Year	Title	Doi	First Author	Affiliation		Platform		Application					Type of ML	Method	Architecture	Type of training data	Type of data applied to:	Synthetic Model	Input Data	Loss Function
				Academia (Company Research center			Application 1	Application 1 note	Application 2	n Applicati 2 note	ion Application Applicati	on				Synthetic/Real		Raw, surface wave, FB, pre-stack and poststack migrated,	
							Journal/Conference/Workshop					211010							velocity model	
	Integrating seismic first- break picking methods with	https://doi.org/10.11							First break Erroneous										Raw seismic Data	
2018	a machine learning approach 3D seismic geometry quality control and corrections by	2998293.1		×		SEG	SEG Technical Program	QC	pick detection Geometry				Supervised	CNN		REal	Real		(Passive)	Log Loss
2019	applying machine learning Application of a convolutional neural	90/geo2018-0617.1 https://doi.org/10.11	Jiang	x	×	SEG	Geophysics	QC	error (source- receiver) Noise				Supervised	CNN		Synthetic	Both		(Zero-offset)	Logloss
2020	network to classification of swell noise attenuation Framework and standalone applications of machine	3425046.1	Farmani		x	SEG	SEG Technical Program International Meeting for	QC	Recognition (Swell Noise) Noise		Multipl		Supervised	CNN	U-Net	Real	Real		(Amplitude)	
2021	learning in seismic processing Automatic Spiky Trace	90/segam2021- 3580418.1 https://doi.org/10.39 97/2214-	Martin		×	SEG	Applied Geoscience & Energy EAGE Annual	QC C	recognition Anomalous Trace	Denoising	(Linear Multiple Groundre	es,	Supervised	CNN	U-NET	Real	Real		Post-stack migrated	
	Removal Using Artificial Neural Network Improving quality control and data understanding of a large OBN survey through	4609.202132017 https://doi.org/10.39	rma					QC	detection First break Erroneous				Supervised							
2020	unsupervised machine learning Visual Identification of Noisy	4609.2020611028 https://doi.org/10.39	s		×	EAGE	EAGE Annual	ОС	pick detection Noise				ed	means/PCA)		Real	Real			
2020	Seismic Records with Machine Learning Application of Convolutional Neural	4609.202011503 https://doi.org/10.39	Walpole		×	EAGE	EAGE Annual	QC QC	recognition Noise				Supervised	CNN	U-NET	Real	Real		Raw seismic Data	
2020	Network in Automated Swell Noise Attenuation Application of unsupervised	4609.202010329			^	EAGE		ų.	recognition				,.		UNEI	Real	Real		(Amplitude)	
2019	machine learning to the processing of a land mega- survey Leveraging a Supervised	97/2214- 4609.2019X6104	Hou		×	EAGE	Subsurface intelligence workshop	QC	Trace detection	Anisotropy	′		Unsupervis ed	Multiple(K means/PCA)		Real	Real		Raw seismic Data (Amplitude)	
2019	Machine Learning Toolkit For Better Seismic Processing Quality Control	https://doi.org/10.39 97/2214- 4609.201901618	Chambef ort	х	x	EAGE	EAGE Annual	QC	Multiple (cycle skipping potential, rig noise, data reduction)				Supervised	KNN		Real	Real		Raw seismic Data (Amplitude)	
2019	Automatic QC of denoise processing using a machine learning classification	https://doi.org/10.39 97/1365-2397.n0055				EAGE	First Break	QC	Anomalous Trace detection											
2018	based on deep residual networks	https://doi.org/10.11 90/IGC2018-435	Zhang	x	x	SEG	International Geophysical Conference, Beijing, China	Denoising	Random				Supervised	Residual NN		real	Real		Migrated	
2018	Deep learning for denoising	https://doi.org/10.11 90/IGC2018-113	Yu	x		SEG	INTERNATIONAL GEOPHYSICAL CONFERENCE	Denoising	multiple (random, linear, multiples)				Supervised	CNN		Synthetic	Both		Raw/post-stack	
2018	Separating ground-roll from land seismic record via convolutional neural network	https://doi.org/10.11 90/AIML2018-16.1	Jia	х	x	SEG	SEG Maximizing Asset Value Through Artificial Intelligence and Machine Learning, Beijing, China	Denoising	Groundroll				Supervised	CNN		Synthetic	Both		Raw seismic Data (Amplitude)	
2019	Ground-roll noise attenuation based on convolutional neural network Generative-adversarial	https://doi.org/10.11 90/frur2019_18.1	Li		x	SEG	Workshop: Fractured Reservoir & Unconventional Resources Forum: Prospects and Challenges in the Era of Big Data, Lanzhou, China,	Denoising	Groundroll				Supervised	Residual NN		Both	Both		Raw seismic Data (Amplitude)	
2018	network-based fast-noise removal on land-seismic data Automated ambient-noise	https://doi.org/10.11 90/segam2018- 2995310.1	Xie		x	SEG	SEG Technical Program	Denoising	Random				Semi- supervised	GAN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2018	processing applied to fiber- optic seismic acquisition (DAS) Noise attenuation for	https://doi.org/10.11 90/segam2018- 2997880.1	Huot	х		SEG	SEG Technical Program	Denoising	Random (DAS)				Supervise/U nsupervised			Real	Real		Raw seismic Data (Amplitude)	
2018	seismic image using a deep- residual learning Random-noise suppression	https://doi.org/10.11 90/segam2018- 2997974.1 https://doi.org/10.11	Zhang	x		SEG	SEG Technical Program	Denoising	Random				Supervised	CNN		Synthetic	Both		Raw seismic Data (Amplitude)	MSE
2018	in seismic data: What can deep learning do? Seismic random noise attenuation in f-x domain using complex-valued	90/segam2018- 2998114.1 https://doi.org/10.11 90/segam2019-	Liu	x		SEG SEG	SEG Technical Program SEG Technical Program	Denoising	Random				Supervised	DnCNN	ComplexNet	Synthetic	Synthetic		f-X	
2018	residual convolutional neural network Seismic data denoising by deep-residual networks	3216543.1 https://doi.org/10.11 90/segam2018-	Jin	×	×	SEG	SEG Technical Program	Denoising	Random				Supervised	DRN (deep residual		Real	Real		Raw seismic Data	MSE
2018	Deep learning for ground- roll noise attenuation	2998619.1 https://doi.org/10.11 90/segam2018- 2981295.1	Li		x	SEG	SEG Technical Program	Denoising	Groundroll				Supervised	network) CNN	dnCNN	Real	Real		(Amplitude) Raw seismic Data (Amplitude)	MSE
2019	Learning seismic image enhancement from pairs of 3D partial and full image volumes	https://doi.org/10.11 90/segam2019- 3200259.1	Wang		x	SEG	SEG Technical Program	Denoising	Enhanced stacking				Supervised	CNN	U-Net	Synthetic	Synthetic	SEAM	Pre-stack migrated	
2019	Swell-noise attenuation: A deep learning approach Applying machine learning	https://doi.org/10.11 90/tle38120934.1	Zhao	x	x	SEG	TLE	Denoising	Swell multiple				Supervised	CNN	Only2Noise	Synthetic	Both		Raw seismic Data (Amplitude)	MSE
2019	to 3D seismic image denoising and enhancement White noise attenuation of seismic trace by integrating	https://doi.org/10.11 90/INT-2018-0224.1	Wang		x	SEG	Interpretation	Denoising	(random, coherent, migration)				Supervised	CNN	U-net	Real	Synthetic		Post-stack migrated	
2019	variational mode decomposition with convolutional neural network	https://doi.org/10.11 90/geo2018-0635.1		x	x	SEG	Geophysics	Denoising	Random				Supervised	CNN		Real	Real		Post-stack migrated	
2020	Ground-roll attenuation using generative adversarial networks Deep denoising autoencoder	https://doi.org/10.11 90/geo2019-0414.1	Yuan	x		SEG	Geophysics	Denoising	Groundroll				Supervised	GAN Denoising		Synthetic	Both		Raw seismic Data (Amplitude)	
2020	for seismic random noise attenuation Ground roll attenuation	https://doi.org/10.11 90/geo2019-0468.1	M. Saad	x		SEG	Geophysics	Denoising	Random				Unsupervis ed	auto- encoder (DAE)		Synthetic	Both		Post-stack migrated	
2020	based on conditional and cycle generative adversarial networks	90/iwmg2019_23.1	31	х		SEG	SEG 2019 Workshop: Mathematical Geophysics: Traditional vs Learning, Beijing	Denoising	Groundroll				Supervised	cGAN		Both	Both		Raw seismic Data (Amplitude)	
2020	Must we have labels for denoising seismic data based on deep learning? Separating primaries and multiples using hyperbolic	https://doi.org/10.11		х	×	SEG	SEG Technical Program	Denoising	Random				Unsupervis ed Semi-	Auto- encoder	U-net	Real	Real		Pre-stack migrated Raw seismic Data	MSE
2020	Radon transform with deep learning Deep learning for simultaneous seismic image	90/segam2020- 3419762.1 https://doi.org/10.11	Kaur	х		SEG	SEG Technical Program	Denoising	reflection				supervised	GAN					(Amplitude)	multiple
2020	super-resolution and denoising Ground roll attenuation	90/segam2020- 3426412.1 https://doi.org/10.11	Li	x	,	SEG	SEG Technical Program	Denoising	Random				Supervised Unsupervis	CNN Auto-	U-net	Synthetic	Both		Post-stack migrated Raw seismic Data	(L1+MS- SSIM)
2020	with an unsupervised deep learning approach Physics-guided self- supervised learning for	90/segam2020- 3425792.1 https://doi.org/10.11 90/segam2021-	Guo Zi	x	x	SEG	SEG Technical Program International Meeting for	Denoising	Groundroll				ed	encoder	multiple (U-	Real	Real	BP 2004	(Amplitude) Raw seismic Data	MSE
2021	monochromatic noise removal Deep learning-based seismic surface-related multiple adaptive subtraction with	3583068.1 https://doi.org/10.11 90/segam2021-		х		SEG	Applied Geoscience & Energy International Meeting for Applied Geography	Denoising	Multiple				Supervised	CNN	NET, ResNet) U-NET	Synthetic	Both		(Amplitude) Raw seismic Data	
2021	complete and representative training of neural networks: A generalization study using	3584041.1 https://doi.org/10.11 90/geo2020-0193.1	Zhana	x		SEG	Applied Geoscience & Energy Geophysics	Denoising	reflection				Supervised	CNN	U-NET	Synthetic	Both		(Amplitude) Raw seismic Data (Amplitude)	MSE
2021	double noise injection and natural images An innovative strategy for seismic swell noise removal	https://doi.org/10.11				SEG	International Meeting for	Denoising	Seismic interference				Supervised	CNN	U-NET				Raw seismic Data	
2021	using deep neural networks Seismic noise attenuation by applying a deep learning method without noise-free	3592770.1		x		SEG	Applied Geoscience & Energy International Meeting for Applied Geoscience & Energy	Denoising	(Marine)				Supervised	CNN	-	Both	Both		(Amplitude) Raw seismic Data (Amplitude)	MSE
2021	labels Physics-constrained deep learning for ground-roll	https://doi.org/10.11 90/segam2021-	Pham		x	SEG	International Meeting for	Denoising	Groundroll				Supervised/ Unsupervis	CNN					-	Logloss
2021	attenuation Seismic noise attenuation by signal reconstruction: an unsupervised machine	3583447.1	Con	х	x	EAGE	Applied Geoscience & Energy Geophysical Prospecting	Denoising	Random				ed Unsupervis ed	Not specified						- "
