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3 Please cite as:

4 Schoenball, M., F. R. Walsh, M. Weingarten, and W. L. Ellsworth, 2017, How faults wake up: the Guthrie-
5 Langston, Oklahoma earthquakes: The Leading Edge, *in press*.

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8 How faults wake up: the Guthrie-Langston, Oklahoma earthquakes

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13

14 Abstract

15 Large-scale wastewater disposal has led to a fast-paced reawakening of faults in the Oklahoma/Kansas
16 region. High resolution earthquake relocations show that the inventory of ancient basement faults in the
17 study region differs from results of seismic surveys and geologic mapping focused on the sedimentary
18 cover. We analyze the evolution of seismic activity in the Guthrie-Langston sequence in central
19 Oklahoma in greater detail. Here, seismic activity has reactivated a network of at least 12 sub-vertical
20 faults in an area less than 10 km across. Recorded activity began in late 2013 and peaked about 6
21 months later and includes two M4 earthquakes. These earthquakes characteristically occur at about 4
22 km depth below the top of the basement and do not reach the sedimentary cover. The sequence shows
23 a radial growth pattern despite being no closer than 10 km to significant wastewater disposal activity.
24 Hydrologic modeling suggests that activity initiated with a time lag of several years relative to early
25 injection activity. Once initiated, earthquake interactions contribute to the propagation of seismicity
26 along the reactivated faults. As a result, the spatio-temporal evolution of the seismicity mimics a
27 diffusive pattern that is typically thought to be associated with injection activity. Analysis of the Fault
28 Slip Potential shows that most faults are critically stressed in the contemporary stress field. Activity on
29 some faults, for which we find low slip probability, suggest a significant contribution of geomechanical
30 heterogeneities to the reawakening of these ancient basement faults.

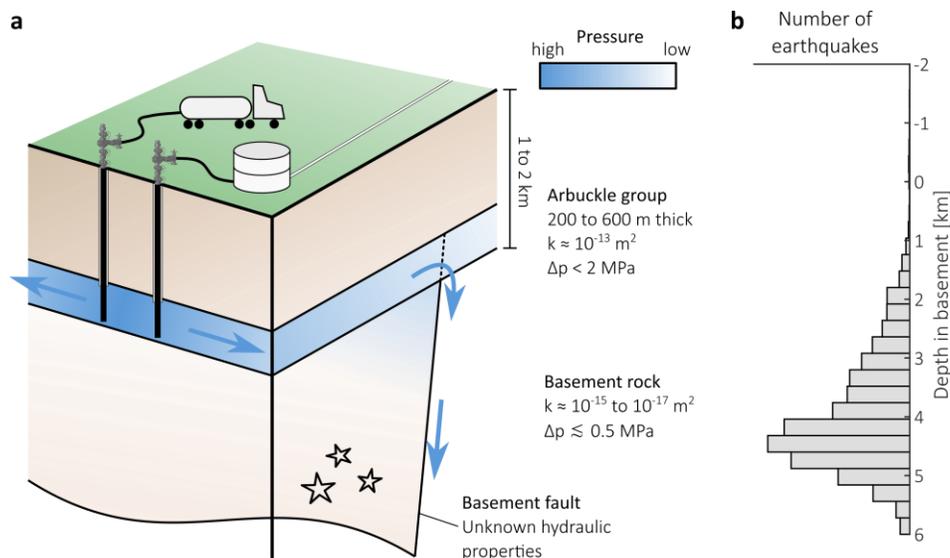
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33 Introduction

34 Since about 2009, the induced seismicity crisis in Oklahoma has produced a carpet of earthquakes that
35 spans an area about 200 km across, stretching from Oklahoma City into southern Kansas. It is now
36 generally accepted that the uptick of seismicity is caused by large-scale wastewater injection into the
37 Arbuckle Group (Ellsworth, et al. 2015, Walsh and Zoback, 2015, Weingarten et al., 2015). Recent efforts
38 to precisely relocate the activity – made possible through waveform data provided by private companies
39 – show that the carpet of earthquakes is composed of discrete basement faults. This high-resolution
40 image of the earthquakes provides unprecedented insights into the regional network of ancient
41 basement faults in this previously quiescent intraplate region (Schoenball and Ellsworth, 2017).

42 In Figure 1 we summarize the current understanding of the link between wastewater and induced
43 earthquakes in Oklahoma and Kansas. Wastewater is disposed into over 800 UIC class II wells. Wells are
44 drilled into the Arbuckle Group and sometimes reached into the basement. Fluids are transported by
45 trucks or through pipelines to disposal wells and injected into the high permeability Arbuckle Group.
46 Addition of fluid creates a far-reaching plume of modestly elevated pore pressure (< 2 MPa) relative to
47 the natural underpressured state of the Arbuckle. Permeable pathways from the Arbuckle into the
48 basement raises the pressure in hydrologically connected basement faults, reducing their strength
49 through the well-known effective stress relation (Raleigh et al., 1976). Earthquake sequences have been
50 observed several 10s of kilometers away from large injectors elsewhere in Oklahoma (Keranen et al.,
51 2014) where modeled effective stress changes at hypocentral depth are less than 0.5 MPa. Because of
52 the many active disposal wells and the far-reaching pressure perturbation, it is generally impossible to
53 associate induced sequences with injection activity of specific wells.



