Research Article

Using microseismicity to estimate formation permeability for geological storage of CO₂

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Abstract

In this paper, we investigate two approaches for estimating formation permeability based on microseismic data. The two approaches differ in terms of the mechanism that triggers the seismicity: pore pressure triggering mechanism and so-called seepage force (or effective stress) triggering mechanism. Based on microseismic data from a hydraulic fracture experiment using water and supercritical CO_2 injection, we estimate permeability using the two different approaches. The microseismic data comes from two hydraulic stimulation treatments that were performed on two formation intervals having similar geological, geomechanical and *in situ* stress conditions, yet different injection fluid used. Both approaches (pore pressure triggering and the seepage force triggering) provide estimates of permeability within the same order of magnitude. However, the seepage force mechanism (i.e. effective stress perturbation) provides more consistent estimates of permeability between the two different injection fluids.

Keywords: CO₂ injection; fracture stimulation; microseismicity; permeability

1. Introduction

Fracture stimulation has been applied for the past 60 years to enhance recovery from hydrocarbon reservoirs, with an estimated 70% of wells being fracture stimulated, and hence is a key factor in the economic exploitation of unconventional reserves, such as tight-gas and shale-gas reservoirs [1]. Over the past 20 years, microseismic monitoring has developed into one of the most effective methods of monitoring fracture stimulation, and hence is routinely applied to monitor fracture stimulation programs.

The spatial and temporal variations in microseismicity can be used to monitor changes in the stress field, and hence potentially be used to monitor perturbations in fluid pathways as well as top seal and well bore integrity. Furthermore, microseismicity has been used also to characterise spatial and temporal variations within the reservoir and surrounding rock mass by monitoring changes in seismic attributes between the source and receiver [e.g. shear-wave splitting analysis to characterise fracture induced anisotropy, 2-4]. Additional information can be gained by evaluating microseismic failure mechanisms to characterise the rock mass at the source and provide a measure of the strength, orientation and type of elastic failure to potentially quantify damage [e.g. 5-7].

Although microseismicity can provide fairly accurate temporal and spatial locations of brittle failure, how the measured microseismicity relates to the evolution of the induced pressure front and effective stress field as well as creation and enhancement of cracks and fractures is still not well constrained. Examination of the distribution of microseismic events is can help characterize the flow and mechanical properties of the stimulated reservoir. In particular, by assuming that seismicity is triggered by the diffusion of pore pressure from the injection point, Shapiro [8] has shown that the permeability of a formation can be estimated from the rate of increase in event-injection-well distance through time. This spatio-temporal behaviour is

commonly visualized on a so-called r-t plot, (where r is injection-well-to-event distance, and t is time). This method has shown potential for predicting apparent formation permeability and hydrocarbon production for various fracture stimulation case studies [e.g., 9].

Although this pore pressure diffusion approach has shown some promise in estimating reservoir permeability, there are some non-physical aspects to the theory, such as weak pore pressure perturbation triggering seismicity [10]. As an alternative to the pore-pressure diffusion approach, Rozhko [11] introduces the concept of seepage force triggering to predict the *r*-*t* response of seismicity, which considers diffusion of effective stress perturbations as the driving force of microseismicity.

In this paper we compare both methods, predicting formation permeability by modelling observed *r*-*t* behaviour during hydraulic fracturing, where water and supercritical CO_2 have been used as the injected fluids. By estimating permeability using microseismic monitoring, we hope to explore the potential of using microseismic monitoring to constrain formation permeability limitations for large-scale CO_2 injection sites.

2. Models describing spatio-temporal evolution of seismicity

2.1. Pore pressure triggering

The *r*-*t* pore pressure triggering approach is based on the concept that the spatial and temporal evolution of microseismicity is hydraulically induced and characterised in terms of a low frequency pore pressure relaxation mechanism described by Biot [12]. The key assumption for application of this approach to hydraulic fracture induced microseismic data is that the tectonic stress in the subsurface is close to the critical stress needed for brittle failure [e.g. 13]. As such, increasing fluid pressure (i.e., injecting fluid) within the reservoir results in a transient increase of the reservoir pore pressure and a decrease in effective stress. If the decrease in effective stress is sufficient it can lead to relaxation of normal stresses along pre-existing fractures and hence slip along the fracture and associated microseismicity. The following derivations are from Shapiro [8] and so the reader is referred this manuscript for a more detailed presentation of the approach.