	learning approach																			

	Adaptive subtraction using a	https://doi.org/10.39						** ***										Raw seismic Data	multiplie(L1
2021	convolutional neural network	97/1365- 2397.fb2021066 https://doi.org/10.39	Kumar		X EA		Denoising	Multiple reflection				Supervised	CNN	U-NET	Both	Real	Sigsbee	Raw seismic Data (Amplitude)	norm, L2 norm)
2020	Attenuating Random Noises in DAS Seismic Seismic Random Noise	97/2214- 4609.202010721 https://doi.ore/10.39	Zhang		X EA	GE EAGE Annual	Denoising	(DAS)				Supervised	CNN						MSE
2020	Unsupervised Sparse Machine Learning Random Noise Attenuation	97/2214- 4609.202010042	Gao	х	X EA	GE EAGE Annual	Denoising	Random				Unsupervis ed			Real	Real		Post-stack migrated	
2020	for Desert Seismic Data Using the Complex Diffusion Coupled with Deep Learning	97/2214-	Zhang	x	EA	GE EAGE Annual	Denoising	Random				Supervised	CNN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2020		https://doi.org/10.11 11/1365-2478.12985	Kaur	x	EA	GE Geophysical Prospecting	Denoising	Groundroll										Raw seismic Data (Amplitude)	multiple (Adversial, cyclic, self distance,
2020	Deep Learning for Migration Artifacts Attenuation	https://doi.org/10.39 97/2214- 4609.202011932	Klochikhi na		X EA	GE First Break	Denoising	Migration induced				Supervised	CNN	U-NET	Synthetic	Both		Post-stack migrated	identity) L2 norm
2020	Leveraging deep learning for seismic image denoising	https://doi.org/10.39 97/1365- 2397.fb2020048	Klochikhi na		X EA	GE First Break	Denoising	Migration induced				Supervised	CNN	U-NET	Synthetic	Both		Post-stack migrated	L2 norm
2020	An Adaptive Anomalous Amplitude Attenuation Method Based on Deep Neural Network	https://doi.org/10.39 97/2214- 4609.202011320	Tian	х	EA	GE EAGE Annual	Denoising	Anomalous amplitude				Supervised	FCN	VGG 16	Synthetic	Both		Raw seismic Data (Amplitude)	Logloss
2020	Multiples Elimination with Denoising Convolutional Neural Networks: A Case Study in South China Sea	https://doi.org/10.39 97/2214- 4609.202010256	Ye	х	X EA	GE EAGE Annual	Denoising	Multiple reflection				Supervised	CNN	user defined	Real	Real		Raw seismic Data (Amplitude)	MSE
2020	Edge-Aware Image Conditioning with a Siamese Neural Network Ground Roll Suppression		Aharchao u		X EA	GE EAGE Annual	Denoising	Enhanced stacking				Supervised	CNN	VGG 16	Real			Raw seismic Data (Passive)	Xontrastive Loss
2020	Using Convolutional Neural Networks Seismic High Amplitude	97/2214- 4609.202011650 https://doi.org/10.39	Oliviera		X EA	GE EAGE Annual	Denoising	Groundroll				Supervised	GAN	Pix2Pix	Both	Both		Raw seismic Data (Amplitude)	MSE Normalized
2019	Noise Attenuation Based on the Deep Learning Method Deep Learning Application	97/2214- 4609.201901356 https://doi.org/10.39	Zhu	x	EA	Conference on Bassachia	Denoising	Random				Supervised	CNN		Real	Real		(Amplitude)	squared difference
2019	in Time-Frequency Analysis for Noise Attenuation Seismic Image Denoising Using Convolutional Neural	97/2214- 4609.201977037 https://doi.org/10.39	Hamidi	x	EA EA	Geoscience	Denoising	Random				Supervised	CNN	U-NET	Synthetic	Synthetic			MSE
2019	Network with Residual Learning Approach Using Convolutional Neural Networks for Denoising and	97/2214- 4609.201900851 https://doi.org/10.39	Wu	х	EA	GE EAGE Annual	Denoising	Random				Supervised	CNN-Resnet		Both	Real		Post-stack migrated	L1 norm
2019	Deblending of Marine Seismic Data Deep Learning for	97/2214- 4609.201900844	Slang	x	X EA	GE EAGE Annual	Denoising	interference (Marine) Multiple	Event separation	deblending		Supervised	CNN		Real	Real		Raw seismic Data (Amplitude)	
2018	Attenuating Random and Coherence Noise Simultaneously	https://doi.org/10.39 97/2214- 4609.201800939	Yu	x	EA	GE EAGE Annual	Denoising	(cycle skipping potential, rig noise, data reduction)				Supervised	CNN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2018	PDF SCoherent Linear Noises Attenuation From 3D Seismic Data Using Artificial Neural Network: Application To Algerian Sahara	https://doi.org/10.39 97/2214- 4609.201802176	Ouadfeul	х	EA	GE ECMOR XVI	Denoising	multiple (linear, groundroll)				Supervised	MLP	user defined	Real	Real		Raw seismic Data (Amplitude)	MSE
2019	Deep learning for denoising	https://doi.org/10.11 90/geo2018-0668.1	Yu	x	SI	G Geophysics	Denoising	multiple (random, linear, multiples,				Supervised	CNN		Synthetic	Synthetic			multiple (MSE, Normalized squared
2014	How to Teach a Neural Network to Identify Seismic Interference	https://doi.org/10.39 97/2214-	Rentsch	x	X EA	GE EAGE Annual	Denoising	groundroll) Seismic interference				Supervised	MLP		Synthetic	Real		Raw seismic Data (Amplitude)	difference)
2018	Random Noise Attenuation Using Convolutional Neural Networks	4609.20141445 https://doi.org/10.39 97/2214- 4609.201801390	Liu	х	EA	GE EAGE Annual	Denoising	(Marine) Random				Supervised	CNN	U-NET	Synthetic	Synthetic		Raw seismic Data (Amplitude)	L1 norm
2018	Generative Adversarial	https://doi.org/10.11 90/AIML2018-11.1	Chang		x si	G Workshop: SEG Maximizing Asset Value Through Artificial Intelligence and Machine Learning, Beijing, China	Trace Interpolation					Semi- supervised	GAN		Real	Real		Migrated	
2019	Interpolation of regularly sampled prestack seismic data with self-supervised learning	https://doi.org/10.11 90/segam2019- 3215774.1	Sen		X SI		Trace Interpolation					Self- learning	Variational auto encoder		real	Real		Raw seismic Data (Amplitude)	
2018	Generative adversarial networks in seismic data processing	https://doi.org/10.11 90/segam2018- 2996002.1	Alwon		X SI	G SEG Technical Program	Trace Interpolation		Denoising	Random		Semi- supervised	GAN						
2018	Seismic data interpolation through convolutional autoencoder What can machine learning	https://doi.org/10.11 90/segam2018- 2995428.1	Mandelli	х	Si	G SEG Technical Program	Trace Interpolation					Supervised	CNN	U-Net	Real	Real		Raw seismic Data (Amplitude)	MSE
2017	processing? An interpolation application	https://doi.org/10.11 90/geo2016-0300.1	Jia	х	SI	G Geophysics	Trace Interpolation					Supervised	SVR		Synthetic	Both		Raw seismic Data (Amplitude)	
2019	Deep-learning-based seismic data interpolation: A preliminary result Seismic trace interpolation	https://doi.org/10.11 90/geo2017-0495.1	*******	х	SI	G Geophysics	Trace Interpolation					Supervised	CNN	ResNet					
2020	sampled data using convolutional autoencoder	https://doi.org/10.11 90/geo2018-0699.1												HUJIVEE	Synthetic	Both		Raw seismic Data (Amplitude)	
2020	image denoising be used for		wang	х	SI	G Geophysics	Trace Interpolation					Supervised	Auto- encoder	NEAVEL	Synthetic	Both		(Amplitude) Raw seismic Data (Amplitude)	
2020	seismic data interpolation? De-aliasing using the U-Net		Zhang	x	SI	G Geophysics	Trace Interpolation					Supervised	encoder		Synthetic Synthetic	Both		(Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	
	De-aliasing using the U-Net image segmentation algorithm Seismic data reconstruction based on super resolution	90/geo2019-0243.1 https://doi.org/10.11 90/segam2020- 3425878.1	Zhang Yvas	x		G Geophysics G SEG Technical Program SEG 2020 Workshop:	Trace Interpolation Trace Interpolation						CNN CNN SRCNN	U-Net	Synthetic	Both		(Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	
2020	De-aliasing using the U-Net image segmentation algorithm Seismic data reconstruction based on super resolution convolutional neural network Crossline interpolation with	90/geo2019-0243.1 https://doi.org/10.11 90/segam2020- 3425878.1 https://doi.org/10.11 90/bwds2020_11.1	Zhang Yvas Jun	x x	SI	G Geophysics G SEG Technical Program SEG 2020 Workshop:	Trace Interpolation Trace Interpolation Trace					Supervised	encoder CNN CNN		Synthetic Synthetic	Both		(Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data	MSE
2020	De-aliasing using the U-Net image segmentation algorithm Seismic data reconstruction based on super resolution convolutional neural network Crossline interpolation with the traces-to-trace approach using machine learning	90/geo2019-0243.1 https://doi.org/10.11 90/segam2020- 3425878.1 https://doi.org/10.11 90/bwds2020_11.1	Zhang Yvas Jun	x x x	SI X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide-azimuth Deepwater Seismic technilogy, Beijing, China	Trace Interpolation Trace Interpolation Trace					Supervised Supervised	CNN CNN SRCNN (Super resolution CNN)	U-Net	Synthetic Synthetic Synthetic	Both Both	SEAM	(Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	
	De aliasing using the U-Net image sigmentation algorithm Seismic data reconstruction based on super resolution convolutional neural network Crossline interpolation with the tracest or trace approach using machine learning Reconstruction of Irregula Missing Seismic Data Using Conditional Generative Adversarial Networks	90/geo2019-0243.1 https://doi.org/10.11 90/segam2020- 3425878.1 https://doi.org/10.11 90/bwds2020_11.1 https://doi.org/10.11 90/segam2020- 3428348.1	Zhang Yvas Jun Yeeh	x x x	X SI	G Geophydcs G SEG Technical Program SEG 2020 Wortschop: Broadband and Wide-aimuth Deepwater Seincit echnilogy, Beijing, China G SEG Technical Program	Interpolation Trace Interpolation Trace Interpolation Trace Interpolation Trace Interpolation					Supervised Supervised	CNN CNN SRCNN (Super resolution CNN)		Synthetic Synthetic Synthetic Real	Both Both Real	SEAM	(Amplitude) Raw seismic Data (Amplitude)	MSE multiple (L1 norm.Advers isl)
2020	De alisang using the U-Net image segmentation algorithm Seismic data reconstruction based on super resolution convolutional neural network Crossline interpolation with the traces to-trace approach using machine learning Reconstruction of irregular Missing Seismic Data Using Conditional Generalive Conditional Generalive Conditional Generalive Conditional Generalive Conditional Conference Conditional Conference Conferen	90/geo2019-0243.1 https://doi.org/10.11 90/segam2020- 3425878.1 https://doi.org/10.11 90/bwds2020_11.1 https://doi.org/10.11 90/segam2020- 3428348.1 https://doi.org/10.11	Zhang Yvas Jun Yeeh Wei	x x x x	SI X SI SI	G Geophysics G SEG Technical Program SEG 2030 Workshop: Broadband and Wide simuth Deepwater Seismic technilogy. Beijing, China G SEG Technical Program G Geophysics	Interpolation Trace Interpolation Trace Interpolation Trace Interpolation Trace Interpolation Trace Interpolation					Supervised Supervised Supervised	CNN CNN SRCNN (Super resolution CNN) RNN-LSTM	U-Net multiple (U-NET,	Synthetic Synthetic Synthetic Real	Both Both Real	SEAM	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm,Advers
