coupled inversion applied in this study significantly enhances both the reliability of the velocity model and the spatial accuracy of MEQ hypocenter determinations [42].

2.5 Double-Difference Relocation

Final hypocenter relocation was carried out using the Double-Difference algorithm in MATLAB to enhance spatial accuracy. The method minimizes differential travel-time residuals between pairs of nearby events recorded at common stations, assuming that closely spaced earthquakes share similar raypaths. This

approach effectively reduces common-mode errors caused by unmodeled velocity heterogeneity and yields precise relative locations. The processing used $\text{MAXDIST} \leq 15$ km, $\text{MAXNGH} \leq 10$, $\text{MINLINK} = 8$, and achieved an average RMS residual of 0.07 s [43]. Similar to findings by Muttaqy *et al.* (2023) in Central and East Java and by McBrearty and Beroza (2025), the use of waveform cross-correlation and differential times improves clustering consistency and fault delineation while minimizing depth uncertainty. Accordingly, the HypoDD method in this study provides a robust refinement of MEQ hypocenter distributions, ensuring reliable spatial patterns for subsequent geological and geothermal interpretation [14], [25].

2.6 Software Utilized

All inversion and relocation analyses were conducted using MATLAB based routines, providing a flexible and reproducible framework. MATLAB enables precise control of inversion parameters and ensures numerical stability throughout the processing workflow [44]. Lomax and Savvaidis (2022) highlighted its effectiveness for both probabilistic and deterministic earthquake localization [45]. Waveform visualization and phase picking were performed using *Seisgram2K*, which supports high-resolution inspection and interactive identification of P and S wave arrivals [46].

3.0 RESULTS AND ANALYSIS

Each stage, from waveform processing to final hypocenter relocation, was designed to enhance spatial resolution and minimize residual errors. Key metrics such as RMS residuals, depth distribution, and spatial clustering confirm the improved accuracy and reliability of the relocated hypocenters.

3.1 1-D Velocity Model Refinement

The one-dimensional (1-D) P-wave velocity model was refined using the Coupled Velocity-Hypocenter inversion, which simultaneously updates velocity parameters and event locations to minimize travel-time misfits. This approach was applied to the clustered MEQ dataset with station corrections to account for near-surface heterogeneity [47]. Figure 3 illustrates that while both initial and refined models share similar shallow gradients down to 2.3 km, notable improvements appear between 2.3 and 10 km, where the refined model yields smoother velocity transitions and enhanced vertical resolution. The P-wave velocity increases from 3.0 km/s near the surface to 5.5 km/s at depth, representing the transition from volcanic deposits to consolidated

basement rocks [48]. The updated model, adopted for subsequent HypoDD relocation, substantially reduced RMS residuals and improved hypocenter clustering, thereby enhancing the structural interpretation of fault and fluid-related seismicity. Comparable studies by Huang *et al.* (2019), Okamoto *et al.* (2018), and Katsumata & Nishimiya (2025) similarly demonstrate that coupled inversion and refined local velocity structures significantly improve the accuracy of microseismic imaging in geothermal fields [42], [49].

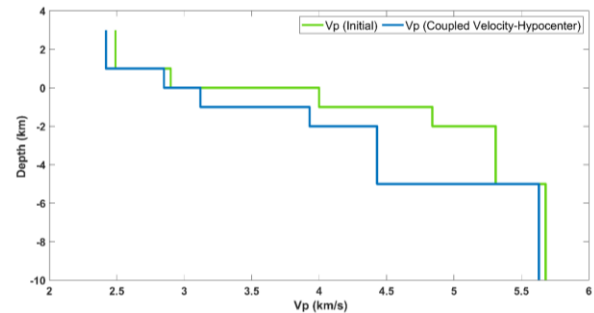


Figure 3 The comparison between the final 1D velocity model for the cluster area, obtained by Coupled Velocity-Hypocenter method with the initial velocity mode

