Year	Title	Doi	First Author	Affiliation		Platform		Application		Type of ML	Method	Architecture	Type of training	Synthetic Model	Type of data applied to:	Input Data		Loss Function
				Academia Co	Resear cent			Main	Note	Supervised			Synthetic	Model 1	арриев то:	Raw seismic Data	Attributes	
2018	Automatic fault detection with Convolutional Neutral	https://doi.org/10.11 90/IGC2018-192	Ma		x	SEG	Conference International Geophysical Conference, Qingdao, China	Fault detection		Supervised	CNN		Synthetic		Both	Post stack migrated	Variable	Logloss
2018	Networks Seismic fault detection using deep learning technology	https://doi.org/10.11 90/IGC2018-434	Chang		x	SEG	China International Geophysical Conference, Beijing, China	Fault detection		Supervised	CNN	U-Net	Synthetic		Both	Post stack migrated		MSE
2018	Automatic fault detection based on the unsupervised seismic	https://doi.org/10.11 90/AIML2018-15.1	Liu	x		SEG	SEG Maximizing Asset Value Through Artificial Intelligence and Machine	Fault detection		Unsupervised	Multiple (Mean Shift, Brich)		Real		Real	Attributes	Manually selected	
2018	attributes clustering Application of deep learning in seismic data fault recognition	https://doi.org/10.11 90/IGC2018-431	Liang	×		SEG	Learning, Beijing, China International Geophysical Conference, Beijing, China	Fault detection		Supervised	CNN							
2018	3D seismic waveform of channels extraction by artificial intelligence	https://doi.org/10.11 90/segam2019- 3216216.1	Liu		x	SEG	SEG Technical Program	Channel detection		Supervised	CNN		Real		Real			Wasserstein
2019	FaultNet: A deep CNN model for 3D automated fault picking	https://doi.org/10.11 90/segam2019- 3215930.1	Zhang	x		SEG	SEG Technical Program	Fault detection		Supervised	CNN	FaultNet	Synthetic		Both	Post stack migrated		
2017	Seismic-fault detection based on multiattribute support vector machine analysis	https://doi.org/10.11 90/segam2017- 17748277.1	Di	x	x	SEG	SEG Technical Program	Fault detection		Supervised	SVM		Real		Real	Attributes	Manual selection	
2019	3D convolutional neural networks for efficient fault detection and orientation estimation	https://doi.org/10.11 90/segam2019- 3216307.1	Zhao		x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Synthetic		Both	Attributes	Manual selection	
2019	Building realistic structure models to train convolutional neural networks for seismic structural in-terpretation	doi/10.1190/segam20 19-3214282.1#		x		SEG	SEG Technical Program	Fault detection		Supervised	CNN		Synthetic		Synthetic	Post stack migrated		
2018	Patch-level MLP classification for improved fault detection	https://doi.org/10.11 90/segam2018- 2996921.1	Di	x	x	SEG	SEG Technical Program	Fault detection		Supervised	MLP		Real		Real	Attributes	Manually selected	
2018	A deep-learning method for automatic fault detection		MA		x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Real		Both	Post stack migrated		Logloss
2018	A fault-detection workflow using deep learning and image processing	https://doi.org/10.11 90/segam2018- 2997005.1	Zhao		x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Synthetic	SEAM	Both	Post stack migrated		
2018	Statistical imaging of faults in 3D seismic volumes using a machine learning approach	https://doi.org/10.11 90/segam2017- 17589633.1	Guitton	x		SEG	SEG Technical Program	Fault detection		Supervised	SVM		Synthetic		Both	Post stack migrated		
2018	Machine-learning methods in geoscience	https://doi.org/10.11 90/segam2018- 2997218.1	Maniar		x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Real		Real	Post-stack migrated		
2017	Automated fault detection without seismic processing	https://doi.org/10.11 90/tle36030208.1	Araya-polo	. х	x	SEG	TLE	Fault detection		Supervised	CNN		Synthetic		Synthetic	Raw Data		Wasserstein
2017	A scalable deep learning platform for identifying geologic features from seismic attributes	https://doi.org/10.11 90/tle36030249.1	Huang	х	x	SEG	TLE	Fault detection		Supervised	Multiple (CNN, SVM)		Synthetic		Synthetic	Attributes	Manually selected	Logloss
2018	Using generative adversarial networks to improve deep-learning fault interpretation networks	https://doi.org/10.11 90/tle37080578.1	Lu		x	SEG	πE	Fault detection		Semi- supervised	GAN		REAL		Real	Post-stack migrated		
2019	Applications of supervised deep learning for seismic interpretation and inversion		Zheng		x	SEG	TLE	Fault detection		Supervised	CNN		Synthetic		Both	Post stack migrated		
2019	Improving seismic fault detection by super- attribute-based classification	https://doi.org/10.11 90/INT-2018-0188.1	Di	x	x	SEG	Interpretation	Fault detection		Supervised	Multiple (SVM, MLP)		Real		Real	Attributes	Not specified	
2019	Seismic structure interpretation based on machine learning: A case study in coal mining		Li	x		SEG	Interpretation	Fault detection		Supervised	Random Forest		Both		Both	Post stack migrated		
2020	Common-azimuth seismic data fault analysis using residual Unet	https://doi.org/10.11 90/INT-2019-0173.1	Liu	х		SEG	Interpretation	Fault detection		Supervised	FCN	U-net	Synthetic		Both	Post stack migrated		Logloss
2020	Comparing convolutional neural networking and image processing seismic fault detection methods	90/segam 2020.	Qi	x		SEG	SEG Technical Program	Fault detection		Supervised	CNN		Synthetic		Both	Post stack migrated		
2020	Seismic fault detection based on 3D Unet++ model	https://doi.org/10.11 90/segam2020- 3426516.1	Yang	x	x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Synthetic		Synthetic	Post stack migrated		
2020	Channel simulation and deep learning for channel interpretation in 3D seismic images	https://doi.org/10.11 90/segam2020- 3426477.1	Gao	х		SEG	SEG Technical Program	Channel detection		Supervised	CNN		Synthetic		Synthetic	Post stack migrated		Log Loss
2020	Structure enhanced least- squares migration by deep learning based structural preconditioning	https://doi.org/10.11 90/segam2020- 3426676.1	Cheng		x	SEG	SEG Technical Program	Fault detection		Supervised	CNN		Real		REal	Post stack migrated		
2020	Uncertainty estimation using Bayesian convolutional neural network for automatic channel detection	https://doi.org/10.11 90/segam2020- 3427239.1	Pham	x		SEG	SEG Technical Program	Channel detection		Supervised	Bayesian- CNN	U-net	Synthetic		Real	Post stack migrated		
2021	Uncertainty and interpretability analysis of encoder-decoder architecture for channel detection	https://doi.org/10.11 90/geo2020-0409.1	Pham	x		SEG	Geophysics	Channel detection		Supervised	Bayesian- CNN	U-Net	Synthetic		Both	Post-stack Depth Migrated		
2021	Uncertainty quantification in fault detection using convolutional neural	https://doi.org/10.11 90/geo2020-0424.1	Feng	x		SEG	Geophysics	Fault detection		Supervised	CNN	U-Net	Synthetic		Synthetic	Post stack migrated		
2021	networks 3D scattering wavelet transform CNN for seismic fault detection	https://doi.org/10.11 90/segam2021- 3594177.1	Shen	x	x	SEG	First International Meeting for Applied Geoscience & Energy	Fault detection		Supervised	CNN	U-Net	Synthetic		Real	Post stack migrated		
2021	Generative models for the transfer of knowledge in seismic interpretation with deep learning		Durall	x	х	SEG	TLE	Fault detection		Semi- supervised	GAN		Synthetic		Real	Post stack migrated		multiple (Adversarial, Translation, Cyclic, Identity)
2019	FaultSeg3D: Using synthetic data sets to train an end-to-end convolutional neural network for 3D seismic	https://doi.org/10.11 90/geo2018-0646.1	Wu	x	x	SEG	Geophysics	Fault detection		Supervised	CNN	U-Net	Synthetic		Both	Post stack migrated		LOG
2021	fault segmentation  Fault interpretation of  listric faults in the Gulf of  Mexico to train CNN fault  prediction – A case study  of human interpretation  vs automatic fault  interpretation	https://doi.org/10.11	Vera- Arroyo	x		SEG	First International Meeting for Applied Geoscience & Energy	Fault detection		Supervised	CNN	U-Net			Real	Post stack migrated		

2021	Improved 3D neural network architecture for fault interpretation on field data	https://doi.org/10.11 90/segam2021- 3583602.1	Wang		х	SEG	First International Meeting for Applied Geoscience & Energy	Fault detection	Supe	ervised	CNN	U-Net	Synthetic	Both	Post stack migrated		
2021	Focusing unfocused faults with deep learning and residual migration Assisted fault	https://doi.org/10.11 90/segam2021- 3593317.1	Jennings	x	x	SEG	First International Meeting for Applied Geoscience & Energy	Fault detection	Supe	ervised	CNN	U-Net	Synthetic	Both	Angle stacks		
2021	identification and surface extraction by machine learning – A case study from Oman	https://doi.org/10.11 90/segam2021- 3581804.1	Jiang		x	SEG	First International Meeting for Applied Geoscience & Energy	Fault detection	Supe	ervised	CNN		Both	Real	Attributes	Manually selected	