2020	De aliasing using the U-Net image segmentation algorithm Seismid data reconstruction based on super resolution convolutional neural network Crossline interpolation with the traces to trace approach using machine learning. Reconstruction of irregular Mixing Seismic Data Using Conditional Generative Adversarial Networks Deep seismic plant bring data using convolutional of seismic activations of the conditional convolutional of the conditional conditional demonstruction of seismic selection of training data sets to reconstruct seismic data using convolutional used to reconstruct seismic data using convolutional data using a convolutional of the conditional data using a convolutional of the control of th	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020- 3425878.1 https://doi.org/10.11 90/gedm2020_11.1 190/gedm2020_11.1 90/gedm2020- 3428348.1 https://doi.org/10.11 90/geo2020-05644.1	Zhang Yvas Jun Yeeh Wei	x x x x x	SI X SI X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide saimuth Deepwater Seismic technilogy, Beijing, China G SEG Technical Program G Geophysics G Geophysics	Interpolation Trace Interpolation					Supervised Supervised Supervised Supervised	CNN CNN SRCNN (Super resolution CNN) RNN-LSTM GAN	U-Net multiple (U-NET, Pix2Pix)	Synthetic Synthetic Real Synthetic Synthetic	Both Both Real Synthetic	SEAM	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm,Advers
2020 2021 2021	De alissing using the U-Net image segmentation algorithm algorithm algorithm algorithm and the segment of the s	90/geo2019-0243.1 https://doi.org/10.11 90/gepm2020- 3425878.1 90/gepm2020- 11.1 90/gepm2020- 3425848.1 https://doi.org/10.11 90/gepm2020- 3425848.1 https://doi.org/10.11 90/gep02020-0564.1 https://doi.org/10.11 90/geo2020-0564.1	Zhang Yvas Jun Yeeh Wei Liu Park	x x	SI X SI SI X SI	G Geophysics G SEG Technical Program SEG 2030 Workshop: Broadband and Wide saimuth Deepwater Seintire Inchnitory, Berjing, China G SEG Technical Program G Geophysics G Geophysics G Geophysics	Interpolation Trace Interpolation					Supervised Supervised Supervised Supervised Supervised Supervised	CNN CNN SRCNN (Super resolution CNN) RNN-LSTM GAN CNN	U-Net multiple (U-NET, Pix2 Pix) U-NET	Synthetic Synthetic Real Synthetic Synthetic Synthetic Synthetic	Both Both Real Synthetic Both	SEAM	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm,Advers
2021 2021 2021	De alisang using the U-Net image segmentation algorithm algorithm algorithm algorithm community of the properties of the	90/geo2019-0243.1 https://doi.org/10.11 90/geo2019-0243.1 https://doi.org/10.11 90/geo2019-0213.1 https://doi.org/10.11 90/geo2019-0301.1 https://doi.org/10.11 90/geo2019-0301.1 https://doi.org/10.11 90/geo2019-0301.1 https://doi.org/10.11 90/geo2019-0301.1 https://doi.org/10.11	Zhang Yvas Jun Yeeh Wei Liu Park Fang	x x x	X SI X SI X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide azimuth Deepwater Service technilogy, Beijing, China G SEG Technical Program Geophysics G Geophysics G Geophysics G Geophysics G Geophysics	Interpolation Trace Interpolation					Supervised Supervised Supervised Supervised Supervised Supervised	encoder CNN CNN SRCNN (Super resolution CNN) RNN-LSTM GAN CNN	U-Net multiple (U-NET, Pis.2Pis.) U-NET	Synthetic Synthetic Real Synthetic Synthetic Synthetic Synthetic Synthetic	Both Both Real Synthetic Both Both		(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm,Advers ial)
2020 2021 2021 2021 2021	De alissing using the U-Net image segmentation algorithm selection algorithm algorithm and the selection and the selection algorithm and the selection a	90/geo2019-0243.1 https://doi.org/10.11 90/gegam2020-03425878.1 https://doi.org/10.11 90/gewam2020-03425843.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1	Zhang Zhang Ywas Jun Yeeh Wei Liu Park Fang Fang Yuan	x x x	X SI X SI X SI X SI X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide azimuth Deepwards Scient technilogy. Belge, Chila G SEG Technical Program G Geophysics G SEG Technical Program	Interpolation Trace Interpolation					Supervised Supervised Supervised Supervised Supervised Supervised Supervised Supervised	encoder CNN CNN SRCNN (Super resolution CNN) GAN CNN CNN CNN CNN	U-Net multiple (U-NET, Pis.2Pis.) U-NET	Synthetic Synthetic Real Synthetic Synthetic Synthetic Synthetic Synthetic Both	Both Both Real Synthetic Both Both Both		(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm.Advers isl)
2020 2021 2021 2021 2021 2021	De alissing using the U-Net image segmentation algorithm seismind data reconstruction based on super resolution common and the seismind data reconstruction based on super resolution metwork. Crossline interpolation with the traces to trace approach using machine learning in the learning and the learning the seismind of irregulate Missing seisming Data Bulley Conditional Generative Adversarial Networks. Deep-seismin prior-based Adversarial Networks. Deep-seismin prior-based reconstruction of seismin data using convolutional United Section of training data sets to reconstruction of seismin data using a convolutional United Section of training data sets to reconstruct seismin data using a convolutional United Sections of training data sets to reconstruct seismin data using a convolutional United Sections of training data sets to reconstruct seismin data using a deep-learning-based prediction-cross filter interpolation in the post his familialing promises. Sesmin data interpolation Middlilling promises.	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020- 3425878.1 https://doi.org/10.11 90/gegm2020- 3425848.1 https://doi.org/10.11 90/gegm2020- 3425848.1 https://doi.org/10.11 90/gegm2020-05644.1 90/geo2019-0570.1 90/geo2019-0570.1 190/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-051.1	Zhang Yvas Jun Yeeh Wei Liu Park Fang Fang Hou	x x x	SI X SI SI X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide saimuth Deepwater Seismic technilogy, Beijing, China G SEG Technical Program G Geophysics G Geophysics G Geophysics G Geophysics G Geophysics G SEG Technical Program	Interpolation Trace Interpolation		Denoting	Random	Event deblending	Supervised Supervised Supervised Supervised Supervised Supervised Supervised Supervised Unsupervised	encoder CNN CNN SRCNN (Super reciolation of cont) GAN CNN CNN CNN CNN Not	U-Net multiple (U-NET, Pis.2Pis.) U-NET	Synthetic Synthetic Real Synthetic Synthetic Synthetic Synthetic Both Synthetic	Both Both Real Synthetic Both Both Both Both	Marmousi SEAM	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm.Advers isl)
2020 2021 2021 2021 2021 2021 2021	De aliasing using the U-Net image segmentation algorithm algorithm algorithm algorithm and algorithm a	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020-03425878.1 https://doi.org/10.11 90/gem2020-03425878.1 https://doi.org/10.11 90/gem2020-03425878.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0515.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1 https://doi.org/10.11 90/geo2019-0615.1	Zhang Yvas Jun Yeeh Wei Liu Park Fang Fang Yuan	x x x	X SI	G Geophysics G SEG Technical Program SEG 2010 Workshop: Broadband and Wide azimuth Deepwater Search ice thinlings, Beijing, China G SEG Technical Program G Geophysics G G G G G G G G G G G G G G G G G G G	Interpolation Trace Interpolation		Denoting	Random		Supervised	encoder CNN CNN SRCNN SRCNN (Suppe resolution CNN) GAN CNN CNN CNN CNN Not specified	U-Net multiple (U-NET, PUZPIs) U-NET	Synthetic Synthetic Synthetic Real Synthetic Real Synthetic Both Synthetic Synthetic	Both Both Both Both Real Synthetic Both Both Soth Soth	Marmousi	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm,Advers ist) Log loss
2020 2021 2021 2021 2021 2021 2021	De alissing using the U-Net image segmentation algorithm	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020-03425878.1 https://doi.org/10.11 90/gegm2020-03425845.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 11/13/65-2478-13055 https://doi.org/10.11	Zhang Yeas Jun Yeeh Wei Liu Park Fang Fang Yuan Hou Kaur Kuijpers	x x x x x x	X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide azimuth Deepwater Science it chinilogy, Bejing, China G SEG Technical Program G Geophysics G G Geophysics G G Geophysics G G Geophysics G G G G G G G G G G G G G G G G G G G	Interpolation Trace Interpolation		Denoting	Random		Supervised	encoder CNN SRCNN GSUPEr GAN CNN CNN CNN CNN CNN CNN CNN	U-Net multiple (U-NET, Piv2Piu) U-NET U-NET U-NET	Synthetic Synthetic Synthetic Real Synthetic Synthetic Synthetic Both Synthetic Synthetic Real	Both Both Real Synthetic Both Both Both Synthetic Real	Marmousi SEAM Marmousi/8	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm.Advers isl) Logloss MSE multiple (Adversia), cyclic)
2020 2021 2021 2021 2021 2021 2021 2021	De aliasing using the U-Net image segmentation algorithm seismind data reconstruction based on super resolution based on super resolution common and the season of the sea	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020- 3425878.1 https://doi.org/10.11 90/gegm2020- 3425878.1 https://doi.org/10.11 90/gegm2020- 90/gegm2020- 90/gegm2020- 90/gegm2020- https://doi.org/10.11 90/geo2019-0570.1 190/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0515.1 https://doi.org/10.11 90/gegm2011- 3584206.1 https://doi.org/10.11 11/165-2478.0 https://doi.org/10.11 11/1365-2478.0 https://doi.org/10.11 11/1365-2478.0 https://doi.org/10.11 11/1365-2478.0 https://doi.org/10.11 11/1365-2478.0 https://doi.org/10.11 11/1365-2478.0 https://doi.org/10.11 90/gegm2011- 35801211 https://doi.org/10.11 90/gegm2011- 35801211 https://doi.org/10.11 90/gegm2011- 4609.20011046	Zhang Yeas Jun Yeeh Wei Liu Park Fang Fang Yuan Hou Kaur Kuijpers	x x x x x x	X SI	Geophysics	Interpolation Trace Interpolation		Denotising	Bandom Bandom Bandom		Supervised	encoder CNN CNN Signer Gan CNN CNN CNN CNN CNN CNN CNN C	U-Net multiple (U-NET, Pis2Pis) U-NET U-NET	Synthetic Synthetic Real Synthetic Real Synthetic Both Synthetic Both Synthetic	Both Both Feal Synthetic Both Both Both Both Both Both Both Both	Marmousi/B SEAM Marmousi/B P 2004	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm. Advers in) Logioss MSE multiple (L1 norm. Adversia), cyclic)