3.2 Initial MEQ Hypocenter (Geiger Method)

The initial micro-earthquake (MEQ) hypocenters were determined using the classical Geiger travel-time inversion method applied to three months of continuous seismic data from eleven broadband stations within the geothermal field. A total of 457 events were successfully located, with 88.84 % situated within the survey boundary and 11.16 % outside due to edge-array limitations. The RMS residuals ranged from 0.01 s to 38.97 s (average 0.86 s), consistent with the sensitivity of the Geiger algorithm to arrival-time and velocity-model uncertainties as noted by Cheng *et al.* (2018) [38]. The resulting epicenters (Figures 4-6) show broad spatial dispersion around production and injection wells, with only partial alignment to mapped fault traces—a pattern commonly produced by irregular station geometry [39]. Hypocentral depths vary from -41.1 km to +2.3 km, with 37.1 % of events concentrated between -0.5 km and +0.5 km, corresponding to the upper geothermal reservoir. This vertical spread reflects uncorrected velocity heterogeneity, a known limitation of conventional Geiger localization, whereas the shallow clustering indicates active microseismicity within thermally and hydraulically influenced zones. Collectively, these preliminary results form a first-order baseline for subsequent Double-Difference relocation and velocity refinement.

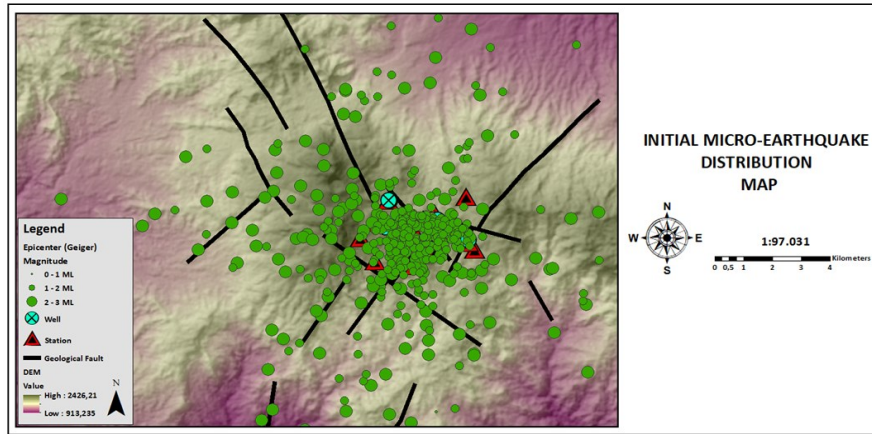


Figure 4 The distribution map of initial MEQ epicenter location as a result of Geiger method

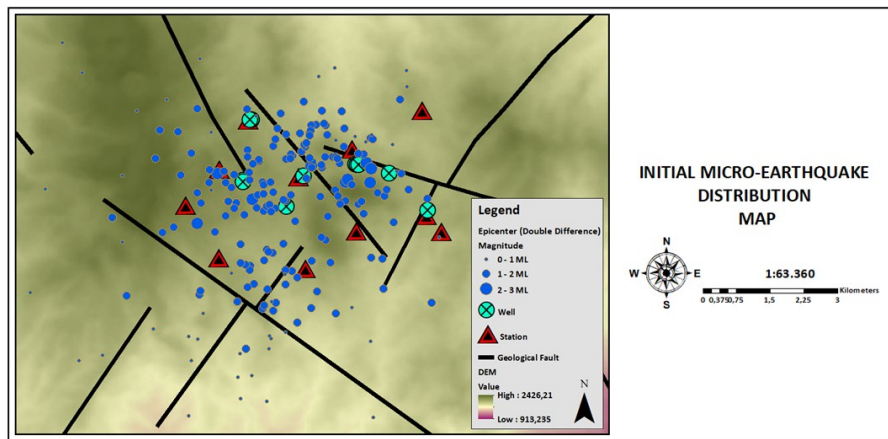


Figure 5 Zoomed-in area in figure 4, the whole location of MEQ epicenter location as a result of Geiger method

