

Deep Learning Discrimination of Earthquakes and Quarry Blasts in Southern California

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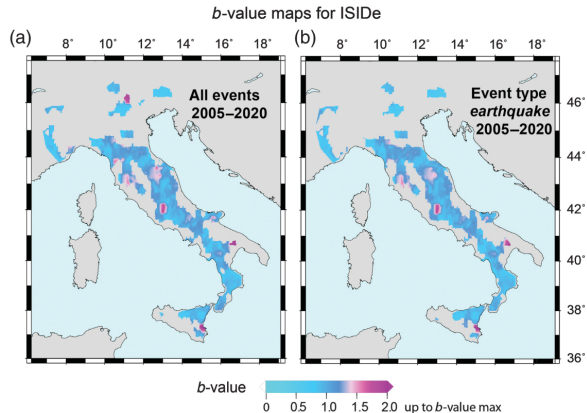
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DAMS
DAta Mining for Seismology



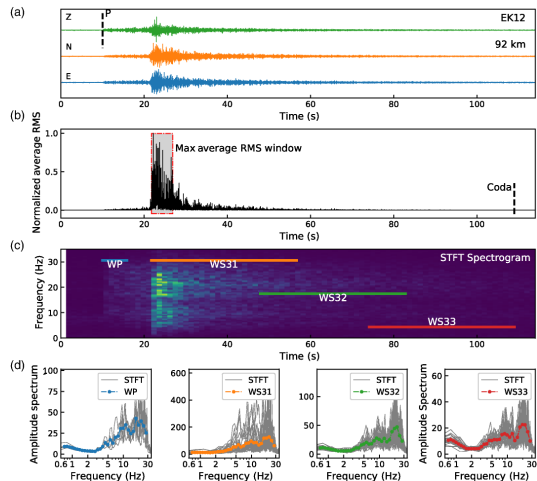
- Background
 - Manual discrimination
 - Artificial Neural Network
- Data
- Method
 - Data preprocess
 - **Data augmentation**
 - CNN architecture
- Results
- Summary

- Manual discrimination between natural earthquakes and anthropogenic activity is a laborious task.
- An automatic discriminator is essential to **seismic hazard analysis** and **fine classification catalog**.



L. Gulia *et al.* SRL, 2021

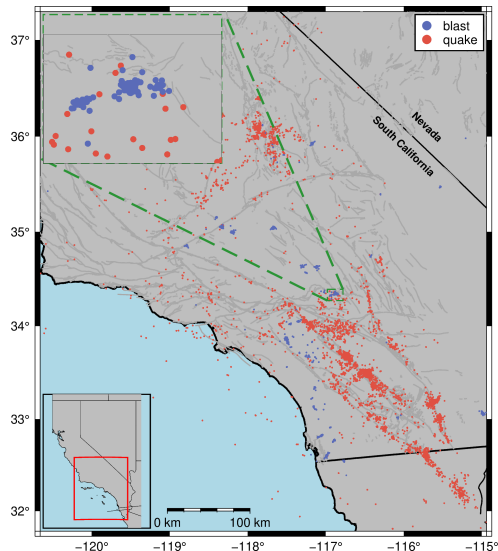
- ANN can achieve automatic discrimination with high accuracy (97%).
- CNN could be a more generalized discriminator than ANN, due to:
 - Feature extraction is done by CNN itself.

