Lateral migration patterns toward or away from injection wells for earthquake clusters in Oklahoma

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Abstract

Exploring the connections between injection wells and seismic migration patterns is key to understanding processes controlling growth of fluid-injection induced seismicity. Numerous seismic clusters in Oklahoma have been associated with wastewater disposal operations, providing a unique opportunity to investigate migration directions of each cluster with respect to the injection-well locations. We introduce new directivity migration parameters to identify and quantify lateral migration toward or away from the injection wells. We take into account cumulative volume and injection rate from multiple injection wells. Our results suggest a relationship between migration patterns and the cluster-well distances, and unclear relationship with injected volume and equivalent magnitudes. Migration away from injection wells is found for distances shorter than 5-13 km, while an opposite migration towards the wells is observed for larger distances, suggesting an increasing influence of poroelastic stress changes. This finding is more stable when considering cumulative injected volume instead of injection rate.

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13 Key Points (140 characters)

- We introduce new parameters to analyze lateral seismicity migration patterns toward or away from multiple associated fluid-injection wells
- Cluster-well distance appears as the main factor in migration behavior in comparison
 with injected volumes or equivalent magnitudes
- Migration away from injection wells is found for distances shorter than 5-13 km,
 migration toward the wells at larger distances
- 20

21 Abstract (150 words)

Exploring the connections between injection wells and seismic migration patterns is key to 22 understanding processes controlling growth of fluid-injection induced seismicity. Numerous 23 seismic clusters in Oklahoma have been associated with wastewater disposal operations, 24 providing a unique opportunity to investigate migration directions of each cluster with respect to 25 the injection-well locations. We introduce new directivity migration parameters to identify and 26 quantify lateral migration toward or away from the injection wells. We take into account 27 cumulative volume and injection rate from multiple injection wells. Our results suggest a 28 relationship between migration patterns and the cluster-well distances, and unclear relationship 29 with injected volume and equivalent magnitudes. Migration away from injection wells is found 30 for distances shorter than 5-13 km, while an opposite migration towards the wells is observed for 31 larger distances, suggesting an increasing influence of poroelastic stress changes. This finding is 32 more stable when considering cumulative injected volume instead of injection rate. 33

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36 Plain Language Summary

37 Oklahoma seismicity has been linked to wastewater injection and provides one of the most important datasets to explore connections between injection wells and induced seismicity. This 38 induced seismicity can be associated in different groups (or clusters) that reveal specific 39 40 spatiotemporal relationships from which preferred spatial migration directions can be identified. We analyzed seismic migration directions of these clusters with the aim of understanding the 41 growth of the rupture process and its location with respect to the closest injection wells. We 42 43 introduce new techniques and parameters to quantify lateral migration patterns toward or away from injection-well locations. Different variables such as the cluster-well distance, injected 44 volumes and magnitudes are considered to assess their influence in these migration behaviors. 45 We identify the main pattern depending on the cluster-well distances. At shorter distances (up to 46 13 km), we observe dominantly migration away from injection wells (particularly for distances 47 shorter than 5 km), whereas at larger distances we observe migration toward the wells. 48

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51 **1 Introduction**

In the last decade, seismic activity observed in Oklahoma has attracted considerable public 52 attention because the annual rate of earthquakes increased since 2009 due to wastewater injection 53 (Ellsworth, 2013; Weingarten et al., 2015; Hincks et al., 2018). Exploring spatio-temporal 54 relations between injection wells and seismic migration patterns is key to understanding the 55 processes controlling the growth of injection-induced seismicity. Earthquakes tend to migrate 56 away from the fluid source following the diffusion of pore pressure, from which hydraulic 57 diffusivity properties can be modeled (Shapiro et al., 2005). However, plausible lateral migration 58 59 patterns toward or away from injection wells in large-scale fluid-injection stimulated areas such as Oklahoma remain unclear (Haffener et al., 2018). Yet, if lateral migration patterns exist and 60 can be tied to (controllable) injection processes, important implications for (time-dependent) 61 fluid-induced seismic hazard assessment arise. This study investigates such properties through a 62 comprehensive migration analysis with respect to multiple injection wells for the Oklahoma 63 seismic clusters. These clusters are defined by applying clustering techniques that associate 64 seismic events into specific groups (or clusters) with specific spatiotemporal relationships, 65 deciphering also different fault structures (Ester et al., 1996; Wang et al., 2013; Zaliapin and 66 67 Ben-Zion, 2013; Cheng and Chen, 2018). The identification and characterization of these clusters has been well studied in natural and tectonic contexts revealing interesting event 68 migration features such as, for instance, event triggering due to fluid flow (Vidale and Shearer, 69 2006; Chen et al., 2012; Passarelli et al., 2018). 70

Recent efforts for improving the existing earthquake catalogues for Oklahoma identified seismicity clusters distributed over the area due to the activation of hundreds of previously unknown faults (Schoenball and Ellsworth, 2017). Analyzing the spatiotemporal evolution of these clusters reveals that seismicity tends to initiate at shallower depth and migrates deeper along faults as the sequence proceeds (Schoenball and Ellsworth, 2017b). Although 40 - 50% of individual clusters exhibit statistically significant diffusive migration, no clear migration patterns along-strike are observed (Haffener et al., 2018). On the other hand, preferred rupture

propagation direction involving directivity effects have been identified for the largest induced 78 79 Oklahoma earthquakes (López-Comino and Cesca, 2018; Lui and Huang, 2019). A common pattern reflecting rupture propagation toward or away from injection wells is difficult to 80 81 establish, also due to the variety in rupture styles. The 2011 Mw 5.7 Prague and 2016 Mw 5.0 Cushing earthquakes ruptured away from the injection wells, whereas the 2016 Mw 5.1 Fairview 82 earthquake ruptured toward the injection. Lui and Huang (2019) attributed the difference in 83 rupture directions to expected pressurization of the fault zone, which relates to the distance away 84 from injection zones and total injected volume. 85

Induced seismicity in Oklahoma provides a unique opportunity to systematically compare the 86 migration direction of each seismic cluster with respect to injection well locations. Each cluster 87 will be characterized by a so-called migration vector calculated using an enhanced migration 88 technique based on previous work (Haffener et al., 2018). Seismicity occurs in a region with 89 many high-rate disposal wells and high-pressure perturbations causing difficulties to establish 90 91 appropriate associations among earthquakes and wells. In this context, we introduce a new methodology to calculate a so-called well vector associated to multiple injection wells. Finally, 92 we explore plausible lateral migration patterns depending on different parameters such as the 93 distance from injection wells, injected volumes weightings and equivalent magnitude of each 94 cluster. 95

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97 **2 Data**

We use a relocated earthquake catalog, recorded between 2010 and 2016, with enhanced spatial 98 resolution, a magnitude of completeness (M_c) of 2.5, and a minimum magnitude of 2.0 (Chen, 99 2016). We consider individual clusters with at least 20 events identified by Haffener et al., 100 101 (2018). This resulted in 60 clusters after aftershocks were removed using the space-time windowing method proposed by Uhrhammer (1986) to avoid the space-time imprint of 102 aftershocks. The injection data used in this study are obtained from Oklahoma Corporation 103 Commission websites with monthly data from 1995 to 2017 with a total number of 876 disposal 104 wells. Considering maximum well distances of 50 km, the number of injection wells involved in 105 this study is 836 (Figure 1). 106

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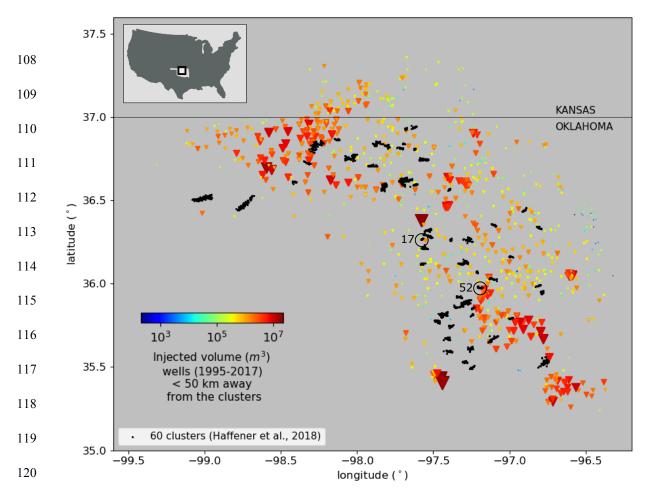


Figure 1. Map of wells (inverted triangles) within a radius of 50 km from the average location in each cluster and 60 seismic clusters in Oklahoma (black dots) detected by a nearest-neighbor approach after aftershocks were removed using a space-time windowing method (Haffener et al., 2018). Wells are scaled (color and size) according to the total injected volume between 1995 – 2017. Two selected clusters (17 and 52) analyzed in Figure 2 and S1 are indicated.