2020 2021 2021 2021 2021 2021 2021 2021	De aliasing using the U-Net image segmentation algorithm seismine data reconstruction based on super resolution to based on super resolution convolutions. The second seco	90/geo2019-0243.1 https://doi.org/10.11 90/gegm2020-03425878.1 https://doi.org/10.11 90/gegm2020-03425878.1 https://doi.org/10.11 90/gegm2020-0342584.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-0570.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 90/geo2019-051.1 https://doi.org/10.11 11/13/65-2478.10055 https://doi.org/10.11 11/13/65-2478.10055 https://doi.org/10.11 11/13/65-2478.10055 https://doi.org/10.11 11/13/65-2478.10055 https://doi.org/10.11	Zhang Yvas Jun Yeeh Wei Liu Park Fang Fang Yuan Hou Kaur Kuijpers Oocharen ko Garg	x x x x x x x x	X SI	G Geophysics G SEG Technical Program SEG 2020 Workshop: Broadband and Wide azimuth Deepwater Seintic technilogy, Belging, China G SEG Technical Program G Geophysics G G G G G G G G G G G G G G G G G G G	Interpolation Trace Interpolation					Supervised	encoder CNN CNN SRCNN SSCNN SUper resolution CNN GAN CNN CNN CNN CNN CNN CNN CNN CNN RNN-STM CNN CNN Not specified CNN RNN RNN RNN RNN RNN RNN RNN RNN RNN	U-Net multiple (U-NET, Fix2Fix) U-NET U-NET U-NET U-NET GenNet	Synthetic Synthetic Real Synthetic Synthetic Synthetic Both Synthetic Synthetic Synthetic Real Synthetic Real Synthetic Real Synthetic	Both Both Both Real Synthetic Both Both Soth Both Both Both Both	Marmousi SEAM Marmousi/8	(Amplitude) Raw seismic Data (Amplitude)	multiple (L1 norm, Advers lat) Logloss MSE multiple (Adversial, cyclic) multiple (L1 norm,

2018	A Quantitative Comparison of Two Convolutional Neural Network Architectures - Seismic Data Interpolation as Example	https://doi.org/10.39 97/2214- 4609.201801395	Wang	x		EAGE	EAGE Annual	Trace Interpolation			Superi	ised CNN	user defined	Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2018	Intelligent intercelation by	https://doi.org/10.11 90/geo2017-0294.1	Jia	x		SEG	Geophysics	Trace Interpolation			Super	ised		Synthetic				
2020	Can learning from natural image denoising be used for seismic data interpolation?	https://doi.org/10.11 90/geo2019-0243.1	Zhang	х		SEG	Geophysics	Trace Interpolation			Super	ised CNN	user defined	Synthetic				MSE
2020	Synthesizing seismic	https://doi.org/10.11 90/segam2020- 3415521.1	Durall	х		SEG	SEG Technical Program	Event separation	Diffraction		Sen supen			Both	Both		Pre-stack migrated	
2020	Extraction of diffraction events from seismic data using deep learning-based approach	https://doi.org/10.11 90/segam2020- 3424217.1	Kim	х		SEG	SEG Technical Program	Event separation	Diffraction		Superi	ised CNN	U-Net	Synthetic	Synthetic	SMAART JV (Gulf of Mexico)	Raw seismic Data (Amplitude)	
2020	A convolutional neural network approach to deblending seismic data	https://doi.org/10.11 90/geo2019-0173.1	Sun	х	х	SEG	Geophysics	Event separation	Deblending		Super	ised CNN		Real	Real		Raw seismic Data (Amplitude)	
2021	Wavefield decomposition for diffraction separation with convolutional neural networks	https://doi.org/10.11 90/segam2021- 3584040.1	Bauer	х		SEG	International Meeting for Applied Geoscience & Energy	Event separation			Super	ised CNN	U-NET	Synthetic	Both		Raw seismic Data (Amplitude)	
2021	Building training data set for deep learning-based P- and S- wave separation: Field data case	https://doi.org/10.11 90/segam2021- 3583348.1	Wei	х		SEG	International Meeting for Applied Geoscience & Energy	Event separation	Multiple (deblending, P- and S- wave)		Super	ised FCNN		Synthetic	Both		Raw seismic Data (Amplitude)	
2019	Multi-Domain Diffraction Identification Using Deep Learning	https://doi.org/10.39 97/2214- 4609.201901208	Lowney	x	×	X EAGE	EAGE Annual	Event separation			Super	ised CNN		Synthetic			Raw seismic Data (Zero-offset)	
2020	Direct Diffraction Separation by Deep Learning on Pre-Migrated Seismic Data	97/2214- 4609.202010719	Lowney	х	×	X EAGE	EAGE Annual	Event separation	Diffraction		Super	ised GAN	multiple (U- NET, PaatchGan)	Real	Real		Raw seismic Data (Zero-offset)	
2020	and deep rearring	11/1365-24/8.12889	Tschanne n	х		X EAGE	Geophysical Prospecting	Event separation	Diffraction		Super	ised CNN		Synthetic	Both		Pre-stack migrated	Logloss
2020	Deblending Based Deep Learning Based U-net B-S Separation from Multi-	97/2214- 4609.202010691	Li	x		EAGE	EAGE Annual	Event separation	Deblending		Super	ised CNN	U-NET	Both	Real		Raw seismic Data (Amplitude)	
2020	Using Deep Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.202010617	Xiong	х		EAGE	EAGE Annual	Event separation	P- and S- wave		Super	ised CNN	U-NET	Synthetic	Synthetic	Marmousi II/Hess	Raw seismic Data (multi-component, angle gathers)	L1 norm
2020	Deep Convolutional Neural Networks: Opportunities and Challenges	97/2214- 4609.202010647	Hou		×	EAGE	EAGE Annual	Event separation	Deblending		Super	ised CNN	U-NET	Real	Real		Raw seismic Data (Amplitude)	multiplle (L1 norm, L2 norm)
2019	reflectivity using iterative deep neural networks	https://doi.org/10.11 90/segam2019- 3216178.1	Chen	х		SEG	SEG Technical Program	Wavelet estimation			Super	ised DNN		Synthetic	Both		Raw seismic Data (Amplitude)	
2020	Deep convolutional neural network and sparse least- squares migration Seismic nonstationary	https://doi.org/10.11 90/geo2019-0412.1 https://doi.org/10.11	Liu	х	х	SEG	Geophysics	Migration	Least-square		Super	ised CNN		Synthetic	Both			
2021	deconvolution with physics- guided autoencoder	90/segam2021- 3582130.1	Phan	х		SEG	International Meeting for Applied Geoscience & Energy	Deconvolution			Unsup ec	ervis Auto- encoder		Synthetic	Synthetic		Post-stack migrated	
2021	squares reverse time migration via a convolutional network	https://doi.org/10.11 90/segam2021- 3583931.1	Huang	х		SEG	International Meeting for Applied Geoscience & Energy	Migration	Least-square		Super	ised CNN	U-NET	Synthetic	Synthetic	Marmousi	Raw seismic Data (Zero-offset)	
2021	Nonstretching NMO correction using deep learning Deep learning based least-	https://doi.org/10.11 90/segam2021- 3580806.1 https://doi.org/10.11	Kaur	х		SEG	International Meeting for Applied Geoscience & Energy	NMO correction			Super	ised GAN		Synthetic	Synthetic		Raw seismic Data (CMP)	
2021	squares reverse-time migration	90/segam2021- 3581036.1	Torres	х		SEG	International Meeting for Applied Geoscience & Energy	Migration	Least-square		Super	ised CNN	U-NET	Synthetic	Synthetic	Marmousi	Raw seismic Data (Zero-offset)	MSE
2021	deep-learning framework Deep learning for multitrace		Vamaraju		x	SEG	Geophysics	Sparse-Spike	Least-square		Super			Synthetic	Synthetic	Marmousi/SE G salt model		Huber
2021		/doi/epub/10.1190/g eo2020-0342.1 https://doi.org/10.39 97/2214-	Chai	x		SEG	Geophysics EAGE Annual	deconvolution (SSD)			Super	CNN	user defined VGG 16	Synthetic	Both			MSE
2019	Using Convolutional Neural Network Extraction of the Seismic Wavelet Based on Deep	4609.202010730 https://doi.org/10.39 97/2214-	Lu	×		EAGE	EAGE Annual	Wavelet				multiple ised (GAN, DN		Synthetic	Real		Post-stack migrated	Normalized squared
2019	Neural Networks Neural Network Least Squares Migration	4609.201901378 https://doi.org/10.39 97/2214-	Liu	x		EAGE	EAGE/SBGF on Least square	estimation Migration	Least-square		Unsup	convoluti	0	Synthetic	Synthetic			difference
2018	Neural Network Least Squares Migration	4609.201900831 https://doi.org/10.39 97/2214- 4609.201803061	Liu	x		EAGE	migration EAGE/SBGF on Least square migration		Least-square		Unsup ec	Convoluti		Synthetic	Synthetic			MSE
2019	An implementation of the seismic resolution enhancing network based on GAN	https://doi.org/10.11	Zhang		x	SEG	SEG Technical Program	Frequency extrapolation	High frequency		Sen supen	i- GAN		Real	Real		Post-stack	
2018	Deep learning-enabled seismic image enhancement	https://doi.org/10.11 90/segam2018- 2996943.1	Halpert		×	SEG	SEG Technical Program	Frequency extrapolation	High frequency		Superv	ised GAN		Synthetic	Synthetic	SEAM	Post-stack migrated	
2021	Deep learning spectral enhancement considering features of seismic field data	https://doi.org/10.11	Choi	х	х	SEG	Geophysics	Frequency extrapolation	High frequency		Super	ised CNN	U-NET	Synthetic	Both		Post-stack migrated	
		90/geo2020-0017.1											UNLI	Synthetic	both		•	
2020	Improving Seismic Resolution by a Sequential Convolutional Neural Network	90/geo2020-0017.1 https://doi.org/10.39 97/2214- 4609.202010836	Yuan	х	x	EAGE	EAGE Annual	Frequency extrapolation	High frequency		Super	SCNN ised (Sequentia CNN)		Synthetic	Synthetic		Raw seismic Data (Amplitude)	MSE
2020	Resolution by a Sequential Convolutional Neural Network Extrapolated full waveform inversion with convolutional neural	https://doi.org/10.39 97/2214-		x	x	EAGE	EAGE Annual SEG Technical Program				Super	ised (Sequentia CNN)					Raw seismic Data	MSE
	Resolution by a Sequential Convolutional Neural Network Extrapolated full waveform inversion with convolutional neural networks Self-supervised learning for low frequency extension of seismic data	https://doi.org/10.39 97/2214- 4609.202010836 https://doi.org/10.11 90/segam2019- 3197987.1	Yuan	x	x			extrapolation	frequency			ised (Sequentia CNN)		Synthetic	Synthetic		Raw seismic Data	MSE
2019	Resolution by a Sequential Convolutional Neural Network Extrapolated full waveform inversion with convolutional neural networks Self-supervised learning for low frequency extension of	https://doi.org/10.39 97/2214- 4609.202010836 https://doi.org/10.11 90/segam/2019- 3187987.1 https://doi.org/10.11 90/segam/2020- 3427086.1	Yuan	x x		SEG	SEG Technical Program	Frequency extrapolation	Low Frequency		Super	ised (Sequentii CNN) ised CNN	ıl	Synthetic Synthetic	Synthetic Synthetic Both	Marmousi/O ver thrust 3D model	Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	MSE