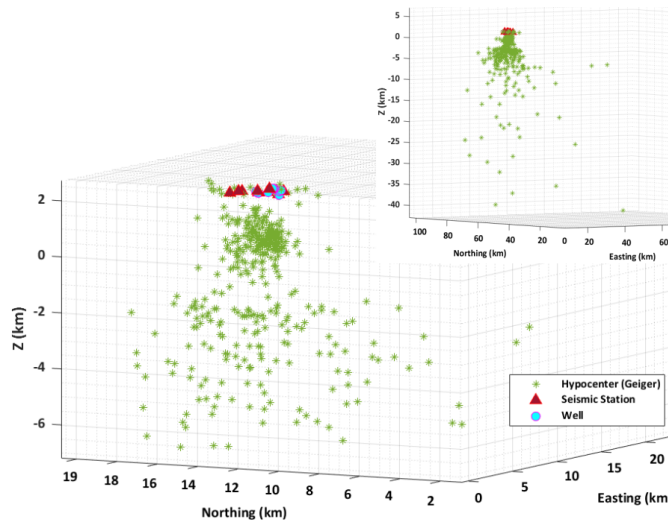


Figure 6 The distribution of initial MEQ hypocenter based on Geiger method, represented by the green dots

3.3 Relocated MEQ Hypocenter (Double-Difference Method)

The relocation of micro-earthquake (MEQ) hypocenters was conducted using the Double-Difference (HypoDD) algorithm with the refined 1-D

velocity model derived from the previous inversion stage. Optimized parameters $MAXDIST \leq 15$ km, $MAXNGH \leq 10$, $MINLINK = 8$, and RMS residual 0.07 s, were applied to minimize travel-time bias and ensure stable convergence [43]. As noted by Muttaqy et al. (2023), these configurations enhance relative

accuracy by correlating residuals among neighboring events that share similar raypaths. The relocated epicenters (Figures 7-9) display tighter clustering and improved spatial coherence compared to the Geiger results, with most events concentrated near production and injection wells, reflecting fluid-induced seismicity within the geothermal reservoir. Peripheral clusters along mapped fault traces indicate fracture reactivation, consistent with regional microseismic patterns in East

Java volcanic fields [25]. Vertically, the hypocenters range from -2.6 km to +2.1 km, with 94.3% of events confined between -1.5 km and +0.5 km, marking the reservoir zone. The 93.66% reduction in vertical scatter confirms the enhanced precision achieved through HypoDD relocation, in agreement with the findings of McBrearty and Beroza (2025), who demonstrated the method's capability to suppress common-mode velocity errors and refine focal depth accuracy [14].

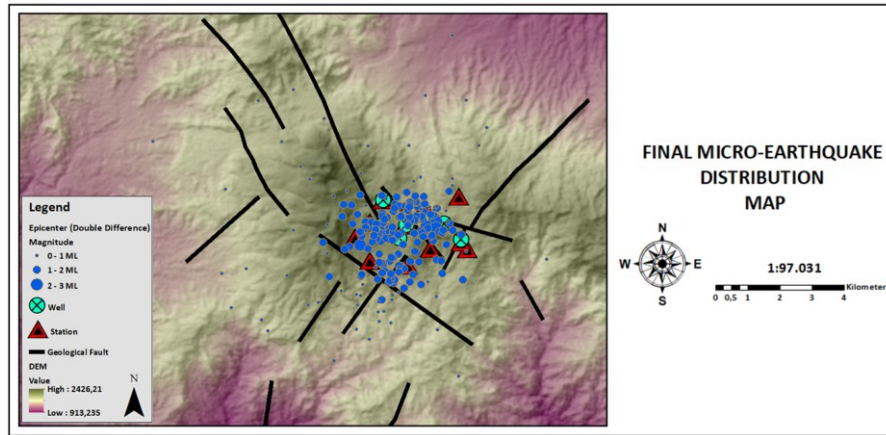


Figure 7 The distribution map of MEQ epicenter location as a result of Double-Difference method

