



What Is Core?

- Core: intact, cylindrical piece of rock coming from wellbore
- Brought to surface for analysis
- Collected during drilling of exploratory wells
 - At novel sites or untapped reservoir depths
- Core analysis evaluates economic viability of exploratory wells and surrounding reservoirs

The Problem

- Analysis is done at a maximum of one-foot intervals which generates a representative data set
 - Reservoir may not be homogenous
- Manual core analysis prone to human errors and time-consuming

The Solution

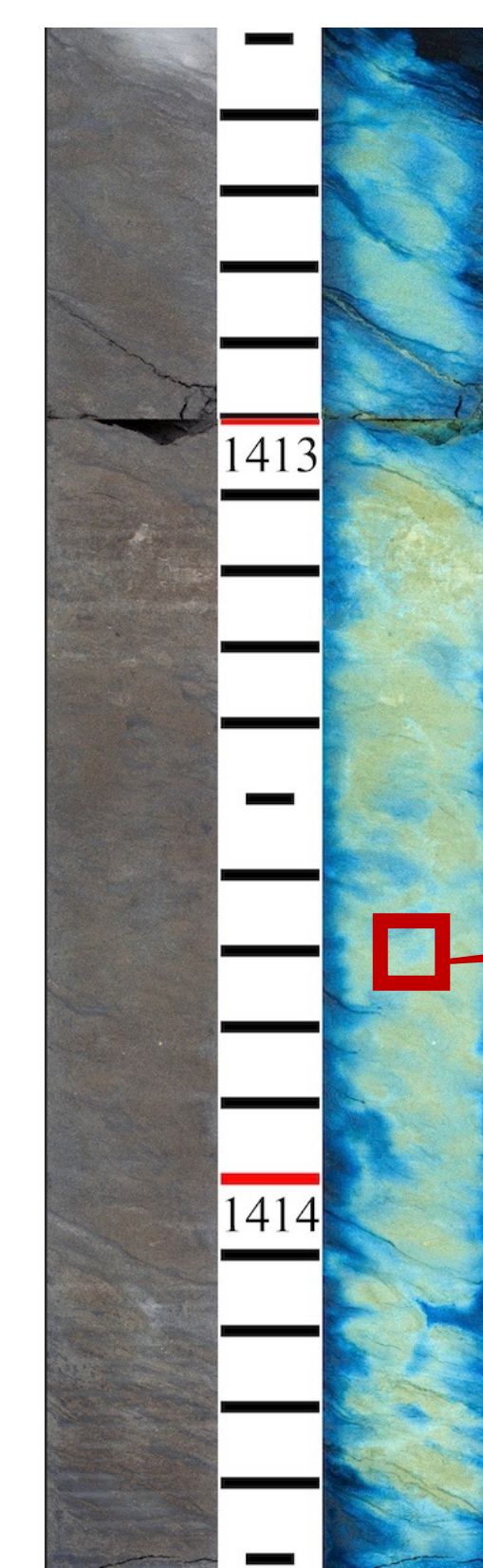
- Artificial intelligence can be used to overcome many of the flaws of conventional core analysis
 - Convolutional neural networks can analyze core images and generate a continuous analysis
 - Convolutional neural networks can generate an analysis of the core almost instantly and save time
 - Convolutional neural networks can analyze core images at no cost
 - Convolutional neural networks can analyze core images and eliminate human error and subjectivity

Industry Partners



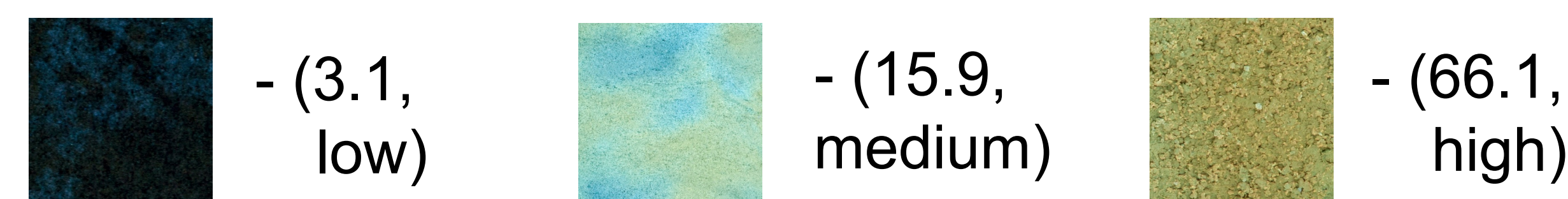
Implementation

To train the neural networks to make predictions about a core's oil saturation level and saturation category, we collected a large inventory of core slab images from our partner companies along with the corresponding core analysis reports. For every oil saturation report value, a crop of the core slab was taken at the equivalent depth.