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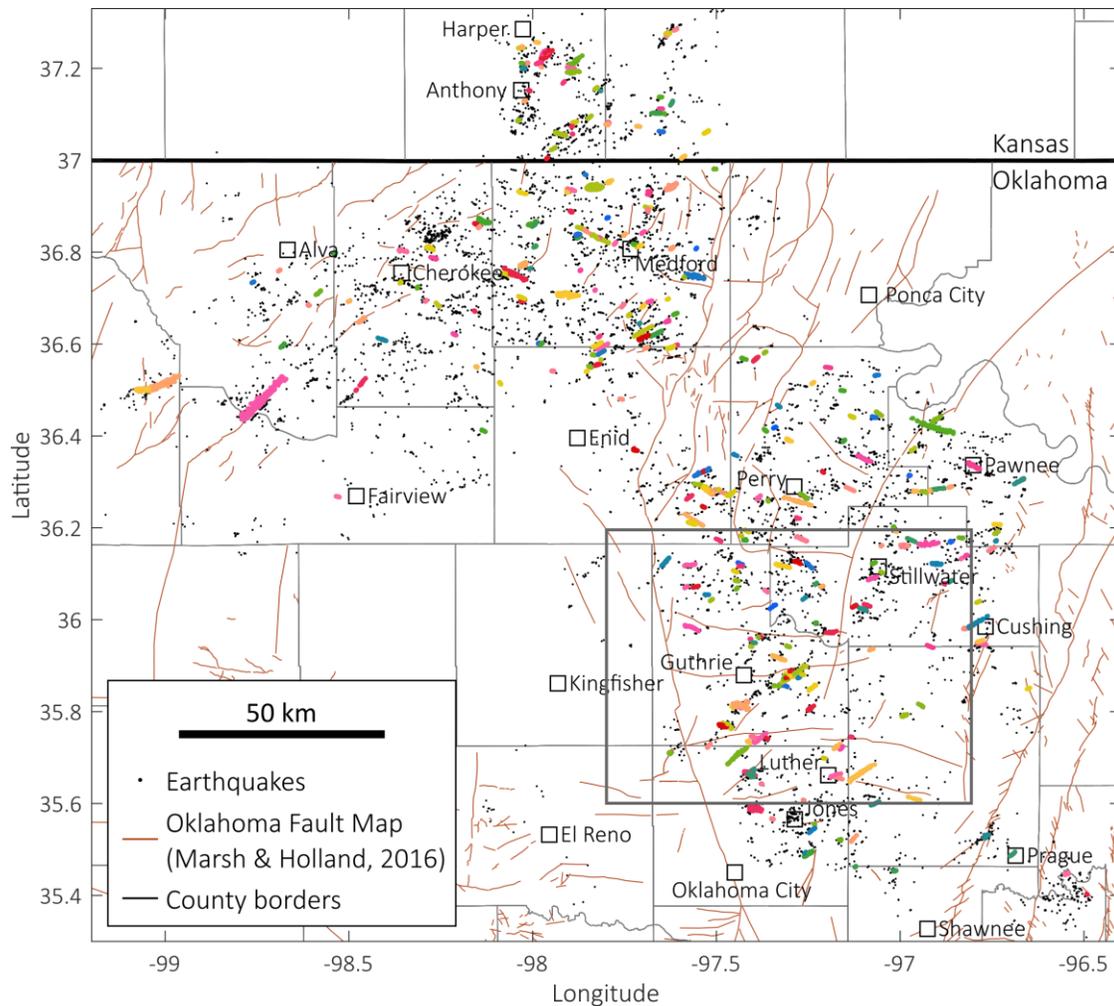
55 *Figure 1: a) Conceptual model of induced seismicity in Oklahoma and southern Kansas. b) Observed focal depths of induced*
56 *earthquakes in Oklahoma and southern Kansas relative to the base of the Arbuckle Group/top of basement (from Schoenball*
57 *and Ellsworth, 2017).*

58 Walsh and Zoback (2016) developed a probabilistic method to estimate the potential for fault
59 reactivation based on geomechanical theory and Monte Carlo sampling of the relevant input parameter
60 distributions. Based on known fault orientation and assumptions of the geomechanical conditions, they

61 estimate the Fault Slip Potential (FSP) as a proxy for the probability of reactivating specific faults through
62 injection operations.

63 Here we compare the fault structures resolved from precise earthquake relocations with the known
64 inventory of basement faults. We model pore pressure changes in the Arbuckle Group and at
65 hypocentral depths and test the FSP framework by applying it to these faults, and focus on a sequence
66 of earthquakes between Guthrie and Langston, Oklahoma.

67 Regional fault network



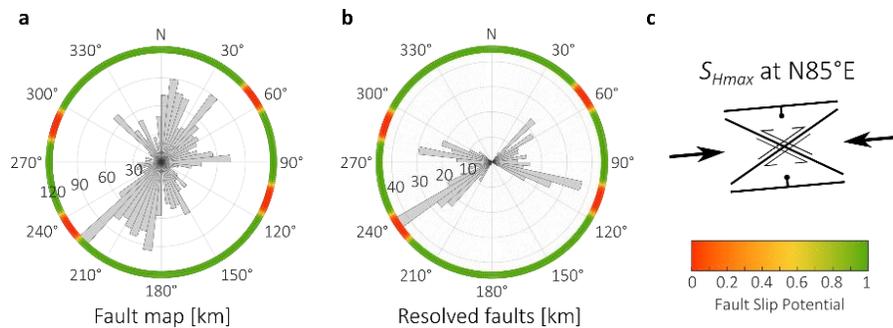
68
69 *Figure 2: Map of relocated earthquakes in the Oklahoma and southern Kansas area. Earthquakes on interpreted faults are*
70 *drawn in distinct colors. Brown lines are faults from Marsh & Holland (2016). The box shows the area of Figure 5.*

71 The refined earthquake relocations of Schoenball and Ellsworth (2017) are shown in Figure 2.
72 Earthquakes cluster along tight lineations that we interpreted as individual basement faults. Both near-
73 vertical and dipping structures are found, with most displaying strike-slip movement. Earthquakes
74 generally occur in the basement, with the distribution of hypocentral depth peaking at 4 km below the
75 top of basement (Figure 1b). Hypocenters in the sedimentary section are extremely rare.