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127 **3** Comprehensive migration analysis with respect to multiple injection wells

In our study, we propose a comprehensive migration analysis based on previous approaches involving different numbers of temporal bins of equal duration spanning the period of the entire sequence (Haffener et al., 2018) (Figure 2 and S1). Earthquakes in each cluster are divided into 10 temporal bins (Figure 2a). A spatial bin point is calculated by averaging epicentral locations in each bin, delineating the migration line of each cluster (Figure 2b).

Next, we define the migration vector $(\overline{v_m})$, as the direction from the 1st spatial bin point to the averaged location of the remaining spatial bin points. Each cluster is then characterized by the azimuth Φ and length r of the migration vector (Figure S2). The notation Φ_0 and r_0 indicates that all events in each cluster were used to calculate the azimuth and length (Figure 2b). To assess uncertainties associated with the calculation of the migration vector, we applied a bootstrap analysis. For each cluster, we calculated 100 migration vectors, randomly removing 10% of events in each repetition (Figure S3). The final Φ and r are then defined from the average locations of the heads and tails of all migration vectors (Figure 2c). We define the associated uncertainties as $\varepsilon_{\Phi} = \Delta \Phi/2$, where $\Delta \Phi$ is the maximum difference of azimuths calculated from the bootstrap analysis, and $\varepsilon_{\rm r}$ as the standard deviation of *r*. Significant changes of Φ between repetitions indicate that the cluster does not have a prevailing direction of migration (Figure S1). Therefore, we only consider clusters with $\Delta \Phi < 45^{\circ}$ in further analysis. Based on this criterion, the migration vectors for 24 clusters were excluded (Figure S4).

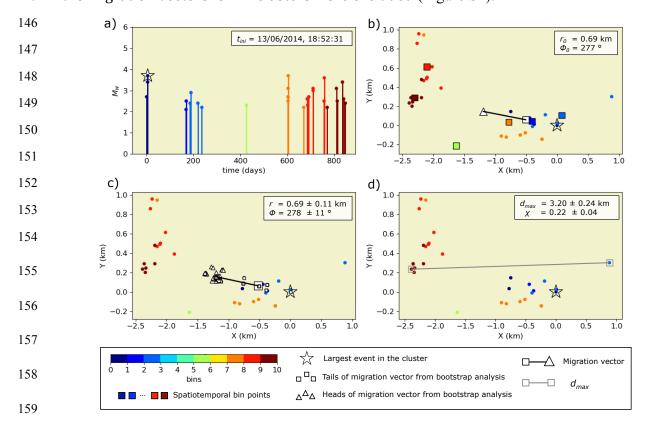


Figure 2. Migration analysis for cluster 52 (see Figure 1) showing results for a stable migration 160 vector. a) Temporal evolution of the seismic sequence from t_{ini} ; the color scale indicates 161 association of seismic events with temporal bins and the star depicts the largest event in the 162 cluster. b) The migration vector (black line) defined from tail (white square) to head (white 163 triangle) and the spatiotemporal evolution of migration (color-coded squares indicate the 164 spatiotemporal bin points). r_0 and Φ_0 represent the length and azimuth of the migration vector 165 calculated using all events in the cluster. c) Bootstrap analysis to calculate the final length r and 166 azimuth Φ of the migration vector and their uncertainties. Small white triangles and small white 167 squares depict the heads and tails of 100 migration vectors for the bootstrap analysis. The final 168 migration vector is depicted by a black line from the tail (large white square) to the head (large 169 170 white triangle). d) The maximum cluster length (d_{max}) and the migration coefficient (χ) are shown with the uncertainties obtained from the bootstrap analysis. d_{max} (gray line), is defined by 171 172 the two seismic events farthest from each other (open gray squares).

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174 Individual clusters are divided into two groups, according to their spatial migration behavior: 175 migration and non-migration groups, or here so-called strong or weak migration groups. Some authors obtain a statistical significance (s_m) ranging from 0.5 to 1.0 to identify each migration group according a fixed threshold value around 0.8 - 0.85 (Chen et al., 2012). Using such criteria, Oklahoma clusters reveal almost a parity division with around 40 - 50 % clusters belonging to group with strong migration (Haffener et al., 2018). We propose a simple way to quantify this property by calculating the ratio of the length of the migration vector (r) to the maximum length of the cluster (d_{max}) (Figure 2d):

183
$$\chi = \frac{r}{d_{max}}$$
(Eq.1)

184

The migration coefficient (χ) increases from 0 (no migration) to 1 (strong migration), reaching the maximum value only in the case of migration from one end of the cluster to the other. Uncertainties for χ -values are calculated using the bootstrap analysis. A similar distribution for χ is obtained using different bins to calculate the migration vector where a value of 0.2 yields similar results as using s_m to establish the separation among different migration groups (Figure S5).

The association of seismic clusters to specific wells is crucial for determining whether clusters migrate toward or away from the fluid-injection point. Multiple injection points and the long history of injection volumes in Oklahoma complicate the individual associations for each cluster. Similar areas in Alberta (Canada) had addressed this issue through spatiotemporal association filters, discarding wells potentially not associated with earthquake clusters based on a set of association criteria, for instance, epicenters of all temporally associated earthquakes must be within 5 km of the well pad surface location (Schultz et al., 2018).

Here, we propose a new methodology representing multiple injection wells around each cluster 198 using a well vector $(\overline{v_w})$ defined as the vector from the 1st spatial bin point of the cluster (used 199 previously to define the migration vector) to an injection midpoint (Figure 3). The well vector is 200 also characterized by the azimuth Φ_w and length r_w (Figure S2). The injection midpoint is 201 determined as the weighted centroid of locations of wells, taking into account the spatial and 202 temporal distribution of the volume of injected fluids into individual wells and the expansion of 203 the diffusion front. Injected fluids can be associated with a cluster only if they have sufficient 204 time to reach the location of the cluster. Considering a linear diffusion model, we approximate 205 this time by the diffusion front (Shapiro et al., 2005) 206

207
$$t_D = \frac{d^2}{4\pi D},$$
 (Eq.2)

where D is diffusion coefficient and d is the distance between the well and cluster. For the 208 analysis, we use a representative value for the diffusion coefficient for Oklahoma area (D = 1.5209 m^2/s , Haffener et al., 2018). This corresponds, for example, to a delay time t_D for the diffusion 210 front of about 18 months for a well that is located at 30 km from a cluster (Figure S6). To 211 account for the effects of diffusion, a well is associated with a cluster only if the fluid-injection 212 started more than $t_{\rm D}$ ago. Next, at each time instant t of the seismic sequence, the weight of an 213 individual well i is adjusted according to reported cumulative injected volume $V_i(t-t_D)$ and the 214 injection rate volume $\Delta V_i(t-t_D)$. Note that we consider ΔV as the volume injected each month and 215

therefore *t* increases in steps of 30 days, consistently with the reporting period of injected volumes. Finally, the individual weights are adjusted to account for the expansion of the diffusion front such as a geometrical spreading effect. Assuming dominantly horizontal diffusion, we consider 2D geometrical spreading, which leads to the weights in the forms of $V_j(t-t_D)/d_i$ and $\Delta V_i(t-t_D)/d_i$. To avoid singularities, we consider d = 1 km for wells with d < 1 km.

