

# Unconventional Reservoir Characterization: Permeability Prediction in Shale Gas Reservoirs from Well-Logs Data Using Neural Network. Application to the Barnett Shale (USA)

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#### Abstract

Here, we suggest the use of the artificial neural network for permeability prediction in horizontally drilled well in unconventional shale gas reservoirs. Prediction of Permeability in shale gas reservoirs is a complicated task that requires new models where Darcy's fluid flow model is not suitable. The proposed idea is based on the training of neural network machine using the set of well-logs data as an input and the measured permeability, from Javadbour model, as an output. In this case, the multilayer perceptron neural network machine was used with Levenberg Marquardt algorithm. Application to two horizontal wells drilled in the Barnett shale formation exhibits the power of the neural network model to enhance unconventional reservoir characterization.

#### Keywords

Artificial neural network • Permeability • Shale gas • Unconventional reservoir • Well-logs

### 1 Introduction

Currently, unconventional reservoir characterization is an important task and challenge in the petroleum domain. It requires advanced technologies for their evaluation and extraction (Zee Ma and Holditch 2016; Aliouane and Ouad-feul 2013). Shale gas unconventional reservoirs and their

permeability are still poorly understood where unconventional gas is found in reservoirs with relatively low permeabilities (less than 1 mD) and hence cannot be extracted via conventional methods. For that, we suggest the use of artificial intelligence such as neural network (ANN) for permeability prediction. Since, the ANN's techniques have been widely used to estimate conventional permeability (Aliouane et al. 2018). These techniques need just to be trained by training data. In our study, the well-logs of two horizontal wells, H1 and H2 in the Barnet shale from the Worth Basin, were used as input of neural network MLP model both for training and generalization. The obtained results will be compared to those obtained by Javadbour's relationship.

## 2 Data Description and Permeability Estimation Using Javadbour's Model

The raw well-logs data of two horizontal wells drilled in the Barnett shale located in the Ft Worth basin are exploited: total natural gamma (GR), bulk density (RHOB), neutron porosity (NPOR), transit time of P wave (DTCO), and transit time of S wave (DTSM). Figure 1 shows the raw well-logs data in tracks 2, 3, 4, 5, and 6. The recordings show that this reservoir is radioactive and tight by its low values of DTCO and DTSM.

Concerning the permeability computation, many authors have proposed many empirical relationships considering that the permeability in shale gas reservoirs is thought to be analogous to very tight sandstones (Moore et al. 2016). Such reservoirs consist of extremely tight formations through which transport of gas occurs by different mechanisms depending on the flow and porous formation conditions (Javadpour 2009; Civan et al. 2011; Haghshenas et al. 2013). Jenkins (2015) has shown the difficulty to estimate this parameter, mainly, in anisotropic permeability due to reservoir pressure change where it requires the confining stress method. In our study, permeability has been estimated using

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**Fig. 1** Recorded well-logs data in the horizontal well H01 and the estimated permeability using the Javadbour model. Track 1: Depth (ft), track 2: GR (API), track 3: RHOB (g/cc), track 04: NPOR (%), track

the Javadbour equation (Javadpour 2009) in the Barnet shale of the two horizontal wells. The results of one well (H01) are shown in Fig. 1 track 7.

### 3 Permeability Prediction by Multilayer Perceptron Neural Network

The multilayer perceptron neural network machine (Aminian and Ameri 2005) with Levenberg Marquardt training algorithm (Aliouane et al. 2014) has been implemented for the prediction of the permeability in shale gas reservoirs. The 05 raw well-logs data, cited below, are used as an input in the input layer with 05 neurons, and the estimated permeability from Javadbour's equation is used as an output to train a neural network MLP for the output layer with one neuron (K\_Predicted). The neural machine is trained in a supervised mode, and weights of connections are calculated during the training stage. The raw well-logs data are then injected in this machine, and an output is calculated using the weight of connections (Fig. 2).

#### 4 Results Analysis and Conclusion

We can observe that the proposed MLP machine is able to provide very good results that are very close to the reality (Fig. 2). It is clear that the implanted machine is able to provide very good and realistic results.

05: DTCO (ms/ft), track 06: DTSM (ms/ft), track 07: estimated permeability using Javadpour model



**Fig. 2** Estimated permeability using the MLP with Levenberg Marquardt training algorithm (red color) compared with the estimated the Javadbour model (blue color) for the generalized well 01H

The implantation of this neural network machine enabled us to predict the permeability without the need to use Javadpour's relationship or another mathematical relationship that requires laboratory measurements of many rock parameters which are expensive and require financial support. We suggest testing the multilayer perceptron neural network machine to data of other horizontal wells drilled in the Barnett shale and that are close to the pilot well 01H to check its efficiency and generalize to weights of connection.

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