76 Also shown in Figure 2 are the Oklahoma faults compiled by Marsh & Holland (2016). This map was
77 compiled from interpretation of reflection seismic data and geologic mapping. Almost none of the

78 earthquake sequences are associated with any of the mapped faults. Furthermore, we notice that the
 79 trends of mapped fault structures differ from the trends that are apparent from the earthquake
 80 locations. To further study the network of faults, we applied the DBSCAN algorithm (Ester et al., 1996) to
 81 objectively identify individual faults in the basement (Schoenball and Ellsworth, 2017). For each fault, we
 82 measure strike and dip using principal component analysis. More than 300 faults could be characterized
 83 in this way. We compare the strike of fault segments weighted by fault length with the mapped faults in
 84 Figure 3. For the Oklahoma Fault Map, we only consider fault segments that are at least partially within
 85 the area that has seen widespread seismicity in the last few years.

86 There is a clear difference in the dominant fault trends between both fault maps. In the Oklahoma Fault
 87 Map a large-scale NNE-SSW trend, related to the Nemaha Uplift and Midcontinent Rift System,
 88 predominates. This trend is absent in the faults illuminated by the earthquakes. Those faults show a
 89 clear pattern of conjugate faulting, that are favorably aligned for slip within the contemporary tectonic
 90 stress state. The predominant fault strikes from earthquake locations are in rough agreement with what
 91 would be expected from strike-slip faulting with the observed stress orientation (Figure 3). Strike
 92 directions that are associated with the Nemaha Uplift are stable in the contemporary stress field and,
 93 from a geomechanical perspective, are highly unlikely to reactivate regardless of the fluid pressure rise
 94 (Walsh and Zoback, 2016).

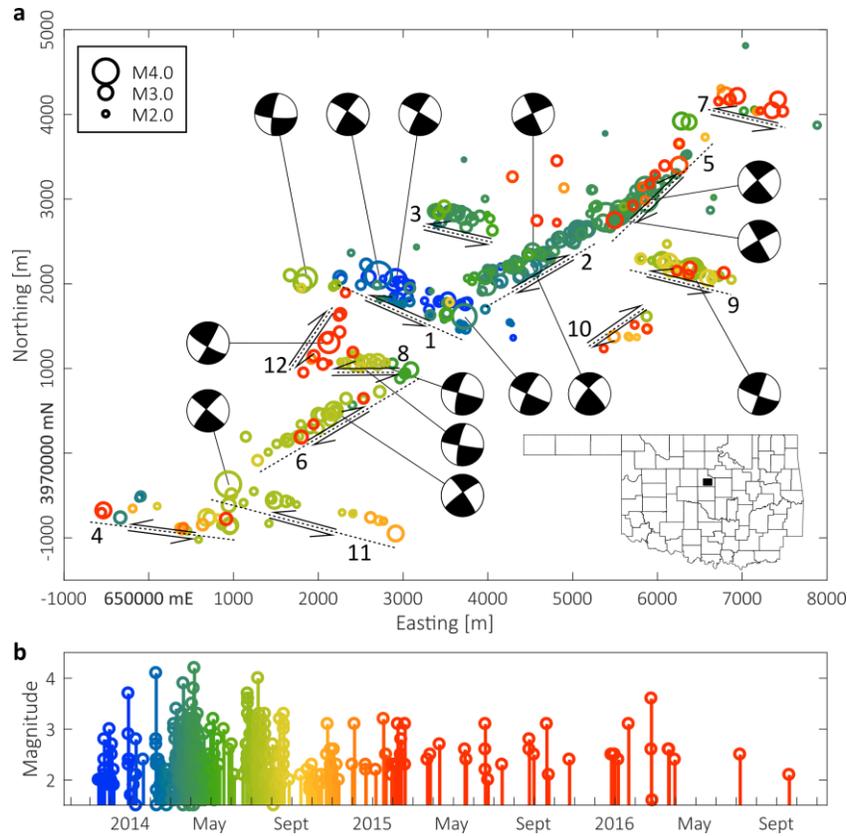


95
 96 *Figure 3: Comparison of fault strikes from (a) the Oklahoma Fault Map (Marsh and Holland, 2016) and (b) resolved from*
 97 *earthquake hypocenters. The colors show the fault slip potential for vertical faults for comparison. (c) shows the orientation of*
 98 *critically stressed fault assuming S_{Hmax} oriented at N85°E (Alt and Zoback, 2017).*

