

Machine Learning in Geophysics/Geology: Some Examples

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SBGf/SEG
WORKSHOP
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RIO DE JANEIRO, BRAZIL
MACHINE LEARNING

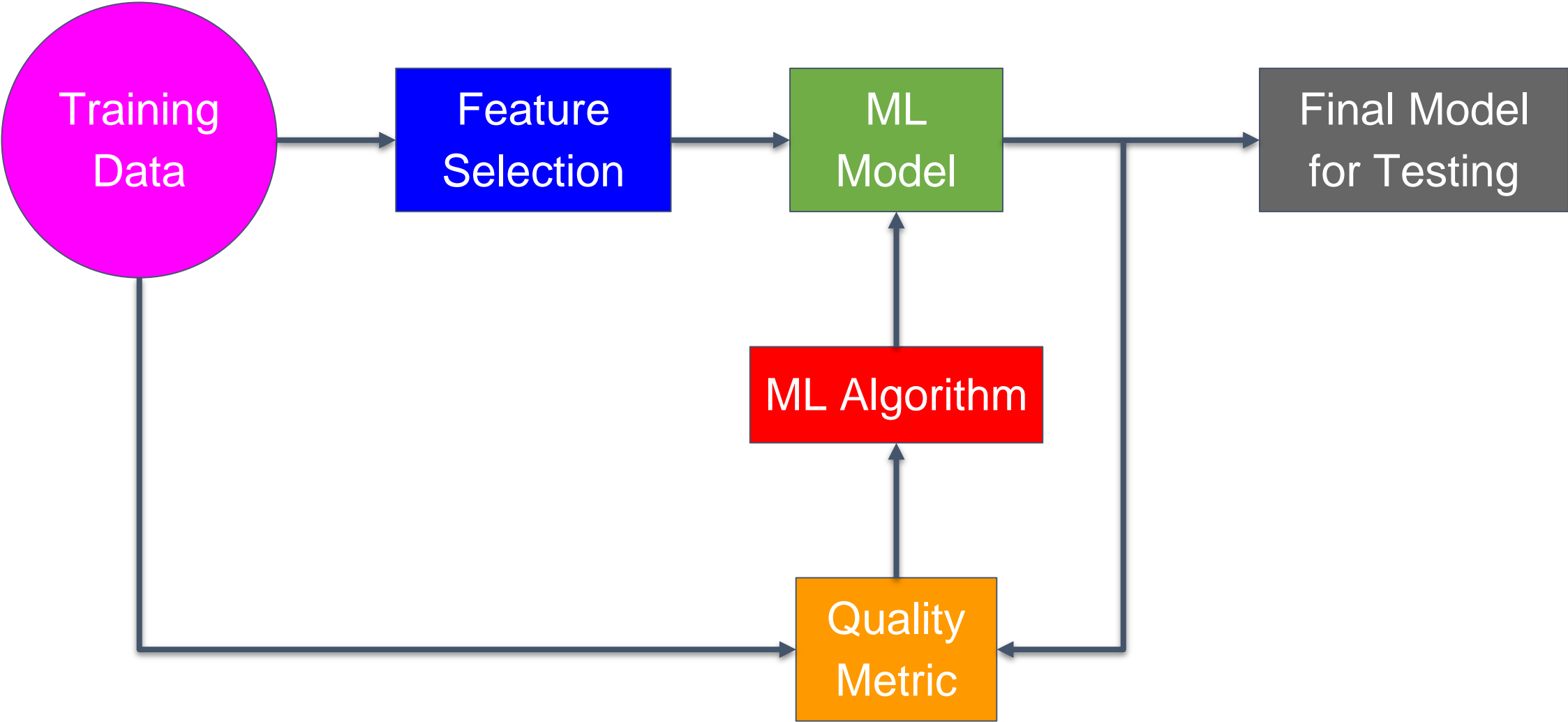
The banner features a dark blue background with a large green arrow pointing right. The text is arranged vertically and horizontally. A green bar at the bottom contains the text 'MACHINE LEARNING'. To the right is a wireframe globe graphic.



‘Machine learning is a field of computer science that gives computer systems the ability to "learn" (i.e. progressively improve performance on a specific task) with data, without being explicitly programmed.’

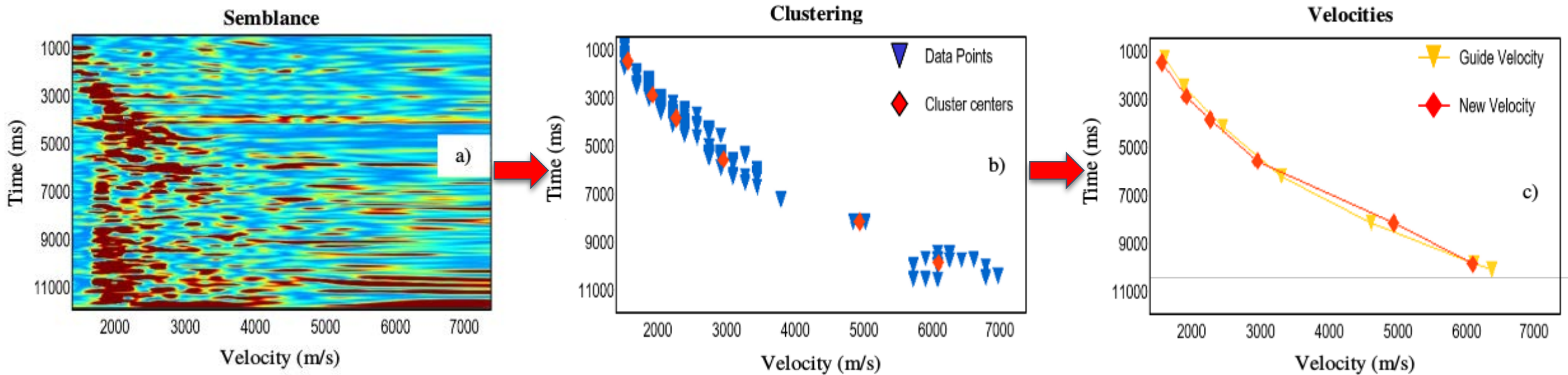
Wikipedia

Supervised Learning



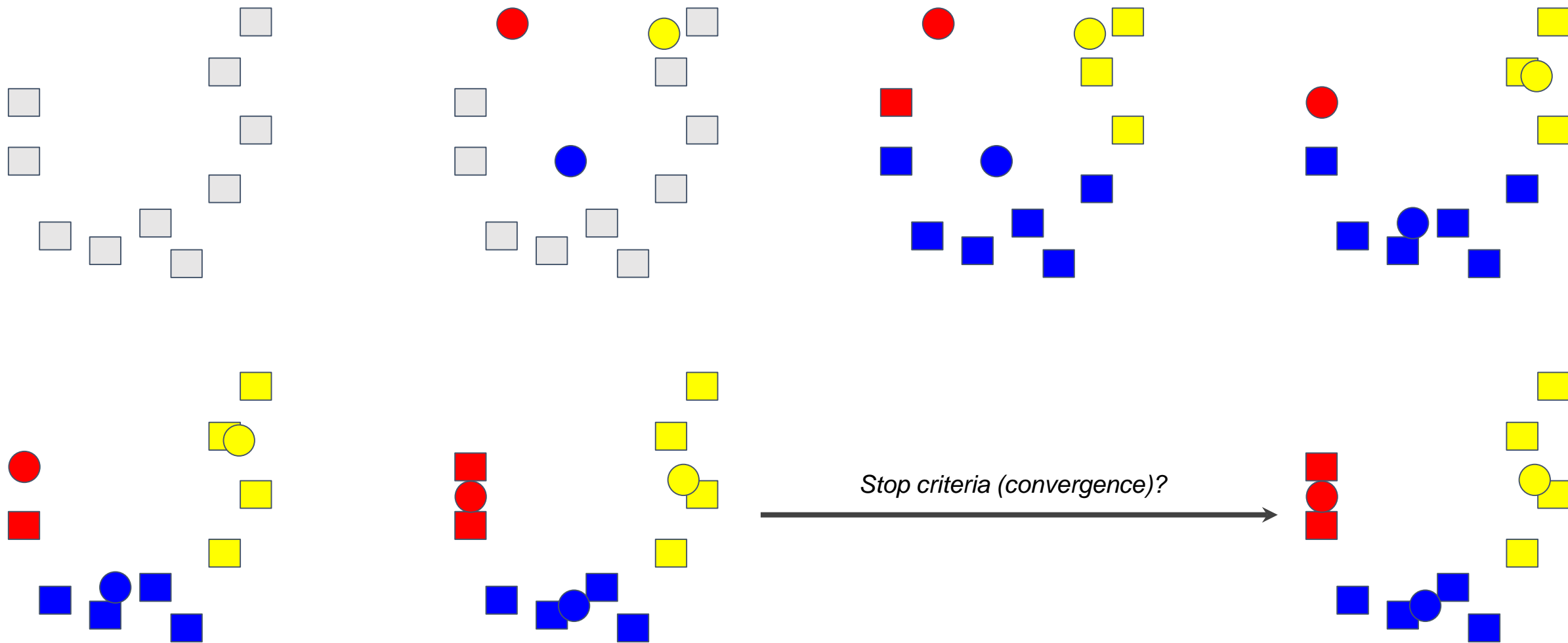
- Unlabeled data
- No evaluation of the accuracy
- Approaches to unsupervised learning include:
 - Clustering
 - Anomaly detection
 - Neural Networks
 - And a few others...