2021	Fault detection from 3D seismic data using Artificial Intelligence	https://doi.org/10.39 97/2214- 4609.202132013	Lowell		x	EAGE	Workshop on Machine Learning	Fault detection					Real	Real	Post stack migrated		
2020	Robust Evaluation of Fault Prediction Results: Machine Learning Using Synthetic Seismic	https://doi.org/10.39 97/2214- 4609.202032015	Sarajaervi		x	EAGE	Digitalization conference	Fault detection	Supe	ervised	CNN	U-Net	Synthetic	Real	Post stack migrated		
2020	Unsupervised Machine Learning Techniques for Subtle Fault Detection Interpretational	https://doi.org/10.39 97/2214- 4609.202010597	Hussein	х		EAGE	EAGE (annual)	Fault detection	Unsu	pervised	Multiple (SOM, PCA)		Real		Attributes	Manually selected	
2020	applications of artificial intelligence-based seismic fault delineation	97/1365- 2397.fb2020020	Han		х	EAGE	First Break	Fault detection									
2018	SEISMIC FAULT DETECTION WITH CONVOLUTIONAL NEURAL NETWORK	https://doi.org/10.11 90/geo2017-0666.1	Xiong		x	EAGE	Geophysics	Fault detection	Supe	ervised	CNN	user defined	Real	Both			LOG
2020	Seismic Attribute-Guided Automatic Fault Prediction by Deep Learning	https://doi.org/10.39 97/2214- 4609.202010542	Jiang		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	multiple (CNN, GAN)	U-Net	Real	Real	Attributes	Random Forest	
2020	Least-Squares Migration with Deep Learning Based Structural Constraints	https://doi.org/10.39 97/2214- 4609.202011134	Cheng		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	CNN	U-Net	Synthetic	Both	Pre-stack pre- migrated		
2020	Enhancing Fault Interpretation Efficiency and Accuracy with Deep Convolutional Neural Network and Elastic Cloud Compute	https://doi.org/10.39 97/2214- 4609.202032100	Manral		х	EAGE	Digitalization conference	Fault detection	Supe	ervised	CNN	U-Net	Real		Post stack migrated		
2020	Deep Bayesian Neural Networks for Fault Identification and Uncertainty Quantification	https://doi.org/10.39 97/2214- 4609.202032036	Mosser		х	EAGE	Digitalization conference	Fault detection	Supe	ervised	Bayesian- CNN	U-Net	Synthetic	Real	Post stack migrated		LOG
2020	Deep Bayesian Neural Networks for Fault Identification and Uncertainty Quantification	https://doi.org/10.39 97/2214- 4609.202011775	Mosser		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	Bayesian- CNN	U-Net	Synthetic	Both	Post stack migrated		
2019	Aspects of automated seismic interpretation using supervised and unsupervised machine learning	https://doi.org/10.39 97/2214- 4609.2019X610101	Bugge		x	EAGE	Subsurface Intelligence workshop	Fault detection	Super	emi- rvised/U iervised	Multiple (DBSCAN; GAN)		Real	Real	Post stack migrated		
2019	Micro-Fault System Detection by Machine Learning	https://doi.org/10.39 97/2214- 4609.201901330	Du		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	SVM	user defined	Real	Real	Attributes	Manually selected	
2019	Deep Neural Networks: Ready for a Day to Day Use in Seismic Interpretation?	https://doi.org/10.39 97/2214- 4609.201901388	Guillon		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	CNN		Both	Real	Post stack migrated		
2019	Automatic fault interpretation from seismic data via convolutional neural	https://doi.org/10.39 97/2214- 4609.2019X610105	Egorov		x	EAGE	Subsurface Intelligence workshop	Fault detection	Supe	ervised	CNN	V-net	Synthetic	Synthetic	Post stack migrated		LOG
2019	networks U_Net & Residual Neural Networks for Seismic Fault Interpretation Fast 3D Seismic	https://doi.org/10.39 97/2214- 4609.201901387	Chang		x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	CNN	U-Net	Real	Real	Post stack migrated		Normalized squared difference
2018	Interpretation with Unsupervised Deep Learning: Application to a Potash Network in the North Sea	https://doi.org/10.39 97/2214- 4609.201800738	Veilard		x	EAGE	EAGE (annual)	Fault detection									
2018	Seismic Fault Detection from Post-Stack Amplitude by Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.201800733	Di	x		EAGE	EAGE (annual)	Fault detection	Supe	ervised	CNN	user defined	Real	Real	Post stack migrated		
2012	Detecting Faults by Integration of Seismic Attributes Using Neural Networks in One of Iranian Oil Fields	https://doi.org/10.39 97/2214- 4609.20148890	Al Sadat Mirkamali	х		EAGE	EAGE (annual)	Fault detection	Unsu	pervised			Real	Real			
2012	Fracture and Carbonate Reservoir Characterization using Sequential Hybrid Seismic Rock Physics, Statistic and Artificial Neural Network: Case Study of North Tiaka Field	https://doi.org/10.39 97/2214-4609- pdb.287.1180851	Hasanusi		х	EAGE	Geo 2012	Fault detection	Supe	ervised							
2013	Seismic data conditioning and neural network-based attribute selection for enhanced fault detection	https://doi.org/10.11 44/petgeo2011-001	Chehrazi	х		SEG	Petroleum Geoscience	Fault detection	Supe	ervised	MLP						
2014	Machine-learning Based Automated Fault Detection in Seismic Traces	https://doi.org/10.39 97/2214- 4609.20141500	Zhang	x	x	EAGE	EAGE (annual)	Fault detection	Supe	ervised			Synthetic	Synthetic			
2017	Machine Learning Based Workflows in Exploration and Production	https://doi.org/10.39 97/2214- 4609.201701656	Limbeck		x	SEG	SEG (annual)	Fault detection	Supe	ervised							
2018	3D Convolutional Neural Networks for Fault Interpretation	https://doi.org/10.39 97/2214- 4609.201800732	Guitton	x	x	EAGE	EAGE (annual)	Fault detection	Supe	ervised	CNN	VGG 16	Real	Real	Post stack migrated		
2019	Constrained automatic tops picking using convolutional neural networks	https://doi.org/10.11 90/segam2019- 3215174.1	Godwin		x	SEG	SEG Technical Program	Horizon tracking	Supe	ervised	CNN						
2019	Semi-automated seismic horizon interpretation using encoder-decoder convolutional neural network	https://doi.org/10.11 90/segam2019- 3215818.1	Wu	х		SEG	SEG Technical Program	Horizon tracking	Supe	ervised	CNN	U-Net	Real	Real	Post stack migrated		
2018	Application of deep learning for seismic horizon interpretation Multiresolution neural	https://doi.org/10.11 90/segam2018- 2998176.1	Lowell		x	SEG	SEG Technical Program	Horizon tracking	Supe	ervised	DAG		Real	Real	Post stack migrated		
2019	networks for tracking seismic horizons from few training images		Peters	x	х	SEG	Interpretation	Horizon tracking	Supe	ervised	CNN	U-Net	Real	Real	Post stack migrated		MSE
2019	Semiautomated seismic horizon interpretation using the encoder- decoder convolutional neural network	https://doi.org/10.11 90/geo2018-0672.1	Wu	х	х	SEG	Geophysics	Horizon tracking	Supe	ervised	CNN		Real	Real	Post stack migrated		Logloss

2020	Uncertainty quantification in imaging and automatic horizon	https://doi.org/10.11 90/segam2020-	Siahkoohi	x		SEG	SEG Technical Program	Horizon		Supervised	Bayesian-		Real		Real	Post stack		
	tracking – A Bayesian deep-prior based approach Waveform embedding: Automatic horizon	3417560.1						tracking			CNN					migrated		
2020	picking with unsupervised deep learning Seismic horizon	https://doi.org/10.11 90/geo2019-0438.1	Shi	x		SEG	Geophysics	Horizon tracking		Unsupervised	Auto- encoder					Post stack migrated		
2021	identification using semisupervised learning with virtual adversarial training	https://doi.org/10.11 90/segam2021- 3594274.1	Wang	x		SEG	First International Meeting for Applied Geoscience & Energy	Horizon tracking		Semi- supervised	Virtual adversarial learning (VAT)		Real		Real	Post stack migrated		
2021	Machine learning assisted seismic interpretation of top and base carbonate for Earth model building: A case study of carbonate platforms from southern Gulf of Mexico	https://doi.org/10.11 90/segam2021- 3592749.1	Fernandez		х	SEG	First International Meeting for Applied Geoscience & Energy	Horizon tracking		Supervised	CNN	U-Net	Real		Real			
2020	Fully Reversible Neural Networks for Large-Scale 3D Seismic Horizon Tracking	https://doi.org/10.39 97/2214- 4609.202012019	Peters	х	х	EAGE	EAGE (annual)	Horizon tracking		Supervised			Real		Real			MSE
2019	Can Machines Learn to Pick Horizons in Post Stack Data?	https://doi.org/10.39 97/2214- 4609.201901975	Yalcinoglu		x	EAGE	EAGE (annual)	Horizon tracking		Supervised	DBN		Real		Real	Post stack migrated	Manually selected	
2018	Deep Learning Based Horizon Interpretation	https://doi.org/10.39 97/2214- 4609.201800923	Lowell		x	EAGE	EAGE (annual)	Horizon tracking		Supervised	DAG		Real		Real	Post stack migrated		
2019	Multivariate attribute selection in seismic facies classification	https://doi.org/10.11 90/segam2019- 3216101.1	Kim	x		SEG	SEG Technical Program	Salt/Geo- body identificatio n	Attributes without seismic section (attribute selection)				Real		Real	Attributes	Wrapper analysis	