Following this procedure, we obtain one injection midpoint for each considered time instant t 221 and their average location then defines the final injection midpoint based on weights from 222 223 cumulative injected volume and injection rate volumes, respectively. The procedure for cluster 52 is illustrated in Figure 3 for cumulative injected volume and in Fig. S7 for injection rate 224 volumes. We also define the associated uncertainties as $\varepsilon_{\Phi w} = \Delta \Phi_w/2$, where $\Delta \Phi_w$ is the 225 maximum difference of azimuths of individual well vectors, and ε_{rw} is the standard deviation of 226 r_w . Like for the migration vector, also here cases with $\Delta \Phi_w > 45^\circ$ are considered unstable. Using 227 this criterion, we found 36 stable cases when using the cumulative injected-volume weighting 228 and 22 when using the injection-rate volume weighting (Figure S4). A summary of all calculated 229 parameters is shown in the Table S1. 230

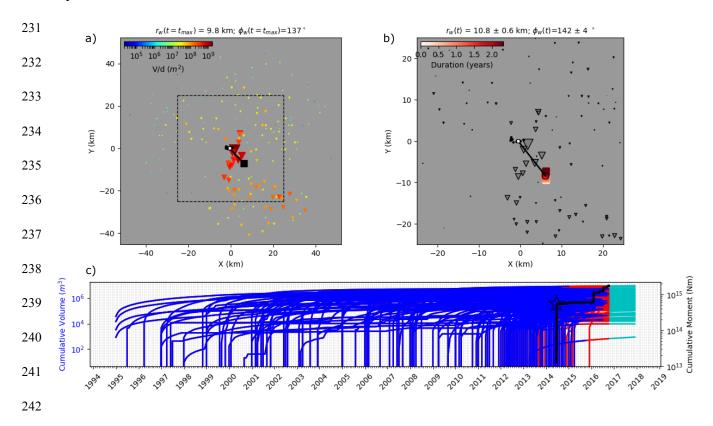


Figure 3. Calculating the well vector for cluster 52 considering the cumulative injected volume weighting. a) Situation at the final time of the seismic sequence t_{max} . The well vector (black line) is defined from the tail of the migration vector (white square defined in Fig 2c) to the injection midpoint (black square). Length (r_w) and azimuth (Φ_w) of the well vector are indicated in the figure header. Wells (inverted triangles) are scaled (color and size) according V(t_{max} - t_D)/d. b) Location of injection midpoints during the seismic sequence (color-coded squares). The final injection midpoint is shown with an open black circle, the final well vector by the black line.

Wells are scaled in size as in a). Only the dashed rectangle from a) is shown. c) Cumulative injected volume for the wells associated with the cluster (blue lines) and cumulative seismic moment for the seismic sequence (black line; the largest event is indicated by the star). Red lines indicate the volume that did not affect the cluster due to diffusion constraints, cyan lines indicate data available after the end of seismic sequence.

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4 Results

To summarize our results of the comprehensive migration analysis considering multiple injection wells, we define the direction toward or away from injection wells by a parameter κ that represents the angle between the migration vector and the well vector:

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$$\boldsymbol{\kappa} = \angle (\overrightarrow{\boldsymbol{v}_m}, \overrightarrow{\boldsymbol{v}_w}). \tag{Eq. 3}$$

 κ -values range from 0° to 180°, with κ -values closer to 0° indicating an alignment among the 261 migration vector and the well vector for a migration direction toward the injection wells. ĸ-262 values closer to 180° indicate the opposite behavior, i.e., migration away from injection wells. 263 We note that the migration vector is also affected by fault geometry as it is most likely oriented 264 along the fault strike. For this reason, we define $\kappa < 60^{\circ}$ as migration toward wells and $\kappa > 120^{\circ}$ 265 as migration away. For intermediate cases ($60^\circ < \kappa < 120^\circ$) the migration and well vectors are 266 close to perpendicular, making these cluster a less appropriate choice to decide whether 267 seismicity migrates toward or away from the wells. 268

In Figure 4 we compare κ as a function of: i) length of the well vector, ii) the total weights 269 assigned to the multiple associated wells based on cumulative injected volumes and injection rate 270 volumes, and iii) the equivalent magnitude (sum of the seismic moments of the events in a 271 272 cluster expressed as moment magnitude following Hanks and Kanamori, 1979). For this comparison and following analysis, we only consider clusters with strong migration ($\gamma > 0.2$): 25 273 of 36 clusters for the cumulative volume weighting (Figure 4a) and 14 of 22 clusters for the 274 injection rate volume weighting (Figure 4b). Average values and their errors are calculated for 275 clusters migrating toward and away from the wells in order to identify potential lateral migration 276 patterns. Depending on r_w, larger differences among these average toward (13 and 16 km) or 277 away (8 and 10 km) values are observed for both weightings. However, no significant changes 278 279 are appreciated depending on the weighted volumes and the equivalent magnitude. Additionally, the histograms of κ -values for cumulative injected-volume weighting indicate that a small 280 majority of clusters (60%) documents a migration away from the wells (Fig 4a4). 281

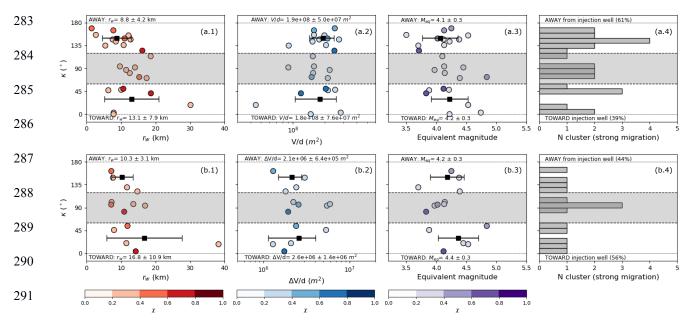


Figure 4. Lateral migration patterns toward or away from injection wells characterized by κ-292 values using 10 temporal bins in the comprehensive migration analysis (for diffusion coefficient 293 of 1.5 m²/s). κ -values for strong migration clusters ($\chi > 0.2$) are plotted as circles scaled in color 294 according the migration coefficient (χ) , considering the cumulative volume weighting (a) and the 295 injection rate volume weighting (b). Results are shown for each cluster according to the length of 296 the well vector (a.1 and b.1), the total weights assigned to the multiple associated wells in 297 relation to cumulative injected volumes and injection rate volumes (a.2 and b.2) and the 298 equivalent magnitude (a.3 and b.3). Average values and error bars (black squares and lines) are 299 indicated for propagation toward ($\kappa < 60^{\circ}$) and away ($\kappa > 120^{\circ}$) from the injection point (see 300 labels). Histograms are also shown including percentages values (a.4 and b.4). Intermediate cases 301 $(60^{\circ} < \kappa < 120^{\circ})$ are not considered (grav background separated by black dashed lines). 302

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5 Discussion and conclusions

A comprehensive migration analysis is applied to decipher the potential relationship between direction of lateral earthquake migration of induced seismic events and the location of multiple injection wells. We introduced a new parameter, κ , to quantify the direction of lateral migration toward or away from the injection point based on the angle between the migration vector and the well vector. This parameter facilitates the identification of these lateral migration patterns and it can be used to compare results in other fluid-injection stimulated areas.

A representative migration line, obtained by joining spatiotemporal bin points identified for a cluster, yields complex shapes/patterns for clusters with no predominant migration direction or with bilateral migration. To compare results for a large number of clusters, we approximate the trajectory by the migration vector. The migration starting point (generally assumed to be the epicenter of the first recorded seismic event in the cluster) is a relevant parameter to obtain a representative migration vector. However, we consider that the migration starts at the 1st spatial bin point, which more accurately defines the first activated fault area than the location of just the 318 single seismic event. Because the complete spatiotemporal history of the seismic sequence must

319 contribute to defining the head of the migration vector, the average epicentral location of the 320 remaining spatial bin points is used.