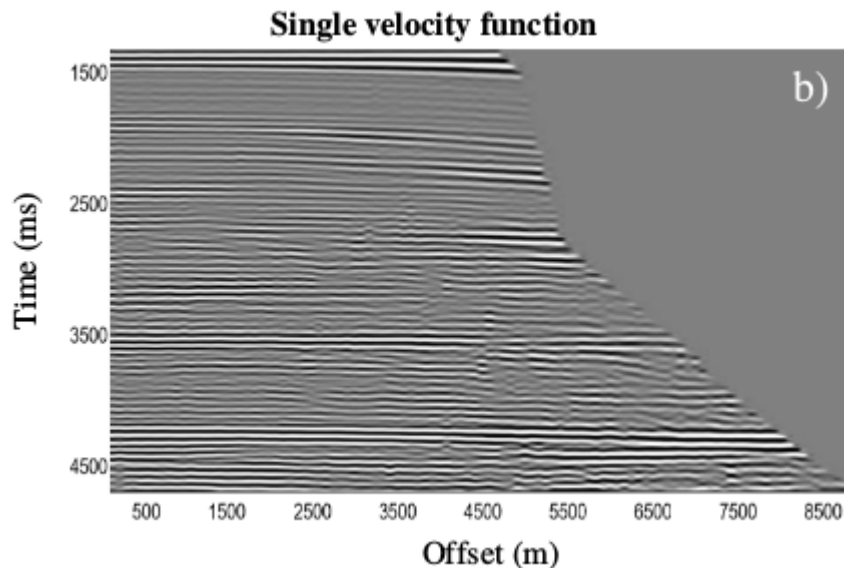
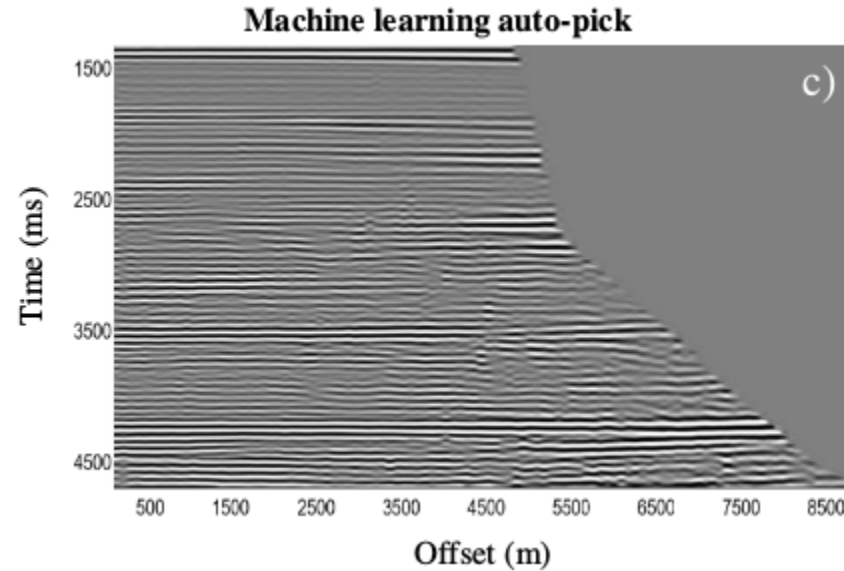
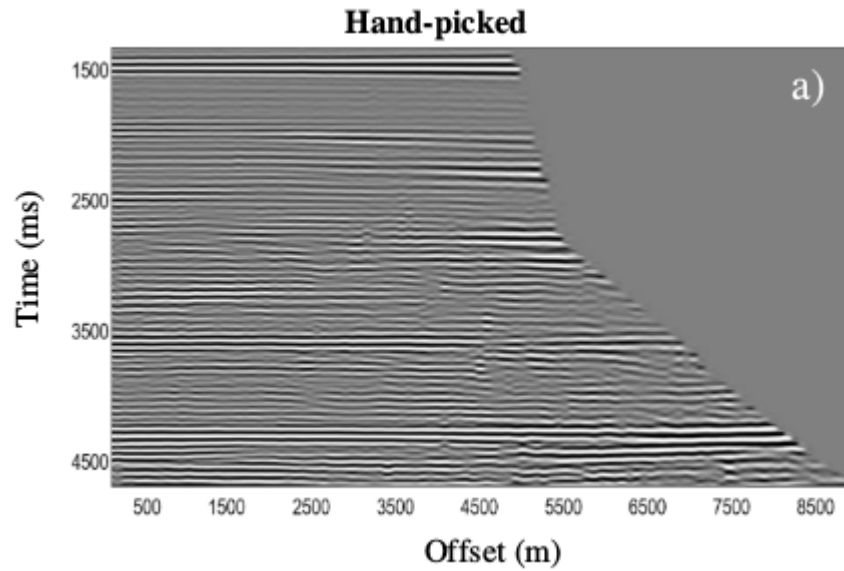
Smith, K. J., 2017, **Machine learning assisted auto-picking**,
SEG International Exposition and 87th Annual Meeting, 5686-
5690



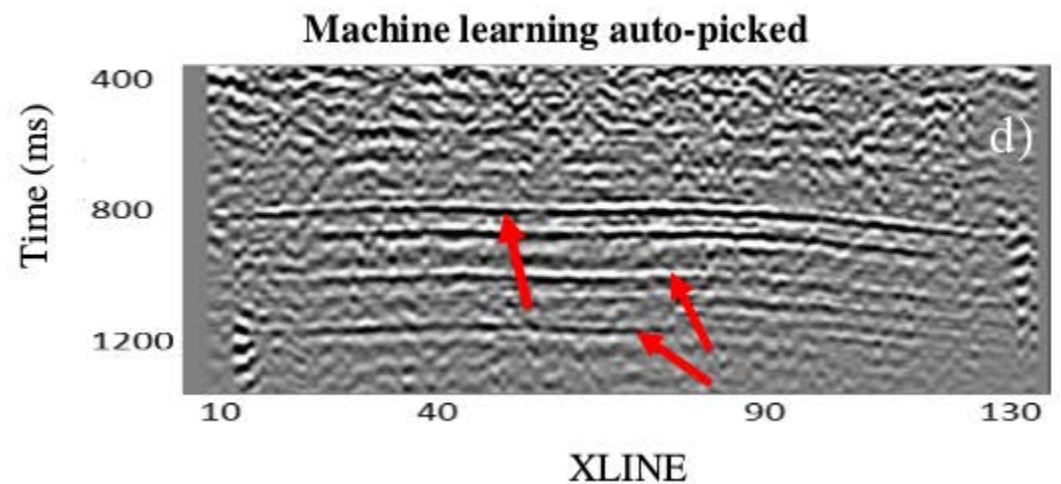
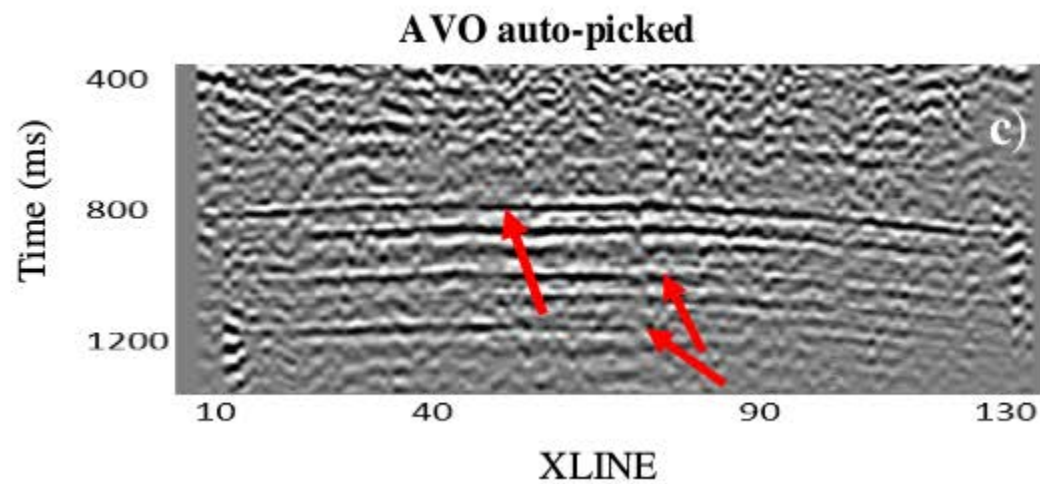
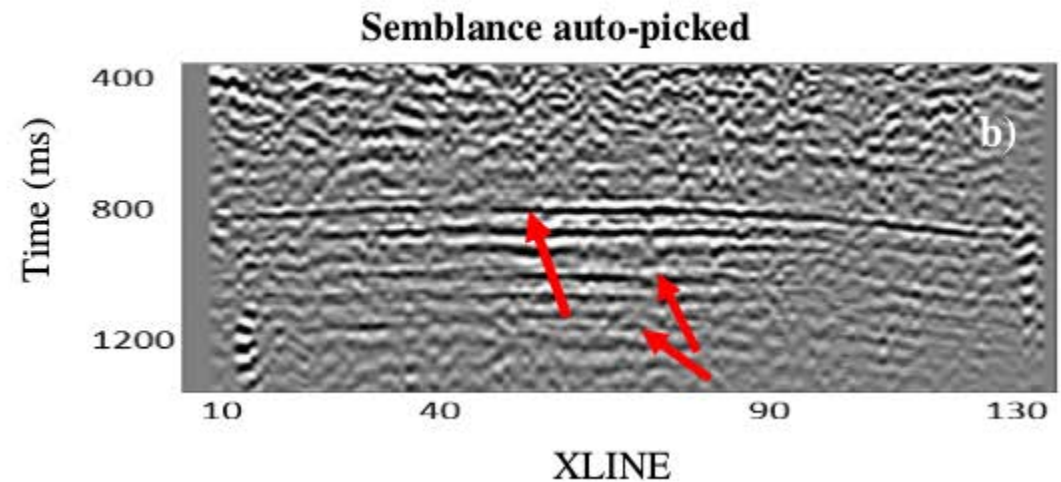
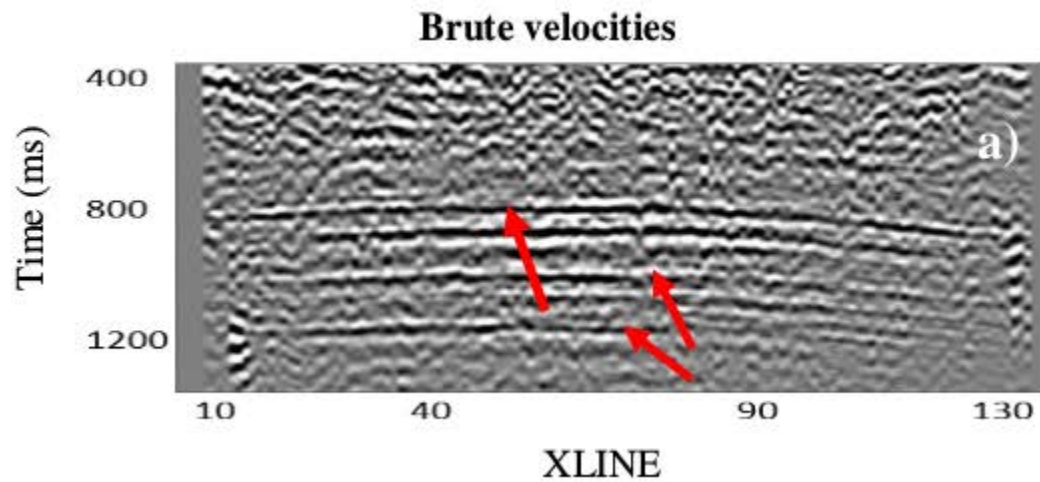
- Guide velocity (one CMP manually picked)
- Semblance
- Threshold (red or blue)
- Filter using others attributes (not clear)
- Group picks into “clusters”
- Determine clusters centers as the picking velocity
- Include new estimation in the guide velocity