99 **Case study: The Guthrie-Langston sequence**

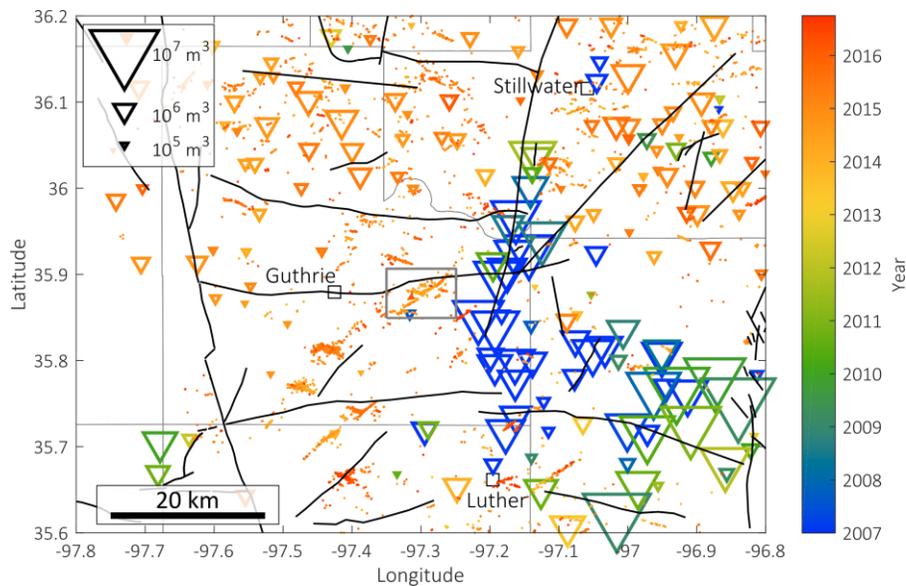
100 The sequence of earthquakes that began in late 2013 between Guthrie and Langston in central
 101 Oklahoma is particularly rich in earthquakes (Benz et al., 2015) and resolved fault structures (Figure 4).
 102 We want to emphasize however, that many of the observations that we detail below are not specific to
 103 this sequence, but are found for other sequences in Oklahoma and Kansas as well.

104 We summarize the injection history and seismic activity in the Guthrie region in Figure 5. Minor
 105 wastewater injection about 10 km east of the Guthrie-Langston sequence occurred at least since 1997.
 106 Significant wastewater disposal with injection rates greater than 100,000 m³ per month in single wells
 107 began in 2001. Injection in this area peaked between 2002 and 2007 and declined thereafter. Most of
 108 these wells are located along a N-S striking fault (Marsh and Holland, 2016) that potentially acted as a
 109 high permeable fluid conduit allowing for large injection volumes. North of Guthrie, large-scale injection
 110 started in 2012 and peaked in 2015. The monthly and cumulative injection volumes in the north never
 111 surpassed the volumes injected to the east.



112

113 *Figure 4: Evolution of the earthquake Guthrie-Langston sequence. (a) Map view with earthquakes colored by order of occurrence*
 114 *as in (b). Fault trends interpreted from the distribution of hypocenters are shown in dashed lines with the sense of slip indicated*
 115 *by arrows. Focal mechanisms are courtesy of Robert Herrmann (see Herrmann et al. 2011). The inset in the bottom right shows*
 116 *the location of the map in the state of Oklahoma. (b) shows the temporal evolution of the sequence with colors the same as in*
 117 *(a).*