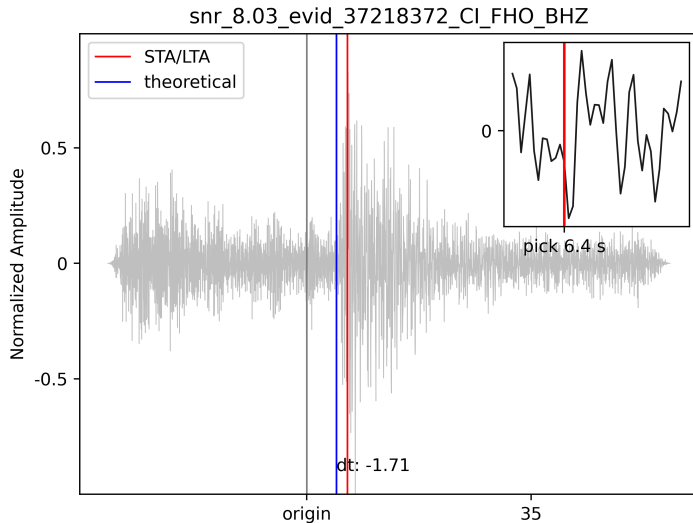


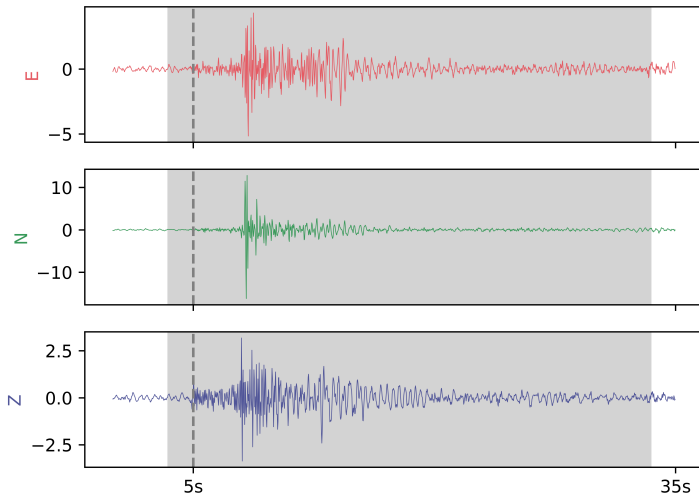
F. Miao *et al.* *SRL*, 2020

- Southern California Seismic Network

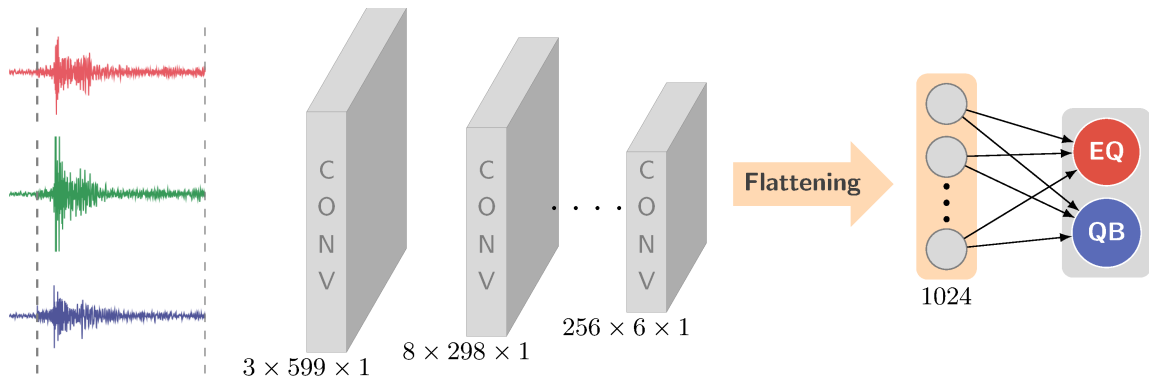
Type	#Event	#Recording
Quake	36,053	427,607
Blast	6,690	128,161
Aggregation	42,743	555,768