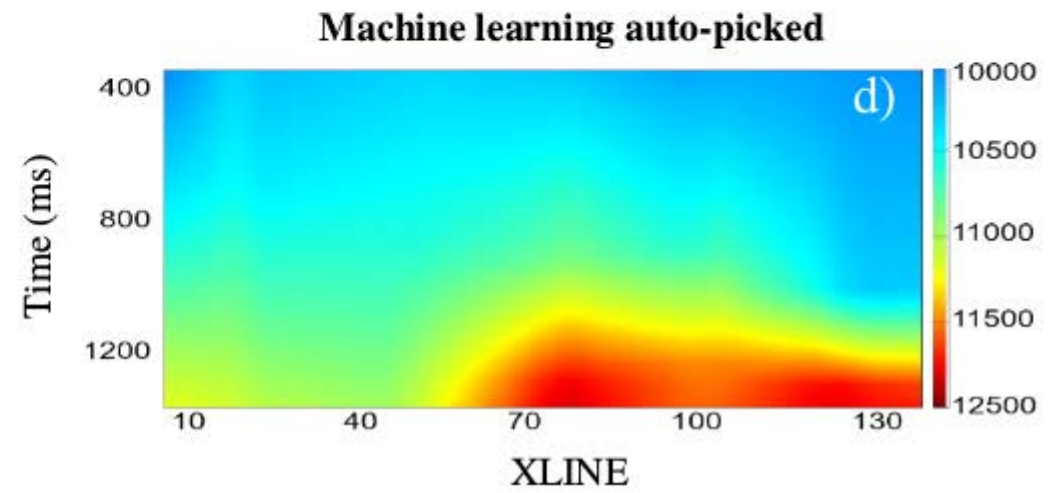
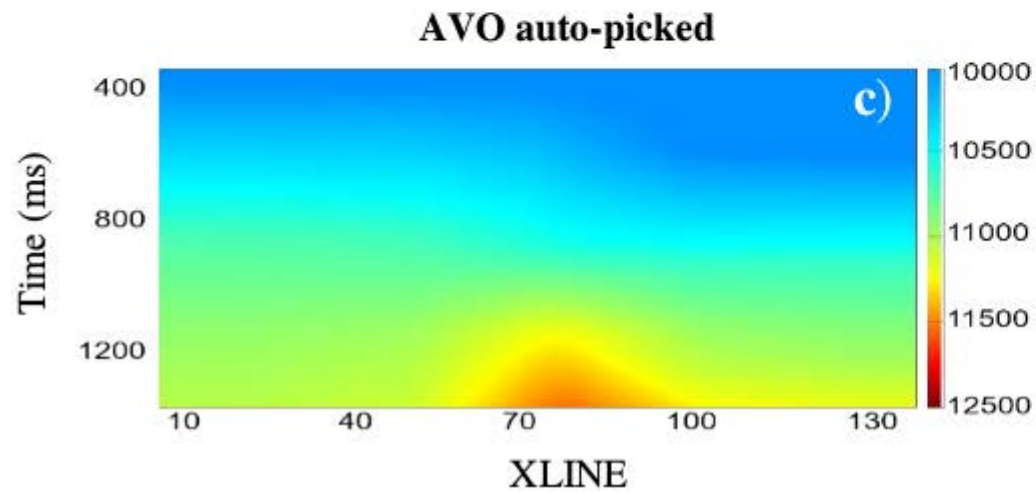
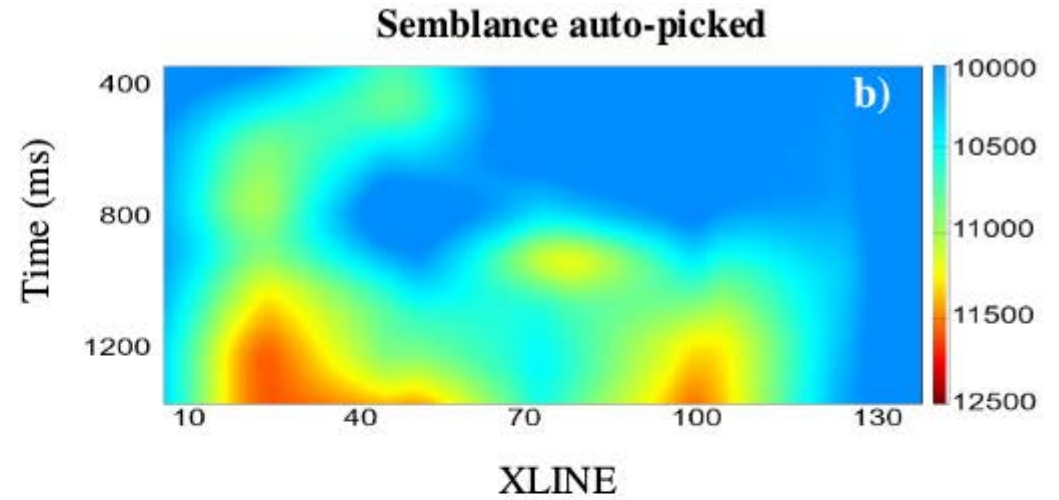
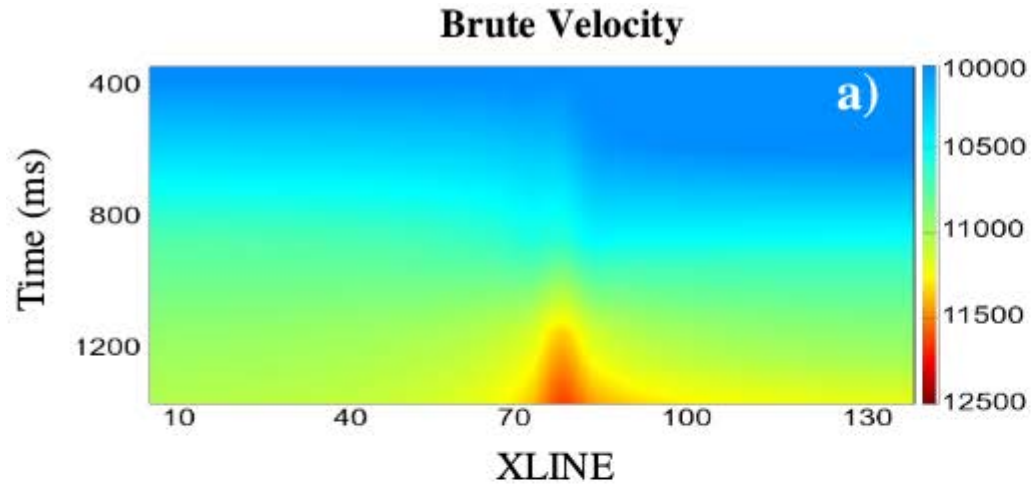
Clustering > *k-means* (or *k*-nearest neighbours)?





- Marine data
- CMP gather - NMO correction
- *“The gather from the machine-picked velocity is virtually indistinguishable from the gather with the hand-picked velocity”*





Paolo, B., Lipari, V., Tubaro, S., 2017, **A machine learning approach to facies classification using well logs**, SEG *International Exposition and 87th Annual Meeting*, 2137-2142

One of the proposed solutions of a Machine Learning Contest in 2016:

<https://github.com/seg/2016-ml-contest>

It was a contest to classify facies using the given well logs. The author's solution can be downloaded from:

<https://bitbucket.org/polimi-ispl/>

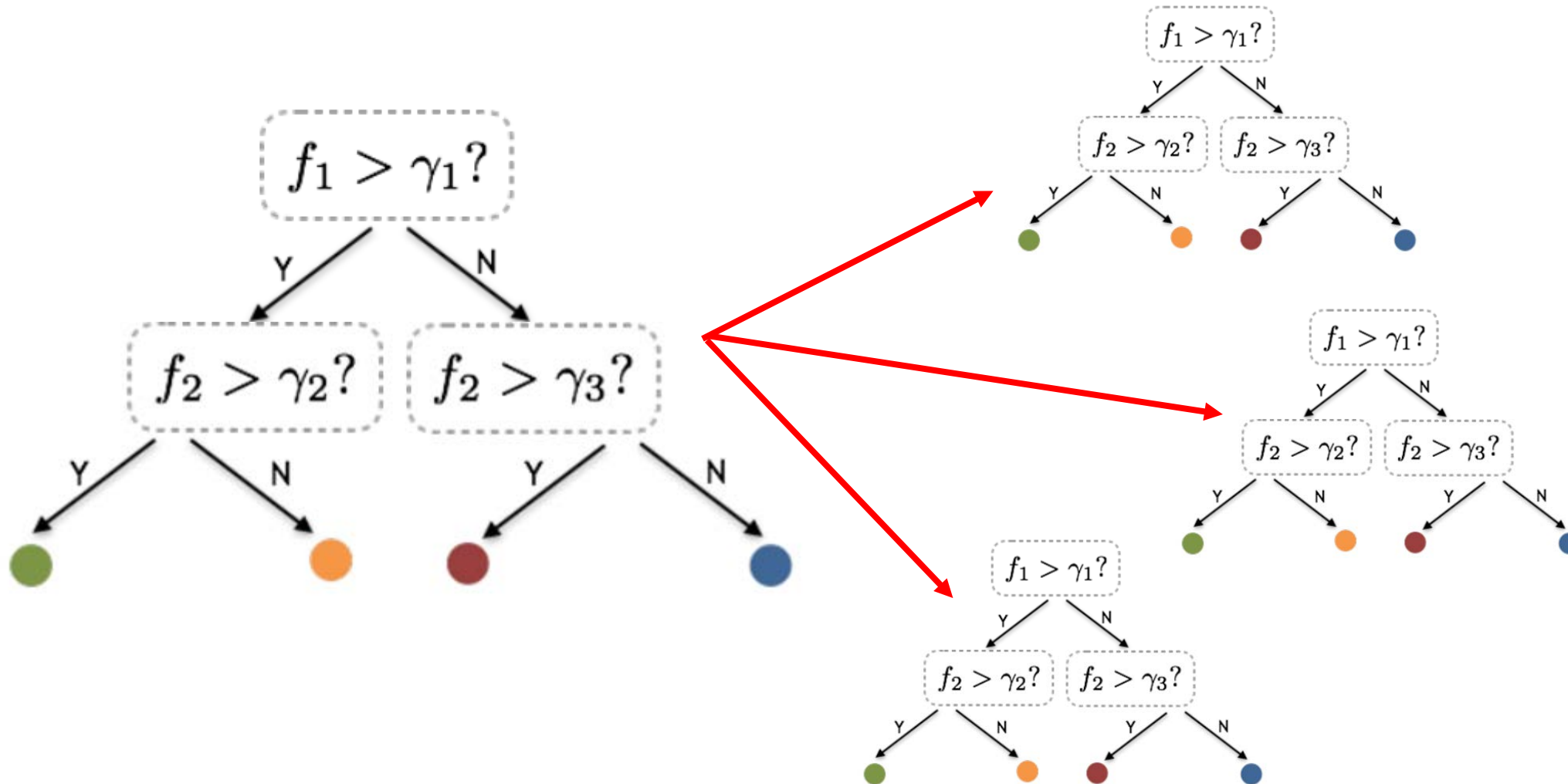
Well Logs ($f_{d,w}$)

