Application of Multiple-point Statistics and Image Logging Data in Modelling Coal Cleats Distribution

Fengde Zhou ^{ab}; Saswata Mukherjee ^{ab}; Steven Tyson ^{ab}, Joan Esterle ^{ab} ^a School of Earth Sciences , The University of Queensland ^b Centre for Coal Seam Gas, The University of Queensland

CCSG Research Review, 9 December 2016 PROJECT: UNDERSTANDING FAULTS AND FRACTURES IN THE SURAT BASIN

ABSTRACT

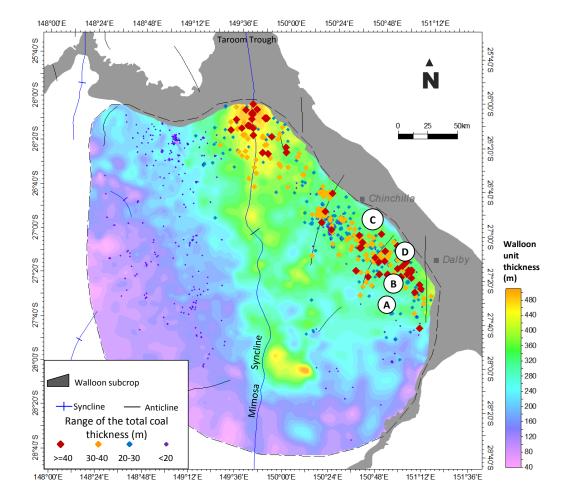
Coal has a complex porosity system that ranges from molecular to macroscopic pore sizes, the latter of which are commonly known as fractures. The orthogonal coal fractures or cleats contribute directly to coal permeability under variable stress conditions, and are the main conduits for gas and water flow in coal. In modelling and reservoir simulation, they are described by distributions referred to as Discrete Fracture Network models (DFN), and used to analyse near-wellbore coal properties, coal seam gas production, and well testing performance. This study presents a workflow to build a grid based DFN near the wellbore using the multi-point facies simulation (MPS) method. Core from three wells were measured for face and butt cleats spacing. An image log from one well was interpreted for cleats azimuth which was used in MPS as hard data. Firstly, we coded a program to generate cleats distribution based on the cleat spacing observed in core. Then the generated cleats were imported into Petrel to be used as the training image. We then built a 2D grid with cells number of 1000×1000 and grid size of 1 mm in x- and y- direction which is used in MPS modelling. Finally, ten realisations of cleats distribution near the wellbore were generated. Results show that this presented workflow is efficient in modelling cleats distribution by integrating image logs interpretations and core observed cleats.