2019	Resolution by a Sequential Convolutional Neural Network Extrapolated full waveform inversion with convolutional neural networks Self-supervised learning for low frequency extension of sessmic data Data-driven low frequency signal recovery using deep- learning prediction in full- waveform inversion Physics-guided self- supervised learning for low frequency data prediction in full-	Stylestation of J. 1 https://doi.org/10.39 https://doi.org/10.31 90/segam.2019- 3197827.1 https://doi.org/10.11 90/segam.2019- 3427086.1 https://doi.org/10.11 90/segam.2020- 0159.1 90/segam.2020- 0159.1	Yuan Sun Wang	x	x	SEG SEG	SEG Technical Program SEG Technical Program	extrapolation Frequency extrapolation Frequency extrapolation Frequency	Low Frequency Low Frequency		Supen	ised (Sequentii CNN) ised CNN ised CNN ised CNN	ıl	Synthetic Synthetic	Synthetic Synthetic Both	ver thrust 3D	Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data	MSE
2019 2020 2020	Resolution by a Sequential Convolutional Neural Network Extrapolated full waveform inversion with convolutional neural networks Seft-supervised learning for low frequency extension of seismic data Data-driven low-frequency signal recovery using deep- learning predictions in full- waveform inversion Physics-guided self- supervised learning for low	https://doi.org/10.39 https://doi.org/10.39 97/2214- 4609.202010836 https://doi.org/10.11 90/segam2019- 3197987.1 https://doi.org/10.11 90/segam2020- 3427086.1 https://doi.org/10.11 90/segam2020- 342396.1	Yuan Sun Wang Fang	x	x	SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics	extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation	Low Frequency Low Frequency Low Frequency		Superi Superi Superi See	ised (Sequentii CNN) ised CNN ised CNN ised CNN ised CNN ised CNN	ıl	Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic	ver thrust 3D	Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data	MSE MSE
2019 2020 2020 2020	Resolution by a Sequential Connolutional Neural Connolutional Neural Section 1 Section	Stylestatocol.7. https://doi.org/10.39 97/2144 4609.202010836 https://doi.org/10.11 90/segm/2019- 3197867.1 100/segm/2020- 34272085.1 https://doi.org/10.11 90/segm/2020- 3423396.1 https://doi.org/10.11 90/segm/2020- 3423396.1 https://doi.org/10.11 90/segm/2020- 3423396.1	Yuan Sun Wang Fang Hu Sun	x	x	SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program	extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation	frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency	Toro.	Supen Supen Supen supen	ised (Sequentii CNN) ised CNN ised CNN ised CNN ised CNN ised CNN	ıl	Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020	Resolution by a Sequential Connolutional Neural Connolutional Neural Section 1 Section	https://doi.org/10.19 ///2011 ////2011 ////2011 ////2011 //////////	Yuan Sun Wang Fang Hu Sun	x	x	SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program	extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation Frequency extrapolation	Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency	Trace Event Interpolatio separation n	Supern Supern Supern Sen supern	(Sequentic CNN) Issed CNN	ıl	Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020	Resolution by a Sequential Connoction by a Sequential Connoction and Neural Extra good and Secretary	https://doi.org/10.19 9/92144 669.20010836 4669.20010836 4669.20010836 1190/segm2019- 3197897.1 https://doi.org/10.11 90/segm2020- 3427086.1 https://doi.org/10.11 90/segm2019- 342396.1 https://doi.org/10.11 90/segm2019- 342396.1 https://doi.org/10.11	Yuan Sun Wang Fang Hu Sun	x	x	SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics	extrapolation Frequency	frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency	Interpolatio Event n separation	Supern Supern Supern Sen supern	(Sequentic CNN) Issed CNN	u-Net	Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020	Resolution by a Sequential Connolutional Neural Neural Neural Connolutional Neural Neural Neural Connolutional Neural Neural Connolutional Neural Neural Connolutional Neural Neural Connolutional Neural Neu	https://doi.org/10.19 9/92144 669.20010836 4669.20010836 4669.20010836 1190/segm2019- 3197897.1 https://doi.org/10.11 90/segm2020- 3427086.1 https://doi.org/10.11 90/segm2019- 342396.1 https://doi.org/10.11 90/segm2019- 342396.1 https://doi.org/10.11	Yuan Sun Wang Fang Hu Sun Nakayam a	x	x	SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics	extrapolation Frequency	frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency	Internolatio	Super Super Super Super Super Super deblending Super	(Sequentii CNN) sised CNN sised CNN sised CNN control	u-Net	Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020 2020	Resolution by a Sequential Convolution by a Sequential Convolutional Neural Extrapolated full valuedrom Return of Convolutional Resolution of Resolution of Resolution of Resolution Resolu	https://doi.org/10.39 97/2144 4669-2001010836 4669-2001010836 4669-2001010836 109/eegam2019-31897887.1 https://doi.org/10.11 90/eegam2019-3187987.1 https://doi.org/10.11 90/eegam2020-3427986.1 https://doi.org/10.11 90/eegam2020-3423996.1 https://doi.org/10.11 90/eegam2020-3189906.1 https://doi.org/10.11 90/eegam2020-3189906.1	Yuan Sun Wang Fang Hu Sun Nakayam a	x	x x	\$6G \$6G \$6G \$6G \$6G	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program	extrapolation Frequency extrapolation	frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency Low Frequency	Interpolatio Event n separation Trace Event Interpolatio eventica	Supern Supern Supern Supern deblending Supern deblending Supern		U-Net U-Net	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw selsmic Data (Amplitude)	
2019 2020 2020 2020 2020 2020 2020	Resolution by a Sequential Connolution by a Sequential Connolution Neural Connolution Neu	https://doi.org/10.39 9/0214 4669.20030836 4669.20030836 4669.20030836 4669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836 14669.20030836	Yuan Sun Wang Fang Hu Sun Nakayam a	x x x x	x	SEG SEG SEG SEG SEG SEG	SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics	extrapolation Frequency extrapolation	Low Frequency	Interpolatio Event n separation Trace Event Interpolatio eventica	Super Super Super Super Super deblending Super		U-Net U-Net	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both Both	Synthetic Both Synthetic Both Both	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020 2020	Resolution by a Sequential Connolution by a Sequential Connolution in Neural Extra gold and Secretary Advanced Secretary and Secretary Advanced Secretar	https://doi.org/10.19 // https://doi.org/10.19 // https://doi.org/10.11 // https://doi.org/10.19 // https:// https://doi.org/10.19 // https://doi.org/10.19	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao u	x	x x	\$6G \$6G \$6G \$6G \$6G	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program	extrapolation Frequency extrapolation	frequency Low Frequency	Interpolatio Event n separation Trace Event Interpolatio eventica	Supern Supern Supern Supern deblending Supern deblending Supern		U-Net U-Net	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Synthetic Both Synthetic Synthetic Synthetic Synthetic	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020 2020 2020	Resolution by a Sequential Connolution by a Sequential Connolution Neural Connolution Neu	https://doi.org/10.39 9/7214- 4669.2001010836 4669.2001010836 4669.2001010836 1https://doi.org/10.11 90/segm2019- 3187987.1 https://doi.org/10.11 90/segm2020- 327086.1 https://doi.org/10.11 90/segm2020- 327086.1 https://doi.org/10.11 90/segm2020- 3339936.1 https://doi.org/10.11 90/segm2020- 3339936.1 https://doi.org/10.11 90/segm2020- 3339936.1 https://doi.org/10.11	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao u	x	x x	SEG SEG SEG SEG SEG SEG	SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics	extrapolation Frequency	frequency Low Low Frequency Low Low Frequency Low Low Low Low Low Low Low Lo	Interpolatio Event n separation Trace Event Interpolatio eventica	Super Super Super Super Super deblending Super	(Sequentic CNN) Isseed CNN	U-Net U-Net	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both	Synthetic Both Synthetic Both Both	ver thrust 3D model Marmousi/B	Raw seismic Data (Amplitude)	
2019 2020 2020 2020 2020 2020 2020 2020	Resolution by a Sequential Connolutional Neural Extraoplated Biological Part of the Ward International Part of Technology and Exhaust International Part of Technology and Exh	https://doi.org/10.39 https://doi.org/10.39 https://doi.org/10.39 https://doi.org/10.31 90/segam2019- 3197987.1 https://doi.org/10.11 90/segam2020- 3427088.1 https://doi.org/10.11 90/segam2020- 3423396.1 https://doi.org/10.11 90/segam2020- 329306.1 https://doi.org/10.11 90/segam2020- 339306.1 https://doi.org/10.11 90/segam2020- 339306.1 https://doi.org/10.11 90/segam2020- 339306.1 https://doi.org/10.31 97/214- 4609.20101111 https://doi.org/10.31 97/214- 6609.20101111	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao u	x	x x	SEG SEG SEG SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics EAGE Annual	extrapolation Frequency	frequency Low Frequency	Interpolatio Event n separation Trace Event Interpolatio eventica	Supern Supern Supern Supern Supern deblending Supern Supern Supern Supern	(Sequenti-CNN) Issed CNN	U-Net U-Net U-NET	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both Real	Synthetic Both Synthetic Synthetic Synthetic Both Both	ver thrust 3D model Marmousl/B P 2004	Raw seismic Data (Amplitude)	MSE
2019 2020 2020 2020 2020 2020 2021 2021	Resolution by a Sequential Connolutional Neural Neu	https://doi.org/10.39 3/9/214- 4669.20019836 4669.20019836 4669.20019836 4669.20019836 4669.20019836 4669.20019836 4669.20019836 4669.2001983 4669.2001983 4669.2001983 4669.2001983 4669.2001983 4669.2001983 4669.2001983	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao UPlotnitsk ii Fabien-	x	x x	SEG SEG SEG SEG SEG SEG SEG SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics EAGE Annual EAGE Annual	extrapolation Frequency extrapolation	frequency Low Frequency	Interpolatio Event n separation Trace Event Interpolatio eventica	Supern Supern Supern Supern deblending Supern deblending Supern Supern Supern	(Sequentic CNN) Issed CNN Issed	U-Net U-Net U-NET	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both Both	Synthetic Both Synthetic Synthetic Both Synthetic Both Synthetic	wer thrust 3D model Marmousi/B P 2004	Raw seismic Data (Amplitude)	MSE