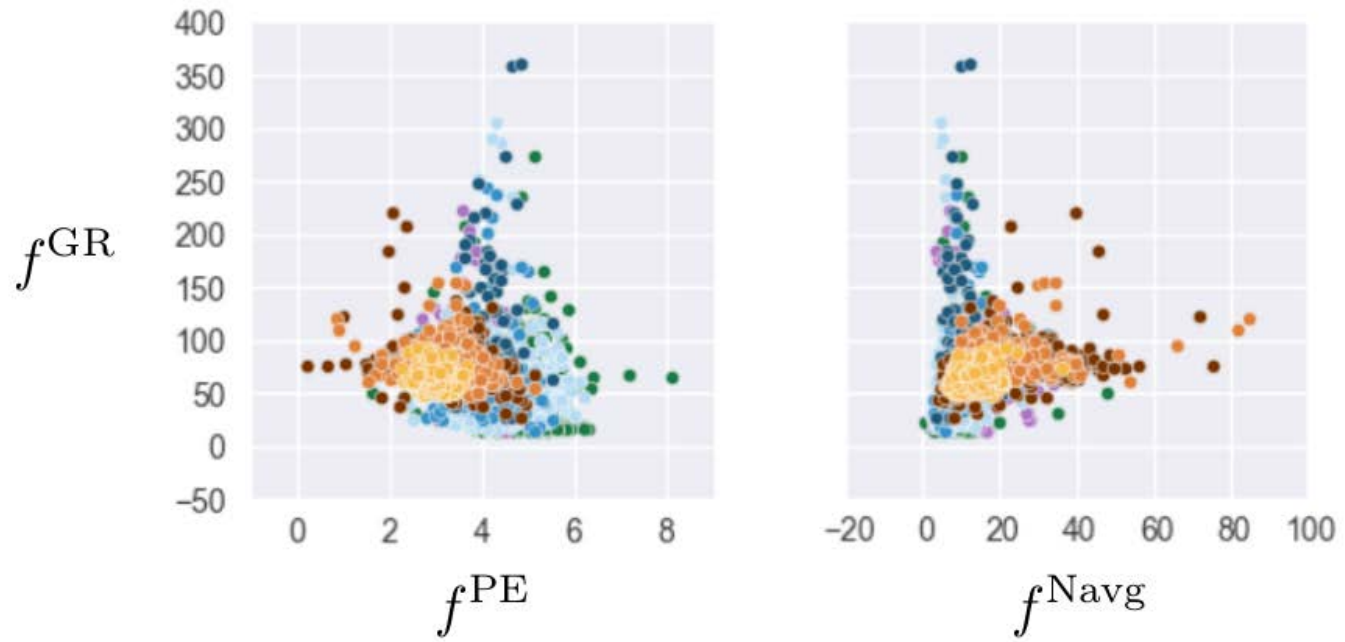
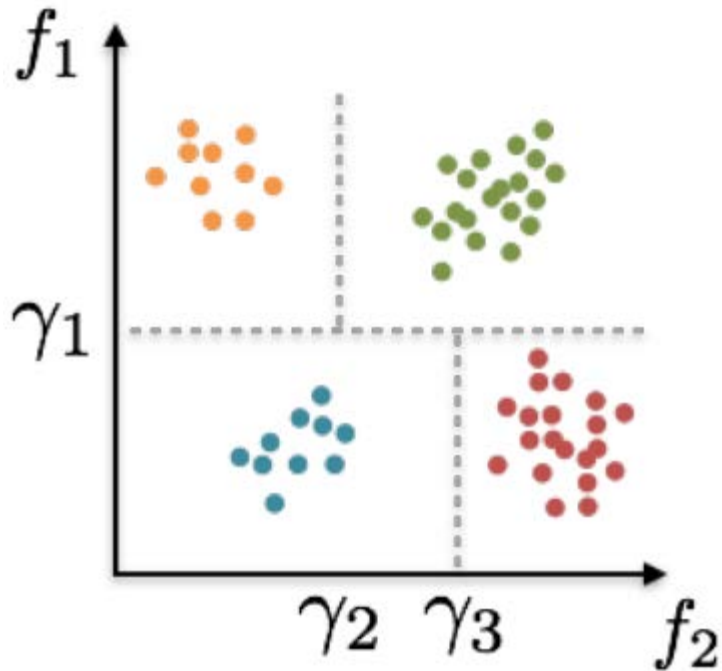
- Gamma ray
- Resistivity
- Photoelectric effect
- Neutron-density porosity difference
- Average neutron-density porosity
- Nonmarine/marine indicator
- Relative position

Facies ($c_{d,w}$)

- Nonmarine sandstone (SS)
- Nonmarine coarse siltstone (CSiS)
- Nonmarine fine siltstone (FSiS)
- Marine siltstone and shale (SiSh)
- Mudstone (MS)
- Wackestone (WS)
- Dolomite (D)
- Packstone-grainstone (PS)
- Phylloid-algal bafflestone (BS)

Classification > Gradient boosting classifier (an ensemble of decision trees)