We introduced a simple way to quantify the strong/weak migration through a migration 321 coefficient χ , computed from the length of the migration vector and the total length of the 322 cluster. χ -values for all clusters show an asymmetric (positively skewed) distribution with a long 323 tail toward the largest χ -values (Figure S5). A median value of this distribution ($\chi \sim 0.2$) 324 provides similar results as using s_m criterions to divide the Oklahoma clusters into strong/weak 325 migration groups. Clusters with γ -values larger than 0.2 show an observable migration, but 326 below this value the length of the migration vector is too short to observe any predominant 327 migration directions. 328

We also propose a new strategy to define representative well vectors associated with multiple 329 injection points surrounding a seismic cluster, considering two types of weighting. The 330 cumulative volume weighting may better represent cumulative effects of pore pressure build up 331 from the beginning of the injection, while injection rate volume weighting may better represent 332 effect of pore pressure variations. From an operational point of view, injection rate weighting 333 may vary significantly over a short time scale, which can cause significant changes of the 334 335 direction of the well vector during the course of the cluster (Figure S7), as documented by 22 identified unstable well vectors. In contrast, the cumulative volume weighting provides more 336 stable results with no unstable well vectors (Figure S4). Also, different directions of the well 337 vector can be found for each weighting in the same cluster (Fig 3b and S7b). 338

Regardless of the influence of the weighting on well vector orientation, we observe similar 339 patterns when comparing propagation towards (small κ) and away (large κ) from the wells, 340 depending on the r_w (Figure 4a.1 and 4b.1) and the equivalent magnitude (Figure 4a.3 and 4b.3). 341 Significant differences are observed only according to r_w, revealing the cluster-well distances as 342 a key factor to control these processes. Migration away from injection wells is found for 343 distances shorter than 5-13 km, while an opposite migration towards the wells is observed for 344 larger distances. Both distributions overlap, indicating that there is no monocausal relationship 345 with distance, but the general trend is clear. Accordingly, at shorter cluster-well distances where 346 hydraulic connections between faults and injection wells can be involved, the seismicity is 347 triggered by propagating pore-pressure front (Shapiro et al., 2005). The cumulative injected 348 volume weighting provides the most stable results revealing clearly this pattern for cluster with 349 $r_w < 5$ km. For larger distances, the previous assumption becomes questionable. Outside of the 350 high-pressure zone, poroelastically-induced Coulomb-stress-changes should surpass pore 351 pressure changes, providing a plausible triggering mechanism in the far-field of injection wells 352 (Goebel et al., 2017). The transition from pore-pressure dominance to poroelastic stress based on 353 distance could explain the changes in migration pattern at further distances, which is observed 354 for r_w in the range between 5 - 20 km. For $r_w > 20$ km, our only observation corresponds to the 355 Woodward cluster, confirming this behavior. 356

The observed patterns remain stable for different choices of the diffusion coefficient (D = 1.25 - 1.25)

1.75 m²/s, Fig S8 and S9), which further supports robustness of the well vector-based approach.

Also, different choices of temporal binning (5, 10, 15 and 20) yield similar and consistent results (Figs S10 and S11). Considering only 2 bins in the injection rate volume weighting, we obtain

- 361 similar results such as the statistical parameter d_s which assesses the distance separation between
- the centers of the first half and second half of each cluster (Chen and Shearer, 2011; Chen et al.,
- 2012). Overall, the comparison indicates that 10 bins is a reasonable choice for our study.

Folesky et al. (2016) analyzed the directivity effects of the largest seismic events associated with the stimulation of geothermal reservoir in Basel (Switzerland) and found that the preferred rupture propagation depends on magnitude. They found that events with $M_L > ~2$ propagated backward into the perturbed volume while smaller events propagated away from the well. Our analysis, with minimum equivalent magnitudes around 3.5, shows a different situation, with a significant number of clusters migrating away from the injection wells, and no clear dependence on magnitude.

In conclusion, albeit the main migration pattern in Oklahoma reflects a downward migration from the Arbuckle layer to the basement (Schoenball and Ellsworth, 2017b, Haffener et al., 2018), we found clear lateral migration patterns involving the cluster-well distance as the main factor to control preferred migration directions toward or away from the injection wells. While clusters closest to the wells show a predominant migration away from the wells attributed to pore-pressure changes, we also observe an opposite behavior toward the wells for larger distances that could be controlled by poroelastic stress changes.

378

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Supporting Information for

Lateral migration patterns toward or away from injection wells for earthquake clusters in Oklahoma

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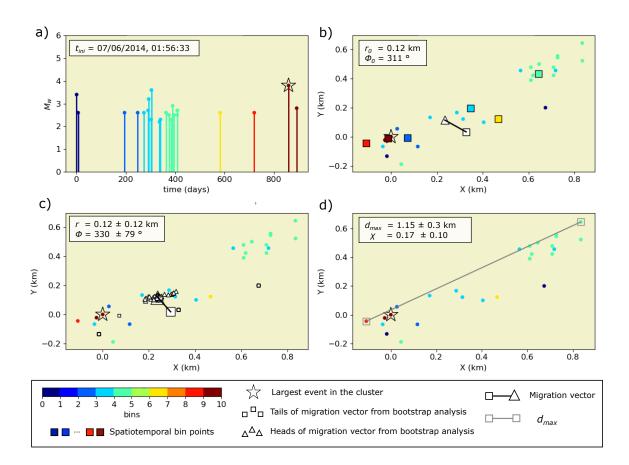


Figure S1. Migration analysis for cluster 17 (see Figure 1) showing results for an unstable migration vector. a) Temporal evolution of the seismic sequence from t_{ini} ; the color scale indicates association of seismic events with temporal bins and the star depicts the largest event in the cluster. b) The migration vector (black line) defined from tail (white square) to head (white triangle) and the spatiotemporal evolution of migration (color-coded squares indicate the spatiotemporal bin points). r_0 and Φ_0 represent the length and azimuth of the migration vector calculated using all events in the cluster. c) Bootstrap analysis to calculate the final length r and azimuth Φ of the migration vector and their uncertainties. Small white triangles and small white squares depict the heads and tails of 100 migration vectors for the bootstrap analysis. The final migration vector is depicted by a black line from the tail (large white square) to the head (large white triangle). d) The maximum cluster length (d_{max}) and the migration coefficient (χ) are shown with the uncertainties obtained from the bootstrap analysis. d_{max} (gray line) is defined by the two seismic events farthest from each other (open gray squares).

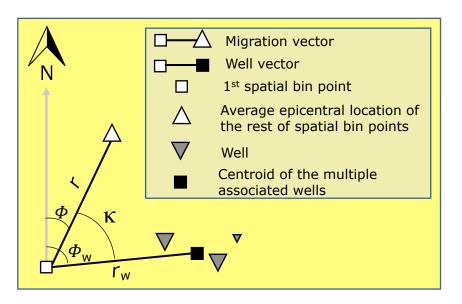


Figure S2. Sketch showing the parameters for calculating the migration and well vector for the comprehensive migration analysis with respect to multiple wells.

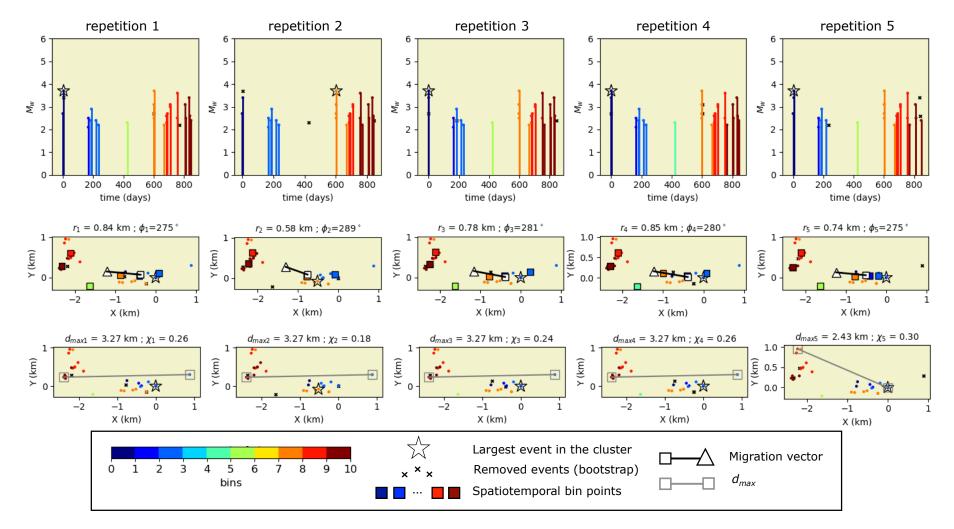


Figure S3. Bootstrap analysis for 5 repetitions (repetitions per columns) to calculate migration parameters for cluster 52. This procedure is repeated 100 times randomly removing the 10% of the events (black x-points). (First row) Temporal evolution of the seismic sequence for each repetition; the color scale indicates different temporal bins associated for each seismic event and the star shows the largest event in the cluster. (Second row) The migration vector (black line) defined from tail (white square) to head (white triangle) and the spatiotemporal evolution of migration (color-coded squares indicate the spatiotemporal bin points). r_i and Φ_i represent the length and azimuth of the migration vector for each *i* repetition. (Third row) The total length (d_{max}) of the cluster (gray line) is defined by the two seismic events farthest from each other (open gray squares) and the migration coefficient (χ) is also shown for each repetition.