2019 2020 2020 2020 2020 2020 2021 2020 2020 2020 2020	Resolution by a Sequential Convolutional Neural Extrapolated full value of the Convolutional Neural Extrapolated full value of the Convolutional neural networks Self-supervised learning for the Neural Neur	https://doi.org/10.19 970214- 4669.20030836 Https://doi.org/10.11 90/segm2019- 3197897.1 https://doi.org/10.11 90/segm2020- 327086.1 https://doi.org/10.11 90/segm2020- 327086.1 https://doi.org/10.11 90/segm2020- 3423396.1 https://doi.org/10.11 90/segm2020- 3423396.1 https://doi.org/10.11 90/segm2020- 3423396.1 https://doi.org/10.11 90/segm2020- 3399306.1 https://doi.org/10.19 97/214- 4609.20011111 https://doi.org/10.39 97/214- 4609.20011118	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao UPlotnitsk ii Fablen-Ouellet Kazel	x	x x	SEG SEG SEG SEG SEG SEG SEG SEG EAGE EAGE	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program SEG Technic	extrapolation Frequency extrapolation	frequency Low	Interpolatio Event n separation Trace Event Interpolatio eventica	Super	issed (Sequentitic CNN) issed CNN	U-Net U-NET Seq2Seq U-NET	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both Both Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Both Synthetic Synthetic Both Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	werthrust 30 model Marmousi/II P 2004	Raw sesimic Data (Amplitude) Raw sesimic Data (Amplitude)	MSE
2019 2020 2020 2020 2020 2020 2020 2020	Resolution by a Sequential Connolution by a Sequential Connolution Neural Neur	https://doi.org/10.19 9/10.11 4669.20.010.1836 https://doi.org/10.11 90/segam2019- 3197987.1 https://doi.org/10.11 90/segam2020- 32.7086.1 https://doi.org/10.11 90/segam2020- 32.7086.1 https://doi.org/10.11 90/segam2020- 32.3396.1 https://doi.org/10.11 90/segam2020- 33.399306.1 https://doi.org/10.19 97/214- 4609.20031038 https://doi.org/10.39 97/214- 4609.20031023	Yuan Sun Wang Fang Hu Sun Nakayam a Hu Aharchao u Plotoitsk II Fabien- Ouellet Kazei	x	x x	SEG	SEG Technical Program SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program Geophysics SEG Technical Program SEG Technic	extrapolation Frequency extrapolation	frequency Low	Interpolatio Event n separation Trace Event Interpolatio eventica	Super		U-Net U-Net U-NET Seq25eq U-NET	Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Both Both Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	Synthetic Both Synthetic Synthetic Both Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic Synthetic	werthrust 30 model Marmousi/II P 2004	Raw seismic Data (Amplitude) Raw seismic Data (Amplitude)	MSE R2

2019	Classifying geological structure elements from seismic images using deep learning	https://doi.org/10.11 90/segam2019- 3216823.1	Xu	х		SEG	SEG Technical Program	VMB	Raw	Supervised	PINN (Physics- informed NN)						
2017	Deep learning prior models from seismic images for full- waveform inversion	https://doi.org/10.11 90/segam2017- 17627643.1	Lewis		x	SEG	SEG Technical Program	VMB	FWI	Supervised	CNN	AlexNet	Synthetic	Both		Post-stack migrated	
2019	Deep learning-driven velocity model building workflow	https://doi.org/10.11 90/tle38110872a1.1	Araya- Polo	х	x	SEG	TLE	VMB	Raw	Semi- supervised	GAN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2019	Deep-learning inversion: A next-generation seismic velocity model building	https://doi.org/10.11 90/geo2018-0249.1	Yang	х		SEG	Geophysics	VMB	Raw	Supervised	FCN	U-net	Synthetic	Both	SEG Salt model	Raw seismic Data (Amplitude)	
2020	method CycleFCN: A physics- informed data-driven seismic waveform inversion	https://doi.org/10.11 90/segam2020-w13-	Jin	x		sEG	SEG Technical Program	VMB	Raw	Supervised	PINN (Physics- informed		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
	method A theory-guided deep- learning formulation and	05.1 https://doi.org/10.11									NN)					Raw seismic Data	
2020	optimization of seismic waveform inversion Application of deep learning	90/geo2019-0138.1	Sun	х		SEG	Geophysics	VMB	Raw	Supervised	RNN		Synthetic	Synthetic	Marmousi	(Amplitude)	
2020	optimization algorithm in full waveform inversion	https://doi.org/10.11 90/iwmg2019_03.1	You	х		SEG	SEG Technical Program	VMB	Raw	Supervised	DNN		Synthetic	Synthetic	Marmousi		
2020	Seismic velocity estimation: A deep recurrent neural- network approach	https://doi.org/10.11 90/geo2018-0786.1	Fabien- Ouellet	x		SEG	Geophysics	VMB	Raw	Supervised	RNN-LSTM		Synthetic	Both		Raw seismic Data (CMP)	Lo
2020	deep learning: From general synthetics to field data	https://doi.org/10.11 90/segam2020- 3428324.1	Kazei	х		SEG	SEG Technical Program	VMB	Raw	Supervised	CNN	VGG	Synthetic	Both	Marmousi	Raw seismic Data (CMP)	
2020	application Elastic near-surface model estimation from full waveforms by deep learning	https://doi.org/10.11 90/segam2020-w13- 06.1	Kazei	х	x	SEG	SEG Technical Program	VMB	Raw	Supervised	CNN		Synthetic	Synthetic	SEAM	Raw seismic Data (Amplitude)	
2020	ML-descent: An optimization algorithm for full-waveform inversion		Sun	х		SEG	Geophysics	VMB	Latent space	Supervised	RNN-LSTM		Synthetic	Synthetic	Marmousi/O ver thrust 3D	Raw seismic Data (Amplitude)	L2
2021	using machine learning Seismic inversion via closed- loop fully convolutional	https://doi.org/10.11	Wang	v		SEG	Geophysics	VMB	Raw	Supervised	CNN		Both	Both	model	Post-stack migrated	
	residual network and transfer learning Reparameterized full-	90/geo2020-0297.1 https://doi.org/10.11	-	^											Marmousi	Raw seismic Data	
2021	deep neural networks Deep-Learning Inversion of	90/geo2019-0382.1 https://doi.org/10.11	He	х		SEG	Geophysics Transactions on Geoscience	VMB	Raw	Supervised	DNN		Synthetic	Synthetic	II/BP 2004	(Amplitude)	
2021	Deep-learning seismic full-	09/TGRS.2019.29534 73 https://doi.org/10.11	Li	х		IEEE	and Remote Sensing	VMB	Raw	Supervised	CNN					(Amplitude) Raw seismic Data	
2021	realistic structural models	90/geo2019-0435.1 https://doi.org/10.11	Liu	х		SEG	Geophysics	VMB	Raw	Supervised	CNN	user defined	Synthetic	Synthetic		(Amplitude)	
2021	frequency-stepping approach Physics-guided deep	90/segam2021- 3584519.1	Alzahrani			SEG	International Meeting for Applied Geoscience & Energy	VMB	Raw	Supervised	RNN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2021	learning for seismic inversion with hybrid training and uncertainty	https://doi.org/10.11 90/geo2020-0312.1	Sun	х		SEG	Geophysics	VMB	Raw	Supervised	multiple [CNN, DNN, PINN)		Synthetic	Synthetic	SEG/EAGE salt model	Raw seismic Data (Amplitude)	
2021	analysis Metallic deposits imaging based on U-net deep	90/segam2021-	Wan	x		SEG	International Meeting for Applied Geoscience & Energy	VMB	Raw	Supervised	CNN	U-NET	Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2021	learning method Deep learning for joint geophysical inversion of seismic and MT data sets	3582815.1 https://doi.org/10.11 90/segam2021- 3583955.1	Singh	х		SEG	International Meeting for Applied Geoscience & Energy	VMB	Raw	Supervised	ANN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2020	ML-Misfit: Learning a Robust Misfit Function for Full- Waveform Inversion Using	https://doi.org/10.39 97/2214- 4609.202010466	Sun	х		EAGE	EAGE Annual	VMB	Raw	Supervised			Synthetic	Synthetic	Marmousi	Raw seismic Data (Amplitude)	
2020	Elastic Full Wavelollii	https://doi.org/10.39 97/2214-	Li	×		EAGE	EAGE Annual	VMB	Raw		DNN		Synthetic	Both		Raw seismic Data	
1010	Inversion Assisted by Deep Learning Deep Learning Tomography by Mapping Full Seismic	4609.202010281	L			DAGE	Distriction	VIVID	1000		Dies		Synthetic	botii	Marmousi/	(Amplitude)	
2020	Waveforms to Vertical Velocity Profiles	97/2214- 4609.202011980	Kazei	х		EAGE	EAGE Annual	VMB	Pre-stack	Supervised	CNN		Synthetic	Synthetic	Marmousi II/Over thrust 3D model	Raw seismic Data (CMP)	
2020	A Deep-Learning inversion method for seismic velocity model building		Targino	х		EAGE	Conference on Machine Learning in Americas	VMB	Raw	Supervised	CNN	U-NET	Synthetic	Synthetic	Marmousi II	Raw seismic Data (Amplitude)	
2020	Optimizing Deep Convolutional Neural Networks for 2D Full Waveform Inversion	https://doi.org/10.39 97/2214- 4609.202011454	Puzyrev	х		EAGE	EAGE Annual	VMB	Raw	Supervised	CNN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2019	Elastic Pre-stack Seismic Inversion in Stratified Media Using Machine Learning	https://doi.org/10.39 97/2214- 4609.201901524	Zheng		x	EAGE	EAGE Annual	VMB	Raw	Supervised	CNN		Synthetic	Real		Raw seismic Data (Amplitude)	
2019	from Raw Shot Gathers	https://doi.org/10.39 97/2214-	Oye		x	EAGE	PESGB	VMB	Raw	Supervised	CNN		Synthetic	Both	Marmousi	Raw seismic Data (Amplitude)	
2019	Using Machine Learning Physics-Based Machine Learning Inversion of Subsurface Elastic Properties	4609.201901522 https://doi.org/10.39 97/2214-	Costa Nogueira		×	EAGE	EAGE Annual	VMB	Raw	Supervised	PINN (Physics-		Synthetic	Synthetic		Raw seismic Data	L2
	Automated Valority	4609.201901147 https://doi.org/10.39	Junior								informed NN)				Over thrust	(Amplitude) Raw seismic Data	
2019	Based Seismic-to-Velocity Mapping Seismic Inversion with Deep	4609.201901523	Duque	х		EAGE	EAGE Annual	VMB	Raw	Supervised	GAN		Synthetic	Synthetic	3D model	(Amplitude)	Lo
2019	Neural Networks: a Feasibility Analysis Tomography: a Deep	97/2214- 4609.201900765 https://doi.org/10.39	Puzyrev	х		EAGE	EAGE Annual HPCUP (HIGH PERFORMANCE	VMB	Raw	Supervised	CNN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2018	Learning vs Full-Waveform Inversion Comparison Pre-Stack Seismic Inversion	97/2214- 4609.201803073 https://doi.org/10.39	Farris	х	×	EAGE	COMPUTING FOR UPSTREAM)	VMB	Sembelance	Supervised			Synthetic	Synthetic		Velocity Sembelance Raw seismic Data	. 1
2018	With Deep Learning Rapid Seismic Domain Transfer: Seismic Velocity	97/2214- 4609.201803008 https://doi.org/10.39	Zheng		×	EAGE	PESGB	VMB	Raw	Supervised	CNN		Synthetic	Synthetic		(Amplitude)	
2018	Inversion and Modeling Using Deep Generative Neural Networks	97/2214- 4609.201800734	Mosser	х	x	EAGE	EAGE Annual	VMB	Migrated	Semisuperv ised	GAN		Synthetic	Both	Marmousi	Pre-stack migrated	
2020	geometry with	https://doi.org/10.11 90/geo2018-0591.1	Wang	х		SEG	Geophysics	VMB	Raw	Supervised	FCN	VMB	Synthetic	Synthetic			Nor sq diff
2019	velocity-model building	https://doi.org/10.11 90/geo2018-0249.1	Yang	х		SEG	Geophysics	VMB	Raw	Supervised	FCN	VMB	Synthetic	Synthetic		Raw seismic Data (Amplitude)	Nor sq diff