2019	Interactive tracking of seismic geobodies using deep learning flood-filling network	https://doi.org/10.11 90/segam2019- 3214810.1	Shi	х		SEG	SEG Technical Program	Salt/Geo- body identificatio n	xiction	Supervised	CNN	U-Net						
2019	Bayesian deep learning for seismic facies classification and its uncertainty estimation	https://doi.org/10.11 90/segam2019- 3216870.1	Mukhopad hyay	х		SEG	SEG Technical Program	Salt/Geo- body identificatio n	Seismic Section	Supervised	Bayesian- CNN	SegNet	Real		Real	Post stack migrated		Log Loss
2019	Supervised seismic facies classification using probabilistic neural networks: Which attributes should the interpreter use?	https://doi.org/10.11 90/segam2019- 3216841.1	Lubo- Robles	x		SEG	SEG Technical Program	Salt/Geo- body identificatio n	Attributes without seismic section	Supervised	PNN		Real		Real	Attributes	PNN (probabilist ic NN)	
2019	Multi-channel convolutional neural network workflow for automatic salt interpretation	https://doi.org/10.11 90/segam2019- 3216596.1	Ye		x	SEG	SEG Technical Program	Salt/Geo- body identificatio n		Semi- supervised	CNN		Real		Real	Post stack migrated		
2017	Geobody interpretation through multiattribute surveys, natural clusters, and machine learning	https://doi.org/10.11 90/segam2017- 17790202.1	Smith		х	SEG	SEG Technical Program	Salt/Geo- body identificatio n		Unsupervised	SOM		Synthetic		Synthetic	Attributes	Not specified	
2020	Deep learning for salt body detection applied to 3D Gulf of Mexico data	https://doi.org/10.11 90/segam2020- 3417484.1	Consolvo		х	SEG	SEG Technical Program	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real		Real	Post stack migrated		
2020	Enrich the interpretation of seismic image segmentation by estimating epistemic uncertainty	https://doi.org/10.11 90/segam2020- 3424987.1	Zhao		х	SEG	SEG Technical Program	Salt/Geo- body identificatio n		Supervised	Bayesian- CNN	U-Net	Synthetic	SEAM	Synthetic	Post stack migrated		
2021	Subsurface salt recognition via deep learning – An iterative semisupervised approach	https://doi.org/10.11 90/segam2021- 3583057.1	Jia	x	х	SEG	First International Meeting for Applied Geoscience & Energy	Salt/Geo- body identificatio n		Self-learning	CNN	U-Net	Real		Real	Post stack migrated		
2021	Deep learning for probabilistic salt segmentation using Bayesian inference	https://doi.org/10.11 90/segam2021- 3594897.1	Konuk	x	x	SEG	First International Meeting for Applied Geoscience & Energy	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Synthetic	BP 2004	Synthetic	Post stack migrated	Not specified	
2021	machines Pseudo three- dimensional deep learning approach for top and bottom of salt	https://doi.org/10.11 90/segam2021- 3594858.1	Kaul		x	SEG	First International Meeting for Applied Geoscience & Energy	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real		Real			LOG
2021	detection Iterative Salt Model Building Using Machine Learning	https://doi.org/10.39 97/2214- 4609.202181011	Warren		x	EAGE	Digital Subsurface Conference in Latin America	Salt/Geo- body identificatio		Supervised	CNN	SaltNet	Real		Real	Post stack migrated		
2021	Deep Learning Techniques Revolutionize E&P – Two practical applications	https://doi.org/10.39 97/2214- 4609.202130009	Jaikla	х	x	EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real		Real	Post stack migrated		
2020	Deep Learning for Salt Body Detection: A Practical Approach	https://doi.org/10.39 97/2214- 4609.202010270	Naeini		х	EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real			Post stack migrated		
2020	Analysis of Seismic Attributes to Assist in the Classification of Salt by Multi-channel Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.202032049	Jiang		х	EAGE	Digitalization conference	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Synthetic	SEAM	Synthetic	Attributes	Not specified	
2020	Enhanced Automatic Segmentation of Salt Bodies from Seismic Images Using Wavelet Convolutional Neural	https://doi.org/10.39 97/2214- 4609.202011987	Zhou	x		EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real			Post stack migrated		
2020	Networks  A comparison of seismic saltbody interpretation via neural networks at sample and pattern levels	https://doi.org/10.11 11/1365-2478.12865	Di	x	х	EAGE	GP	Salt/Geo- body identificatio n		Supervised	Multiple (CNN, MLP)		Synthetic		Synthetic	Attributes	Not specified	LOG
2020	U-SaltNet: A Neural Network for Salt Interpretation	https://doi.org/10.39 97/2214- 4609.202011847	Zhou		х	EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	CNN	U-Net	Real		Real	Post-stack Depth Migrated		Lovasz hinge
2019	Augmented Adversarial Training: Improving Robustness of Neural Network Based Geologic Interpretation	https://doi.org/10.39 97/2214- 4609.201901507	Sen		x	EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	CNN-resnet		Real		Real	Post stack migrated		
2018	Automatic Salt Detection with Machine Learning	https://doi.org/10.39 97/2214- 4609.201800917	Wang	x		EAGE	EAGE (annual)	Salt/Geo- body identificatio n		Supervised	FCNN	U-Net	Synthetic		Synthetic	pre-stack migrated	Attenuatio n	LOG
2017	Salt Classification Using Deep Learning	https://doi.org/10.39 97/2214- 4609.201700918	Waldeland	x		EAGE	EAGE (annual)	Salt/Geo- body identificatio		Supervised	CNN	user defined	Real		Real	Post stack migrated		LOG
2017	Seismic Lithofacies Computation Method	https://doi.org/10.11 90/IGC2017-165	Liu		x	SEG	International Geophysical Conference, Qingdao, China	n LithoFacies	Attribute without seismic section	Supervised	MLP		Real		Real	Post stack migrated		
2017	Based on Deep Learning Automated input attribute weighting for unsupervised seismic facies analysis	https://doi.org/10.11 90/segam2017- 17740318.1	Zhao	x		SEG		LithoFacies	Attributes without seismic section	Unsupervised	SOM		Real		Real	Attributes	Manual weighting	

2019	Seismic facies analysis based on 3D-CNNand seismic stratigraphicslice	https://doi.org/10.11 90/segam2019- 3215951.1	Wang	х			SEG	SEG Technical Program	LithoFacies	Seismic Section	Supervised	CNN		Real		Real	Attributes	Root mean square amplitude	
2019	Facies classification with weak and strong supervision: A comparative study	https://doi.org/10.11 90/segam2019- 3216766.1	Alaudah		x		SEG	SEG Technical Program	LithoFacies	Seismic section	Supervised	CNN	SegNet	Real		Real	Post stack migrated		
2019	Convolutional recurrent neural networks based waveform classification in seismic facies analysis	https://doi.org/10.11 90/segam2019- 3215237.1	Lei		х		SEG	SEG Technical Program	LithoFacies	Seismic section	Supervised	RNN-LSTM		REal		Real	Post stack migrated		
2019	3D seismic facies classification using convolutional neural network and semi- supervised generative adversarial network	https://doi.org/10.11 90/segam2019- 3216797.1	Liu		x		SEG	SEG Technical Program	LithoFacies	Seismic section	Supervised/Se mi-supervised	Multiple (CNN, GAN, PSVM)		Synthetic		Synthetic	Post stack migrated		
2018	Machine-learning based technique for lithology and fluid content prediction: Case study from offshore West Africa	https://doi.org/10.11 90/segam2018- 2996428.1	Zhang		x		SEG	SEG Technical Program	LithoFacies	Attributes with seismic section	Supervised	Random Forest		Synthetic		Both	Post stack migrated	Not specified	
2018	facies prediction and reservoir-property inversion	https://doi.org/10.11 90/segam2018- 2996374.1	Jin		x		SEG	SEG Technical Program	LithoFacies	Impedance (P- and S-)	Supervised	Multiple (KNN, SVM, Random Forest, DNN)		Real		Real	S- and P- impedance		
2018	Seismic facies classification using some unsupervised machine- learning methods	https://doi.org/10.11 90/segam2018- 2997356.1	Chopra	x	x		SEG	SEG Technical Program	LithoFacies	Attributes without seismic section	Unsupervised	Multiple (SOM, GTM, K-means, PCA)		Real		Real	Attributes	Manually selected	
2015	A comparison of classification techniques for seismic facies recognition	https://doi.org/10.11 90/INT-2015-0044.1	Zhao	x	x		SEG	Interpretation	LithoFacies	Attributes without seismic section	Supervised/U nsupervised	Multiple (K- means, SOM, GTM, ANN, SVM)		Real		REal	Attributes	Manually selected	
2018	Unsupervised seismic facies analysis via deep convolutional autoencoders	https://doi.org/10.11 90/geo2017-0524.1	Qian	x			SEG	Geophysics	LithoFacies	Seismic section	Unsupervised	Autoencoder		Botth		Both	Pre-stack migrated		
2019	Convolutional neural networks as aid in core lithofacies classification A seismic facies	https://doi.org/10.11 90/INT-2018-0245.1	de Lima	х			SEG	Interpretation	LithoFacies	Seismic section	Supervised	CNN	ResNetV2	Real		Real	Post stack migrated		