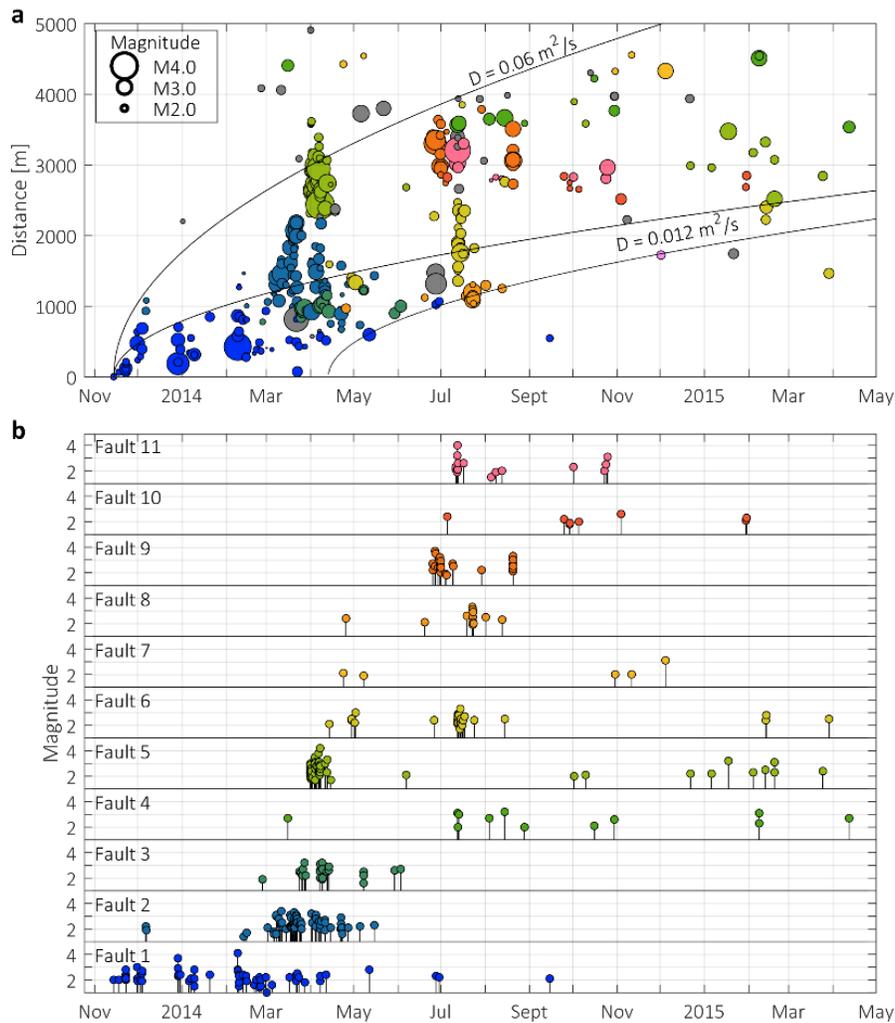


118

119 *Figure 5: Injection wells and earthquakes around the Guthrie area. Earthquakes (dots) are colored by their time, injection wells*
 120 *(triangles) are sized by their cumulative injection volume between 1995 and 2016 and colored according to the period of highest*
 121 *injection (mean injection time weighted by volume). The box shows the area of Figure 4.*

122 The first earthquakes were detected by the Oklahoma Geological Survey in late 2013 on a ESE striking
 123 fault (Fault 1) and activity propagated towards nearby Fault 2, to the northeast of Fault 1. Seismic
 124 activity reached a maximum rate in March and April 2014 when Faults 2 and 5 were in their most active
 125 phases (Figure 6). This was also the time when the two largest events of M4.2 and M4.1 occurred.
 126 Overall, 398 earthquakes were recorded through November 2016 when the catalog ends.

127 A delay from initiation to the highest rate of activity is observed in many sequences throughout
 128 Oklahoma (Schoenball and Ellsworth, 2017). Pre-shock activity typically builds over the course of a
 129 sequence, but sequences never start with the largest event. This pattern is distinctly different from
 130 bursts of natural seismicity, where we typically observe the mainshock preceded by only a small number
 131 of foreshocks, if any. Hence, the seismicity rate is highest early-on in a sequence. The occurrence
 132 pattern of the induced earthquakes suggests that these sequences are initiated by different processes.
 133 The rise of activity to its peak can be interpreted as a probing of the criticality of faults by the
 134 anthropogenic stressing (Dempsey and Suckale, 2016). More and larger asperities of faults activate as
 135 forcing continues.



136

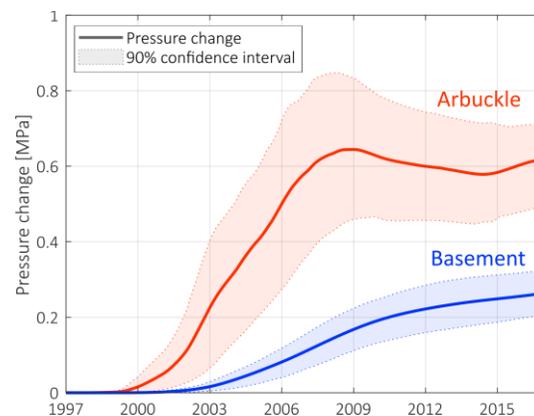
137 *Figure 6: (a) Spatio-temporal evolution of the Guthrie-Langston sequence. Events are grouped by a distinct color for each fault.*
 138 *(b) shows the magnitude and timing of earthquakes grouped for each fault. Fault 12 is not shown because only one event falls in*
 139 *the time window shown.*

140 A remarkable feature of the Guthrie-Langston sequence is the radial growth readily visible in Figure 4.
141 The sequence eventually spreads northeast and southwest by 4 km involving 12 distinct faults that
142 activate in succession. Later activity on two faults to the north and to the ESE (Figure 5) may be
143 interpreted as a continuation of the radial growth pattern to even greater distances.

144 The radial expansion of the sequence is surprising since the closest class II injection well is 3 km from the
145 geometric origin of the sequence and this well only injected a small volume and mostly before 2009
146 (Figure 5). This contradicts the classic view that pressure diffusion away from an injection well is
147 reflected in the spatio-temporal growth pattern of a seismicity cloud (Shapiro et al., 1997). Rather, the
148 growth pattern we observe in the Guthrie-Langston sequence may result from interactions of
149 earthquakes through static stress transfer due to the displacement of each event. As earthquakes occur,
150 neighboring faults are loaded by the displacements and may also activate (Figure 4a). The clear radial
151 migration pattern does not seem to originate from any injection wells. Instead, we infer that other
152 structures act as fluid conduit from the Arbuckle to the fault that initially reactivated (Figure 1). The
153 sequence then grew, driven by the stress or pressure perturbation originating at the intersection
154 between the first fault and the fluid conduit.

155 Pore pressure modeling of all Arbuckle injection wells within 30 kilometers of the Langston-Guthrie
156 sequence shows how changes in pore pressure are manifested in the reservoir formation (Arbuckle
157 Group) and the underlying basement fault (Figure 7). The modeling follows the conceptual framework
158 shown in Figure 1, after Weingarten and Zoback (2016) and Walsh and Zoback (2015). Permeability in
159 the Arbuckle Group is represented by a spatially heterogeneous, log-normal distribution in 100
160 stochastic realizations (mean $k = 10^{-13 \pm 0.7} \text{ m}^2$). The Arbuckle Group overlies a low-permeability, intact
161 crystalline basement ($k = 10^{-17} \text{ m}^2$) with a permeable fault zone representing the Guthrie-Langston
162 complex ($k = 10^{-15} \text{ m}^2$).