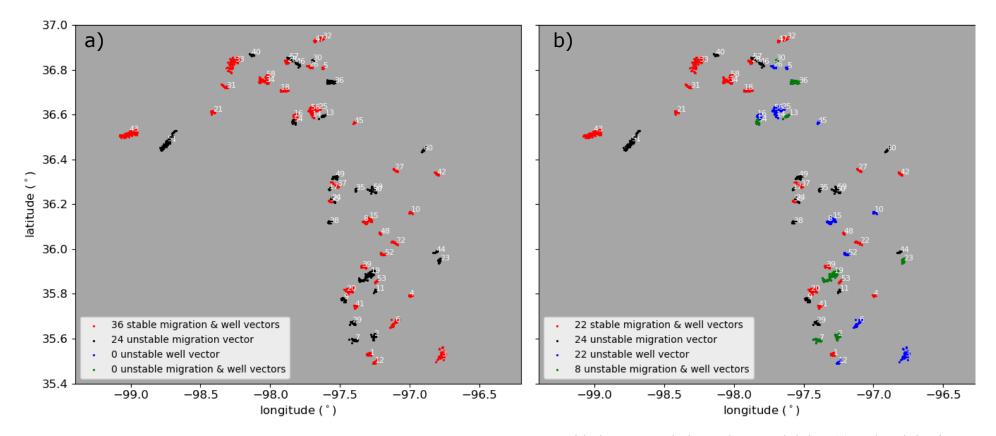


Figure S4. Migration and well vector stability for each cluster (color-coded dots) considering a cumulative volume weighting (a) and an injection rate volume weighting (b). White numbers denote each cluster (see table S1).

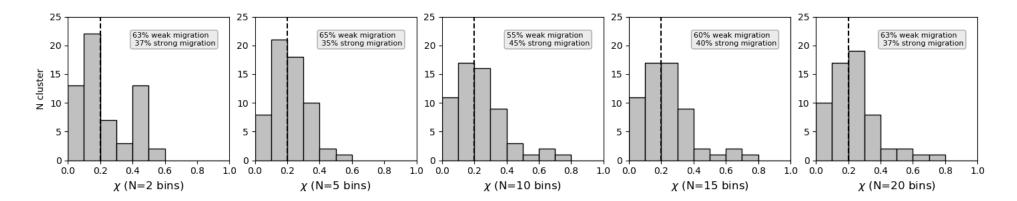


Figure S5. Histograms for the migration coefficient (χ) using different temporal bins (N) in the comprehensive migration analysis. The dashed black lines separates weak ($\chi < 0.2$) and strong ($\chi > 0.2$) migration clusters.

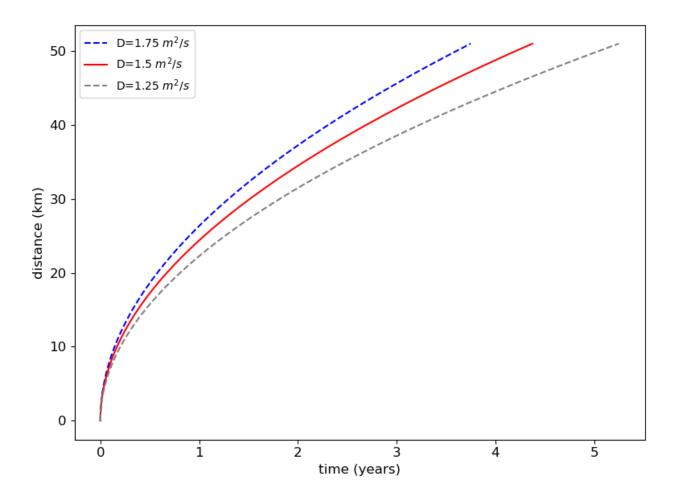


Figure S6. Diffusion fronts according Shapiro et al., 2005 for different diffusion coefficients (D) tested in the comprehensive migration analysis with respect to multiple injection wells.

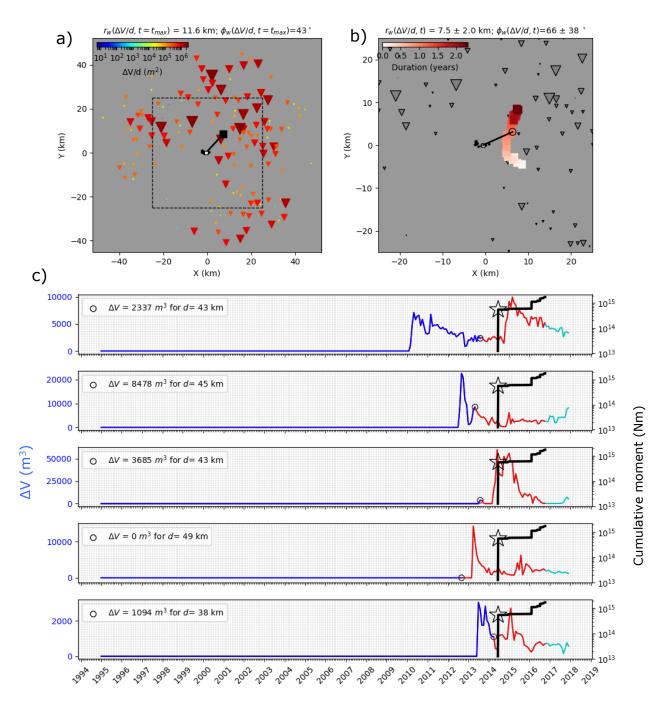


Figure S7. Calculating the well vector for cluster 52 considering the injection rate volume weighting and a diffusion coefficient of 1.5 m²/s. a) Situation at the final time of the seismic sequence t_{max} . The well vector (black line) is defined from the tail of the migration vector (white square defined in Fig 2c) to the injection midpoint (black square). Length (r_w) and azimuth (Φ_w) of the well vector are indicated in the figure header. Wells (inverted triangles) are scaled (color and size) according $\Delta V(t_{max}-t_D)/d$. b) Location of injection midpoints during the seismic sequence (color-coded squares). The final injection midpoint is shown with an open black circle, the final well vector by the black line. Wells are scaled in size as in a). Only the dashed rectangle from a) is shown. c) Injection rate volumes for five wells associated with the cluster (blue lines) and cumulative seismic moment for the seismic sequence (black line; the largest event is indicated by the star). Red lines indicate the volume that did not affect the cluster due to diffusion constraints, cyan lines indicate data available after the end of seismic sequence. For each well the injection rate volume (ΔV) with respect to t_{max} (open black circle) and the well distance (d) are indicated.