2019	classification method based on the convolutional neural network and the probabilistic framework for seismic attributes and spatial classification	https://doi.org/10.11 90/INT-2018-0238.1	Liu	х			SEG	Interpretation	LithoFacies	Attributes without seismic section	Supervised	Multiple (CNN, MLP, SVM, KNN)		Real		Real	Attributes	MIC (maximal informatio n coeficient)	
2019	A machine-learning benchmark for facies classification Rotation invariant CNN	https://doi.org/10.11 90/INT-2018-0249.1	Alaudah	x			SEG	Interpretation	LithoFacies	Seismic section	Supervised	CNN		Real		Real	Post stack migrated		
2020	using scattering transform for seismic facies classification Unsupervised seismic	https://doi.org/10.11 90/segam2020- 3427911.1	Wang	x			SEG	SEG Technical Program	LithoFacies	Seismic Section	Supervised	CNN		Real		REal	Post stack migrated		
2020		https://doi.org/10.11 90/iwmg2019 06.1	Liu	x	x		SEG	SEG Technical Program	LithoFacies	Seismic section	Unsupervised	Agglomerativ e		Synthetic		Synthetic	Post stack migrated		
2020	Data augmentation using CycleGAN for overcoming the imbalance problem in petrophysical facies classification		Kim	x			SEG	SEG Technical Program	LithoFacies	Attributes with seismic section	Semi- supervised	Cycle-Gan		Real		Real	Attributes	Manually selected	
2020	A comparison of deep machine learning and Monte Carlo methods for facies classification from seismic data	https://doi.org/10.11 90/geo2019-0405.1	Grana	х		x	SEG	Geophysics	LithoFacies	Seismic section	Supervised	RNN-LSTM		Synthetic		Synthetic	Pre-stack migrated		
2020	Creating probabilistic 3D models of lithofluid facies using machine-learning algorithms		Keynejad	х	x		SEG	Interpretation	LithoFacies	Attributes without seismic section	Supervised	Multiple (PNN, BT)		Real		Real	Attributes	step-wise regression method	
2020	Applications of machine learning techniques on angle stacks to enhance carbonate reservoir characterization	https://doi.org/10.11 90/segam2020- 3428250.1	Silver	x			SEG	SEG Technical Program	LithoFacies	Attributes without seismic section	Unsupervised	SOM				Real	Attributes	Manually selected	
2021	Lithofacies prediction from core images using Bayesian neural networks	https://doi.org/10.11 90/segam2021- 3582611.1	Xie	x	x		SEG	First International Meeting for Applied Geoscience & Energy	LithoFacies	Seismic section	Supervised	Bayesian- CNN	U-Net	Real		Real	Post stack migrated		MSE
2021	Deep learning network optimization and hyperparameter tuning for seismic lithofacies classification	https://doi.org/10.11 90/tle40070514.1	Jervis		x		SEG	TLE	LithoFacies	Seismic section	Supervised	CNN	VGG 16	Synthetic		Synthetic	Post stack migrated		
2021	Semisupervised facies classification with reconstruction cooperation	https://doi.org/10.11 90/segam2021- 3594315.1	Asghar	x			SEG	First International Meeting for Applied Geoscience & Energy	LithoFacies	Attributes without seismic section	Semi- supervised	CNN		Real		Real	Attributes	Not specified	multiple (MSE, LOG)
2021	Automatic seismic facies interpretation using supervised deep learning Deep-learning-based	https://doi.org/10.11 90/geo2019-0425.1	Zhang	x	х		SEG	Geophysics	LithoFacies	Seismic section	Supervised	CNN		Real		Real	Post stack migrated		
2021	lithofacies prediction for a complex clastic reservoir using multiseismic information	https://doi.org/10.11 90/segam2021- 3580622.1	Xu	x	x		SEG	First International Meeting for Applied Geoscience & Energy	LithoFacies	Attributes with seismic section	Supervised	CNN		Real		Real	pre-stack migrated	Manually selected	
2021	Facies prediction with Bayesian inference using supervised and semisupervised deep learning	https://doi.org/10.11 90/segam2021- 3579316.1	Singh	х	x		SEG	First International Meeting for Applied Geoscience & Energy	LithoFacies	Seismic section	Supervised/Se mi-supervised		U-Net	Real		Real	Post stack migrated		Variational
2021	Are seismic attributes still helpful for deep learning? Impact of sedimentary		Dunham	х			SEG	First International Meeting for Applied Geoscience & Energy	LithoFacies	Attributes without seismic section	Supervised	RNN		Synthetic	SEAM	Synthetic	Angle stacks/attrib utes	Not specified	
2021	facies on machine learning of acoustic impedance from seismic data: Lessons from a geologically realistic 3D model	https://doi.org/10.11 90/INT-2021-0035.1	Zeng	x			SEG	Interpretation	LithoFacies	Attributes without seismic section	Supervised	Random Forest		Synthetic		Synthetic	Post stack migrated	Not specified	
2021	Unsupervised machine learning using 3D seismic data applied to reservoir evaluation and rock type identification	https://doi.org/10.11 90/INT-2020-0108.1	Hussein	x	x		SEG	Interpretation	LithoFacies	Attributes without seismic section	Unsupervised	SOM		Real		Real	Attributes	PCA	
2021	Unsupervised machine learning for time-lapse seismic studies and reservoir monitoring	https://doi.org/10.11 90/INT-2020-0176.1	Hussein	x	x		SEG	Geophysics	LithoFacies	4D data	Unsupervised	SOM		Real		Real	4D	PCA	
2021	Seismic classification: A Thalweg tracking/machine learning approach	https://doi.org/10.39 97/1365- 2397.fb2021020	de Groot		x		EAGE	First Break	LithoFacies	Attributes without seismic section	Supervised	CNN							
2020	Toward a Semisupervised Machine Learning Application to Seismic Facies Classification	https://doi.org/10.39 97/2214- 4609.202011486	Dunham	х			EAGE	EAGE (annual)	LithoFacies	Attributes without seismic section	Semi- supervised			Synthetic	SEAM	Synthetic	Attributes	Not specified	
`2020	Deep Machine Learning Application for Supervised Facies Classification	https://doi.org/10.39 97/2214- 4609.202075030	Amarullah		x		EAGE	EAGE (annual)	LithoFacies	Attributes without seismic section	Supervised	DFNN (Deep Forward NN)		Real		Real	Attributes	Manually selected	MSE

	Interactive Deep Learning																		
2020	Assisted Seismic Interpretation Technology Applied to Reservoir Characterization: A Case	https://doi.org/10.39 97/2214- 4609.202084012	Krueger		x		EAGE	EAGE (annual)	LithoFacies	Attributes with Seismic section				Real			Post stack migrated		
2020	Study from Offshore Santos Basin in Brazil Application of Fuzzy System with Deep Learning in Seismic Facies	https://doi.org/10.39 97/2214-	Zhang		x		EAGE	EAGE (annual)	LithoFacies	Seismic section	Supervised	CNN	U-Net	Synthetic	Dutch Nlog	Synthetic	Post stack migrated		LOG
2020	Analysis  DeepSeismic: a Deep  Learning Library for  Seismic Interpretation	4609.202010742 https://doi.org/10.39 97/2214- 4609.202032086	Salvaris		x		EAGE	Digitalization conference	LithoFacies	Seismic section	Supervised	CNN	multiple (U- Net, Resnet, Hrnet)	Real		Real	Post stack migrated		
2020	Comparing Bayesian and neural network supported lithotype prediction from seismic	https://doi.org/10.39 97/1365- 2397.fb2020053	Klarner		x		EAGE	First Break	LithoFacies	Attributes without seismic section	Supervised		·····ccy	Real		Real			
	data Comparison of Seismic Traces Clustering Efficiency of Different	https://doi.org/10.39	Churochki									Multiple (K- means,					Acoustic	Acoustic	
2019	Unsupervised Machine Learning Algorithms in Forward Seismic Models Investigating the AVO	97/2214- 4609.201901390	n	х	X		EAGE	EAGE (annual)	LithoFacies	Seismic section	Unsupervised	DBSCAN, Agglomerativ e)		Synthetic		Synthetic	Impedance		
2019		https://doi.org/10.39 97/2214- 4609.201901146	Mosser	х	х		EAGE	EAGE (annual)	LithoFacies	Attributes without seismic section									
2019	Elastic impedance based facies classification using support vector machine and deep learning		Nishitsuji	х	x		EAGE	GP	LithoFacies	Attributes without seismic section	Supervised	Multiple (SVM, DNN)		Synthetic		Both	elastic impedance	elastic impedance	L1 norm
2019	Seismic waveform classification based on Kohonen 3D neural networks with RGB visualization	https://doi.org/10.39 97/1365- 2397.2019012	Priezzhev		x	x	EAGE	First Break	LithoFacies	Attributes without seismic section	Supervised/U nsupervised	SOM		Synthetic		Real	pre-stack migrated		
2019	Seismic Facies Recognition and Stratigraphic Traps Characterisation Based	https://doi.org/10.25 23/IPTC-19503-MS	Zhang		x		EAGE	IPTC	LithoFacies	Attributes without seismic section	Supervised/U nsupervised	multiple (SLP, MLP)		Real		Real			
2019	on Neural Network  Probabilistic Seismic Interpretation Using Bayesian Neural Networks	https://doi.org/10.39 97/2214- 4609.201901510	Mosser	x			EAGE	EAGE (annual)	LithoFacies	Attributes with or without seismic section	Supervised	Bayesian- CNN	Bayesian SegNet	Real		Real	Post stack migrated		