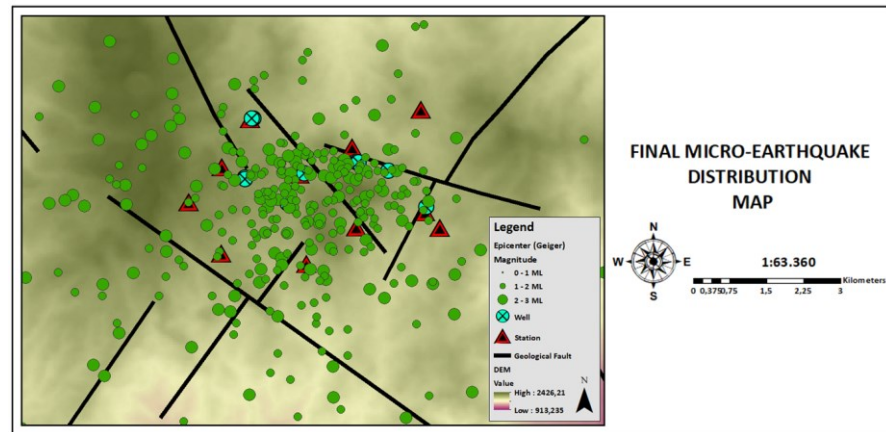


Figure 8 Zoomed-in area in figure 7, the whole location of MEQ epicenter location as a result of Double-Difference method

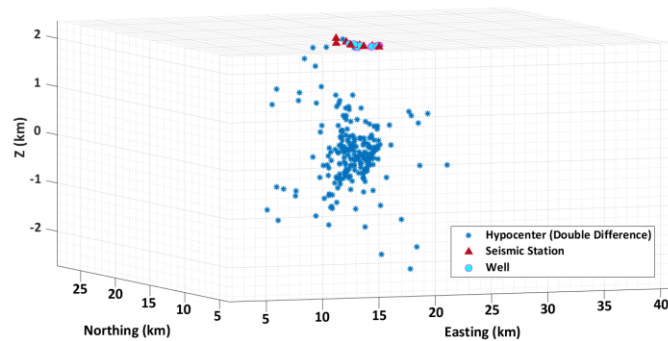


Figure 9 The distribution of MEQ hypocenter from Double-Difference method, represented by the blue dots

3.4 Quality Metrics and RMS Residual Comparison

The reliability of the relocated micro-earthquake (MEQ) hypocenters was assessed using Root Mean Square (RMS) residuals as the primary quality indicator. The comparison between the initial Geiger and the final Double-Difference (HypoDD) solutions (Figure 10) shows a substantial improvement, with RMS values reduced from a wide 0.01–38.97 s range to a narrow 0.00–0.09 s interval (average of 0.07 s). Over 90% of relocated events exhibit $RMS \leq 0.1$ s, confirming strong agreement between observed and calculated arrivals and significantly enhanced model precision.

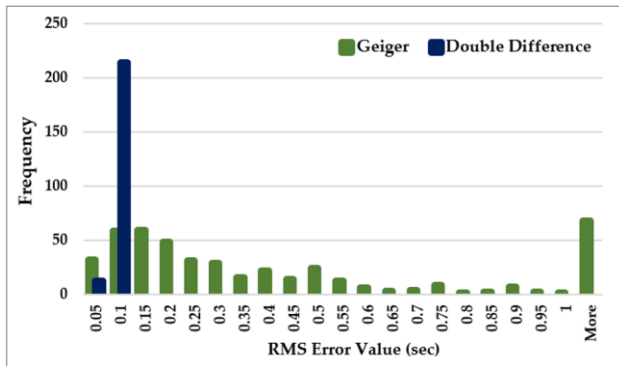


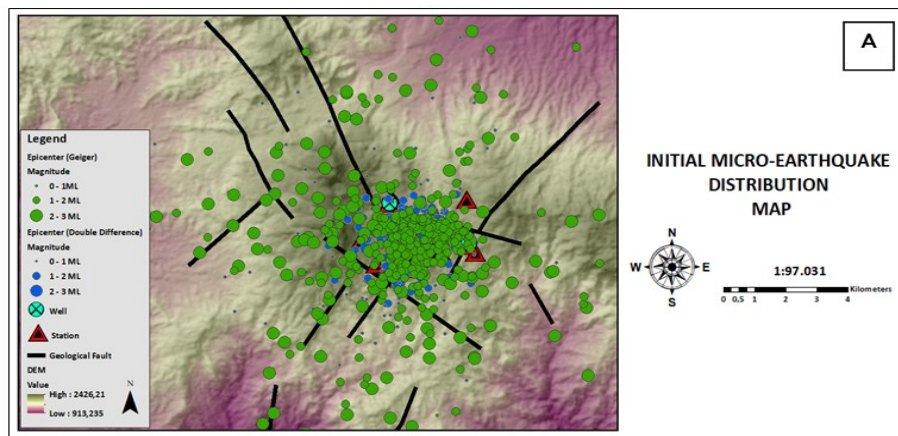
Figure 10 Histogram comparison of the RMS Error MEQ initial (Geiger) and final relocation (Double-Difference) values