Assuming a point-source injector and a homogeneous and isotropic medium, the triggering front (i.e., the distance between outer envelop of the microseismic "cloud" and the fluid injection point) is described by

$$r(t) = \sqrt{4\pi D(t - t_o)}, \qquad (1)$$

where t is observation time, t_0 is injection start time and D is the scalar apparent hydraulic diffusivity. By plotting the microseismic events on a time-distance plot and matching the best-fitting *r*-*t* curve to the triggering front and estimate of the hydraulic diffusivity can be obtained. The calculated apparent diffusivity can then be used to estimate other reservoir and flow parameters, such as formation permeability.

Assuming the injected fluid is incompressible, the fluid volume balance is such that the total injected fluid is equal to sum of the fluid volume within the fracture and lost to the surrounding formation. Further assuming the induced fracture is straight and of fixed height [i.e., the PKN model, see 14] then the fracture half-length is approximated by

$$L(t) \approx \frac{q_i t}{4h_j C_i \sqrt{2t} + 2h_j w},$$
(2)

where q_i is the average injection rate, h_f is the fracture height (either estimated from perforation interval or vertical extent of microseismicity), C_1 is the fluid-loss coefficient and w is the average fracture width. The fluid-loss coefficient C_1 is given by

$$C_i \approx \frac{q_i}{8h_f \sqrt{2\pi D}},\tag{3}$$

Another surface, the back front, characterises the seismically quiet zone after injection stops and tracks the propagation of maximum pore pressure perturbation. The back front is given by

$$r_b(t) = \sqrt{2dDt \left(\frac{t}{t_s} - 1\right) \ln\left(\frac{t}{t - t_s}\right)},\tag{4}$$

where d is the dimension of the pressure diffusion (1D, 2D or 3D) and t_s is the injection shut-off time.

Neglecting induced fracture surface effects such as filter cake permeability damage and effects on pore space and fractures within the vicinity of the fracture treatment, the permeability of the reservoir can be estimated

$$\kappa \approx \frac{q_i^2 \mu_f}{128 h_f^2 \Delta p^2 \phi c_f D},$$
(5)

where c_f and μ_f are the compressibility and viscosity of the reservoir fluid, respectively, Δp is the difference in average injection pressure and the initial (or far-field) reservoir pressure, and ϕ is the reservoir porosity.

2.2. Seepage force triggering

The nonlinear diffusion approach of Shapiro [8] and the so-called Coulomb Failure Stress (CFS) criteria [e.g. 15] applied to microseismicity make the assumption that seismicity is triggered by the propagation of a fluid pressure perturbation front. Based on the above mechanism, it is implied that small changes in pore pressure is sufficient to trigger seismicity and this is often explained by assuming most faults are critically stressed [13,15]. Rozhko [11] argues that induced microseismicity is explained and predicted better by linear diffusion coupled to linear poroelastic deformation rather than the highly nonlinear fluid diffusion mechanism [8] or CFS and critically stressed faults [15].

In the Rozhko [11] approach, the seismicity is triggered by the propagation of an effective stress perturbation front. The role of fluid pressure on rock strength is significant, yet comes about through the Terzaghi [16] effective stress law

$$\underline{\sigma}_{ij} = \sigma_{ij} + \delta_{ij}P \tag{6}$$

where $\underline{\sigma}_{ij}$ is the effective stress tensor, σ_{ij} is the stress tensor, P is pressure and δ_{ij} is the Kronecker delta function. Rozhko [11] refers to this as the so-called seepage force and makes use of the Coulomb yielding criteria (CYS) written

$$CYS = \frac{\sigma_1 - \sigma_3}{2} + \sin\varphi_y \left(\frac{\sigma_1 + \sigma_3}{2} + P\right) + C_y \cos\varphi_y, \tag{7}$$

where CYS is the Coulomb Yielding stress, and σ_1 and σ_3 are the maximum and minimum principal stresses (positive in tension). The parameters ϕ_y and C_y are the friction angle and cohesion during dilatancy, and can be determined from geomechanical triaxial laboratory measurements. Application of CYS stems from laboratory measurements, where observed acoustic emissions during loading have been shown to correlate with the onset of dilatancy. CYS not only describes the onset of dilatancy and hence microseismicity, but also incorporates the Kaiser effect [17]; during unloading deformation is elastic with no additional fracturing and during reloading no additional fracturing and seismicity develops until overcoming the previous loading maximum.