2019	Stratigraphic Segmentation Using Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.201901968	Civitarese		x		EAGE	EAGE (annual)	LithoFacies	Attributes with or without seismic section	Supervised	CNN	user defined	Real		Real	Post stack migrated		
2018	Multi-Modal Machine Learning Fusion for Characterizing Complex Fluvio-Deltaic Tertiary Deposits of North Eastern	https://doi.org/10.39 97/2214- 4609.201801714	Boateng	x			EAGE	EAGE (annual)	LithoFacies	Attributes with or without seismic section	Supervised	MLP		Real		Real	Post stack migrated	Manually selected	
2018	China Solving exploration problems with machine learning	https://doi.org/10.39 97/1365-2397.n0100	Sarcey		x		EAGE	First Break	LithoFacies	seismic section	Unsupervised	SOM		Real			Attributes	Manually selected	
2018	How Machine Learning Is Replacing Conventional Interpretation	97/2214- 4609.201803011	Sacrey		x		EAGE	PESGB	LithoFacies	Attributes with or without seismic section	Unsupervised	SOM		Real		Real	Attributes	PCA	
2018	Deep Learning Applied to Seismic Facies Classification: a Methodology for Training	https://doi.org/10.39 97/2214- 4609.201800237	Chevitarese		x		EAGE	SAINT PETERSBURG	LithoFacies	Attributes with or without seismic section	Supervised	CNN	multiple (VGG 16, AlexNET)	Real		Real	Post stack migrated		LOG
2018	Application of Deep Learning Along Directional Image Gathers for High- Definition Classification	https://doi.org/10.39 97/2214- 4609.201800633	Serfaty		x		EAGE	EAGE (annual)	LithoFacies										
2018	of Subsurface Features Semi-Supervised Deep- Learning Applied To UK North Sea Well And Seismic Data	https://doi.org/10.39 97/2214- 4609.201803013	Nishitsuji	x	x		EAGE	PESGB	LithoFacies	Attributes with or without seismic section	Self-learning	SELF- LEARNING		Real		Real	elastic impedance	elastic impedance	
2018	Lithologic mapping using Random Forests applied to geophysical and remote-sensing data: A demonstration study	https://doi.org/10.11 90/geo2017-0590.1	Kuhn	x			SEG	Geophysics	LithoFacies	Attributes with or without seismic section	Supervised	Random Forest		Real		Real			
2015		https://doi.org/10.39	Hami-		×		EAGE	APGCE		Attributes with	Supervised	DNNA (Democratic				Real			
2015	type, a Neural Network Facies Inversion Machine Learning Applied to 3D Seismic	97/2214- 4609.201526170	Eddine		x		EAGE	APGLE	LithoFacies	or without seismic section	Supervised	network association)		Real		Real			
2019	Data from the Denver- Julesburg Basin Improves Stratigraphic Resolution in the Niobrara	https://doi.org/10.15 530/urtec-2019-337	Laudon		x		SEG	Unconventional Resources Technology Conference	Stratigraphic sequence	Attributes	Unsupervised	SOM		Real		Real	Attributes	PCA	
2018	Seismic data prediction lithology sequence model based on machine learning	https://doi.org/10.11 90/REGE2018-64.1	Li	x			SEG	Reservoir Geophysics, Daqing, China, 5–7 August 2018	Stratigraphic sequence		Supervised	RNN-LSTM		Real		Real	Post stack migrated		
2019	Seismic stratigraphy interpretation via deep convolutional neural networks	https://doi.org/10.11 90/segam2019- 3214745.1	Di		x		SEG	SEG Technical Program	Stratigraphic sequence		Supervised	CNN		Real		Real	Post stack migrated		
2019	Analysis of seismic and texture attributes for stratigraphic segmentation	https://doi.org/10.11 90/segam2019- 3216339.1	Ferreira		x		SEG	SEG Technical Program	Stratigraphic sequence		Supervised/U nsupervised	Multiple (K- means, GMM, Random Forest, SVM)							
2019	Stratigraphy estimation from seismic data using deep learning	https://doi.org/10.11 90/segam2019- 3215869.1	Huot	х	x		SEG	SEG Technical Program	Stratigraphic sequence		Supervised	CNN		Synthetic		Synthetic	Post stack migrated		Logloss
2020	Seismic stratigraphy interpretation by deep convolutional neural networks: A semisupervised workflow	https://doi.org/10.11 90/geo2019-0433.1	Di		x		SEG	Geophysics	Stratigraphic sequence		Supervised	CNN		Real		Real	Post stack migrated		
2020	Turbidite Fan Interpretation in 3D Seismic Data by Point Cloud Segmentation Using Machine Learning	https://doi.org/10.39 97/2214- 4609.202012020	Corlay	x		x	EAGE	EAGE (annual)	Stratigraphic sequence		Unsupervised	DBSCAN		Real		Real	Post stack migrated		
2019	A New Data-Driven Seismic Interpretation Workflow Using Unsupervised Machine Learning and Non-Local Trace Matching	https://doi.org/10.39 97/2214- 4609.201901509	Bugge	х	x		EAGE	EAGE (annual)	Stratigraphic sequence		Unsupervised	DBSCAN		Real		Real	Attributes	Fixed manually	
2019	Semi-Supervised DeepMachine Learning Assisted Seismic Image Segmentation and Stratigraphic Sequence Interpretation	https://doi.org/10.39 97/2214- 4609.201901389	Li		x		EAGE	EAGE (annual)	Stratigraphic sequence		Supervised	CNN		Real		Real	Post stack migrated		
	incopretation																		

2019	Petrophysical properties prediction from pre-stack seismic data using convolutional neural networks	https://doi.org/10.11 90/segam2019- 3215122.1	Das	x		SEG	SEG Technical Program		PP: Porosity, water content, V clay, density, VS, VP	Supervised	CNN	Syr	nthetic		Both	Post stack migrated		
2018	Multiattribute reservoir parameter estimation based on a machine learning technique	https://doi.org/10.11 90/segam2018- 2992296.1	Yuan	х	х	SEG	SEG Technical Program	Petrophysica I properties	PP: Brittleness	Supervised	SVR		Real		Real	Attributes	Manually selected	
2018	Validated artificial neural networks in determining petrophysical properties: A case study from Colombia	https://doi.org/10.11 90/INT-2018-0011.1		×	x	SEG	Interpretation	Petrophysica I properties	PP: density, Vclay	Supervised	ANN		Real		Real	Attributes	Manually selected	
2020	Joint seismic and electromagnetic inversion for reservoir mapping using a deep learning aided feature- oriented approach An unsupervised deep-	https://doi.org/10.11 90/segam2020- 3425261.1	Zhang	x	x	SEG	SEG Technical Program	Petrophysica I properties	PP: Porosity, permeability, water saturation	Supervised	CNN	Syı	nthetic		Synthetic	Post stack migrated		
2020	learning method for porosity estimation based on poststack seismic data	https://doi.org/10.11 90/geo2020-0121.1	Feng	х		SEG		Petrophysica I properties	PP: Porosity	Unsupervised	CNN	Syı	nthetic		Both	Post stack migrated		
2020	A new approach to improve reservoir modeling via machine learning Machine learning for	https://doi.org/10.11 90/tle39030170.1	A Mohamma d	x	x	SEG	TLE	Petrophysica I properties	PP: Porosity, water saturation	Supervised	PNN		Real		Real	Post-stack Depth Migrated	Multi-linear regression	
2020	geophysical characterization of brittleness: Tuscaloosa Marine Shale case study	https://doi.org/10.11 90/INT-2019-0194.1	Mlella	x		SEG	Interpretation	Petrophysica I properties	PP: Brittleness	Supervised	ANN		Real		Real	Attributes	Manually selected	
2020	Porosity prediction using machine learning	https://doi.org/10.11 90/segam2020-w13- 04.1	Jiang	x	х	SEG	SEG Technical Program	Petrophysica I properties	PP: Porosity	Supervised	Multiple (MLP, SVR, Random Forest)	Syı	nthetic		Both	Attributes	Not specified	
2021	Building large-scale density model via a deep- learning-based data- driven method	https://doi.org/10.11 90/geo2019-0332.1	Gao	x	x	SEG	Geophysics	Petrophysica I properties	PP: Density	Supervised	RNN-LSTM	Syı	nthetic I	Marmousi II	Both	Post stack migrated		
2021	A random forest regressor based uncertainty quantification of porosity estimation from multiple seismic attributes	https://doi.org/10.11 90/segam2021- 3582868.1	Zou	x		SEG	First International Meeting for Applied Geoscience & Energy	Petrophysica I properties	PP: Porosity	Supervised	Random Forest		Real		Real	Attributes	Not specified	
2021	Supervised machine learning to predict brittleness using well logs and seismic signal attributes: Methods and application in an unconventional reservoir	https://doi.org/10.11 90/segam2021- 3594773.1	Ore	x		SEG	First International Meeting for Applied Geoscience & Energy	Petrophysica I properties	PP: Brittleness		Multiple (MLP, GB, SVR)				Real	Attributes	Not specified	
2020	New Age, Kolmogorov Full Functional Neural Network Usage for Nonlinear Predictive Seismic Inversion	https://doi.org/10.39 97/2214- 4609.202053084	Priezzhev	х		EAGE	SAINT PETERSBURG	Petrophysica I properties	PP: Porosity, density, VP/VS, V clay	Supervised	Kolmogorov neural network		Real		Real	Angle stacks		
2020	Reservoir characterization utilizing pre-stack inversion and artificial neural network techniques, Offshore Nile Delta, Egypt	https://doi.org/10.39 97/1365- 2397.fb2020086	Othman	х	x	EAGE	First Break	Petrophysica I properties	PP: Porosity, water saturation	Supervised	PNN		Real		Real	Attributes	Manually selected	