This reduction indicates improved waveform coherence and stable inversion performance, consistent with findings by Zhang et al. (2021), who demonstrated that low RMS values reflect high inversion fidelity in microseismic applications. The

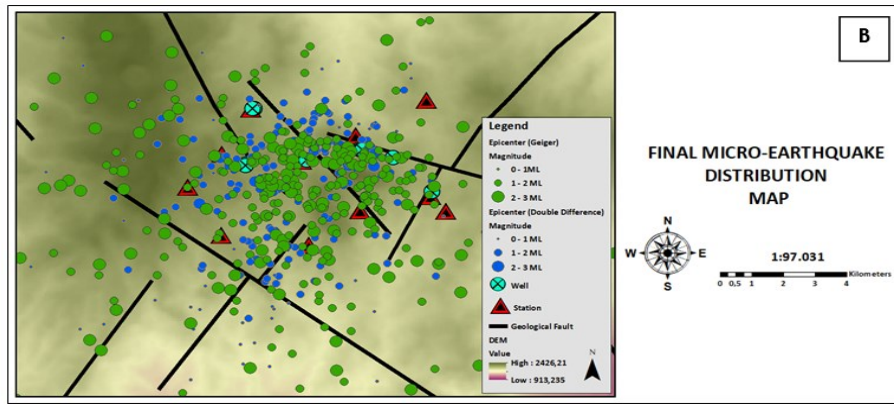
distinct contrast between the scattered Geiger residuals and the compact HypoDD distribution underscores the method's effectiveness in minimizing travel-time discrepancies [47].

3.5 Hypocenter Distribution Maps

The spatial distribution of relocated micro-earthquake (MEQ) events (Figures 11-13) demonstrates the significant improvement achieved through the Double-Difference (HypoDD) relocation. Compared with the widely scattered Geiger solutions, the relocated epicenters exhibit a compact distribution concentrated along major fault zones, indicating enhanced positional accuracy and reduced uncertainty. The close alignment of seismic clusters with mapped faults and well zones suggests strong structural and hydrothermal control within the geothermal reservoir. Three-dimensional visualization confirms that most events are confined between -2.5 km and +0.5 km depth, coinciding with the active reservoir interval and reflecting fluid-induced fracture reactivation. Depth-frequency analysis further reveals that 94% of relocated events occur within -1.5 to +0.5 km, representing a substantial reduction in vertical dispersion. These findings are consistent with Chen et al. (2024), who noted that double-difference relocation enhances spatial coherence and mitigates velocity-model bias, resulting in improved correlation between seismicity and true fault geometries [37]. Collectively, the observed spatial compaction, fault alignment, and depth consistency validate the effectiveness of the HypoDD approach in producing high-fidelity representations of reservoir-scale microseismic activity.



(a)



(b)

Figure 11 The comparison of initial MEQ epicenter location distribution using Geiger method (A) and the final MEQ epicenter location using Double-Difference method, involving a new 1D velocity model of the cluster region (B)

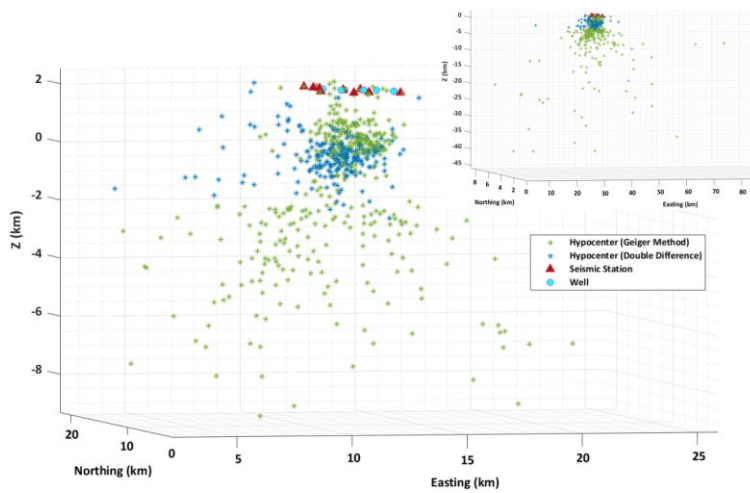


Figure 12 Comparison of the initial (green dots) and final (Double-Difference) (blue dots) MEQ depth distributions