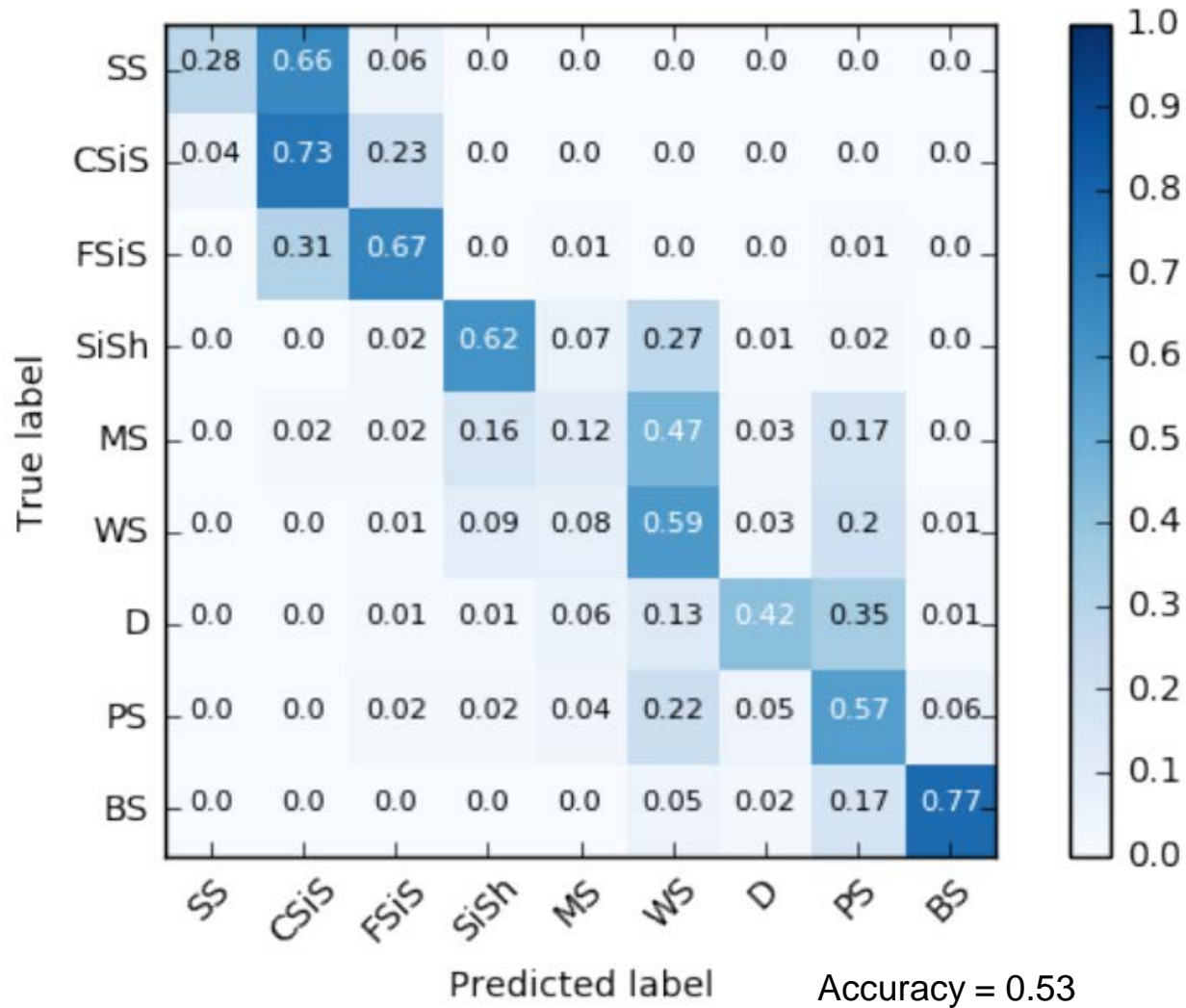




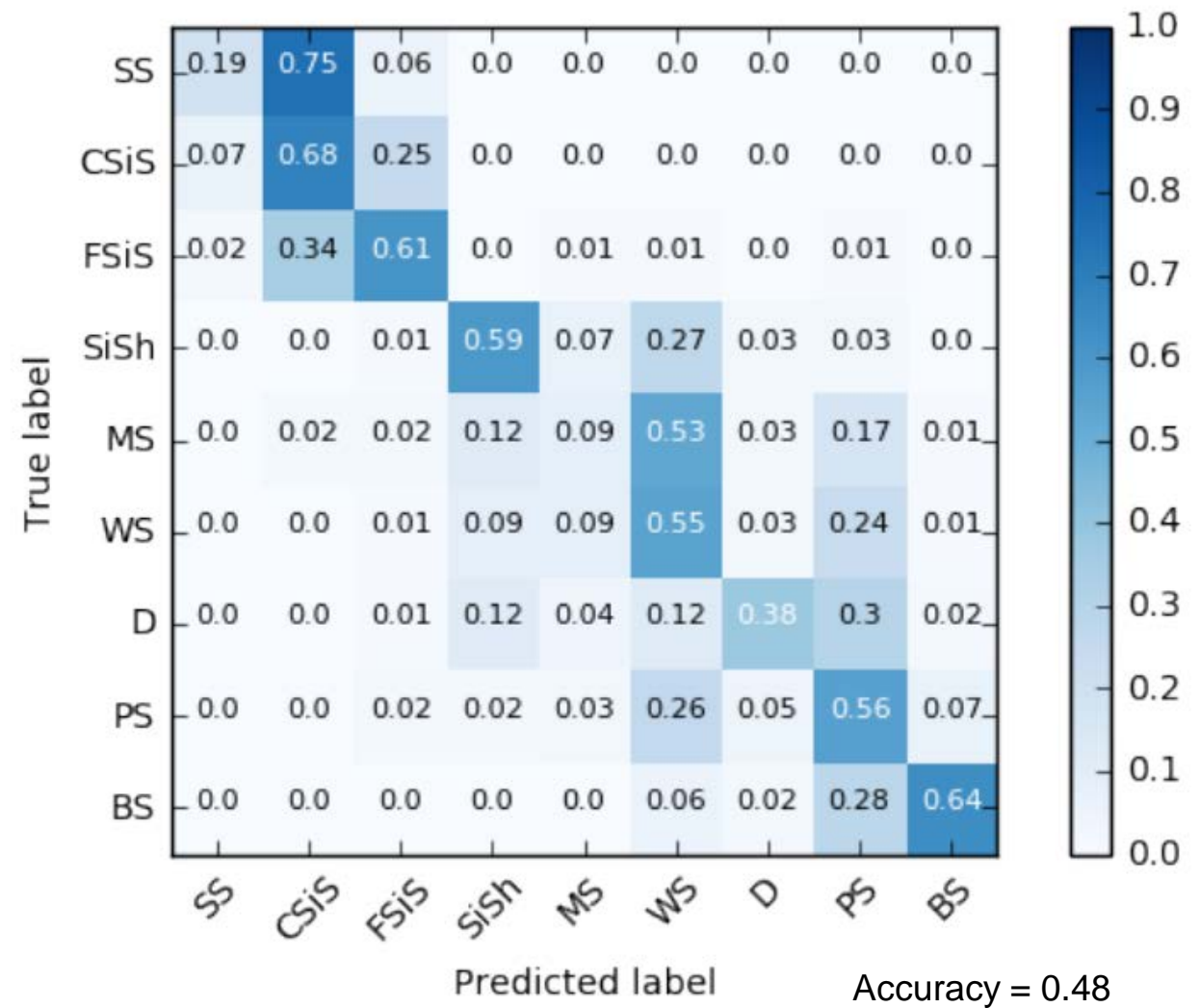
- Facies
- BS
 - PS
 - D
 - WS
 - MS
 - SiSh
 - FSiS
 - CSiS
 - SS

Proposed solution > *Feature augmentation* (generate new features from the available ones)

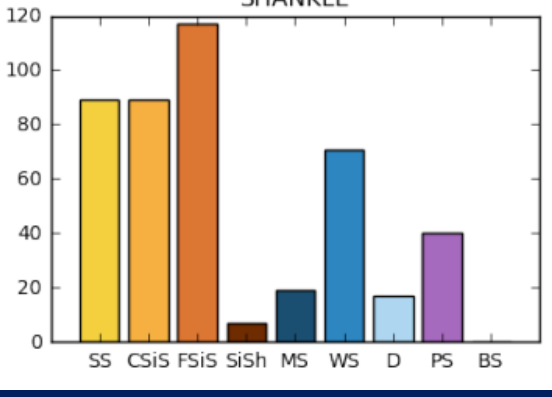
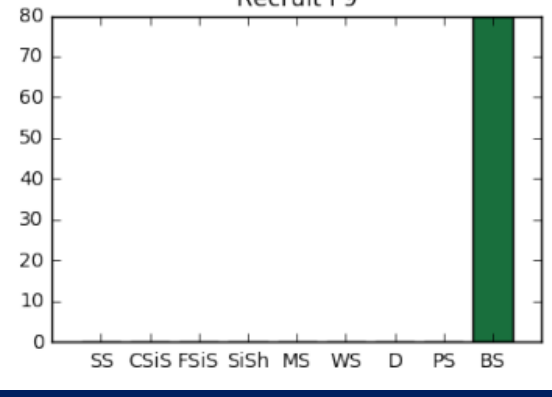
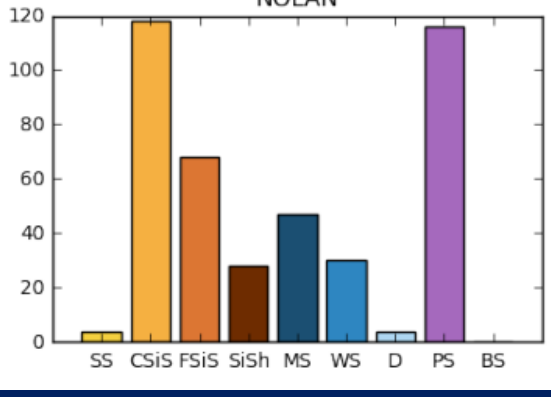
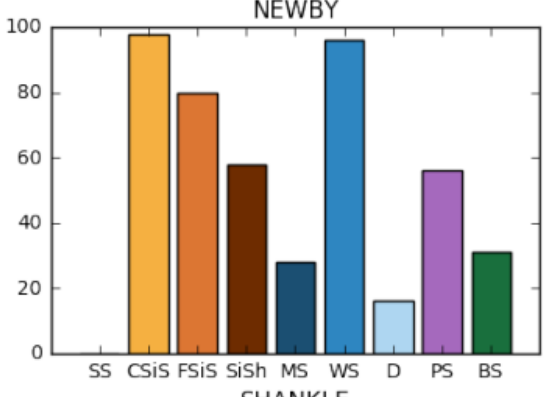
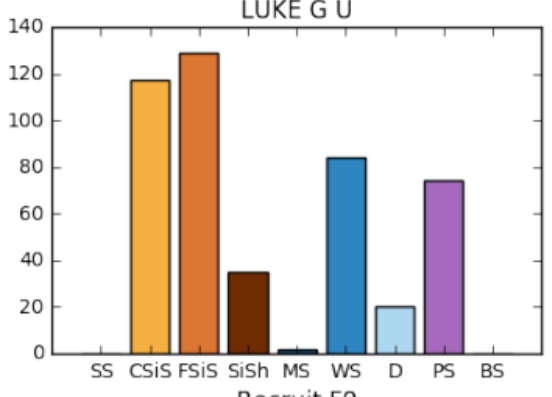
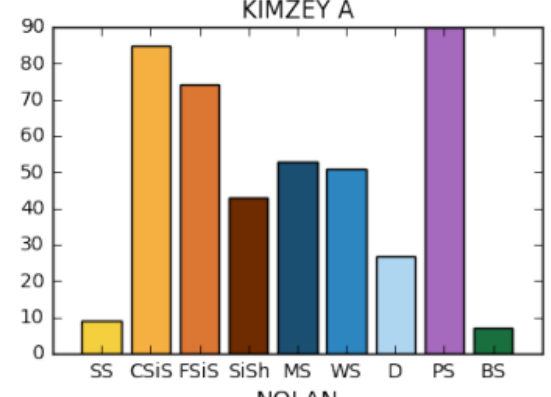
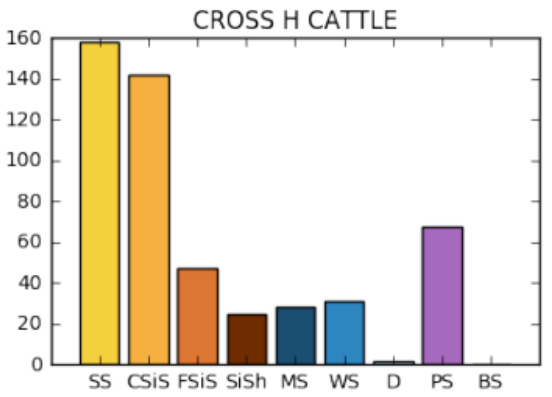
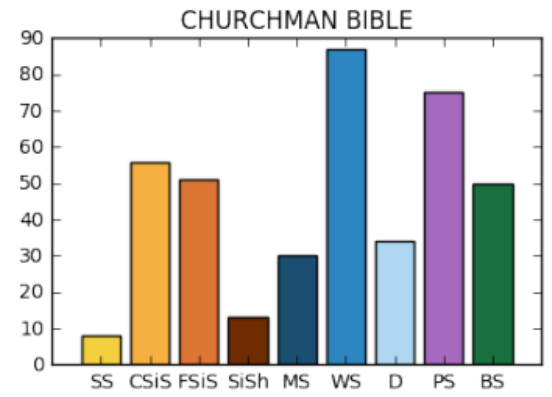
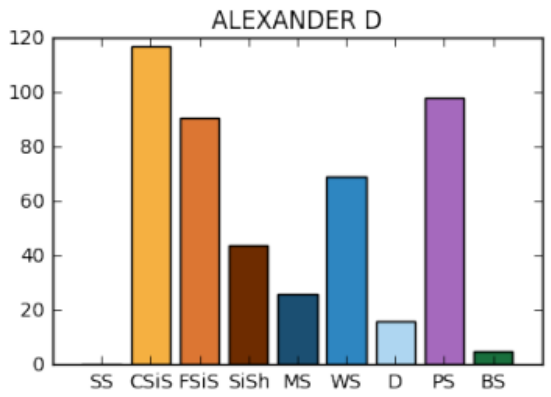
Confusion Matrix