2019	Integrating facies-based Bayesian inversion and supervised machine learning for petro-facies characterization in the Snadd Formation of the Goliat Field, south-western Barents Sea	https://doi.org/10.11 11/1365-2478.12654		x	x :	X EAGE	GP	Petrophysica I properties	PP: Porosity	Supervised	Linear Regression		Real		Real	Attributes	Not specified	
2019	ARRI ICATION OF	https://doi.org/10.39 97/2214- 4609.201903385	Yenwongfai Malik		x	EAGE	APGCE	Petrophysica I properties	PP: Porosity	Supervised	Multiple (DNN, PNN)		Real		Real	Attributes	Not specified	
2019	Machine Learning from Core to Seismic; Porosity Prediction in the Norwegian North Sea	https://doi.org/10.39 97/2214- 4609.201901619	Purves		х	EAGE	EAGE (annual)	Petrophysica I properties	PP: Porosity	Supervised	CNN		Real		Real	Post stack migrated		MSE
2019	Machine-Learning Enhanced AVA Inversion for Flow Model Generation	https://doi.org/10.39 97/2214- 4609.201901611	Hoversten		х	EAGE	EAGE (annual)	Petrophysica I properties	PP: Permeability	Semi- supervised	GAN	U-Net Syr	nthetic		Both	AVA	AVA	
2019	Using Convolutional Neural Networks (CNN) and Seismic Data to Predict Gas Content in Shales Gas Reservoirs	https://doi.org/10.39 97/2214- 4609.201900766	Sheng		x	EAGE	EAGE (annual)	Petrophysica I properties	PP: Gas content	Supervised	CNN		Real		Real	pre-stack migrated attributes	Instantano us amplitude	
2018	Integrated Geophysics and Machine Learning for Risk Mitigation in Exploration Geosciences	97/2214-	Dell'Aversa na		x	EAGE	EAGE (annual)		PP: Exploration risk mitigation (reservoir detection)	Supervised	SVM		Real		Real	Post stack migrated		
2018	Hybrid Strategy for Porosity Distribution Mapping on a Heterogeneous Reservoir Using Artificial Intelligence Method Application of spectral	https://doi.org/10.39 97/2214- 4609.201803052	Singh	x		EAGE	Eastern mediteranian	Petrophysica I properties	PP: Porosity	Supervised					Real	Acoustic Impedance		
2018	decomposition and neural networks to characterise deep turbidite systems in the outer fold and thrust belt of the Niger Delta	https://doi.org/10.11 11/1365-2478.12569	Maminu	х		EAGE	GP	Petrophysica I properties	PP: Porosity, resistivity	Supervised	MLP		Real		Real	Post stack migrated	Manually selected	
2013	Neural Network Applications of Seismic Attributes for Predicting Porosity and Production	https://doi.org/10.39 97/2214-4609- pdb.380.3	Celik		х	EAGE	IPETGAS	Petrophysica I properties	PP: Porosity	Supervised			Real		Real			
2016	Identifying Unconventional Potential Using Seismic Inversion and Neural Networks - An Eagle Ford Shale Study	https://doi.org/10.39 97/2214- 4609.201600769	Refunjol	х	x	EAGE	EAGE (annual)	Petrophysica I properties	PP: Porosity, TOC	Supervised			Real		Real			
2017	Application Of Artificial Neural Networks As A Tool For Properties Prediction Using Seismic Data	https://doi.org/10.39 97/2214- 4609.201702628	Muradov	x		EAGE	AMGE	Petrophysica I properties	PP: V shale	Supervised	MLP		Real		Real	Attributes	Manually selected	
2018	Comparative Study Of Deep Feed Forward Neural Network Application For Seismic Reservoir Characterization	https://doi.org/10.39 97/2214- 4609.201803009	Colwell		x	EAGE	PESGB	Petrophysica I properties	PP: Porosity, V shale, water saturation	Supervised	DFNN (Deep Forward NN)		Real		Real	Post stack migrated		
2018	Effect of rock physics modeling in impedance inversion from seismic data using convolutional neural network	https://doi.org/10.11 90/SEGJ2018-135.1	Das	x		SEG	Proceedings of the 13th SEGJ International Symposium, Tokyo, Japan	Impedance	Acoustic	Supervised	CNN	Syı	nthetic		Synthetic	Post stack migrated		

2019	Estimation of acoustic impedance from seismic data using temporal	https://doi.org/10.11 90/segam2019- 3216840.1	Mustafa	x	x	SEG	SEG Technical Program	Impedance	Acoustic	Supervised	TCN		Synthetic		Synthetic	Post stack migrated		
2019	convolutional network Semi-supervised learning for acoustic impedance inversion		Alfarraj	x	x	SEG	SEG Technical Program	Impedance	Acoustic	Semi- supervised	Multiple (CNN, RNN)		Synthetic	Marmousi 2	Synthetic			
2019	Deep learning with cross- shape deep Boltzmann machine for pre-stack inversion problem	https://doi.org/10.11 90/segam2019- 3215696.1	Phan	x		SEG	SEG Technical Program	Impedance	Acoustic	Supervised	CDBM(Cross- shape Deep Boltzmann Machin)							
2019	Seismic impedance inversion based on cycle- consistent generative adversarial network	https://doi.org/10.11 90/segam2019- 3203757.1	Wang		x	SEG	SEG Technical Program	Impedance		Supervised/Se mi-supervised	cGAN		Synthetic		Both	Post-stack migrated		
2019	Semisupervised sequence modeling for elastic impedance inversion	https://doi.org/10.11 90/INT-2018-0250.1	Alfarraj	х		SEG	Interpretation	Impedance	Elastic	Semi- supervised	Multiple (CNN, RNN)		Synthetic	Marmousi 2	Synthetic			MSE
2019	Convolutional neural network for seismic impedance inversion Wasserstein cycle-	https://doi.org/10.11 90/geo2018-0838.1	Das	х		SEG	Geophysics	Impedance	Acoustic	Supervised	Bayesian- CNN		Synthetic		Both	Post stack migrated		MSE
2020	consistent generative adversarial network for improved seismic impedance inversion: Example on 3D SEAM model	https://doi.org/10.11 90/segam2020- 3425785.1	Cai		x	SEG	SEG Technical Program	Impedance	Acoustic	Semi- supervised	cGan	Alex-net	Synthetic	SEAM	Synthetic	Post stack migrated		
2020	Spatiotemporal modeling of seismic images for acoustic impedance estimation	https://doi.org/10.11 90/segam2020- 3428298.1	Mustafa	x		SEG	SEG Technical Program	Impedance	Acoustic	Supervised	TCN (Temporal convolution al network)		Synthetic		Synthetic	Post stack migrated		
2020	Joint learning for seismic inversion: An acoustic impedance estimation case study	https://doi.org/10.11 90/segam2020- 3428378.1	Mustafa	x	x	SEG	SEG Technical Program	Impedance		Supervised	TCN (Temporal convolution al network)		Synthetic	Multiple (SEAM, MARMOUSI)	Synthetic	Post stack migrated		L2 norm
2021	Deep learning for multidimensional seismic impedance inversion	https://doi.org/10.11 90/geo2020-0564.1	Wu	x	х	SEG	Geophysics	Impedance		Supervised	CNN		Both	SEAM	Both	Post stack migrated		MSE
2021	Joint learning for spatial context-based seismic inversion of multiple data sets for improved generalizability and robustness	https://doi.org/10.11 90/geo2020-0432.1	Mustafa	х		SEG	Geophysics	Impedance		Supervised	TCN		Synthetic	Multiple (SEAM, MARMOUSI 2)	Synthetic	Post stack migrated		MSE
2021	Facies control on machine learning of acoustic impedance	https://doi.org/10.11 90/segam2021- 3574779.1	Zeng	х		SEG	First International Meeting for Applied Geoscience & Energy	Impedance		Supervised	Random Forest		Synthetic		Synthetic			
2020	Theory-Guided Deep Learning for Reservoir Characterization	https://doi.org/10.39 97/2214- 4609.202011797	Downton		x	EAGE	EAGE (annual)	Impedance		Supervised			Synthetic		Both	pre-stack migrated		
2020	Acoustic Impedance Inversion of High Resolution Marine Seismic Data with Deep Neural Network	https://doi.org/10.39 97/2214- 4609.202020169	Dujardin		x	EAGE	NSG	Impedance		Supervised	DNN	user defined	Synthetic	Marmousi	Synthetic	Post stack migrated		
2020	Seismic Acoustic Impedance Estimation by Learning from Sparse Wells via Deep Neural Networks	https://doi.org/10.39 97/2214- 4609.202010486	Di		x	EAGE	EAGE (annual)	Impedance		Semi- supervised	CNN	user defined	Synthetic	SEAM	Synthetic	Post stack migrated		MSE
2020	Seismic Impedance Inversion via Combining Convolutional Neural Network and Geostatistics: An Example from Songliao Basin, China	https://doi.org/10.39 97/2214- 4609.202011022	Ge	x	x	EAGE	EAGE (annual)	Impedance		Supervised	CNN	U-Net	Real		Real	Post stack migrated		L2 norm
2019	Rock-physics based Augmented Machine Learning for Reservoir Characterization	https://doi.org/10.39 97/2214- 4609.2019X610102	Downton		x	EAGE	Subsurface Intelligence workshop	Impedance		Supervised	DNN		Synthetic		Real	Angle stacks		
2019	Seismic Impedance Inversion Based on Sub- Image and Convolutional Neural Network Application of Bi-	https://doi.org/10.39 97/2214- 4609.201901391	Li	х		EAGE	EAGE (annual)	Impedance		Supervised	CNN		Synthetic	Marmousi II	Synthetic	Post stack migrated		
2019	directional Long Short- Term Memory Recurrent Neural Network for Seismic Impedance Inversion	https://doi.org/10.39 97/2214- 4609.201901386	Guo		х	EAGE	EAGE (annual)	Impedance		Supervised	Multiple (CNN, RNN)		Real		Real	Post stack migrated		
2018	Geophysical inversion versus machine learning in inverse problems	https://doi.org/10.11 90/tle37120894.1	Kim	х		SEG	TLE	Reflectivity inversion		Supervised	DNN		Synthetic	Multiple	Both	Post stack migrated		