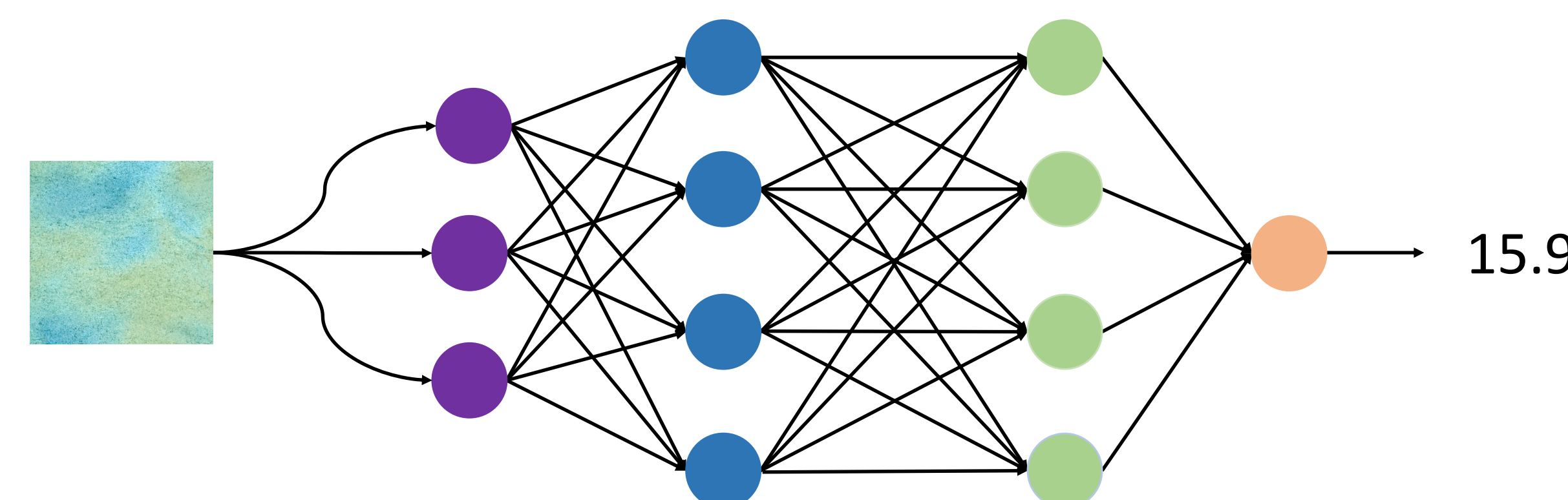


Sample Number	Depth ft	Perm. Kair md	Porosity %	Fluid Saturation			
				Oil %	Water %	OW Ratio	Total %
53	1362.3	1.8	12.3	21.6	72.8	0.30	94.4
54	1380.7	247.5	29.5	18.3	68.4	0.27	86.7
55	1382.3	2865.6	33.3	55.3	17.1	3.23	72.4
56	1386.2	181.3	24.7	35.6	56.8	0.63	92.4
57	1413.7	35.1	22.1	15.9	72.5	0.22	88.4
58	1423.4	1789.5	35.1	58.1	16.9	3.43	75.0
59	1432.8 F/	109.7	29.6	29.9	56.5	0.53	86.4
60	1434.9	46.0	29.2	38.2	52.4	0.73	90.6
61	1437.0	61.2	26.3	43.8	50.2	0.87	94.0
62	1440.9	12.3	28.0	30.8	63.6	0.48	94.3
63	1447.5	32.3	26.4	34.8	58.3	0.60	93.2
64	1449.8	47.3	26.9	39.5	55.9	0.71	95.4