(a) With feature augmentation



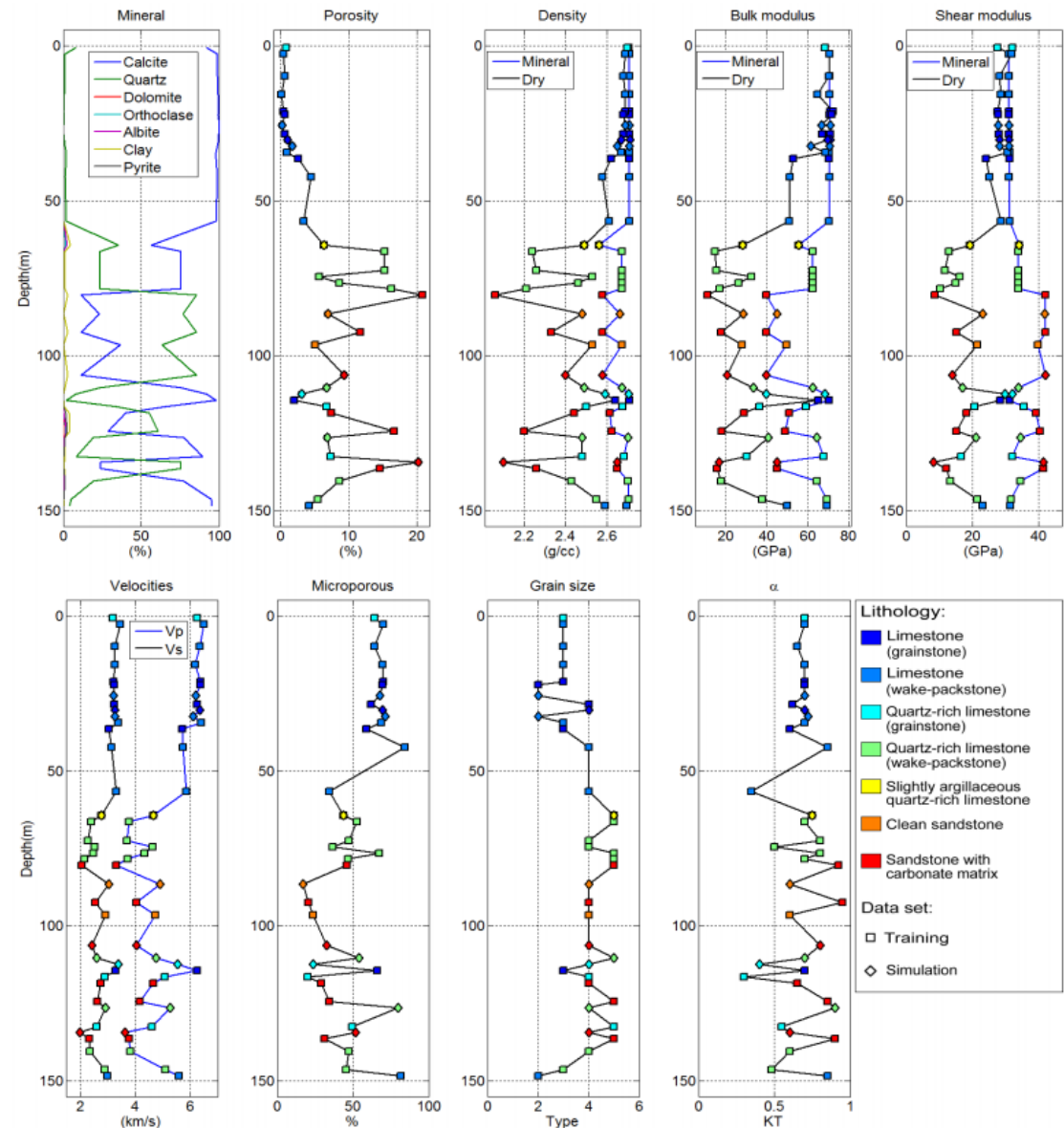
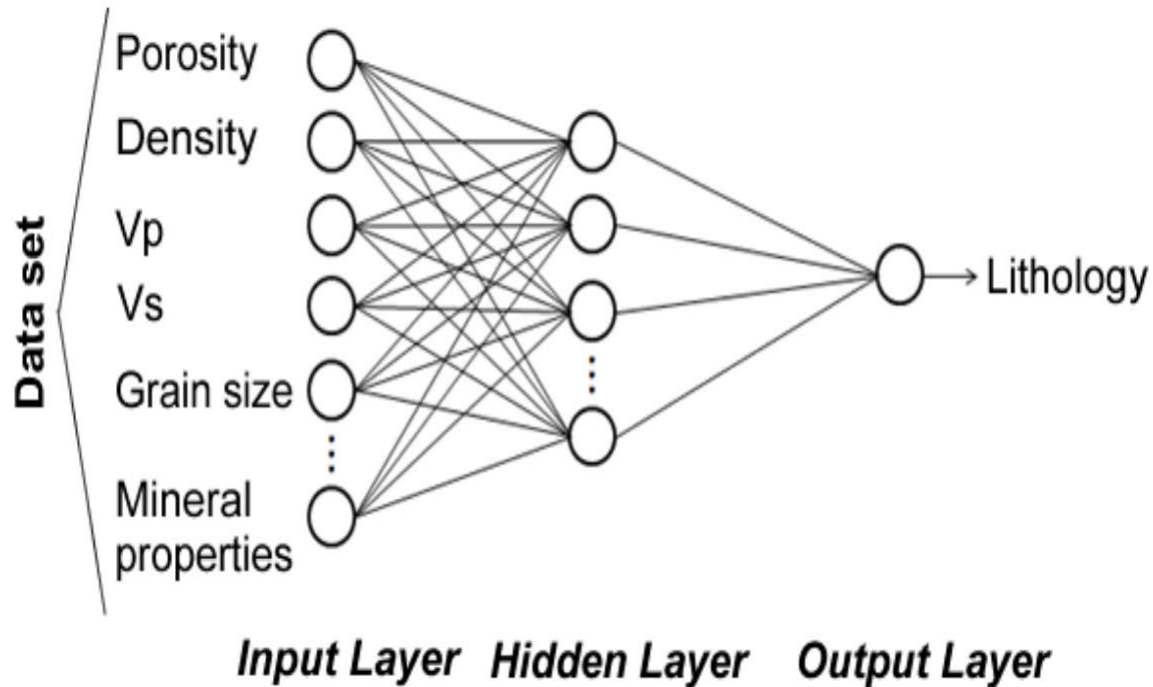
(b) Without feature augmentation



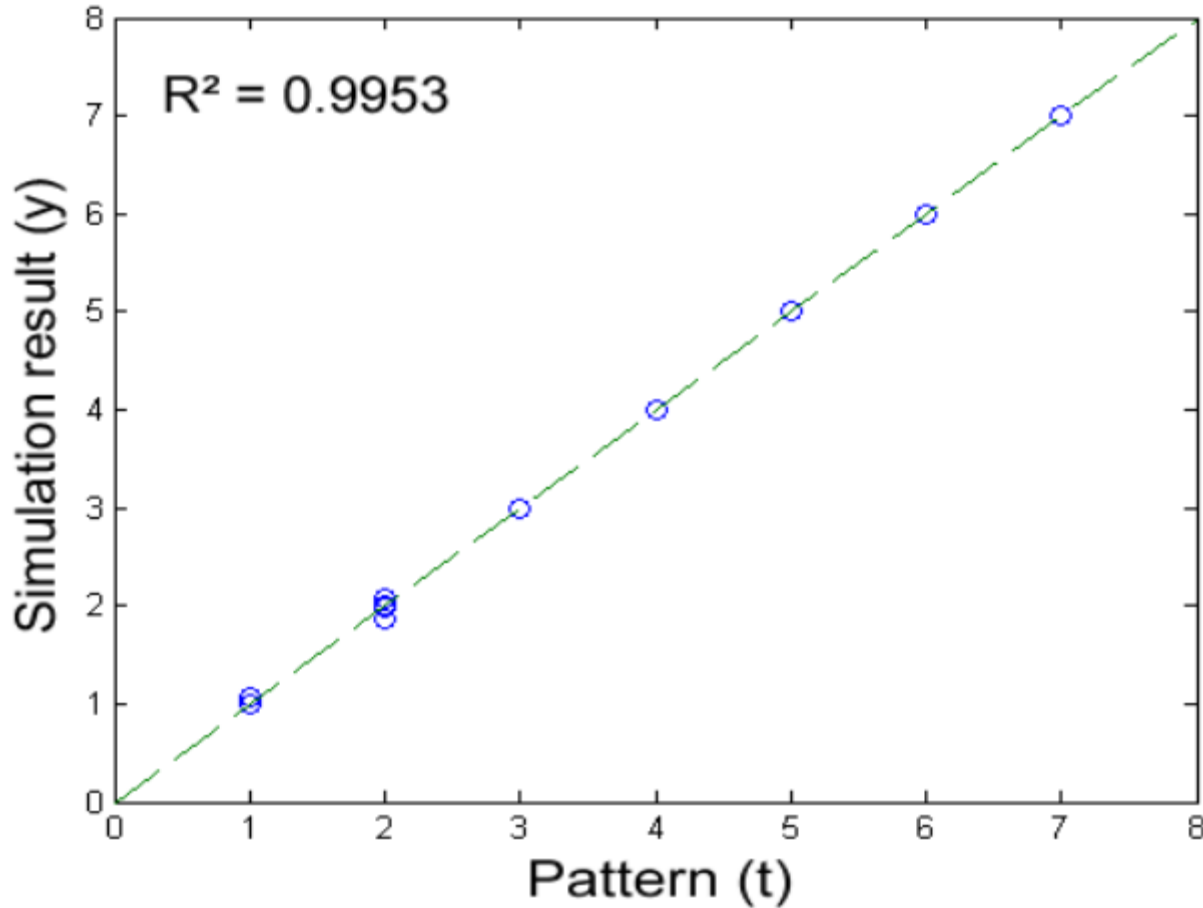
Unbalanced classification

Silva, A., Neto, I. L., Carrasquilha, A., Missagia, R., Ceia, M., Archilha, N., 2013, **Neural network computing for lithology prediction of carbonate- siliciclastic rocks using elastic, mineralogical and petrographic properties**, *Thirteenth International Congress of the Brazilian Geophysical Society*, 1055-1058