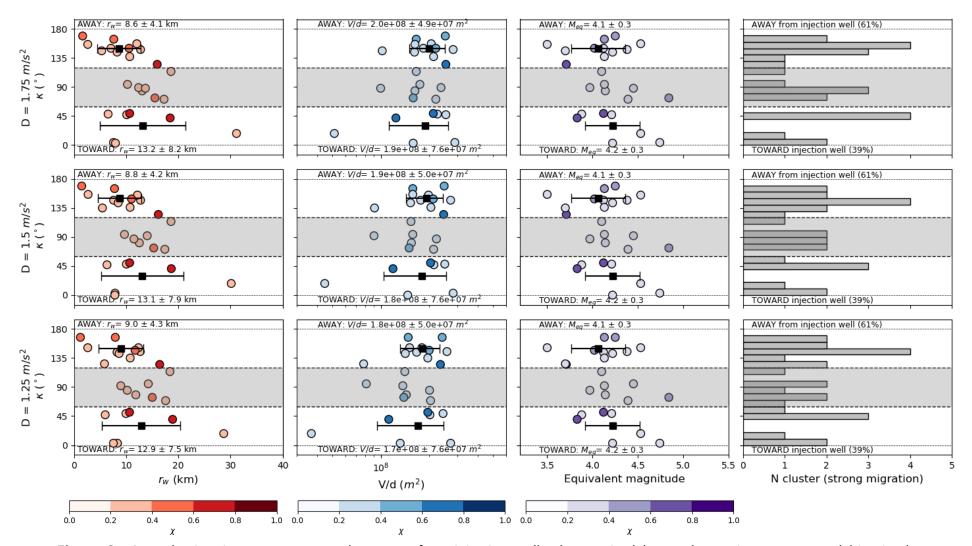


Figure S8. Lateral migration patterns toward or away from injection wells characterized by κ -values using 10 temporal bins in the comprehensive migration analysis and different diffusion coefficients of 1.75 m²/s (first row), 1.5 m²/s (second row) and 1.25 m²/s (third row). κ -values for strong migration clusters ($\chi > 0.2$) are plotted as circles scaled in color according the migration coefficient (χ), considering the cumulative volume weighting. Results are shown for each cluster according to the length of the well vector (first column), the total weights assigned to the multiple associated wells in relation to cumulative injected volumes (second column) and the equivalent magnitude (third column). Average values and error bars (black squares and lines) are indicated for propagation toward ($\kappa < 60^\circ$) and away ($\kappa > 120^\circ$) from the injection point (see labels). Histograms are also shown including percentages values (forth column). Intermediate cases ($60^\circ < \kappa < 120^\circ$) are not considered (gray background separated by black dashed lines).

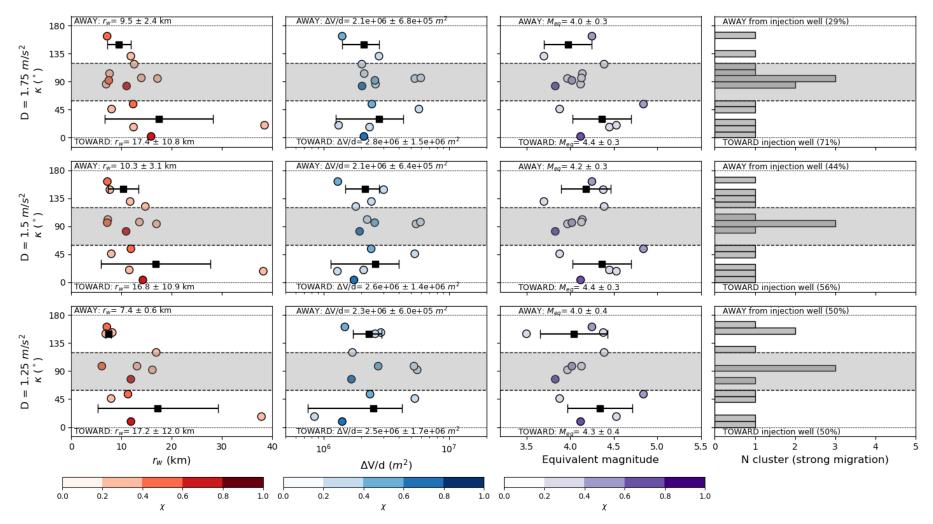


Figure S9. Same as figure S8, but considering injection rate volume weighting. Note that second column show the results according to the total weights assigned to the multiple associated wells in relation to injection rate volumes.

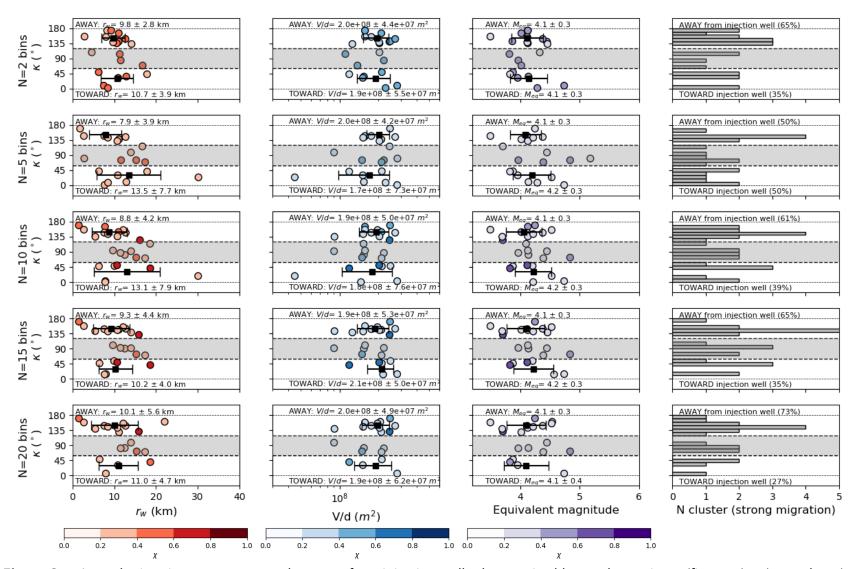


Figure S10. Lateral migration patterns toward or away from injection wells characterized by κ -values using 2 (first row), 5 (second row), 10 (third row), 15 (fourth row) and 20 (fifth row) temporal bins in the comprehensive migration analysis and a coefficient diffusion of 1.5 m²/s. κ -values for strong migration clusters ($\chi > 0.2$) are plotted as circles scaled in color according the migration coefficient (χ), considering the cumulative volume weighting. Results are shown for each cluster according to the length of the well vector (first column), the total weights assigned to the multiple associated wells in relation to cumulative injected volumes (second column) and the equivalent magnitude (third column). Average values and error bars (black squares and lines) are indicated for propagation toward ($\kappa < 60^\circ$) and away ($\kappa > 120^\circ$) from the injection point (see labels). Histograms are also shown including percentages values (forth column). Intermediate cases ($60^\circ < \kappa < 120^\circ$) are not considered (gray background separated by black dashed lines).

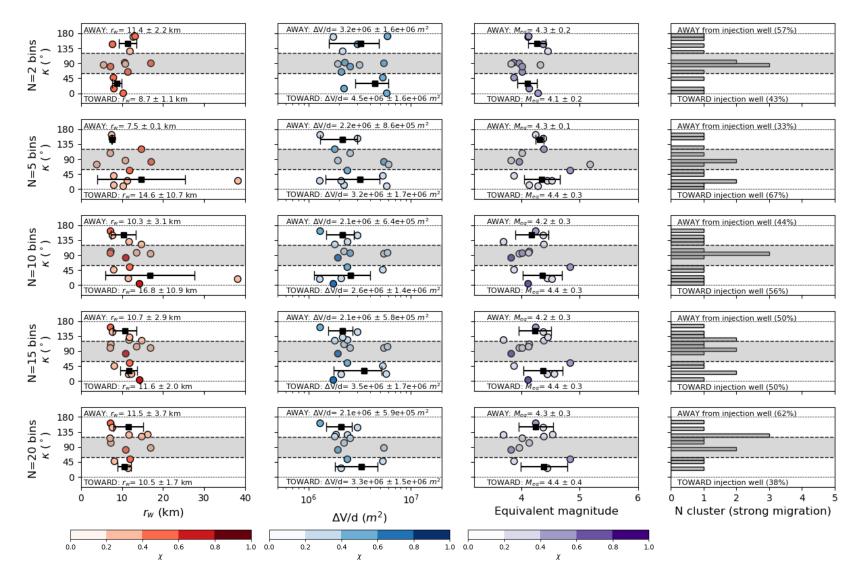


Figure S11. Same as figure S10, but considering injection rate volume weighting. Note that second column show the results according to the total weights assigned to the multiple associated wells in relation to injection rate volumes.