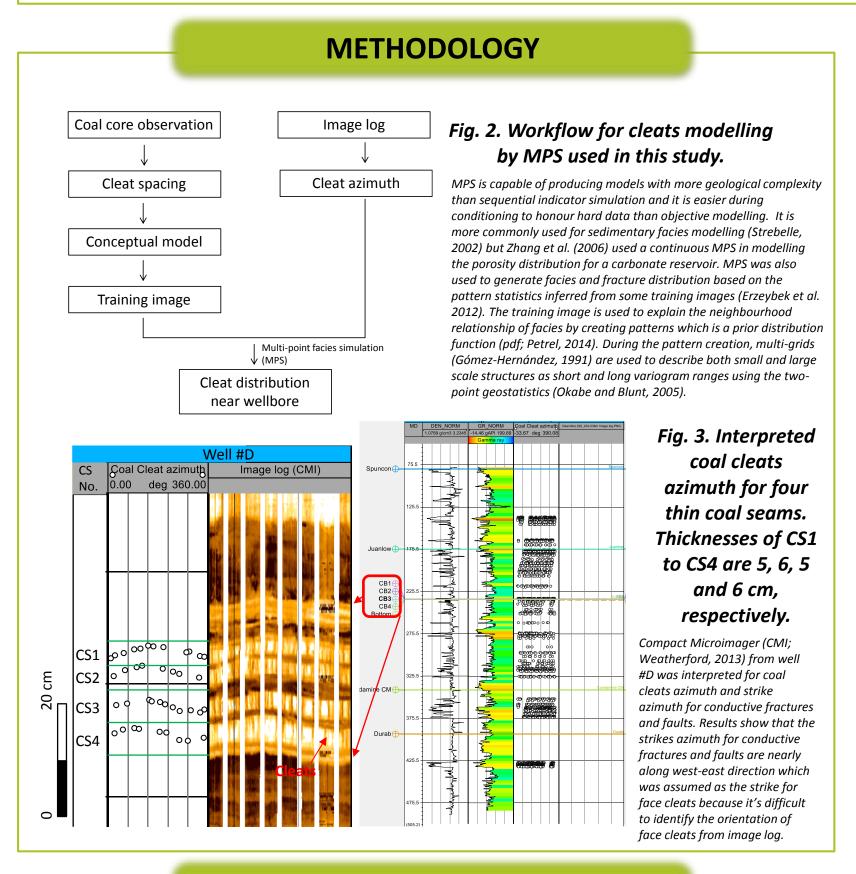
INTRODUCTION

Coal has a complex porous system ranging from micro-, meso- and macro- pores, to visible fractures (butt and face cleats, joints and fractures; Laubach et al. 1998). Butt cleats and orthogonal face cleats, which contribute to the coal permeability, are the main conduits for gas and water flow in coal. After stress magnitude, the permeability is strongly dependent on the cleats spacing, geometry, orientation to stress, aperture size, connectivity, degree of mineralisation and topology (Close and Mavor, 1991; Laubach et al. 1998; Zhou and Yao, 2014; Jing et al. 2016). Discrete fracture models (DFM) or discrete fracture network (DFN) models are commonly used to analyse coal permeability and fluid flow (Gong et al. 2014). Zhou and Yao (2014) studied the effect of cleats aperture and spacing on coal permeability by building 2D cleat models. Gao et al. (2014) analysed the strength and deformability of coal based on $(2m)^3$ DFN models with bedding planes, face cleats and butt cleats. Jing et al. (2016) built DFNs for coal cleats based on micro-computed tomography imaging (μ CT) and then using a Monte Carlo method to assign to the remaining reservoir. This latter point is the problem, as the borehole only confidently intersects the small scale cleat network, and the larger scale joints and faults are under represented.

This study presents a workflow to build a grid based DFN near wellbore using multi-point facies simulation (MPS, a pixelbased sequential simulation algorithm; Strebelle, 2002; Strebelle and Journel, 2001; Rezaee et al. 2015) method and image log data which has a high resolution to 5 mm but much smaller fractures can be imaged if there is a sufficient electrical contrast with the background (Weatherford, 2013).

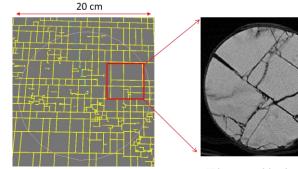
Core from three wells, A, B and C shown in Figure 1 were measured for face and butt cleats spacing (55 points from well A, 138 from well B and 69 from well C). Image logging data from well D was used to interpret cleats, faults and fractures.





DISCUSSION AND FUTURE WORKS

This study presents a workflow to build cleats distribution near a wellbore by integrating the image log interpreted cleats, cleats spacing data from coal core observation and MPS method. The built cleats and matrix model can help to analyse near-wellbore coal properties, coal seam gas production, and well testing performances. More training images from μ CT scanned cleats (Figure 7) will be used in the future. Well testing (Figure 8) or production data will be used to optimise the cleats distribution model.



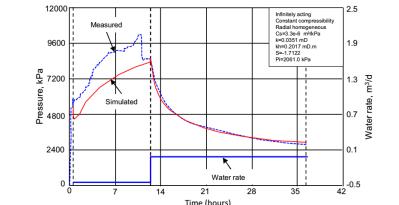




Fig. 4. Histogram of coal cleat spacing from boreholes A, B, and C which were shown in Fig. 1.

Fig. 5. Strikes for conductive fracture and fault which was assumed as the azimuth of face cleats

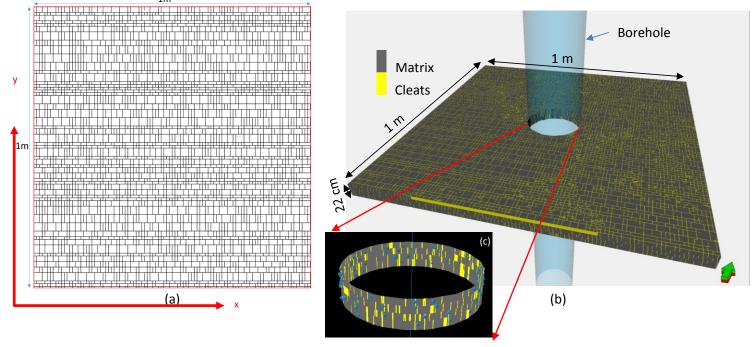


Fig. 6. A training image (a) and one realisation of coal cleats distribution near wellbore (b), and Ring showing cleat distribution and image interpreted cleat points (c).