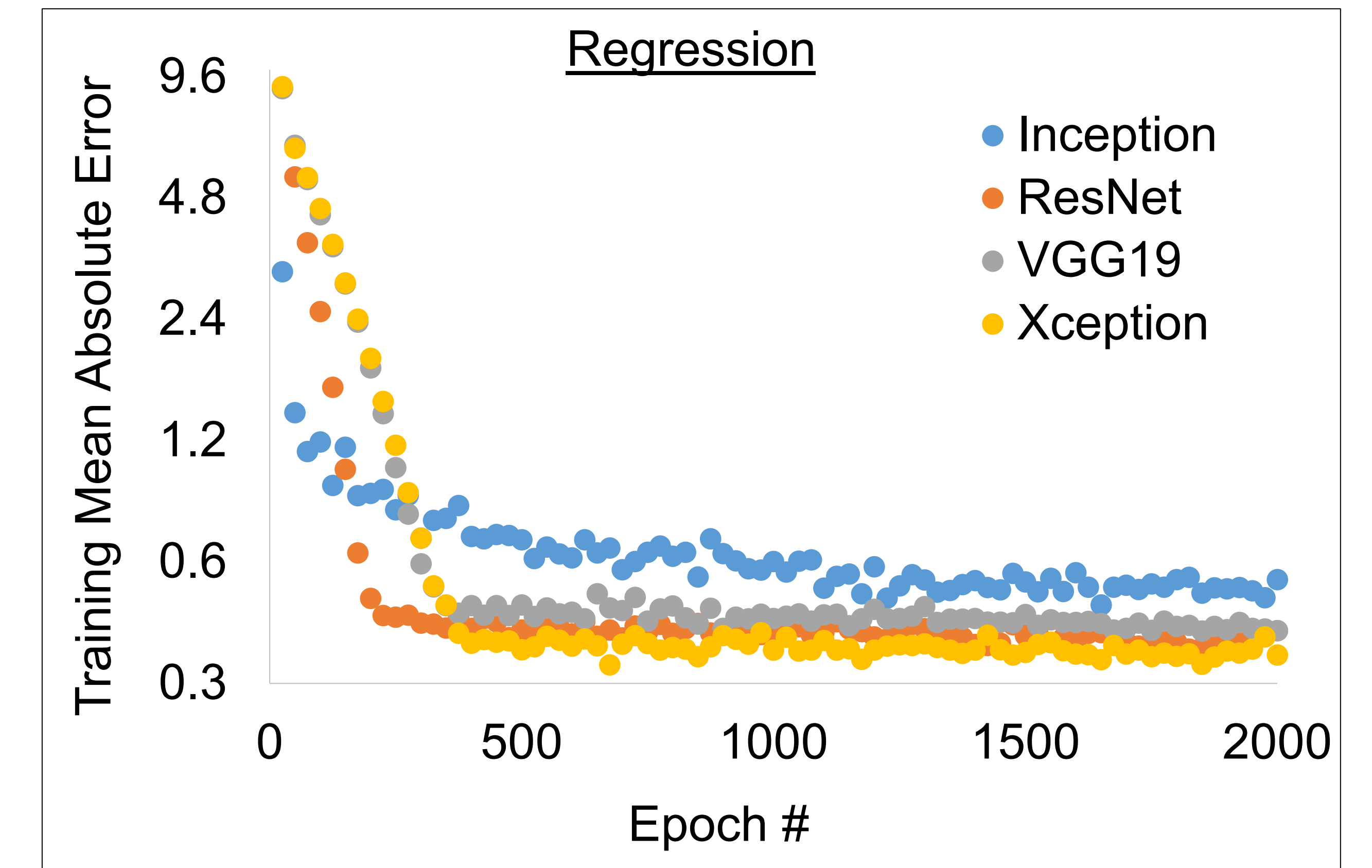
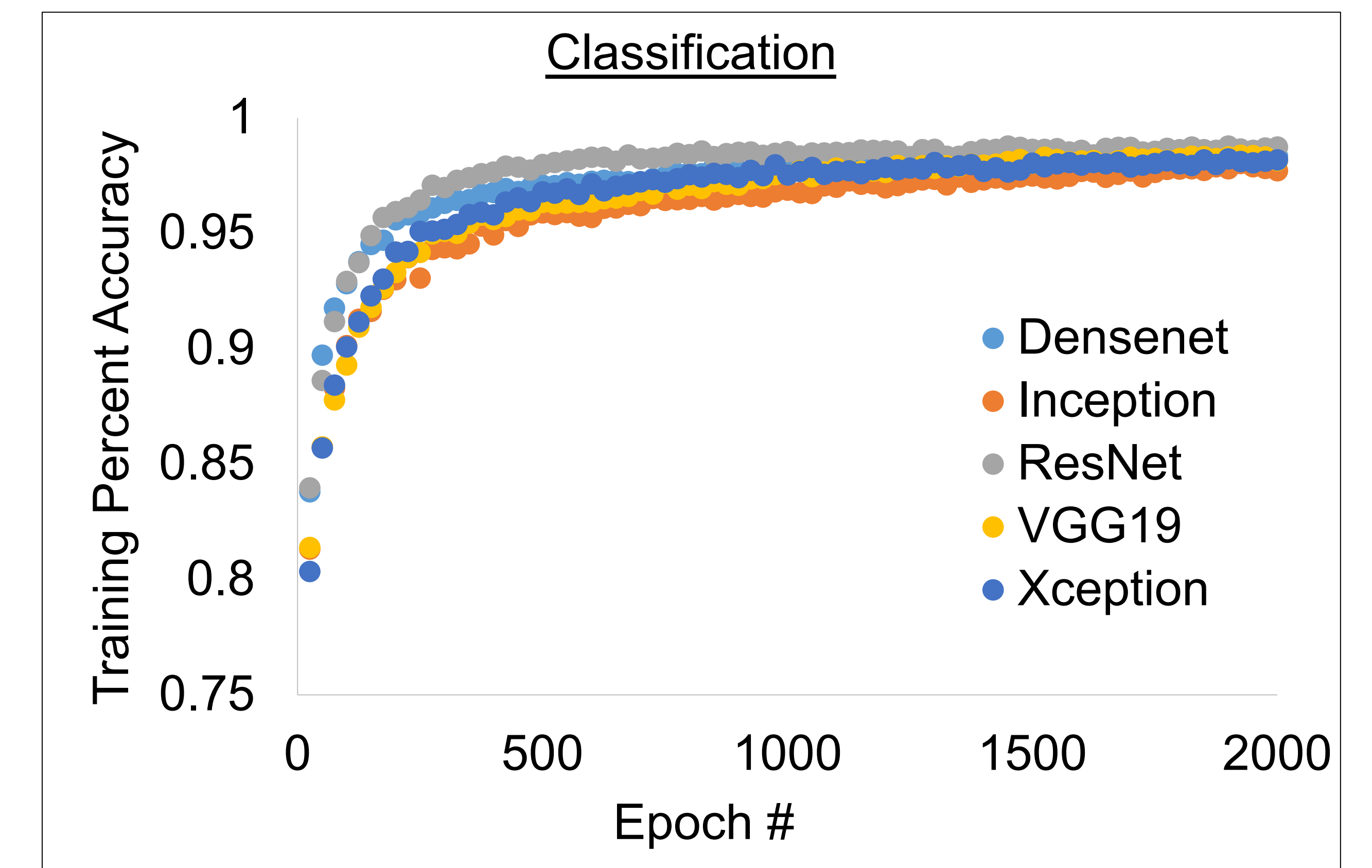
We used neural networks to perform two distinct tasks – classification and regression. For classification, the images were given as input to the neural networks along with a label. For regression, the images were given as input to the neural networks along with the reported saturation value.



The neural networks went through 2000 iterations of training each to refine their predictive capabilities.



Results



- Outcome aligns with expectations – ultraviolet light images generate most accurate results for oil saturation detection
- Classification models categorize core images (high, medium, low) with a high degree of accuracy
 - 87% accuracy (ultraviolet light)
 - 78% accuracy (white light)
- Regression models were trained to generate an average mean absolute error of 9.28 surpassing the ability of a human core analyst
- Future work will involve adding more core data to improve the predictive capabilities of the models and adding an automated lithology describer