	lon (deg)	lat (deg)	N events		фw (deg)		Kanna (dag)	r (km)	dmax (km)	Xi	rw (km)		V/d (m2)	ΔV/d (m2)	Meq (Nm)	Stability (1 = YES & 0 = No)		
Cluster ID				φ (deg)			Kappa (deg)									Migration	Well vector	
					for V weighting	for ΔV weighting	for V weighting for ΔV weightin	3			for V weighting	for ΔV weighting				vector	for V weighting	for ΔV weighting
1	-97,285	35,528	29		22 ± 2	42 ± 9		1 1,01 ± 0,17	2,74 ± 0,07	0,37 ± 0,06	12,53 ± 0,72	7,27 ± 1,39	1,55E+08	2,23E+06	4,14	1	1	1
2	-97,259	35,613	48		42 ± 7	116 ± 62	134 15		3,52 ± 0,05	0,18 ± 0,05	11,1 ± 1,78	3,87 ± 1,87	2,45E+08	1,76E+06	4,45	0	1	0
3 (Prague)	-96,772	35,522 35,792	90 25	22 ± 17 251 ± 3	328 ± 11 60 ± 15	93 ± 89 89 ± 19	54 7 169 16	1 0,64 ± 0,12 2 1,11 ± 0,07	8,36 ± 0,38 2,50 ± 0,10	0,08 ± 0,01 0,44 ± 0,04	2,13 ± 0,44 1,49 ± 0,29	2,18 ± 1,30 7,19 ± 0,95	2,44E+08 2,48E+08	1,16E+06 1,30E+06	5,74	1	1	0
5	-97,001	36,809	30		135 ± 3	137 ± 44	114 11		1,06 ± 0,02	0,24 ± 0,04	1,49 ± 0,29 18,59 ± 3,15	10,72 ± 4,11	1,54E+08	2,52E+06	4,25	1	1	0
6	-97,104	35,675	36		40 ± 3	158 ± 57	3 12		7,10 ± 0,12	0,23 ± 0,02	7,87 ± 0,89	3,38 ± 0,80	2,82E+08	1,74E+06	4,74	1	1	0
7	-97,393	35,593	90		86 ± 7	154 ± 25	159 13	3 0,24 ± 0,08	5,07 ± 0,50	0,05 ± 0,01	9,41 ± 0,59	8,65 ± 1,20	2,24E+08	1,10E+06	4,96	0	1	0
8	-97,33	36,124	20		139 ± 4	24 ± 70		8 0,54 ± 0,12	3,07 ± 0,17	0,18 ± 0,05	10,85 ± 1,45	3,55 ± 2,63	1,52E+08	2,40E+06	3,96	1	1	0
9	-97,475	35,775	53		109 ± 3	140 ± 13		2 0,23 ± 0,07	3,34 ± 0,09	0,07 ± 0,02	22,82 ± 0,82	16,84 ± 4,42	2,08E+08	1,86E+06	4,54	0	1	1
10	-96,992 -97,243	36,165 35.814	23 23		166 ± 0 108 ± 3	160 ± 53 156 ± 17	125 13 157 10		2,19 ± 0,14 1,78 ± 0,15	0,71 ± 0,10 0,16 ± 0,02	16,17 ± 1,96 9,93 ± 0,17	2,10 ± 0,66 10,63 ± 4,11	2,46E+08 2,49E+08	2,53E+06 2,27E+06	3,71	1 0	1	0
11		35,814	23		108 ± 5 12 ± 2	309 ± 47	48 11		3,41 ± 0,18	0,18 ± 0,02	10,05 ± 0,6	3,69 ± 0,57	2,49E+08	1.28E+06	4,21	1	1	0
13	. , .	36,596	42		99 ± 5	66 ± 31	156 17		4,41 ± 0,60	0,1 ± 0,02	12,21 ± 1,12	6,62 ± 2,66	1.78E+08	3,28E+06	4,58	0	1	0
14		36,591	23		103 ± 3	46 ± 66	165 10		2,64 ± 0,19	0,43 ± 0,04	7,78 ± 1,43	1,37 ± 1,26	1,58E+08	2,77E+06	4,13	1	1	0
15		36,138	56	294 ± 5	147 ± 10	19 ± 26		5 1,27 ± 0,09	3,77 ± 0,32	0,34 ± 0,04	7,6 ± 1,7	6,22 ± 1,76	1,77E+08	2,51E+06	4,12	1	1	0
16		36,594	52		99 ± 6	46 ± 41		3 0,87 ± 0,16	3,03 ± 0,05	0,29 ± 0,06	7,83 ± 2,24	2,11 ± 1,29	1,47E+08	3,28E+06	4,22	1	1	0
17		36,267	26		36 ± 3	49 ± 11		3 0,12 ± 0,13	1,16 ± 0,03	0,19 ± 0,11	17,24 ± 1,22	11,87 ± 2,86	1,36E+08	2,27E+06	4,03	0	1	1
18		36,71 35,891	38 123		263 ± 12 104 ± 2	291 ± 13 77 ± 39	6 3 93 12	4 0,91 ± 0,14 0 0,36 ± 0,10	5,18 ± 0,08 10,86 ± 0,03	0,18 ± 0,03 0,03 ± 0,01	5,78 ± 0,59 16,73 ± 0,4	10,22 ± 2,03 13,01 ± 3,84	1,17E+08 1.90E+08	4,68E+06 1,62E+06	4,78	0	1	1
20 (Guthrie)	-97,452	35,851	123	288 ± 11	104 ± 2	152 ± 9	174 13		5,82 ± 0,21	0,14 ± 0,02	20,49 ± 0,8	13,16 ± 3,89	2,04E+08	1,02L+00	4,01	1	1	1
21	-98,411	36,612	21		358 ± 5	5 ± 3		4 0,79 ± 0,07	2,43 ± 0,06	0,32 ± 0,03	11,49 ± 1,47	16,97 ± 1,55	2,22E+08	5,47E+06	3,97	1	1	1
22		36,027	29		145 ± 3	81 ± 21	147 14		4,62 ± 0,11	0,23 ± 0,07	12,76 ± 0,64	7,68 ± 0,41	2,72E+08	3,02E+06	4,38	1	1	1
23		35,948	33		228 ± 5	323 ± 37	106 1	1 0,75 ± 0,12	2,46 ± 0,04	0,31 ± 0,05	5,73 ± 0,46	4,05 ± 1,34	2,99E+08	2,59E+06	4,19	0	1	0
24		36,22	20		50 ± 1	46 ± 7	1	3 0,64 ± 0,12	3,46 ± 0,07	0,2 ± 0,03	17,02 ± 0,79	11,60 ± 2,01	1,38E+08	2,10E+06	4,02	0	1	1
25		36,626 36,813	89 47		103 ± 5 140 ± 6	74 ± 33 202 ± 68	155 17 31 9	6 1,95 ± 0,13 3 0,46 ± 0,12	6,93 ± 0,13 3,56 ± 0,07	0,28 ± 0,02 0,13 ± 0,03	12,16 ± 1,49 10,2 ± 1,32	6,39 ± 3,42 4,26 ± 2,44	1,98E+08 1,86E+08	2,75E+06 3,92E+06	4,53	1	1	0
26		36,813	32		289 ± 3	202 ± 68 267 ± 13	171 14		2,93 ± 0,07	0,13 ± 0,03	10,2 ± 1,32 10,24 ± 0,69	4,26 ± 2,44 5,82 ± 0,99	1,86E+08 1.82E+08	3,92E+06 2.78E+06	4,44	1	1	1
28		36,837	53		265 ± 9	287 ± 7	116 13		2,46 ± 0,00	0,19 ± 0,01	8,16 ± 0,61	12,99 ± 1,82	1,15E+08	3,84E+06	4,06	1	1	1
29		35,672	63		97 ± 6	152 ± 19		6 0,27 ± 0,13	3,91 ± 0,09	0,07 ± 0,03	14,47 ± 0,62	13,00 ± 2,45	2,26E+08	1,47E+06	4,83	0	1	1
30		36,843	20		143 ± 6	209 ± 69	35 10		1,42 ± 0,09	0,06 ± 0,03	10,41 ± 1,12	5,10 ± 2,40	1,72E+08	3,63E+06	3,92	0	1	0