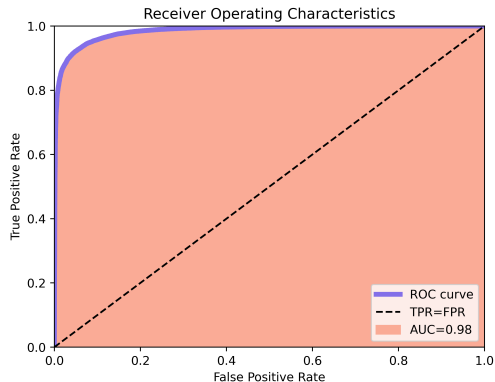


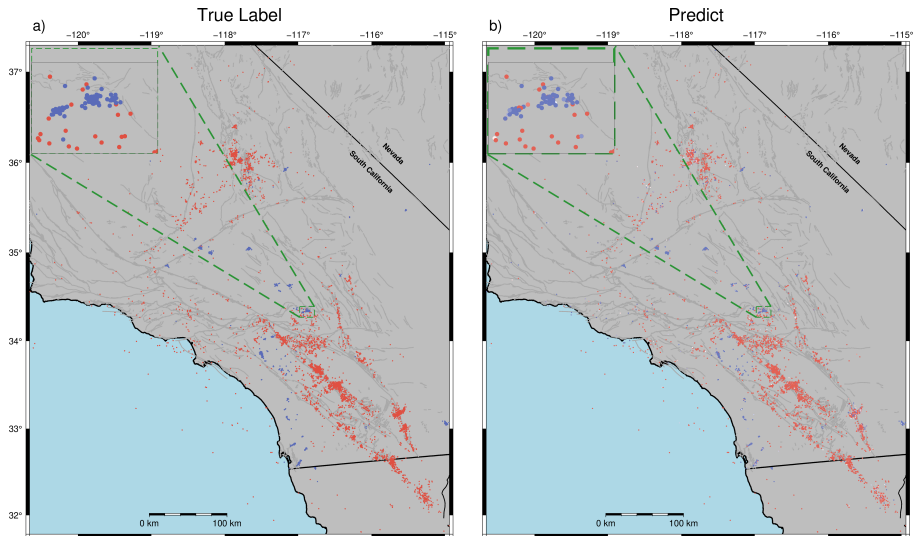
Method: CNN architecture



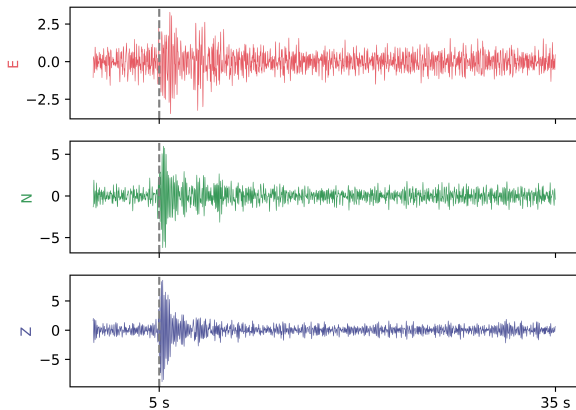
Accuracy: 93%

	Precision	Recall	F1-score	Support
Quake	0.98	0.92	0.95	64,014
Blast	0.78	0.95	0.86	19,186