The seismicity-triggering front (the CYS equivalent to equation 1) is given

$$\Delta CYS = \frac{1}{r} P_{c0} H(t - t_0) \Big[\eta f_D(R_0) + (\eta f_M(R_0) + f_F(R_0)) \sin \varphi_y \Big] + \dots + \frac{1}{r} P_{c1} H(t - t_1) \Big[\eta f_D(R_1) + (\eta f_M(R_1) + f_F(R_1)) \sin \varphi_y \Big]$$
(8)

where r is radial distance, η is the poroelastic stress coefficient, $f_F(R)$, $f_D(R)$ and $f_M(R)$ are non-dimensional functions given by Rozhko [14, equations 7, 10 and 11], $R=r/(4Dt)^{1/2}$, P_{c0} and P_{c1} are the pressure perturbations for time t_0 and t_1 , $R_0=r/[4D_0(t-t_0)]^{1/2}$ and $R_1=r/[4D_1(t-t_1)]^{1/2}$, and D is the pressure diffusivity constant. The seismicity-suppression front (the CYS equivalent to equation 4) is written

$$\Delta CYS_* = \frac{1}{r} P_{c0} H(t - t_0) \Big[\eta f_D(R_0) + (\eta f_M(R_0) + f_F(R_0)) \sin \varphi_y \Big] + \dots + \frac{1}{r} P_{c1} H(t - t_1) \Big[-\eta f_D(R_1) + (\eta f_M(R_1) + f_F(R_1)) \sin \varphi_y \Big]$$
(9)

By fitting the seismicity-triggering and seismicity-suppression fronts to the induced-seismicity, the formation diffusivity can be estimated. Apparent permeability can then be estimated from

$$\kappa = D\mu_f \phi(c_m + c_f), \tag{10}$$

where c_m is the pore volume compressibility.

3. Microseismic data

Verdon et al. [18] compared the microseismicity produced when first water and then CO_2 (in a supercritical state) were used as the injection fluids for hydraulic fracture of a tight gas reservoir, with the purpose of identifying any characteristic differences in event locations and/or magnitudes induced by the different fluids. A total of 9 injection stages were performed in a vertical well, with each stage at a slightly shallower depth than the previous stage. The first 7 stages used water, while the final 2 used supercritical CO_2 . No major lithologic differences have been identified between the stages. The fracture stimulations were monitored with a downhole array of 12 three-component geophones installed in a nearby vertical well. Verdon et al. [18] presented data from Stages 4 (water) and 8 (CO_2). For both fluids, microseismic event locations indicated the formation of fracture networks parallel to the maximum horizontal stress (Figure 1). Event magnitudes showed a weak correlation with injection pressure, while the influence of the differing fluids was found to be minimal.



Figure 1: Map views of event locations during hydraulic fracture stimulation for water (a) and CO_2 (b) fluid injection. The locations of the injection well and monitoring array are also marked. Error bars represent one-standard-deviation errors based on arrival time residuals and particle motion analysis. As such, they do not account for the additional errors introduced by velocity model discrepancies [e.g., 19, 20], and so should be considered a lower bound of the true location error.

4. Permeability estimates from pore-pressure and seepage force triggering

4.1. Pore pressure triggering

Water-gel injection: Figure 1a displays a map view of the recorded microseismicity during the water-gel fracture treatment. The microseismicity follows an approximately linear trend with an absolute correlation coefficient of 0.83 using simple linear regression. The length and width of the microseismic cloud is approximately 220 m and 70 m. In Figure 2, the microseismic events are plotted with respect to distance from injection well and injection time (i.e., *r-t* space). The vertical error bars represent the estimated location errors based on the residuals between the predicted and observed travel-times. Also shown are horizontal error bars that serve as a qualitative (and not quantitative) measure of event measurement confidence and so by no means reflect error in time. In other words, they attempt to present additional information about microseismic event quality to help further scrutinize the *r-t* plot. Three microseismic trigger (or forward front) *r-t* curves are shown for apparent diffusivities of $1.25 \text{ m}^2/\text{s}$, $0.60 \text{ m}^2/\text{s}$ and $0.40 \text{ m}^2/\text{s}$ using equation 1. These curves represent subjective end-member *r-t* curves for the forward front microseismicity. Also shown is an *r-t* curve for the back front for an apparent diffusivity of $0.10 \text{ m}^2/\text{s}$ using equation 4.