2018	picking	https://doi.org/10.11 90/AIML2018-04.1	Hu	x	x	SEG	Workshop: SEG Maximizing Asset Value Through Artificial Intelligence and Machine Learning, Beijing, China	Firstbreak picking	Active	Supervised	CNN	U-Net	Real	Real		Raw seismic Data (Amplitude)	
2018	Application of deep learning in first break picking of seismic data	https://doi.org/10.11 90/AIML2018-05.1	Pu		x	SEG	SEG Maximizing Asset Value Through Artificial Intelligence and Machine Learning, Beijing,	Firstbreak picking	Active	Supervised	CNN		Real	Real		Raw seismic Data (Amplitude)	
2019	First arrival picking using U- net with Lovasz loss and nearest point picking	https://doi.org/10.11 90/segam2019-	Yuan	x	x	SEG	China SEG Technical Program	Firstbreak picking	Active	Supervised	CNN	U-Net	Both	Both		Raw seismic Data (Amplitude)	His
2040	method First-break automatic picking with fully	3214404.1 https://doi.org/10.11	v*-			***		Firstbreak	A		EGA:		p1			Raw seismic Data	
2019	convolutional networks and transfer learning Automatic first arrival	90/segam2019- 3215277.1 https://doi.org/10.11	Xie	х	×	SEG	SEG Technical Program	picking	Active	Supervised	FCN		Real	Real		(Amplitude)	Lo
2019	picking for borehole seismic data using a pixel-level network	90/segam2019- 3216775.1	Ma	х		SEG	SEG Technical Program	Firstbreak picking	Active	Supervised	CNN	U-Net	Real	Real		Raw seismic Data (Amplitude)	
2019	arrivals with a recurrent neural network	3215081.1	Kirschner	х	×	SEG	SEG Technical Program	Firstbreak picking	Passive	Supervised	RNN-LSTM						
2018	picking with deep semisupervised learning neural network	2998106.1	Chun Tsai	x	×	SEG	SEG Technical Program	Firstbreak picking	Active	Semi- supervised	DNN		Real	Real		Raw seismic Data (Amplitude)	
2018	Using a deep convolutional neural network to enhance the accuracy of first-break	https://doi.org/10.11 90/segam2018- 2982650.1	Hollande r		×	SEG	SEG Technical Program	Firstbreak picking	Active	Supervised	CNN		Real	Real		Raw seismic Data (Amplitude)	
2020	picking Automated arrival-time picking using a pixel-level network	https://doi.org/10.11 90/geo2019-0792.1	Ma	х		SEG	Geophysics	Firstbreak picking	Active	Supervised	CNN	U-net	Real	Real		Raw seismic Data (Amplitude)	Lo
	Direct seismic waveform																
2020	classification and first-break	https://doi.org/10.11 90/iwmg2019_07.1	Zhao	x		SEG	SEG 2019 Workshop: Mathematical Geophysics:	Firstbreak	Active	Supervised	FCN		Real	Real		Raw seismic Data (Amplitude)	

2020	Convolutional neural networks for microseismic waveform classification and arrival picking	https://doi.org/10.11 90/geo2019-0267.1	Zhang	x			SEG	Geophysics	Firstbreak picking	Passive	Passive	Microseismic	Supervised	CNN		Synthetic	Both		Raw seismic Data (Amplitude)	
2020	Automated first break picking with constrained pooling networks	https://doi.org/10.11 90/segam2020- 3427812.1	Cova		x		SEG	SEG Technical Program	Firstbreak picking	Active			Supervised	CNN	U-Net	Real	Real		Raw seismic Data (Amplitude)	
2020	Using neural networks to detect microseismicity and pick P-wave arrival times in Oklahoma	https://doi.org/10.11 90/segam2020- 3417918.1	Luo	x			SEg	SEG Technical Program	Firstbreak picking	Passive	Passive	Microseismic	Supervised	CNN		Real	Real		Frequency-time	Logloss
2021	Acquisition/Processing: Al- complemented first-break picking for field low-S/N seismic data Enhancing Seismic Image Quality through an	https://doi.org/10.11 90/tle40060460.1	Woog	х			SEG	TLE	Firstbreak picking	Active			Supervised	CNN		Real	Real		Raw seismic Data (Amplitude)	
2021	Automatic Refraction Static Correction: A Machine Learning Application in Web Based Seismic Processing	97/2214-	Wardaya				EAGE	EAGE Asia Pacific Virtual Geoscience Week	Firstbreak picking	Active						Real	Real		Raw seismic Data (Amplitude)	
2020	Picking in Land Seismic Surveys The First-Break Detection	https://doi.org/10.39 97/2214- 4609.202011754	ov		x		EAGE	EAGE Annual	Firstbreak picking	Active			Supervised	CNN	U-NET	Real			Raw seismic Data (Passive)	
2019	For Real Seismic Data With Use of Convolutional Neural Network	97/2214- 4609.201901614	Loginov	х	x		EAGE	EAGE Annual	Firstbreak picking	Active			Supervised	CNN		Real	Real		Raw seismic Data (Amplitude)	
2018	Can Machines Learn To Pick First Breaks As Humans Do?	97/2214- 4609.201803026	Yalcinogl u		x		EAGE	PESGB	Firstbreak picking	Active				SVM		Real	Real		Raw seismic Data (Amplitude)	
2018	Automatic Seismic First Arrival Picking With Deep- Learning Deep Neural Network and	https://doi.org/10.39 97/2214- 4609.201803023	Xie		x		EAGE	PESGB	Firstbreak picking	Active			Supervised	FCNN	U-NET	Real	Real		Raw seismic Data (Amplitude)	Logloss
2018	Multi-pattern Based Algorithm for Picking First- arrival Traveltimes	https://doi.org/10.39 97/2214- 4609.201801109	Mezyk	х			EAGE	EAGE Annual	Firstbreak picking	Active			Supervised	CNN			Real		Raw seismic Data (Amplitude)	
2011	The Improvement of Neural Network Cascade- correlation Algorithm and its Application in Picking Seismic First Break Novel hybrid artificial neural	https://doi.org/10.39 97/2214- 4609.20149418	Song	х			EAGE	EAGE Annual	Firstbreak picking	Active			Supervised			Real	Real			
2014	network based autopicking workflow for passive seismic data	https://doi.org/10.11	Maity	х		х	EAGE	Geophysical Prospecting	Firstbreak picking	Passive			Supervised	MLP		Synthetic	Synthetic			Log mean square
2020	A density-based spatial clustering application for a fully automatic picking of surface wave dispersion curves Detecting the fundamental	https://doi.org/10.11 90/segam2020- 3423024.1	Rovetta		x		SEG	SEG Technical Program	Surface waves	DC pick			Unsupervis ed	DBSCAN		Synthetic	Synthetic	SEAM	Picked DCs	
2020	mode of energy for surface wave analysis, modelling, and inversion, using a deep convolutional network	https://doi.org/10.11 90/segam2020- 3424584.1	Kaul		×		SEG	SEG Technical Program	Surface waves	DC pick			Supervised	CNN	U-net	Real	Real		Sembelance volume	Dice
2021	Automatic extraction of surface wave dispersion curves using unsupervised learning	https://doi.org/10.11 90/segam2021- 3582711.1	Yao	x			SEG	International Meeting for Applied Geoscience & Energy	Surface waves	DC pick			Unsupervis ed	multiple (Kmeans/PC A)		Real	Real			
2021	Application of a density- based spatial clustering algorithm for fully automatic picking of surface- wave dispersion curves	https://doi.org/10.11 90/tle40090678.1	Rovetta		x		SEG	TLE	Surface waves	DC pick			Unsupervis ed	DBSCAN		Both	Both	SEAM	Raw seismic Data (CMP)	
2021	On application issues of automatic dispersion curves picking by machine learning	3594524.1	Ren	x	×		SEG	International Meeting for Applied Geoscience & Energy	Surface waves	DC pick			Supervised	CNN	U-NET	Synthetic	Real			L2 norm
2021	dispersion curves		Yablokov	x		x	EAGE	Geophysical Prospecting	Surface waves	Inversion			Supervised	ANN		Synthetic	Both		Surface Waves (DC)	MSE
2021	wave dispersion data	https://doi.org/10.10 02/nsg.12163	Aleardi	х			EAGE	NSG (Journal)	Surface waves	Inversion			Supervised	RNN					Surface Waves (f-k)	MSE
2020		https://doi.org/10.11 11/1365-2478.12927		х			EAGE	Geophysical Prospecting	Surface waves	Modal separation			Supervised	SVM						
2020	Machine-Learning-Driven Dispersion Curve Picking for Surface-Wave Analysis, Modelling, and Inversion	https://doi.org/10.39 97/2214- 4609.202010598	Kaul		×		EAGE	EAGE Annual	Surface waves	DC pick			Supervised	CNN	U-NET	Real	Real		Raw seismic Data (Amplitude)	Dice
2020	Near Surface Velocity Estimation from Phase Velocity-Frequency Panels with Deep Learning	https://doi.org/10.39 97/2214- 4609.202010253	Zwartjes		×		EAGE	EAGE Annual	Surface waves	Inversion			Supervised	CNN		Synthetic	Synthetic	SEAM	Surface Waves (f-k)	MSE
2020	Using Convolutional Neural Networks to Expedite the Hamiltonian Monte Carlo Inversion of Rayleigh Wave Dispersion Curves	97/2214- 4609.202020045	Salusti	x			EAGE	NSG	Surface waves	Forward model			Supervised	CNN		Synthetic			Surface Wave (model)	Normalized squared difference
2020	Inversion of Surface Waves Dispersion Curves Using Artificial Neural Network Near Surface	https://doi.org/10.39 97/2214- 4609.202010809	Yablokov	x		х	EAGE	EAGE Annual	Surface waves	Inversion			Supervised	FCNN		Synthetic	Synthetic		Surface Waves (DC)	
2019	Characterization in Southern Oman: Multi-Wave Inversion Guided by Machine Learning Near-Real Time 3D Seismic	https://doi.org/10.39 97/2214- 4609.201900968	Masclet		×		EAGE	EAGE Annual	Surface waves	DC pick			Unsupervis ed	Kmeans		Real	Real		Surface Waves (DC)	
2019	Velocity and Uncertainty Models from Ambient Noise, Gradiometry and Neural Network Inversion	https://doi.org/10.39 97/2214- 4609.201901993	Curtis	х			EAGE	EAGE Annual	Surface waves	Inversion			Supervised	Mixture density network		Synthetic	Synthetic		Surface Waves (DC)	
2019	Learn to Invert: Surface Wave Inversion with Deep Neural Network Detecting microseismic	https://doi.org/10.39 97/2214- 4609.201901965	Hou		×		EAGE	EAGE workshop	Surface waves	Inversion			Supervised	multiple (GAN, DNN)		Real	Real		Surface Waves (DC)	L1 norm
2019	events in downhole distributed acoustic sensing data using convolutional neural networks	https://doi.org/10.11 90/segam2019- 3214863.1	Binder	х			SEG	SEG Technical Program	Passive	Signal detection			Supervised	CNN		Synthetic	Both		Raw seismic Data (Amplitude)	Logioss
2019	Bayesian deep learning and uncertainty quantification applied to induced seismicity locations in the Groningen gas field in the Netherlands: What do we need for safe Al?	https://doi.org/10.11 90/segam2019- 3216455.1	Gu	x			SEG	SEG Technical Program	Passive	Event location			Supervised	Bayesian- CNN						
2018	Automatic microseismic-	https://doi.org/10.11 90/segam2018- 2998279.1	Qu	x			SEG	SEG Technical Program	Passive	Signal detection			Supervised	SVM		Real	Real		Raw seismic Data (Passive)	MSE
2020	Microseismic event or noise: Automatic classification with convolutional neural networks	https://doi.org/10.11	Consolvo		x		SEG	SEG Technical Program	Passive	Microseismic			Supervised	CNN		Real	Real		Raw (zero offset)	
2020	Predict passive seismic events with a convolutional neural network		Wang	x			SEG	SEG Technical Program	Passive	Event detection			Supervised	ANN	VGG	Synthetic	Synthetic		Raw seismic Data (Amplitude)	