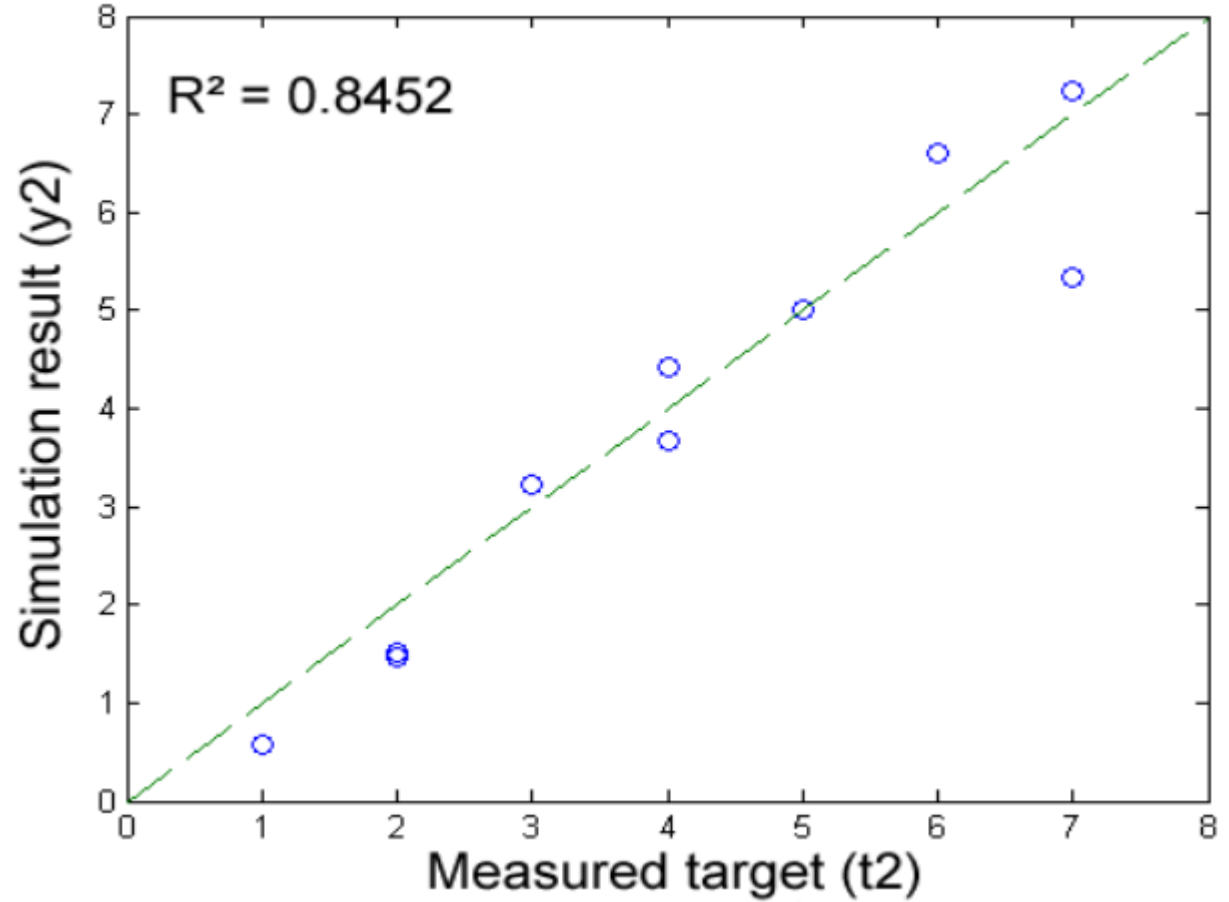
1	Limestone (grainstone)
2	Quartz-rich limestone (grainstone)
3	Quartz-rich limestone (wacke-packstone)
4	Limestone (wacke-packstone)
5	Slightly argillaceous quartz-rich limestone
6	Clean sandstone
7	Sandstone with carbonate matrix



Knowledge of lithology - training dataset



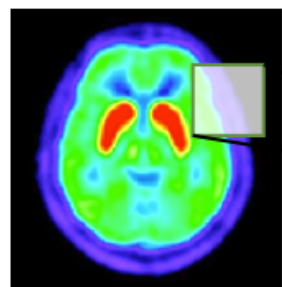
Knowledge of lithology - predicting target



Accuracy = 84.52%

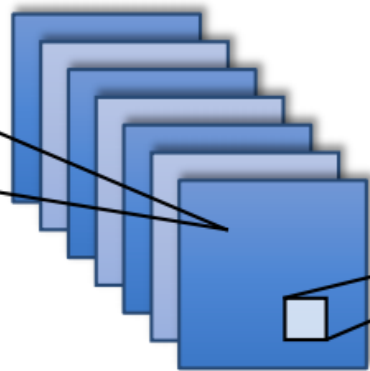
Lewis, W., Vigh, D., 2017, **Deep learning prior models from seismic images for full-waveform inversion**, *SEG International Exposition and 87th Annual Meeting*, 1512-1517

- FWI > issues inverting salt bodies
- Regularization using Tikhonov method
- Uses information from other surveys
- Train a “*Convolutional Neural Networks*” model for salt bodies
- Incorporates in the FWI routine