31		36,721	24		344 ± 0	343 ± 4		6 1,29 ± 0,18	4,21 ± 0,12	0,31 ± 0,04	6,33 ± 0,23	7,97 ± 0,21	2,14E+08	5,34E+06	3,88	1	1	1
32		36,94 36,836	52		146 ± 4	218 ± 14		0 0,98 ± 0,09	3,42 ± 0,20	0,29 ± 0,03	14,03 ± 0,59	11,55 ± 0,74	9,06E+07 2,83E+08	2,09E+06 6.08E+06	4,45	1	1	1
33		36,836	76 98		307 ± 11 290 ± 6	294 ± 11 298 ± 6		1 1,07 ± 0,43 0 1,30 ± 0,07	9,56 ± 0,05 6,37 ± 0,09	0,11 ± 0,05 0,2 ± 0,01	2,38 ± 0,18 7,67 ± 1,14	3,28 ± 0,47 12,30 ± 1,27	2,83E+08 2,19E+08	5,77E+06	5,18	1	1	1
35		36,265	27		26 ± 1	33 ± 17		6 0,02 ± 0,07	1,51 ± 0,05	0,05 ± 0,05	8,07 ± 0,1	7,09 ± 0,58	1,74E+08	2,83E+06	3,94	0	1	1
36		36,742	95		129 ± 2	140 ± 34	124 13		5,29 ± 0,04	0,1 ± 0,01	16,28 ± 0,93	10,49 ± 1,52	1,76E+08	2,03E+06	4,24	0	1	0
37	-97,522	36,281	130		41 ± 4	60 ± 12		4 2,91 ± 0,19	5,57 ± 0,07	0,52 ± 0,04	15,28 ± 0,93	11,87 ± 2,78	1,50E+08	2,40E+06	4,84	1	1	1
38		36,116	22		83 ± 1	56 ± 7		1 0,30 ± 0,08	2,86 ± 0,03	0,11 ± 0,03	15,11 ± 0,45	8,50 ± 0,91	1,08E+08	1,27E+06	4,45	0	1	1
39		35,922	36		111 ± 1	56 ± 19		6 0,51 ± 0,13	3,36 ± 0,05	0,15 ± 0,04	19,66 ± 0,38	8,80 ± 1,05	1,66E+08	1,32E+06	4,62	1	1	1
40		36,868	23		277 ± 1	275 ± 3		0 0,21 ± 0,09	2,89 ± 0,02	0,08 ± 0,03	8,3 ± 0,28	10,10 ± 0,31	2,29E+08	5,83E+06	4,29	0	1	1
41	-97,394 -96,805	35,744 36,334	52 20		101 ± 2 162 ± 1	152 ± 13 203 ± 3	71 12 41 8	2 1,14 ± 0,06 2 1,81 ± 0,09	2,90 ± 0,13 2,93 ± 0,02	0,39 ± 0,02 0,62 ± 0,03	17,41 ± 0,26 18,65 ± 0,45	14,74 ± 2,88 10,92 ± 0,20	2,13E+08 1,19E+08	1,81E+06 1,94E+06	4,39	1	1	1
43 (Woodward)	-99,008	36,525	176	65 ± 11	47 ± 1	47 ± 7		8 2,57 ± 0,29	11,50 ± 0,13	0,02 ± 0,03	30,13 ± 1,26	38,24 ± 1,07	4,47E+07	1,34L+00 1,29E+06	4,53	1	1	1
44 (Cushing)	-96,811	35,988	27		202 ± 2	310 ± 14		7 0,42 ± 0,19	3,23 ± 0,30	0,14 ± 0,05	6,96 ± 0,48	3,15 ± 0,58	3,09E+08	2,47E+06	5,15	0	1	1
45	-97,397	36,564	27	232 ± 10	76 ± 1	21 ± 73	156 14		1,89 ± 0,09	0,3 ± 0,04	2,71 ± 0,06	1,63 ± 1,86	1,58E+08	2,08E+06	3,5	1	1	0
46		36,828	41		258 ± 10	275 ± 12	160 17		5,37 ± 0,11	0,07 ± 0,03	7,09 ± 0,91	12,17 ± 1,71	1,44E+08	4,66E+06	4,5	0	1	1
47		36,928	27		159 ± 7	254 ± 4	135 13		1,49 ± 0,11	0,37 ± 0,03	5,45 ± 0,25	11,73 ± 1,79	9,13E+07	2,41E+06	3,7	1	1	1
48		36,069 36,324	20 55		149 ± 2 27 ± 3	34 ± 8 50 ± 15	149 9 178 15	6 0,47 ± 0,04 5 0,21 ± 0,11	1,17 ± 0,01 4,25 ± 0,07	0,41 ± 0,03 0,06 ± 0,02	11,03 ± 0,54	7,16 ± 0,43	2,12E+08 1,45E+08	2,56E+06 2,48E+06	4,02	0	1	1
50	-97,535	36,324	33		27 ± 3 4 ± 6	28 ± 12		8 0,50 ± 0,21	4,25 ± 0,07 5,66 ± 0,01	0,06 ± 0,02 0,09 ± 0,04	12,82 ± 0,42 4,67 ± 0,05	10,21 ± 1,99 5,50 ± 0,81	2.05E+08	2,48E+06 3.16E+06	4,47	0	1	1
51 (Fairview)	-98,738	36,48	148		19 ± 3	23 ± 6	112 11		13,69 ± 0,18	0,12 ± 0,02	24,84 ± 1,19	28,83 ± 1,60	7,43E+07	2,54E+06	5,31	0	1	1
52 (10) 100		35,975	30	278 ± 11	142 ± 4	66 ± 38	136 14		3,20 ± 0,24	0,22 ± 0,04	10,76 ± 0,58	7,51 ± 1,99	2,05E+08	2,14E+06	4,13	1	1	0
53	-97,234	35,858	21	46 ± 1	96 ± 1	50 ± 6	50	4 1,39 ± 0,09	2,23 ± 0,06	0,63 ± 0,03	10,67 ± 0	14,26 ± 0,56	2,04E+08	1,75E+06	4,12	1	1	1
54		36,568	26		94 ± 5	61 ± 26		1 0,88 ± 0,09	2,63 ± 0,14	0,38 ± 0,04	12,29 ± 1,89	6,25 ± 3,71	1,11E+08	2,78E+06	3,82	0	1	0
55		36,214	28		54 ± 1	47 ± 8	139 13		1,93 ± 0,05	0,2 ± 0,03	14,07 ± 0,38	11,86 ± 2,72	1,26E+08	2,16E+06	4,46	1	1	1
56		36,621 36,849	44		126 ± 3	147 ± 86	143 12 155 17		5,19 ± 0,04	0,3 ± 0,04	8,48 ± 1,61	1,48 ± 0,97	1,53E+08	2,98E+06	4,22	1	1	0
57		36,849	29 20		263 ± 13 286 ± 1	283 ± 11 289 ± 4		5 0,35 ± 0,10 7 0,74 ± 0,07	2,71 ± 0,10 3,03 ± 0,11	0,16 ± 0,04 0,25 ± 0,02	9,1 ± 0,98 9,68 ± 0,56	13,65 ± 2,37 13,57 ± 0,47	1,41E+08 1,58E+08	4,51E+06 5,93E+06	4,11 4,13	0	1	1
59		36,268	35		356 ± 18	289 ± 4 29 ± 20		7 0,52 ± 0,13	2,41 ± 0,12	0,23 ± 0,02	3,72 ± 0,13	5,69 ± 1,76	2.05E+08	3.05E+06	4,15	0	1	1
60		36,441	27		266 ± 1	248 ± 2		6 0,94 ± 0,44	3,47 ± 0,13	0,28 ± 0,12	13,87 ± 0,08	16,43 ± 1,00	1,38E+08	2,44E+06	4,15	0	1	1
														Sun	n stable cases	36	60	38

Table S1. Parameters derived from the comprehensive migration analysis with respect to multiple injection wells for 60 cluster considered inOklahoma using 10 temporal bins and a diffusion coefficient of 1.5 m²/s.