Training image generation:

We coded a program in MATLAB to generate cleats distribution based on the cleat spacing shown in Figure 4. Then the generated cleats were imported into Petrel as FAB data because Petrel cannot generate butt cleats terminated by face cleats (Petrel, 2014). We then built a 1000×1000 2D grid with grid size of 1 mm in x- and y- direction to store cleats as shown in Figure 6a for one realisation because MPS is a pixel-based sequential simulation algorithm.

Cleats modelling by using MPS:

A 3D grid with grid cells of 1000×1000×4 in x-, y- and z-direction was built firstly. The grid size is 1 mm in x- and y- direction and equals coal seam thickness in vertical. The training image (Figure 6a) was used to create pattern by using an ellipsoid search mask (Petrel, 2014) with radius of 50×50×1 in x-, y- and z-direction. We then converted the log interpreted cleats (Figure 3) to a probability model. Finally, MPS was used to generate cleats distribution with distribution pattern from Figure 6a, conditioning data from Figure 3 and face cleat azimuth from Figure 5. Figure 6b shows the modelled cleats and matrix within 50 cm of wellbore for one realisation.



Close, J.C. and Mavor, M.J. [1991]. Influence of coal composition and rank on fracture development in Fruitland coal gas reservoirs of San Juan Basin. In: Schwochow, S. Murray, D.K. Fahy, M.F. (Eds.), Coalbed Methane of Western North America. Rk. Mt. Assoc. Geol., Field Conf. 109-121. Erzeybek, S. Srinivasan, S. and Janson, X. [2012] Multiple-point statistics in a non-gridded domain: Application to karst/fracture network modeling. Quantitative Geology and Geostatistics, 17, 221-237.



Fig. 7. Cleats distribution compared with CT extracted cleats. Note that the position of this CT image is not exact. Fig. 8. Using well testing data to optimise cleats model (from Zhou and Yao, 2014)

Acknowledgements: We thank all the coal seam gas companies that provide data to the Queensland Department of Natural Resources & Mines (DNRM). This research has been supported by the UQ Centre for Coal Seam Gas and its industry members (APLNG, Arrow Energy, QGC and Santos). We thank Schlumberger for providing the license of PetreITM; and Paradigm for providing the license for GeologTM.

Gao, F. Stead, D. and Kang, H. [2014] Numerical investigation of the scale effect and anisotropy in the strength and deformability of coal. International Journal of Coal Geology, 136, 25-37.

Gómez-Hernández, J. [1991]. A stochastic approach to the simulation of block fields conductivity upon data measured at a smaller scale. PhD thesis, Stanford University, Stanford.

Gong, B. Zhang, Y. Fan, Y. and Qin, G. [2014] A novel approach to model enhanced coal bed methane recovery with discrete fracture characterizations in a geochemical simulator. Journal of Petroleum Science and Engineering, 124, 198-208.

Jing, Y. Armstrong, R.T. Ramandi, H.L. and Mostaghimi, P. [2016] Coal cleat reconstruction using micro-computed tomography imaging. Fuel, 181, 286-299. Laubach, S.E. Marrett, R.A. Olson, J.E. and Scott, A.R. [1998] Characteristics and origins of coal cleat: a review. International Journal of Coal Geology, 35, 175-207. Okabe, H. and Blunt, M.J. [2005] Pore space reconstruction using multiple-point statistics. Journal of Petroleum Science and Engineering, 46, 121-137. Petrel, 2014. Petrel Help Center. Schlumberger.

Rezaee, H. Marcotte, D. Tahmasebi, P. and Saucier, A. [2015] Multiple-point geostatistical simulation using enriched pattern databases. Stochastic Environmental Research and Risk Assessment, 29, 893-913.

Strebelle, S.B. [2002] Conditional simulation of complex geological structures using multiple-point statistics. Mathematical Geology, 34, 1-21. Strebelle, S.B. and Journel, A.G. [2001]. Reservoir modeling using multiple-point statistics. SPE paper 71324 presented at the SPE Annual Technical Conference and Exhibition held in New Orleans, Louisiana, 1-11.

Weatherford [2013] Compact Microimager. http://www.weatherford.com/doc/wft268562.

Zhang, T. Bombarde, S. Strebelle, S.B. and Oatney, E. [2006] 3D Porosity Modeling of a Carbonate Reservoir Using Continuous Multiple-Point Statistics Simulation. SPE Journal, 11(3), 375-379.

Zhou, F. and Yao, G. [2014] Sensitivity analysis in permeability estimation using logging and injection-falloff test data for an anthracite coalbed methane reservoir in southeast Qinshui Basin, China. International Journal of Coal Geology, 131, 41-51.