Figure 2: *r*-*t* pore pressure triggering plot for water-gel injection treatment. The circles represent the spatio-temporal location of each microseismic event, with the vertical error bar being the total estimated location error and the horizontal error bar representing a scaled event confidence term. The black, the grey and the light-grey curves are the triggering front r-t curves for diffusivity of 1.25, 0.60 and 0.40 m²/s, respectively. The blue curve is the back front r-t curve for diffusivity of 0.10 m²/s.

Supercritical CO_2 injection: Figure 1b displays a map view of the recorded microseismicity during the supercritical CO_2 fracture treatment. The microseismicity follows a more diffuse trend compared with the water-gel treatment with an absolute correlation coefficient of 0.65. The length and width of the microseismic cloud is approximately 120 m and 50 m. In Figure 3, the microseismic events are plotted in *r*-*t* space. Three forward front *r*-*t* curves are shown for apparent diffusivities of 1.20 m²/s, 0.80 m²/s and 0.30 m²/s and a back front r-t curve for an apparent diffusivity of 0.90 m²/s.



Figure 3: r-t pore pressure triggering plot for supercritical CO_2 injection treatment (refer to Figure 2 for details). The black, the grey and the light-grey curves are the triggering front r-t curves for diffusivity of 1.20, 0.80 and 0.30 m²/s, respectively. The blue curve is the back front r-t curve for diffusivity of 0.90 m²/s.

4.2. Seepage force triggering

Figures 4 and 5 show the same observed *r*-*t* data, but overlain with predictions of seismicity triggering and suppression fronts based on seepage force modelling. In these figures, we assume $\phi_y=30^\circ$ and $\eta=0.30$. For the water treatment (Figure 4), we use an average differential injection pressure of 14 MPa to define P_{c0}=14 MPa and P_{c1}=-14 MPa. The best fitting seismicity-triggering and seismicity-suppression fronts were obtained using the following values: perforation interval of 20 m, $\Delta CYS_h=0.1MPa$, $D_0=2.75x10^3$ m²/h and $D_1=3.85x10^3$ m²/h. For the supercritical CO₂ treatment (Figure 5), we use an average differential injection pressure of 15 MPa to define P_{c0}=15 MPa and P_{c1}= -15 MPa. The best fitting seismicity-triggering and seismicity-suppression fronts were obtained using the following values: perforation using the following values: $P_{c0}=15$ MPa and $P_{c1}=-15$ MPa. The best fitting seismicity-triggering and seismicity-suppression fronts were obtained using the following values: $P_{c0}=15$ MPa and $P_{c1}=-15$ MPa. The best fitting seismicity-triggering and seismicity-suppression fronts were obtained using the following values: $P_{c1}=-15$ MPa, $D_{c2}=1.5 \text{ MPa}$, D_{c2}



Figure 4: r-*t* seepage force predictions for the water treatment injection. The black curve is the seismicity-triggering front and the grey curve is the seismicity-suppression front.



Figure 5: r-*t* seepage force predictions for the supercritical CO_2 injection treatment. The black curve is the seismicity-triggering front and the grey curve is the seismicity-suppression front.

4.3. Permeability estimates

Pore pressure triggering: The average reservoir porosity is assumed to be 10% (an upper end for tight sand reservoirs). The reservoir fluid viscosity and compressibility are estimated to be 1.00×10^{-3} Pa•s and 1.45×10^{-11} Pa⁻¹, respectively, based on typical values for oil given by Dake [21]. For the water-gel treatment, the average injection rate is 0.09 m³/s, the pressure difference 15 MPa and fracture height 70 m. For the supercritical CO₂ treatment, the average injection rate is 0.08 m³/s, the pressure difference 15 MPa and fracture height 100 m. We assume a fracture width of 0.01 m for both fracture stimulations. Table 1 compiles the results for the estimated fluid-loss coefficient (equation 3), fracture half-length (equation 2) and reservoir permeability (equation 5).

Fracture treatment fluid	Diffusivity (m²/s)	C ₁ (m/s ^{1/2})	L(t) (m)	к (mD)
Water-gel	1.25	5.91x10-5	108.80	34.07
	0.60	8.53x10-5	93.19	70.99
	0.40	1.04x10-4	84.33	106.48
CO ₂	1.20	3.68x10-5	79.59	14.54
	0.80	4.73x10-5	73.41	21.81
	0.30	7.72x10-5	57.91	58.15

Table 1: Estimated fluid loss, fracture half-length and formation permeability based on pore pressure triggering.