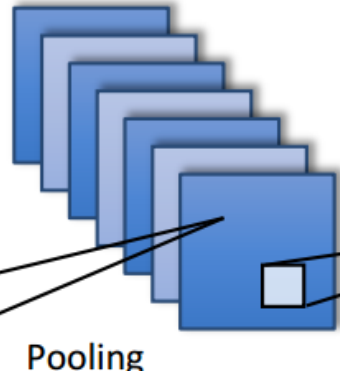


Image

Convolution

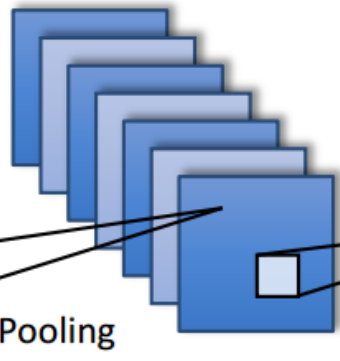


Convolution



Pooling

Convolution



Pooling

Convolution



Pooling

Fully connected



Fully connected



Fully connected

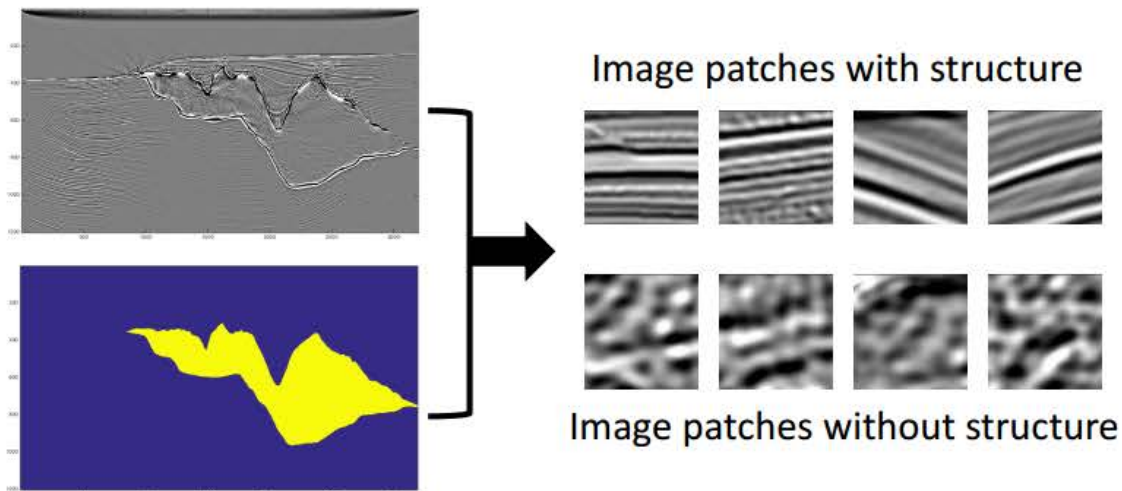


Softmax

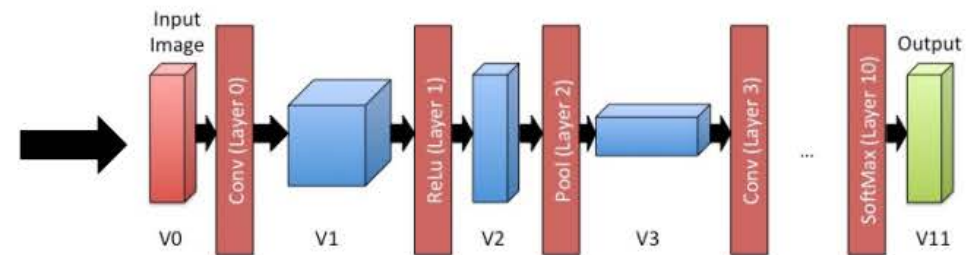


Probability

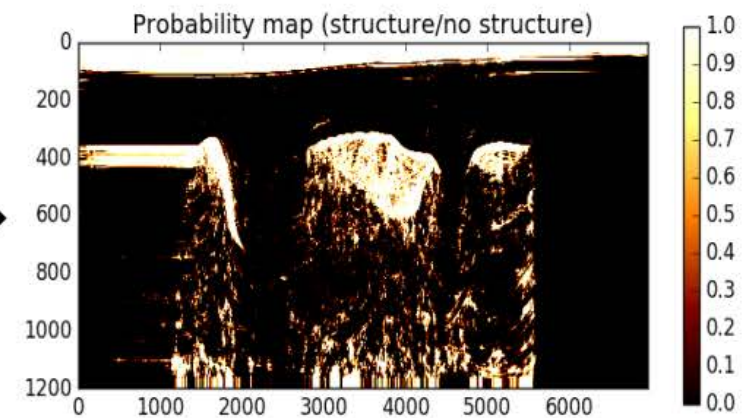
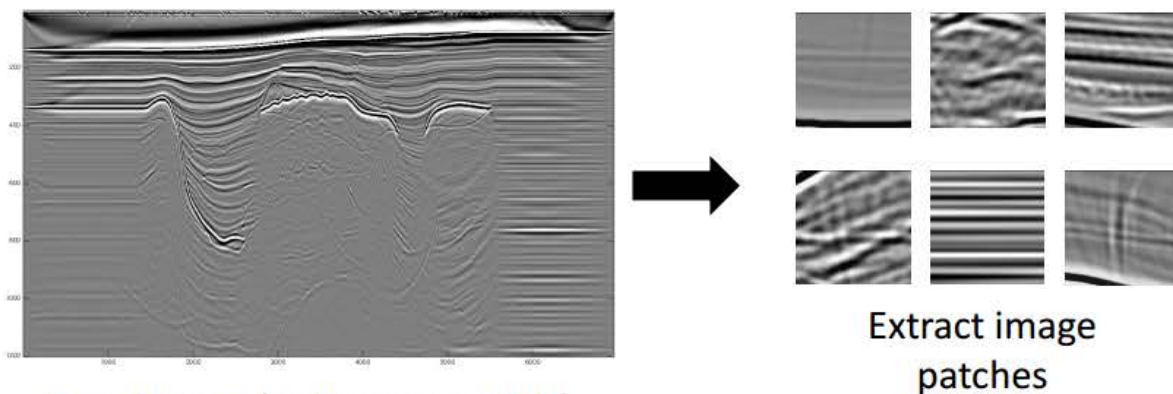
Training dataset (known models)

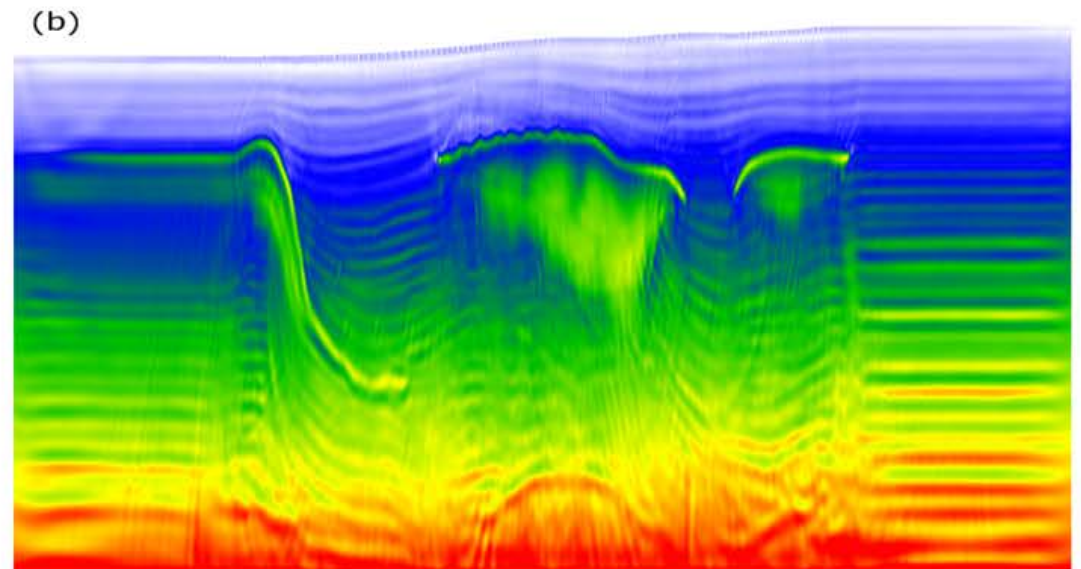
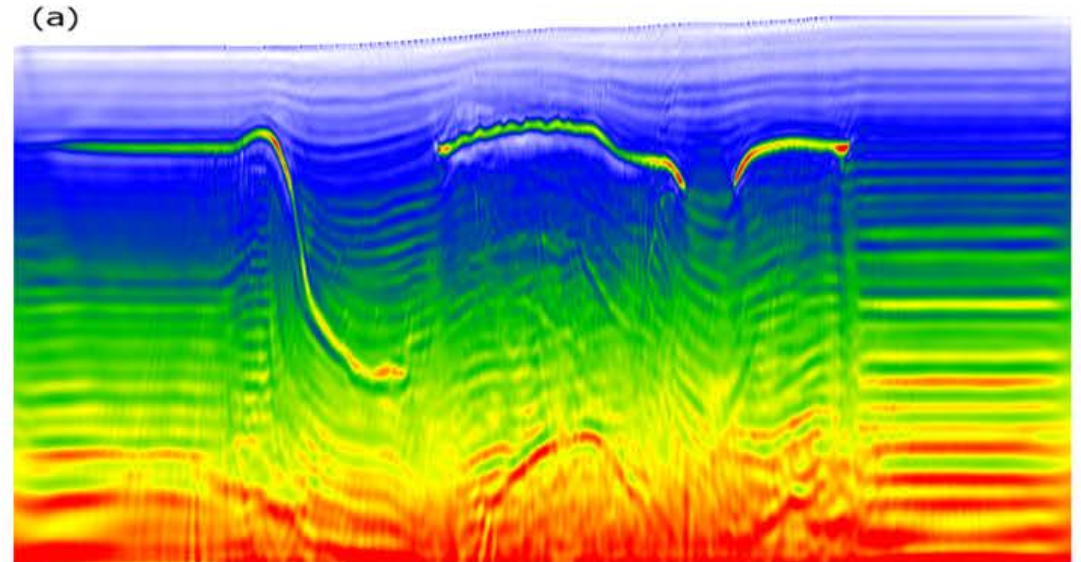
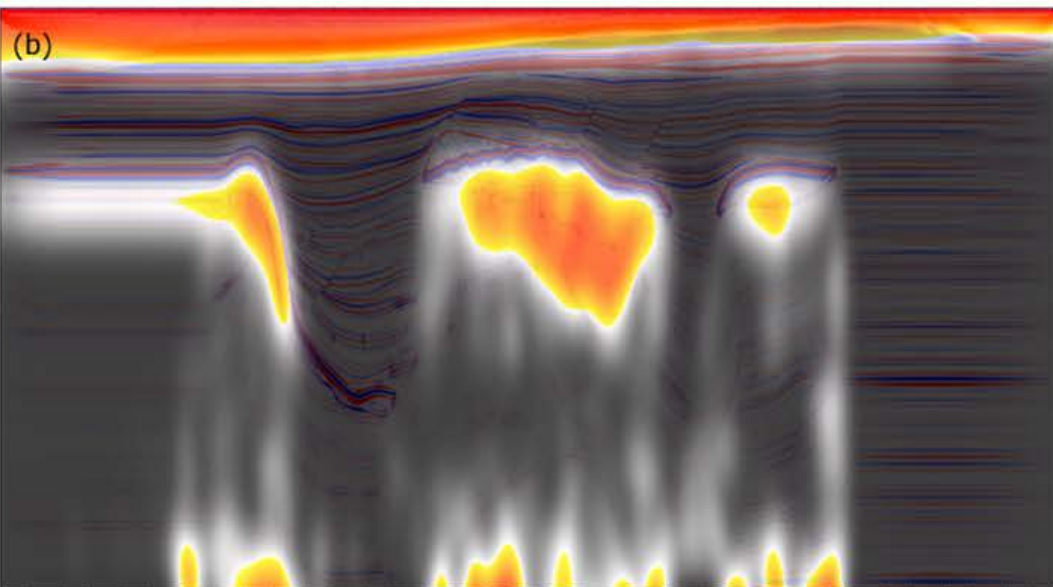
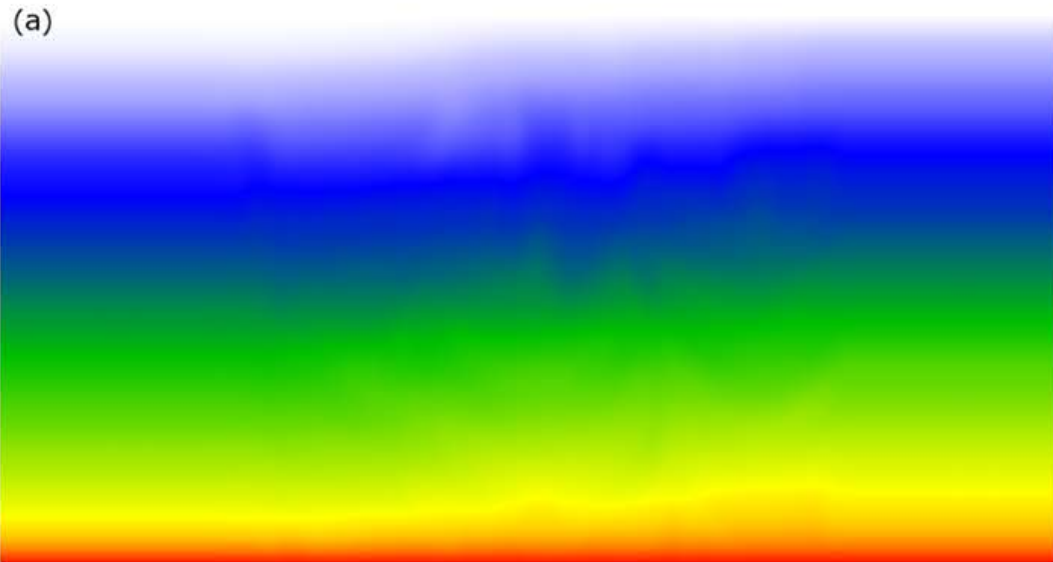


Train deep learning model



Run prediction





- Machine learning is an area in full expansion and applicable on many different sciences.
- Increased computer power, such as use of GPUs.
- Need to handle “big data”.
- Promising applicability in geophysics/geology
- **We should be part of this!**