163 Modeled pressure rise in the Arbuckle Group at the location of the Guthrie-Langston sequence reached
164 between $\approx 0.5 - 0.8 \text{ MPa}$, peaking in 2008 and slowly falling until about 2014 when it began to rise again
165 due to injection activity to the north. Modeled pressure changes at hypocentral depth, however,
166 steadily rose by $\approx 0.2 - 0.3 \text{ MPa}$ since injection began in about 2002 and were still rising at the end of
167 the simulation. The current modeled rate of pressure increase at hypocentral depth is less than
168 previously observed during 2007 – 2009.



169
170 *Figure 7: Modeled pore pressure changes in the Arbuckle Group and at 4.5 km below the top of basement at the location of the*
171 *Fault 1 in the Guthrie-Langston sequence. The shaded pressure interval is obtained from sampling 100 realizations of log-normal*
172 *stochastic permeability distributions in the Arbuckle Group.*

173 One important result of the modeling shows how the permeability contrast between the Arbuckle
174 Group and the permeable basement delays the onset of pressure propagation to hypocentral depths.
175 Modeled pressures peaked in 2008 in the Arbuckle Group, without any observed seismicity in this
176 sequence until late 2013. Pressure diffusion to hypocentral depths takes years to exceed a critical
177 pressure to induce slip. Furthermore, the modeling indicates that measured pressure changes in the
178 Arbuckle Group alone may be insufficient to adequately characterize pressure changes at hypocentral
179 depths, and thus, induced seismic hazard into the future. Therefore, a combination of measured
180 pressures and calibrated models are needed to adequately characterize and manage future induced
181 seismic hazard (Yeck et al., 2016).

182 In Figure 6a we plot the growth of the sequence as the distance from the first earthquake, separately
183 coloring each fault. We see that the initial activity on Fault 1 can be modeled by a diffusive process with
184 $D = 0.012 \text{ m}^2/\text{s}$. Later, activity jumps to other faults and also ahead of the initial triggering front. In order
185 to fit all activity into a single diffusive process, D has to be of the order of $0.06 \text{ m}^2/\text{s}$. Another possible
186 interpretation is that every fault spawns its own sub-sequence with a different diffusivity as can be
187 observed in the r - t plot in Figure 6a.

188 We also observe differences in the temporal behavior of activity on each fault. Once activated, some
189 faults show continuous activity (e.g. Fault 1 has activity over about 6 months), while others have short-
190 lived bursts (e.g. fault 5 was active for just over two weeks). This suggests that activity on some faults is
191 dominated by slow processes such as fluid diffusion, and activity on others is dominated by fast
192 processes such as stress transfer from one rupture to the next. It is particularly noteworthy that none of
193 the individual faults activate with their largest event. Instead, the observed largest magnitudes tend to
194 increase as more events are produced (Figure 6b). This is in agreement with the statistical model of van
195 der Elst et al. (2016) where each earthquake magnitude is an independent sample of the local
196 magnitude-frequency distribution.

197 The complex spatio-temporal behavior is evidence for several processes at play in the development of
198 induced earthquake sequences. We therefore regard estimates of diffusivity based on the spatio-
199 temporal envelope of seismicity to infer the seismogenic diffusivity (Talwani et al., 2007) – a convolution
200 of the hydraulic diffusivity, stress redistribution and processes that sample the heterogeneity and the
201 criticality of the tectonic stress field.

202 After about five months, activity on Fault 1 declined as it migrated away from its origin. As a result, we
203 see a zone of quiescence spreading from the origin (Figure 6a). The envelope of this spreading zone of
204 quiescence can also be approximated by a diffusion process with $D = 0.012 \text{ m}^2/\text{s}$. This back front
205 (Parotidis, 2004) seems to exist also for later activity on other faults. The existence of a back front
206 suggests that seismicity is driven by a stress perturbation that relaxes after it sweeps over the faults. The
207 coincidence of the diffusivity obtained for the initial activity on Fault 1 and the back front suggests both
208 represent the same process. The continuous activity on Fault 1 and the absence of activity bursts
209 indicates that earthquake interactions are less important for driving seismicity on this fault. From these
210 observations, we conclude that the diffusivity observed for the triggering and back fronts is indeed the
211 hydraulic diffusivity of the fault system.

212 Retrospective estimations of fault slip potential

213 The identified fault structures can be used to test the FSP approach of Walsh and Zoback (2016). But
214 first we must assess the precision of the earthquake hypocenters from which the fault structures were
215 derived. Earthquakes can generally be located more precisely in latitude and longitude than in depth.
216 For the relocated catalog, the vertical precision is about a factor 5 to 10 less than the horizontal. To get
217 accurate estimates of an earthquake's depth, we require stations that are close to the epicenter,
218 typically closer than one focal depth. For the Guthrie-Langston sequence, the closest stations are
219 between 5 and 15 km away from the events and seismicity is about 6 km deep. This is not sufficient to
220 resolve the vertical structure in detail. As a result, the dip angles of resolved faults may be systematically
221 biased.