Seepage force triggering: Assuming the pore volume compressibility is negligible with respect to the reservoir fluid compressibility (i.e. $C_m \ll C_f$), the permeability estimates based on equation 10 are summarized in table 2. The assumption of negligible pore volume compressibility suggests our estimates of formation permeability are conservative (i.e. an underestimate).

 Table 2: Estimated formation permeability based on seepage force triggering.

Fracture treatment fluid	Diffusivity (10 ³ m²/h)	к (mD)
Water-gel	D ₀ =2.75	67
	D ₁ =3.85	94
CO ₂	D ₀ =2.15	53
	D ₁ =3.75	92

Estimates of apparent permeability from both approaches are within the same order of magnitude between 10 mD and 100 mD. Note we use the term 'apparent' permeability for two reasons. First, the permeability estimates are typically higher than the true formation permeability through enhancement via hydraulic stimulation. Second, the apparent permeability is indirectly calculated from microseismic data and depends on assumed values of Biot's constant, Poisson's ratio and the internal angle of friction.

For the water-gel fluid injection, the formation permeability estimates based on pore pressure triggering range between 34 mD to 106 mD, whereas those based on seepage force triggering range between 67 mD and 94 mD. For seepage force triggering, there are two estimates for formation permeability. This is because two values of diffusivity were needed to fit the microseismic data: D_0 for the diffusivity during hydraulic stimulation and D_1 for diffusivity due to negative pore pressure perturbation in fractured rock. Thus, the estimate of κ =67 mD during hydraulic stimulation is more representative of formation permeability during fluid injection, whereas the estimate of κ =94 mD is more representative of the formation after fracture damage. For the supercritical CO₂ fluid injection, the formation permeability estimates based on seepage force triggering are lower and range between 14 mD and 58 mD, whereas those based on seepage force triggering are more consistent between the water-gel and supercritical CO₂ injections. The permeability estimates for the formation are nearly equal as would be expected for similar geological formations. As per the

conclusions of Verdon et al. [18], we see little evidence for a different seismic response when CO_2 rather than water is the injected fluid.

As well as examining event locations and magnitudes, Verdon et al. [18] used shear-wave splitting to image the induced fracture networks. Although not robustly constrained, inversions based on the SWS measurements appeared to show that the fracture network created during water injection was slightly more intense. This may account for what differences there are in formation permeability between supercritical CO_2 and water-gel injection, as estimated by the pore-pressure method. However, the permeability estimates for the fractured formation (i.e. based on diffusivity estimates D_1) would suggest that both fluids generate similar fracture density. One possible explanation for the difference might be in terms of the size of fractures generated. The water-gel treatment may generate large fractures that are effectively constant in dimension (e.g. displaying a Gaussian distribution), whereas the supercritical CO_2 treatment may generate fewer large fractures yet many smaller fractures (i.e. skewed distribution). Based on the geometry of the microseismic monitoring array, the detectability limitations would be biased towards larger fractures. This would certainly explain the lower number of events recorded from the supercritical CO_2 injection yet similar fractured formation permeability.

5. Discussion and conclusions

We examined two approaches of estimating formation permeability using microseismic data; the pore pressure triggering and the seepage force triggering mechanisms. Based on microseismic data from a hydraulic fracture experiment injecting water and supercritical CO_2 during different stages, we compared permeability estimates between the two approaches. The two hydraulic stimulation treatments were performed separately on two formation intervals having similar geological, geomechanical and *in situ* stress conditions, and only only differed in terms of the injection fluid used. Both approaches (pore pressure triggering and the seepage force triggering) provided estimates of permeability within the same order of magnitude. However, the seepage force mechanism (i.e. effective stress perturbation) provided more consistent estimates of permeability between the two different injection fluids.

Urbancic et al. [22] monitored significant microseismicity during a field test injecting 10,000 tons CO_2 over a period of one month. They observed that microseismicity can be used to identify the position of the CO_2 plume and, although microseismicity was significant, there was no evidence of reduced cap rock integrity. Verdon et al. [23] demonstrated the added benefit of microseismic monitoring for the geological storage of CO_2 during injection, where, for example, it was shown that microseismic activity is a natural consequence of fluid production and injection, and does not necessarily imply leakage from the storage formation. Our results show that permeability estimates based on the seepage force triggering mechanism technique using microseismic monitoring has strong potential to constrain formation permeability limitations for large scale CO_2 injection.