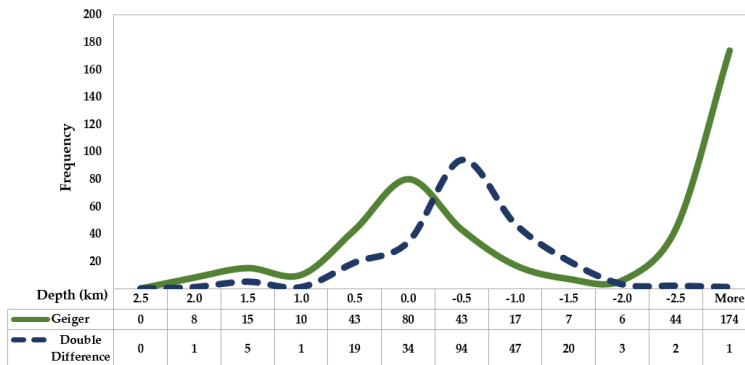


Figure 13 Histogram comparison of RMS error values between Geiger and Double-Difference results

4.0 DISCUSSION

The integration of the refined 1D velocity model with the Double-Difference (HypoDD) relocation significantly enhanced the accuracy and consistency of micro-earthquake (MEQ) locations within the geothermal field. Coupled inversion effectively minimized travel-time misfits, reducing RMS residuals from 38.97 s in the initial Geiger results to an average of 0.07 s in the final solution, indicating

strong model convergence and waveform coherence. The refined velocity structure, capturing the transition from low-velocity volcanic deposits to dense basement rocks, better represents subsurface heterogeneity and improves the precision of hypocentral depth estimates. The narrow RMS distribution (<0.1 s) confirms robust internal consistency between observed and synthetic arrivals, aligning with findings by Zhang *et al.* (2021) [47].

Spatially, the relocated events show distinct clustering along NW-SE fault traces and around injection–production wells, contrasting with the diffuse pattern from the Geiger solution. This alignment demonstrates that seismicity is dominantly controlled by fluid migration and fracture reactivation, with most events concentrated between -1.5 km and +0.5 km, corresponding to the active geothermal reservoir zone. The spatial correlation between seismicity, mapped structures, and geothermal infrastructure supports the interpretation of strong hydro-mechanical coupling driven by fluid-pressure changes, consistent with Chen *et al.* (2024) and Muttaqy *et al.* (2023) [25], [37]. These findings validate the effectiveness of the HypoDD method in delineating fracture-controlled permeability zones, similar to patterns reported in other Indonesian geothermal fields [12], [13]. While the current results provide a reliable high-resolution depiction of reservoir-scale seismicity, future integration of denser station networks and 3-D velocity inversions is recommended to further enhance structural imaging and fault delineation accuracy.

5.0 CONCLUSION

The integrated Coupled Velocity–Hypocenter inversion and Double-Difference (HypoDD) relocation significantly enhanced MEQ hypocenter accuracy, reducing RMS residuals from 38.97 s to 0.07 s and yielding a 93.66 % reduction in vertical deviation. About 94.3 % of events are confined between -1.5 km and +0.5 km, delineating the active geothermal reservoir zone. The relocated seismicity forms compact NW–SE clusters aligned with mapped faults and production–injection wells, indicating strong coupling between stress redistribution, fracture reactivation, and fluid flow. This refined methodological framework provides a reliable, replicable model for precise hypocenter determination, supporting early stress-change detection, improved induced-seismicity assessment, and safer geothermal reservoir management.

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Conflicts of Interest

The authors declare that there is no conflict of interest regarding the publication of this paper.

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