222 Moment tensor solutions for the larger events provide an independent constraint of fault dip. Here we
223 use moment tensor solutions determined by St. Louis University (see Herrmann et al. (2011) for details
224 on their methods). Generally, the strike and, with some exceptions, the dip of faults determined from
225 the hypocenters is in good agreement with one of the nodal planes of moment tensor solutions
226 obtained from waveform modeling (Figure 4). In some cases, the lack of close-by stations provides
227 insufficient coverage to resolve the fault dip. As a result, the resolved fault planes get vertically
228 compressed in the relative relocation step and derived dip angles are unrealistically low. The most
229 obvious example is Fault 1 which has a very well-defined fault plane. The dip resolved from earthquake
230 hypocenters is an unreasonably low 33° . Moment tensors of the two largest events associated with this
231 structure have dips of 80° and 85° . To reconcile this discrepancy, we estimate the minimal along-dip
232 extent of fault reactivation as the rupture length of largest events. We assume a roughly circular rupture
233 area of 1 km across for this M4.2 event. The fault is activated along about 1700 m of strike and all
234 hypocenters associated with this fault are distributed over about 270 m along the dip direction. We
235 estimate the lower bound on the dip using these dimensions to be about 75° . This is in rough agreement
236 with the moment tensors.

237 The distribution of seismometers provides suitable azimuthal coverage to precisely constrain the
238 epicenters of earthquakes and we do not expect a systematic error in the strike of faults. Uncertainties
239 of the resolved fault strikes were estimated from bootstrap resampling of hypocenters used for the
240 principal component analysis.

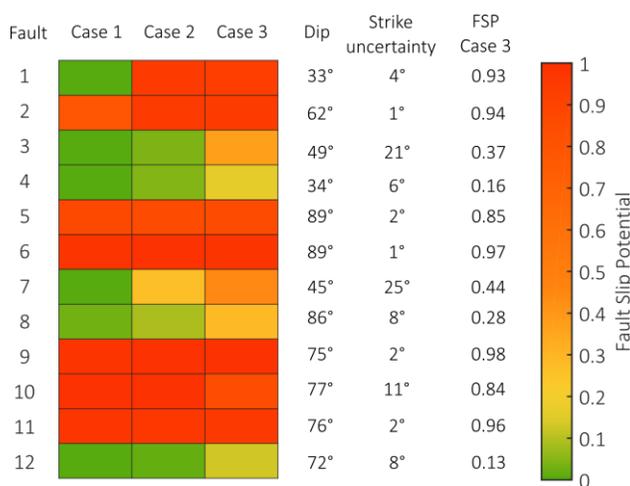
241 Estimation of the Fault Slip Potential requires several steps (Walsh and Zoback, 2016). First, the local
242 stress field has to be characterized. This can be done using stress determinations from borehole data
243 and inversion of focal mechanisms (Zoback et al., 2003, Hardebeck and Michael, 2006). Distributions of
244 the stress measurements can be obtained from bootstrap resampling of inverted focal mechanisms and
245 from statistical analysis of borehole data (Schoenball and Davatzes, 2017). Furthermore, distributions of
246 the coefficient of friction and initial pore pressure are assumed (Nelson et al., 2015, Carpenter et al.,
247 2016). Using Mohr-Coulomb faulting theory and Monte Carlo sampling of the input parameter
248 distributions, the probability of a fault slipping under a given pressure change is estimated (Walsh and
249 Zoback, 2016).

250 The state of stress on each fault is both heterogeneous and uncertain. While most geomechanical
251 modeling software model stress variability but not uncertainty, FSP models uncertainty but not
252 variability. FSP assumes that each mechanical model is spatially uniform and stress is linearly increasing
253 with depth. Uncertainty is modeled by assuming uniform distributions which are taken as the 2nd and

254 98th percentile of distributions (of varying shape) in Walsh and Zoback (2016). The result of FSP is a
 255 distribution of pore pressure to slip on each fault.

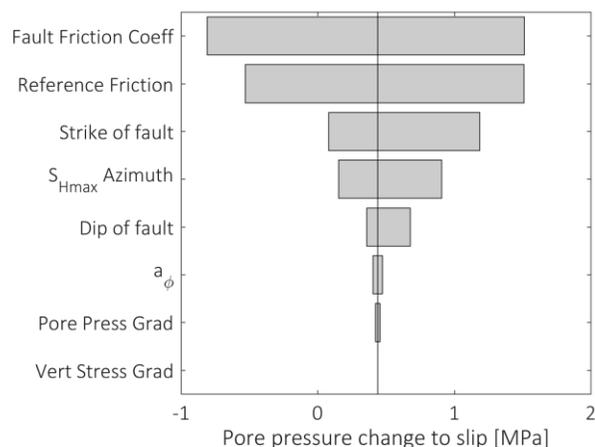
256 In Figure 8 we summarize the estimated FSP for three different assumptions about the orientation of
 257 interpreted faults within the same uncertain stress field. In the first case, we calculate FSP using the
 258 strike and dip as estimated from the relocated earthquake hypocenters with no uncertainty in either
 259 value. Only 6 out of 12 faults have a FSP larger than 0.5 in this case, and 4 have no potential to slip if 2
 260 MPa is added to them. However, as discussed above, we are not confident about the fault dip angles.
 261 We are, however, confident about the resolved strikes. Therefore, in column 2 we compute FSP for the
 262 resolved strike and assume a dip of $85^\circ \pm 5^\circ$. Now, 7 out of 12 faults have a FSP of larger than 0.5 and all
 263 have at least 4% slip potential. In the third case, we also allow the strike to vary as a uniform distribution
 264 within its 2- σ interval. 7 out of 12 faults still have FSP > 0.5, but slip potentials have generally increased.
 265 This analysis demonstrates the importance that the fault orientation relative to the stress field has on
 266 estimates of FSP. Discrepancies as small as 10° or less can have a strong impact on FSP.