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REFERENCES

- [1] Jones, J.R. and L.K. Britt (1997) Design and appraisal of hydraulic fractures, SPE Publications, Texas.
- [2] Teanby, N., J.-M. Kendall, R. H. Jones, and O. I. Barkved (2004) Stress-induced temporal variations in seismic anisotropy observed in microseismic data, *Geophysical Journal International*, **156**, 459-466.

- [3] Verdon, J.P., J-M. Kendall and A. Wüstefeld (2009) Imaging fractures and sedimentary fabrics using shear wave splitting measurements made on passive seismic data, *Geophysical Journal International*, **179**(2), 1245-1254.
- [4] Wüstefeld, A., O. Al-Harrasi, J.P. Verdon, J. Wookey and J-M. Kendall (2010) Strategies for a fully automated passive microseismic anisotropy analysis, *Geophysical Prospecting*, 58(5), 755-773.
- [5] McGarr, A. (1971) Violent deformation of rock near deep-level tabular excavations-seismic events, *Bulletin of the Seismological Society of America*, **61**, 1453-1466.
- [6] Trifu, C.-I., D. A. Angus, and V. Shumila (2000) A fast evaluation of the seismic moment tensor for induced seismicity, *Bulletin of the Seismological Society of America*, **90**(6), 1521-1527.
- [7] Al-Anboori, A., J.-M. Kendall, D. Raymer, and R. Jones (2006) Spatial variations in microseismic focal mechanisms, Yibal Field, Oman, *EAGE Proceedings of the 68th Mtg.*
- [8] Shapiro, S.A. (2008) Microseismicity: A tool for reservoir characterization, EAGE Publications, Netherlands.
- [9] Grechka, V., P. Mazumdar and S.A. Shapiro (2010) Predicting permeability and gas production of hydraulically fractured tight sands from microseismic data, *Geophysics*, **75**(1), B1-B10.
- [10] Shapiro, S.A., J. Kummerow, C. Dinske, G. Asch, E. Rothert, J. Erzinger, H-J. Kumpel and R. Kind (2006) Fluid induced seismicity guided by a continental fault: Injection experiment of 2004/2005 at the German Deep Drilling Site (KTB), *Geophysical Research Letters*, 33, L01309.
- [11] Rozhko, A.Y. (2010) Role of seepage forces on seismicity triggering, *Journal of Geophysical Research*, 115, B11314.
- Biot, M. (1962) Mechanics of deformation and acoustic propagation in porous media, *Journal of Applied Physics*, 33, 1482-1498.
- [13] Crampin, S. and S. Peacock (2005) A review of shear-wave splitting in compliant crack-critical anisotropic Earth, *Wave Motion*, **41**, 59-77.
- [14] Narr, W., D.W. Schechter and L.B. Thompson (1997) *Naturally fractured reservoir characterization*, SPE Publications, Texas.
- [15] Zoback, M.D. and H.-P. Harjes (1997) Injection-induced earthquakes and crustal stress at 9 km depth at the KTB deep drilling site, Germany, *Journal of Geophysical Research*, **102**, 18,477-18,491.
- [16] Terzaghi, K. 1943. Theoretical Soil Mechanics. John Wiley and Sons, New York.
- [17] Kaiser, E.J. (1959) A study of acoustic phenomena in tensile test, Doctoral thesis, Tech. Hochsch. Munchen, Munich, Germany.
- [18] Verdon, J.P., J-M. Kendall and S.C. Maxwell (2010) A comparison of passive seismic monitoring of fracture stimulation from water and CO₂ injection, *Geophysics*, **75**(3), MA1-MA7.
- [19] Eisner, L., P.M. Duncan, W.H. Heigl and W.R. Keller (2009) Uncertainties in passive seismic monitoring, *The Leading Edge*, 28(6), 648-655.
- [20] Usher, P.J., D.A. Angus and J.P. Verdon (2012) Influence of a velocity model and source frequency on microseismic waveforms: some implications for microseismic locations, *Geophysical Prospecting*, pp12.
- [21] Dake, L.P. (2001) The practice of reservoir engineering (revised edition), Elsevier, London.
- [22] Urbancic, T.I., J. Daugherty, S. Bowman and M. Prince (2009) Microseismic monitoring of a carbon sequestration field test, 2nd EAGE Passive Seismic Workshop, Limassol, Cyprus.
- [23] Verdon, J.P., J-M. Kendall, D.J. White and D.A. Angus (2011) Linking microseismic event observations with geomechanical models to minimize the risks of storing CO₂ in geological formations, *Earth and Planetary Science Letters*, **305**, 143-152.