2019	seismic migration	90/INI-2018-0230.1	Vamaraju	x		SEG	Interpretation	Reflectivity inversion	RTM, Kirschhoff	Supervised	RNN		Synthetic	(SEG/EAGE overthrust model, Marmousi)	Synthetic	Synthetic		
2019	Seismic image processing through the generative adversarial network Reflectivity Inversion of	90/INT-2018-0232.1	Picetti	х		SEG	Interpretation	Reflectivity inversion		Semi- supervised	GAN		Synthetic	SEG/EAGE overthrust model	Synthetic	Synthetic		
2020	Nonstationary Seismic Data with Deep Learning Based Data Correction	97/2214- 4609.202011195	Gao	х		EAGE	EAGE (annual)	Reflectivity inversion		Supervised	DNN		Both		Both	Post stack migrated		
2019	Time-lapse seismic inversion for reservoir changes in the Norne Field, North Sea	https://doi.org/10.11 90/segam2019- 3216321.1	BaBalola	x		SEG	SEG Technical Program	4D	pressure, saturation, p- impedance, S- impedance	Supervised	MDN (mixture density NN)		Synthetic		Both	Attributes		
2018	Machine learning to reduce cycle time for time-lapse seismic data assimilation into reservoir management	https://doi.org/10.11 90/segam2018- 2997250.1	Xue		x	SEG	SEG Technical Program	4D	Water content	Supervised	Multiple (MLP, Random Forest)		Synthetic		Both	4D data		
2019	Machine learning to reduce cycle time for time-lapse seismic data assimilation into reservoir management	https://doi.org/10.11 90/INT-2018-0206.1	Zue		x	SEG	Interpretation	4D	Water content	Supervised	ANN		Synthetic		Both	Post stack migrated		
2020	Time-lapse seismic data inversion for estimating reservoir parameters	https://doi.org/10.11 90/segam2020- 3419749.1	Kaur	x		SEG	SEG Technical Program	4D	CO2 saturation	Semi- supervised	GAN		Synthetic		Synthetic	4D data		
2020	using deep learning Cross-equalization of time-lapse seismic data using recurrent neural networks	https://doi.org/10.11 90/segam2020- 3424773.1	Alali	х	x	SEG	SEG Technical Program	4D	All	Supervised	RNN-LSTM		Synthetic		Synthetic	Raw		
2020	Ground-truth uncertainty-aware metrics for machine learning applications on seismic image interpretation: Application to faults and	https://doi.org/10.11 90/tle39100734.1	Guillon		x	SEG	TLE	Multiple (Fault detection,Ho rizon tracking)		Supervised	CNN	U-Net	Real		Real	Post stack migrated		MSE
2018	horizon extraction Seismic-facies analysis based on deep-encoder clustering	https://doi.org/10.11 90/segam2018- 2997763.1	Duan	x		SEG	SEG Technical Program	hoFacies)	Seismic Section	Unsupervised	Encoder Deep network		REal		Real	Attributes	Amplitude spectra	Logloss
2018	Automated interpretation of top and base salt using deep- convolutional networks	https://doi.org/10.11 90/segam2018- 2996306.1	Gramstad		x	SEG	SEG Technical Program	Multiple (Salt/Geo- body identificatio n, Horizon tracking)		Supervised	CNN		Real		Real	Post stack migrated		

2018	Petrophysical-property estimation from seismic data using recurrent neural networks Supervised and	https://doi.org/10.11 90/segam2018- 2995752.1	Alfarraj	x	x		SEG	SEG Technical Program	multiple(Pet rophysical properties, Impedance)	PP: Density	Supervised	RNN-GRU		Real		Real	Post stack migrated		MSE
2015	unsupervised learning: how machines can assist quantitative seismic interpretation	https://doi.org/10.11 90/segam2015- 5924540.1	Zhao	x			SEG	SEG Technical Program	Petrophysica I properties)	Attributes without seismic section, PP: Brittleness	Supervised/U nsupervised	SOM		Real		Real	Attributes	Manually selected	
2015	Geologic pattern recognition from seismic attributes: Principal component analysis and self-organizing maps	https://doi.org/10.11 90/INT-2015-0037.1	Roden		x		SEG	Interpretation	Multiple (LithoFacies, DHI interpretatio n)	Attributes without seismic section	Unsupervised	SOM		Real		Real	Attributes	PCA	
2016	Analysis of porosity, stratigraphy, and structural delineation of a Brazilian carbonate field by machine learning techniques: A case study	https://doi.org/10.11 90/INT-2016-0024.1	Chaves Kuroda	x			SEG	Interpretation	Multiple (LithoFacies, Stratigraphic sequence, Petrophysica I properties)	Attributes without seismic section, PP: Porosity	Supervised	MLP		Real		Real	Attributes	Genetic Algorithm	MSE
2019	Prestack and poststack inversion using a physics- guided convolutional neural network	https://doi.org/10.11 90/INT-2018-0236.1	Biswas	х			SEG	Interpretation	multiple(Pet rophysical properties, Impedance)	Elastic, PP: Density	Supervised	CNN		Synthetic		Both	Angle stacks		
2020	Theory-guided data science-based reservoir prediction of a North Sea oil field Petrophysical properties	https://doi.org/10.11 90/tle39100742.1	Downton		x		SEG	TLE	multiple(Pet rophysical properties, Impedance)	Elastic, PP: Porosity	Supervised	DNN		Synthetic		Both	Post stack migrated		
2020	prediction from prestack seismic data using convolutional neural networks	https://doi.org/10.11 90/geo2019-0650.1	Das	х	х		SEG	Geophysics	multiple(Pet rophysical properties, Impedance)	Acoustic, PP: Porosity, Vclay	Supervised	CNN		Real		Real	Pre-stack migrated		
2020	Accelerating seismic fault and stratigraphy interpretation with deep CNNs: A case study of the Taranaki Basin, New Zealand	https://doi.org/10.11 90/tle39100727.1	Di		х		SEG	The Leading Edge	Multiple (Fault detection, Stratigraphic sequence)		Supervised	CNN		Real		Real	Post stack migrated		
2020	Uncertainty estimation in impedance inversion using Bayesian deep learning	https://doi.org/10.11 90/segam2020- 3428098.1	Choi	x	x		SEG	SEG Technical Program	multiple(Pet rophysical properties, Impedance) Multiple	Elastic, PP: Density	Supervised	Bayesian- CNN		Real		Real	Angle stacks		
2020	Seismic attribute selection for machine- learning-based facies analysis	https://doi.org/10.11 90/geo2019-0223.1	Qį	x			SEG	Geophysics	(Salt/Geo- body identificatio n, LithoFacies)	Attributes without seismic section	Unsupervised	GTM		Real		Real	Attributes	Gaussian mixture model (GMM)	
2020	Machine learning assisted seismic inversion	https://doi.org/10.11 90/segam2020- 3425889.1	Roy		x		SEG	SEG TECHNICAL PROGRAM	Multiple (DHI interpretatio n, Petrophysica I properties)	PP: Porosity	Supervised			Synthetic	SEAM	Both	Attributes	Gradient boosting	
2021	Estimating subsurface properties using a semi- supervised neural networks approach	https://doi.org/10.11 90/geo2021-0192.1	Di		x		SEG	Geophysics	multiple(Pet rophysical properties, Impedance) Multiple	PP: Porosity	Supervised/U nsupervised	CNN		Synthetic		Synthetic	Post stack migrated		MSE
2021	Interactively tracking seismic geobodies with a deep-learning flood- filling network	https://doi.org/10.11 90/geo2020-0042.1	Shi	x			SEG	Geophysics	(Fault detection,Sal t/Geo-body identificatio n)		Supervised	CNN	U-Net	Synthetic		Synthetic	Post stack migrated		
2021	Integrating deep learning insights for complex geological features: A case study from the Santos Basin, offshore Brazil	https://doi.org/10.11 90/segam2021- 3594718.1	Chenin		x		SEG	First International Meeting for Applied Geoscience & Energy	Multiple (Fault detection,Sal t/Geo-body identificatio n)		Supervised	CNN		Real		Real	Post-stack Depth Migrated		
2021	Multiparameter geophysical reservoir characterization augmented by generative networks	https://doi.org/10.11 90/segam2021- 3583673.1	Li		x		SEG	First International Meeting for Applied Geoscience & Energy	multiple(Pet rophysical properties, Impedance)	PP: VP/VS	Semi- supervised	GAN		Synthetic		Synthetic	AVO	AVO	
2021	Channel facies and faults multisegmentation in seismic volumes	https://doi.org/10.11 90/segam2021- 3584046.1	Pham	х			SEG	First International Meeting for Applied Geoscience & Energy	Multiple (Fault detection, LithoFacies) Multiple	Attributes without seismic section	Supervised	CNN		Real		Real			Dice
2021	3D seismic interpretation with deep learning: A brief introduction	https://doi.org/10.11 90/tle40070524.1	Wrona	х		х	SEG	πε	(Fault detection,Sal t/Geo-body identificatio n, Horizon tracking)		Supervised	CNN	U-Net	Real		Real	Post stack migrated		LOG
2021	Imposing interpretational constraints on a seismic interpretation convolutional neural network	https://doi.org/10.11 90/geo2020-0449.1	Di		x		SEG	Geophysics	Multiple (Fault	Seismic section	Supervised	CNN	U-Net	Real		Real	Post stack migrated		LOG
2021	An in-depth analysis of logarithmic data transformation and per- class normalization in machine learning: Application to unsupervised classification of a turbidite system in the Canterbury Basin, New Zealand, and supervised classification of sat in the Eugene Island minibasin, Guilf of Nexico	https://doi.org/10.11 90/INT-2021-0008.1	На	x	х		SEG	Interpretation	Multiple (Salt/Geo- body identificatio n, LithoFacies)	Attributes without seismic section	Supervised/U nsupervised	SOM		Real		Real	Post stack migrated	PCA	