2020	Application of machine learning to microseismic event detection in distributed acoustic sensing data	https://doi.org/10.11 90/geo2019-0774.1	Stork	x	×		SEG	Geophysics	Passive				Supervised	CNN	YOLOV3	Synthetic	Both		Raw seismic Data (Passive)	multiple (squared error, log)
2021	Microseismic event location using artificial neural networks Towards fast DAS passive	https://doi.org/10.11 90/segam2021- 3582729.1	Anikiev	x			SEG	International Meeting for Applied Geoscience & Energy	Passive	Event detection			Supervised	ANN		Synthetic	Both			
2021		https://doi.org/10.39 97/2214- 4609.202131024	Rodrigue z		×		EAGE	EAGE GEOTECH	Passive	Event location			Supervised	CNN	U-NET	Synthetic	Synthetic		Raw seismic Data (Amplitude)	MSE
2021	Automatic microseismic signals classification with Deep Learning using multi- input Convolutional Neural	https://doi.org/10.39 97/2214- 4609.202132010	Rajeul		x		EAGE	EAGE Annual	Passive	Signal detection			Supervised	CNN	GoogLeNet	Real	Real		Raw seismic Data (Passive)	
2020	Networks How to Leverage Advanced TensorFlow and Cloud Computing for Efficient Deep Learning on Large Seismic Datasets	https://doi.org/10.39 97/2214- 4609.202032057	Birnie		×		EAGE	EAGE Annual	Passive	Signal detection			Supervised	CNN	U-NET	Synthetic	Synthetic		Raw seismic Data (Passive)	

	Convolutional neural																		
2020	networks for automated microseismic detection in downhole distributed acoustic sensing data and comparison to a surface geophone array	https://doi.org/10.11 11/1365-2478.13027	Binder	x		E	AGE Geophysical Prospe	ecting	Passive	Signal detection		Supervise	d CNN	user defined	Synthetic	Both		Raw seismic Data (Passive)	Logloss
2020	Enhanced Microseismic Event Detection Using Deep Neural Networks Performance Review of a	https://doi.org/10.39 97/2214- 4609.202010892	Birnie		x	E	AGE EAGE Annual	ı	Passive	Signal detection		Supervise	d CNN	U-NET	Synthetic	Both		Raw seismic Data (Passive)	
2019	Real-Time Machine Learning Based Seismic Catalog Generator in Production	https://doi.org/10.39 97/2214- 4609.201901241	Reynen			E	AGE EAGE Annual	ı	Passive	Event location									
2019	Automatic Microseismic Event Detection Using Deep Learning: a Classification is Detection Method		Zhao	x		E	AGE EAGE Annual	ı	Passive	Signal detection		Supervise	d CNN	user defined	Real	Real		Raw seismic Data (Passive)	
2018	Data-Driven Signal Recognition- A Machine Learning Application For The Real-Time Microseismic Monitoring	https://doi.org/10.39 97/2214- 4609.201803007	Shamsa		x	E	AGE PESGB		Passive	Signal detection		Unsuperv ed	s GMM		Real	Real		Raw seismic Data (Passive)	
2018	Event Detection and Phase Picking Based on Deep Convolutional Neural Networks	https://doi.org/10.39 97/2214- 4609.201801052	Zhu	x		E	AGE EAGE Annual	ı	Passive	Phase detection		Supervise	d CNN	user defined	Real	Real		Raw seismic Data (Amplitude)	
2019	Deep learning guiding first- arrival traveltime tomography	90/segam2019- 3215632.1	Li	х		:	EG SEG Technical Pro	igram	Traveltime tomography	Inversion		Semi- supervise	GAN		Synthetic	Both		Raw seismic Data (Amplitude)	
2018	Automatic velocity picking with convolutional neural networks	90/segam2018-	Ma		x	:	EG SEG Technical Pro	gram	Semblance picking			Supervise			Synthetic	Synthetic		NMO corrected	
2020	Anisotropic eikonal solution using physics-informed neural networks	https://doi.org/10.11 90/segam2020- 3423159.1	Waheed	x			EG SEG Technical Pro	igram	Traveltime tomography	Eikonal Eq.		Supervise	PINN (Physics- informed NN)		Synthetic	Synthetic		Velocity model	
2020	Wavefield reconstruction inversion via machine learned functions	https://doi.org/10.11 90/segam2020- 3427351.1	Song	x		:	EG SEG Technical Pro	igram	Helmholtz equation (Wavefield solution)			Supervise	PINN		Synthetic	Synthetic		Velocity model	MSE
2020	with deep learning constraint	https://doi.org/10.11 90/segam2020- 3426298.1	Guo	x	x		EG SEG Technical Pro	gram	Traveltime tomography	Joint Inversion		Supervise	SRCNN (Deep residual CNN)		Synthetic	Synthetic		Velocity model	
2020	The near-surface velocity reversal and its detection via unsupervised machine learning	90/geo2019-0025.1	Juli	x		;	EG Geophysics		Traveltime tomography	Shingling (velocity reversal detection)			Fuzzy c- mean		Both	Both		Raw seismic Data (Amplitude)	
2021	network approach for ghost removal	https://doi.org/10.11 90/segam2021- 3584059.1	Almuteri	х		:	EG International Meet Applied Geoscience 8		Source deghosting			Supervise	d CNN		Synthetic	Synthetic	Marmousi/Si gsbee	Raw seismic Data (Amplitude)	
2021	Automatic velocity picking from semblances with a new deep-learning regression strategy: Comparison with a classification approach	https://doi.org/10.11	Wang	x		:	EG Geophysics		Semblance picking			Supervise	i CNN	U-NET	Synthetic	Both	Marmousi II/Hess	Raw seismic Data (CMP)	multiple (MSE, Log loss)
2021	Source deghosting of coarsely sampled common- receiver data using a convolutional neural	https://doi.org/10.11 90/geo2020-0186.1	Vrolijk	x		:	EG Geophysics		Source deghosting			Supervise	d CNN		Both	Both		Raw seismic Data (Amplitude)	
2021	network Accelerating E&P Decisions by Applying Artificial Intelligence and Big Data	https://doi.org/10.39 97/2214-	Maver		x	E	Digital Subsurface Co	inference	Data management	Unstructure d data									
2010	Analytics to Unstructured Data Neural-network based multi- azimuth processing	4609.202181002 https://doi.org/10.39 97/2214-	Huck	×		,	AGE EAGE Annual		Multi-Azimuthal	Gutu		Unsuperv	Multiple (MLP/UV0 s (Unsuperv		Synthetic	Real			
2020	Source Deghosting of Coarsely-Sampled Data	4609.20149948 https://doi.org/10.39 97/2214-		x			AGE EAGE Annual		processing			ed Supervise	ed Vector Quantizers		Synthetic	Synthetic	Marmousi		
	Using a Machine-Learning Approach Wavefield Solutions from Machine Learned Functions	4609.202011488							deghosting						-,	-,			
2020	that Approximately Satisfy the Wave Equation Deep-Learning Inversion to Efficiently Handle Big-Data	97/2214- 4609.202010588	Alkhalifa	х		E	AGE EAGE Annual		equation (Wavefield solution)			Supervise	f CNN		Synthetic	Synthetic		Velocity model	
2020	Assimilation: Application to Seismic History Matching Deep Learning for	97/2214- 4609.202035158	Xiao	х		E	AGE ECMOR XVII		Seismic history matching			Supervise	d CNN						
2020	Anisotropy Parameters Estimation in Oil/Gas Fractured Reservoirs Physics-Constrained Deep	https://doi.org/10.39 97/2214- 4609.202010395 https://doi.org/10.39	Sabinin	х		X E	AGE EAGE Annual	ı	Anisotropy			Supervise	PINN		Synthetic	Synthetic		Raw seismic Data (Amplitude)	MSE
2020	Learning for Solving the Eikonal Equation Time-Lapse Cross-	97/2214- 4609.202011764 https://doi.org/10.39	Grubas	х		X E	AGE EAGE Annual		Traveltime tomography	Eikonal Eq.		Supervise	(Physics- informed NN)		Synthetic	Synthetic		Velocity model	
2020	Equalization by Deep Learning Building and Understanding	97/2214- 4609.202011720	Alali			E	AGE EAGE Annual	1	Time-Lapse Cross Equalization										
2020	Deep Neural Networks Components for Seismic Processing: Lessons Learned Understanding How a Deep	https://doi.org/10.39 97/2214- 4609.202011287	Chambef ort		x	E	AGE EAGE annual	ı	Muting										
2020		https://doi.org/10.39 97/2214- 4609.202032076	Messud			E	AGE First EAGE Digitalization Conference and Exhibit		Muting										
2020	Data-To-Data and Gradient- To-Gradient Translations in Geophysics Using Deep	https://doi.org/10.39 97/2214- 4609.202011334	Yao			E	AGE EAGE Annual		Acoustic-elastic data translation		Hydrophone- Geophone data								
2020	Neural Networks Slope Estimation by Convolutional Neural Networks	https://doi.org/10.39 97/2214-	Zu			E	AGE EAGE Annual	ı	local slope		translationů								
2020		4609.202010260 https://doi.org/10.39 97/2214- 4609.202011141	Dhara			E	AGE EAGE Annual	ı	Seismic data registration										
2019	ROCK PHYSICS AT-SCALE, ENABLED BY BIG DATA ANALYTICS & MACHINE	https://doi.org/10.39 97/2214-	Mannini		x	E	AGE APGCE		Data management	Cloud management									
2019	LEARNING Before Machine Learning: handling seismic data with Python and segyio	4609.201903407 https://doi.org/10.39 97/2214- 4609.201901973	Kvalsvik		x	E	AGE EAGE Annual	ı	Data management	Segy preparation for ML									
2019	The Use of the Neural Network for Traveltimes Approximation for Inhomogeneous Velocity Models	https://doi.org/10.39 97/2214- 4609.201901193	Grubas	x		E	AGE EAGE Annual	ı	Traveltime tomography	Traveltime interpolatio n					Synthetic	Synthetic			
2019	Neural Network Travel-Time Tomography	https://doi.org/10.39 97/2214- 4609.201901966	Earp	х		E	AGE EAGE Annual	ı	Traveltime tomography	Inversion		Supervise	Mixture d density network		Synthetic	Synthetic		First breaks	
2018	Machine Learning to Support Technical Document Indexing, a Case Study on Seismic Acquisition Reports	https://doi.org/10.39 97/2214- 4609.201801219	Blondelle		x	E	AGE PESGB		Data management	Data indexing		Supervise	i		Real	Real		Raw seismic Data (Passive)	
2018	Machine Learning To Support Technical Document Indexing, How To Measure The Accuracy?	https://doi.org/10.39 97/2214- 4609.201803012			×	E	AGE PESGB		Data management	Data indexing		Supervise	i		Real	Real			
2018	Unsupervised Machine Learning: K-means Clustering Velocity Semblance Auto-Picking Deep Learning Based 3D	https://doi.org/10.39 97/2214- 4609.201800919	Wei	х	×	E	AGE EAGE Annual	ı	Semblance picking			Unsuperv ed	s Kmeans		Real	Real		Velocity Sembelance	
2018	Velocity Field Nonlinear Multiple Regression Machine Learning can	https://doi.org/10.39 97/2214- 4609.201800922	Wei			E	AGE EAGE Annual	ı	Semblance picking										
2017	extract the information needed for modelling and data analysing from unstructured documents	https://doi.org/10.39 97/2214: 4609.201701654	Blondelle		x	E	AGE EAGE Annual	ı	Data management	Data indexing		Supervise	1						
2021	solution using a high- resolution neural network	https://doi.org/10.11 90/segam2021- 3583898.1 https://doi.org/10.20	Duan	х		:	EG International Meet Applied Geoscience 8	& Energy	statics			Supervise	i		Both			Raw seismic Data (Amplitude)	L1 norm
2018	Machine Learning and Wave Equation Inversion of Skeletonized Data	97/2214- 4609.201801882	Schuster	х		E	AGE EAGE worksho	р	Wave equation inversion										