267 Fault 12 is the last fault to be activated during this sequence and the activated fault with the lowest FSP.
 268 It is misoriented by about 20° from the optimal strike for failure if the maximum horizontal stress is
 269 trending at 82°. The fault orientation is interpreted based on hypocenters of 10 events. The focal
 270 mechanism of the largest M3.6 event has a strike of 29° in close agreement with the interpreted fault
 271 strike of 35°. We therefore trust this fault orientation. This suggests that at scales of smaller faults
 272 heterogeneity of stress might play a significant role. The late activation of the fault and its proximity to
 273 earlier active faults suggests that previous activity on nearby faults may contribute to changing the state
 274 of stress on this fault such that it became reactivated. However, it is unlikely that static stress transfer
 275 rotated the stress state enough to enable Fault 12 to slip. Instead, local heterogeneity of the stress field
 276 may have caused this fault to slip under moderate stress perturbations. Alt and Zoback (2017) found
 277 that the stress orientation is consistent on a large scale in the Oklahoma region. However, stress
 278 rotations of 20° or more are frequently observed locally in borehole data. Such rotations can be
 279 explained by slip on faults and reflect the heterogeneity of the state of stress (Barton and Zoback, 1994,
 280 Sahara et al., 2014, Schoenball and Davatzes, 2017).



281
 282 *Figure 8: Fault Slip Potential after Walsh & Zoback (2016) for faults derived from hypocenter locations for different assumptions*
 283 *of the accuracy of the resolved strike and dip. Case 1 assumes strike and dip as resolved from earthquake hypocenters, Case 2*
 284 *assumes dip of $85^\circ \pm 5^\circ$ and strike as resolved from earthquake hypocenters and Case 3 assumes dip of $85^\circ \pm 5^\circ$ and strike and its*
 285 *2 σ -error.*

286 In Figure 9 we summarize the influence of each parameter of the FSP analysis on fault number 1 under
 287 case 3, with strike of $115^\circ \pm 4^\circ$ and dip of $85^\circ \pm 5^\circ$. This is done by calculating the pore pressure to slip with
 288 each parameter at the center, lower and upper bound of its distribution. Those parameters that provide
 289 the largest variability in answers are ranked at the top. For this fault, it is readily apparent that the
 290 pressure to slip is most sensitive to the frictional properties of the fault and the magnitude of the stress
 291 state (as represented by the reference friction). These are followed in importance by the strike of the
 292 fault relative to the trend of S_{Hmax} . It is not surprising that the pressure to slip is not sensitive to the
 293 vertical stress because in a strike-slip faulting regime, it is the intermediate principal stress. Similarly, the
 294 uncertainty of the relative stress magnitudes, described by the a_ϕ parameter, does not play a significant
 295 role within its uncertainty. This can be used to inform prioritization of which parameters should be
 296 better constrained to decrease uncertainty.



297
 298 *Figure 9: Tornado diagram summarizing the impact of variations of the input parameters on the Fault Slip Potential for fault 1.*

299 **Conclusions**

300 Refined earthquake locations provide an image of the reawakened fault structures at high resolution
 301 enabling us to study their evolution in detail. Sequences of induced earthquakes typically grow to larger
 302 magnitudes after they initiate with minor activity. We do not typically see a mainshock-aftershock
 303 pattern without any prior activity. Improved monitoring can help to anticipate potentially damaging
 304 sequences. Reducing injection activity typically reduces the earthquake activity and lessens the
 305 probability for large magnitude events to occur (Langenbruch and Zoback, 2016).

306 The Guthrie-Langston sequence occurred with a large temporal and spatial separation from the nearest
 307 injection activity. Large scale injection east of the sequence does not show an immediate temporal
 308 correlation with the occurrence of these earthquakes. There is a stronger temporal correlation with the
 309 onset of large-scale injection activity about 15 km north of the sequence. However, stochastic pore
 310 pressure modeling indicates both injection areas contributed to the delayed pressure diffusion to depth.
 311 Modeled pressures in the Arbuckle Group peaked in 2008 and have slowly declined, but pressure at
 312 hypocentral depths in the Guthrie-Langston sequence has steadily increased through the end of 2016.
 313 Due to the large number of disposal wells and large distances between injection sites and seismicity
 314 sequences, it remains difficult to associate activity in isolated sequences to particular wells.

315 Previously, the correlation of injection activity and earthquake occurrence in space and time has been
 316 used as a strong argument to identify man-made sequences (Davis and Frohlich, 1993). We have shown

317 that the Guthrie-Langston sequence grows in a radial pattern, reminiscent of a radial diffusion process
318 originating at an injection well (Shapiro et al., 1997). However, there is no injection well near the
319 sequence origin and different processes must be at play to propagate the seismicity such as static stress
320 transfer.

321 Analysis of the Fault Slip Potential for the reactivated faults has shown significant probability for slip for
322 most reactivated faults. Low FSP values for few faults may indicate to the role largely unknown
323 heterogeneities of the geomechanical conditions such as state of stress and friction play. FSP analysis
324 can only be useful if we have a good understanding of the faults in the area of interest and their
325 geomechanical state. Potential pitfalls in its application include large uncertainties (such as sliding
326 friction, cf. Figure 9), and incomplete sampling of strike-slip basement faults through seismic imaging in
327 quiescent sediments but not in the seismogenic basement. Unfortunately, it remains a geophysical
328 challenge to image ancient sub-vertical faults in igneous basement through active seismic surveys.

329 [Acknowledgements](#)

330 M.S. acknowledges support by the Alexander von Humboldt Foundation. All authors were supported by
331 the Stanford Center for Induced and Triggered Seismicity.

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