2021	Using machine learning to detect horizons, alluvial fans and fluvial channels, as well as vertical faults	https://doi.org/10.39 97/1365- 2397.fb2021052	Sergey		x		SEG	First Break	Multiple (Fault detection,Ho rizon tracking)		Supervised			Real		Real			
2020	A New Era in Geoscience Data Interpretation - Artificial Intelligence Based Solutions	https://doi.org/10.39 97/2214- 4609.202011568	Gorlov		x		EAGE	EAGE (annual)	Multiple (Fault detection,Ho rizon tracking, LithoFacies)	Seismic section	Supervised	CNN	U-Net	Synthetic		Both	Post stack migrated		
2020	Uncertainty quantification for deep learning in geoscience applications	https://doi.org/10.39 97/2214- 4609.202075015	Mosser		x		EAGE	AAPG digital subsurface for Asia Pacific Conference	Multiple (Fault detection, Petrophysica I properties)	PP: VP/VS	Supervised	Bayesian- CNN		Real		Real	pre-stack migrated		
2019	Pseudo-Wells based HitCube 'trace-matching' and Machine Learning Inversions: Seismic Reservoir Characterization in a Challenging Environment	https://doi.org/10.39 97/2214- 4609.2019X610104	Kocsis		x		EAGE	Subsurface Intelligence workshop	multiple(Pet rophysical properties, Impedance)	PP: Porosity, V shale, VP/VS	Supervised	Random Forest		Synthetic		Synthetic	pre-stack migrated		
2020	3D Seismic Based Reservoir Property Prediction Using Deep Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.202011954	Purves		x		EAGE	EAGE (annual)	multiple(Pet rophysical properties, Impedance)	PP: Porosity, density, VP/VS, V clay	Supervised	CNN		Real		Real	Post stack migrated		

2020	Experiences with Machine Learning and Deep Learning Algorithms for Seismic, Wells and Seismic-to- Well Applications	https://doi.org/10.39 97/2214- 4609.202010990	Jaglan		х	EAG	E EAGE (annual)	Multiple (Fault detection, Petrophysica I properties)	PP: Chimney volume	Supervised	CNN	multiple (U- Net, Le-Net)	Both	Real	Post stack migrated		
2019	Combining Machine Learning and a Knowledge Base for Seismic Data Intelligence Expansion		Vital Brazil		x	X EAG	E EAGE (annual)	Multiple (Salt/Geo- body identificatio n, DHI interpretatio n)		Supervised	KNN		Real	Real	Post stack migrated		
2018	Application of Machine Learning Algorithms Using Seismic Data and Well Logs to Predict Reservoir Properties	https://doi.org/10.39 97/2214- 4609.201800920	priezzhev	x		EAG	E EAGE (annual)	multiple(Pet rophysical properties, Impedance)	PP: density	Supervised	Multiple (Random Forest, Linear regression)		Real	Real	Post stack migrated		
2010	Estimation of Reservoir Properties from Seismic Attributes and Well Log Data using Artificial Neural Networks	https://doi.org/10.39 97/2214- 4609.20145507	Sitouah	x		EAG	E EAGE (annual)	Multiple (LithoFacies, Petrophysica I properties)	Attributes with or without seismic section, PP: Porosity, permeability	Supervised	multiple (MLP, GRNN)		Real	Real			
2015	The Application of Probabilistic Neural Network Analysis Technique in the Prediction of Sedimentary Tuff Reservoir	https://doi.org/10.39 97/2214- 4609.201413381	u		x	EAG	E EAGE (annual)	(LithoFacies, Petrophysica	Attributes with or without seismic section, PP: Porosity	Supervised			Real				
2017	Reducing the risk of hydrocarbon exploration and lithology characterization — using a neural network approach in the West Delta Deep Marine Concession Area, Offshore Nile Delta, Egypt	97/1365- 2397.2017016	Abd- Elfattah	x	х	EAG	E First Break	Multiple (LithoFacies, Petrophysica I properties)		Supervised	PNN		Real	Real	Attributes	PCA	
2018	Sweet Spot Interpretation from Multiple Attributes: Machine Learning and Neural Networks Technologies	https://doi.org/10.11 90/IGC2018-387	Guo		х	SEC	International Geophysical Conference, Beijing, China	DHI interpretatio n		Unsupervised	Multiple (SOM, PCA, K means)	-	Real	Real	Attributes	PCA	
2020	3D relative geologic time estimation with deep learning	https://doi.org/10.11 90/segam2020- 3427388.1	Bi	x	x	SEC	SEG Technical Program	RGT volume		Supervised	CNN	U-Net	Synthetic	Synthetic	Post stack migrated		MSE
2021	A deep-learning method for latent space analysis of multiple seismic attributes	https://doi.org/10.11 90/INT-2020-0194.1	Wallet	х	x	SEC	Interpretation	Latent space of the attributes		Unsupervised	Auto- encoder		Real	Real	Attributes	Not specified	
2021	Automatic computation of relative geologic time volume using self- supervised learning	https://doi.org/10.11 90/segam2021- 3581832.1	Kim	x	x	SEC	First International Meeting for Applied Geoscience & Energy	RGT volume			Multiple				Partial stacked migrated		
2021	Evaluating proxies for the drivers of natural gas productivity using machine-learning models	https://doi.org/10.11 90/INT-2020-0200.1	Kumar	x	х	X SEC	Interpretation	Gas Productivity		Supervised	(Random Forest, Linear regression, DNN)		Real	Real	Attributes	Not specified	
2020	Prestack AVO Attributes Inversion Using Convolutional Neural Network Automatic Detection and	https://doi.org/10.39 97/2214- 4609.202011622	Sun	x		EAG	E EAGE (annual)	AVO inversion		Supervised	CNN		Synthetic	Both			MSE
2020	Classification of Unconformities on Seismic Data Using Machine Learning	https://doi.org/10.39 97/2214- 4609.202032084	Alberts	x	x	EAG	E Digitalization conference	Unconformit ies		Supervised	CNN		Synthetic	Both	Post stack migrated		
2020	Seismic Inversion Through the Application of Machine Learning Algorithms Principal Component	https://doi.org/10.39 97/2214- 4609.202053123	Smorodin	x		EAG	E SAINT PETERSBURG	Log estimation	Sonic and density	Supervised							
2020	Analysis and Deep Learning along Directional Image Gathers for High- Definition Classification of Subsurface Features	https://doi.org/10.39 97/2214- 4609.202032023	De Rebit		x	EAG	E Digitalization conference	Diffraction imaging									
2019	Traditional Feature Based vs Direct Machine Learning Based AVO Classification	https://doi.org/10.39 97/2214- 4609.201901148	Das	x		EAG	E EAGE (annual)	AVO classification		Supervised	CNN		Synthetic	Synthetic	pre-stack migrated		LOG
2019	Estimation of groundwater storage from seismic data using	https://doi.org/10.11 11/1365-2478.12831	Timo Lähivaara	х		EAG	E GP	Ground Water Storage		Supervised	CNN	AlexNET	Synthetic	Synthetic	Raw seismic Data		Normalized squared difference
2010	deep learning Carbonate Seismic Geopressure Prediction Using Neural Networks	https://doi.org/10.39 97/2214- 4609.201400796	AlbinHassa n	х		EAG	E EAGE (annual)	Geo-Pressure		Unsupervised	SOM		Real	Real			MSE
2013	Application of Probabilistic Neural Network on Complex Carbonate Reservoir	https://doi.org/10.25 23/IPTC-16952-MS	Sun	х	×	EAG	E IPTC	Risk mitigation (reserveoir		Supervised		user defined	Real	Real			
2017	Prediction Sweet Spot Interpretation from Multiple Attributes: Machine Learning and	https://doi.org/10.39 97/2214- 4609.201702564	Guo		x	EAG	E AMGE	prediction)  DHI interpretatio n		Unsupervised	SOM		Real	Real	Attributes	Manually selected	
2017	Neural Networks Technologies Interpretation of DHI characteristics with	https://doi.org/10.39 97/1365-	Roden		x	EAG	E First Break	DHI interpretatio		Unsupervised	SOM		Real	Real	Attributes	Manually selected	
2017	Machine learning  A growing machine learning approach to	2397.35.5.88069 https://doi.org/10.11	Hami-		,		CCC Technical Decrees	n Review article (Fault detection, Salt/Geo-								Science	
2017	optimize use of prestack and poststack seismic data	90/segam2017- 17767532.1	Eddine		х	SEC	SEG Technical Program	body identificatio n, AVO classification )									
2018	Machine learning in the interpreter's toolbox: Unsupervised, supervised, and deep- learning applications	https://doi.org/10.11 90/segam2018- 2997015.1	Bandura		х	SEC	SEG Technical Program	Review article									
2018	Successful leveraging of image processing and machine learning in seismic structural interpretation: A review	https://doi.org/10.11 90/tle37060451.1	Wang	х		SEC	πε	Review article (Fault detection, Salt/Geo- body identificatio n, Horizon tracking)		Supervised	CNN						
2018	Machine learning and learning from machines	https://doi.org/10.11 90/tle37120886.1	Naeini		x	SEC	TLE	Review article (Fault detection)									
2020	Domain Expertise, Deep Learning and Cloud: How to Build Powerful Workflows for Exploration	https://doi.org/10.39 97/2214- 4609.202032063	Vynnytska		x	EAG	E Digitalization conference	Review article									

2019	Drivers of imbalance in machine learning uptake across geology and geophysics	https://doi.org/10.39 97/1365-2397.n0056	Samuel R.	x	EAGE	First Break	Review article
2018	Is Machine Learning taking productivity in petroleum geoscience on a Moore's Law trajectory?	https://doi.org/10.39 97/1365-2397.n0145	Larsen	x	EAGE	First Break	Review article