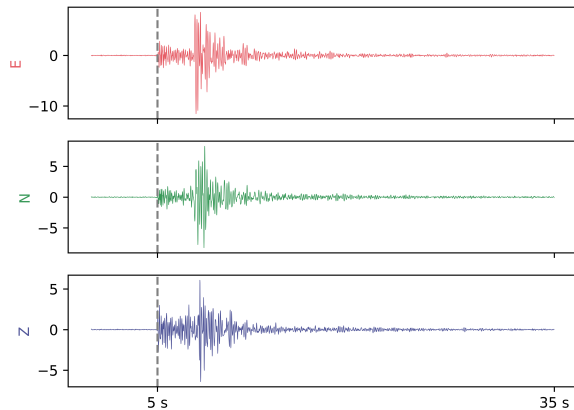




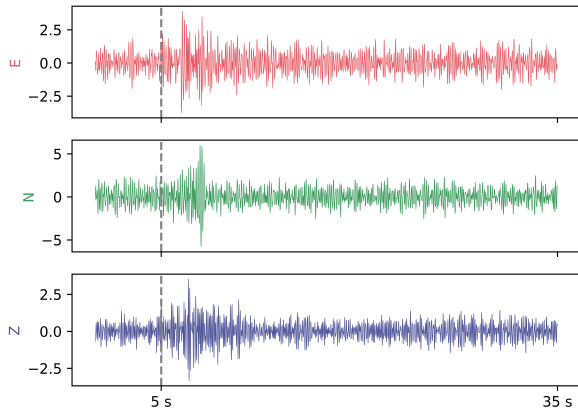
True label: blast; Predict: blast



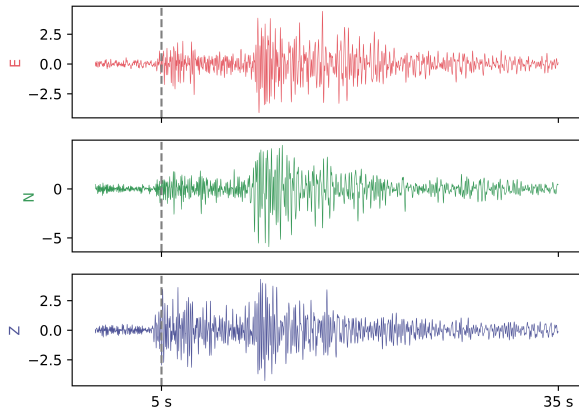
True label: quake; Predict: quake



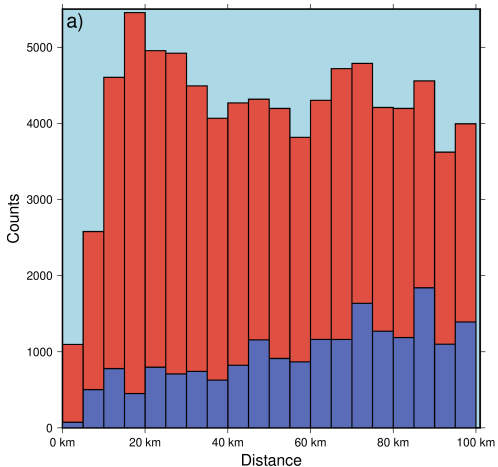
True label: quake; Predict: blast



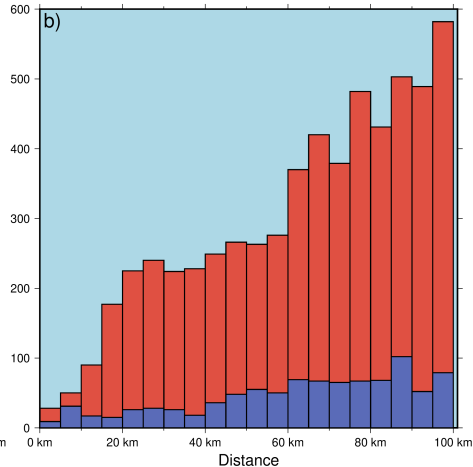
True label: blast; Predict: quake



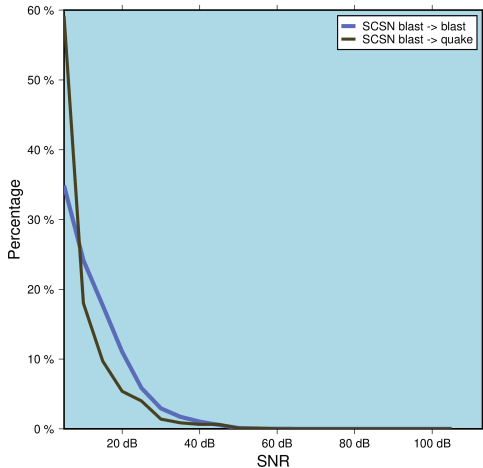
Distribution of all data



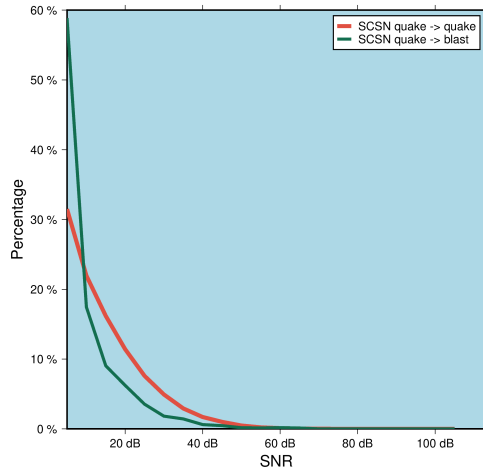
Distribution of false predictions



Distribution of blast

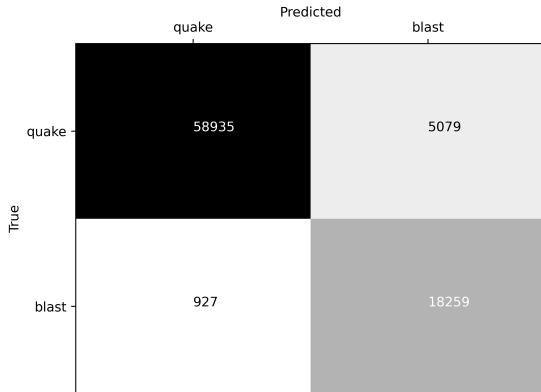
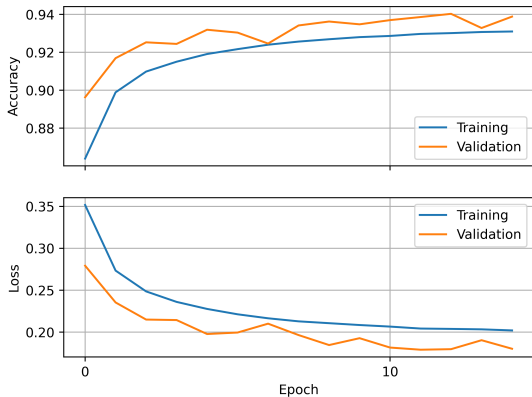


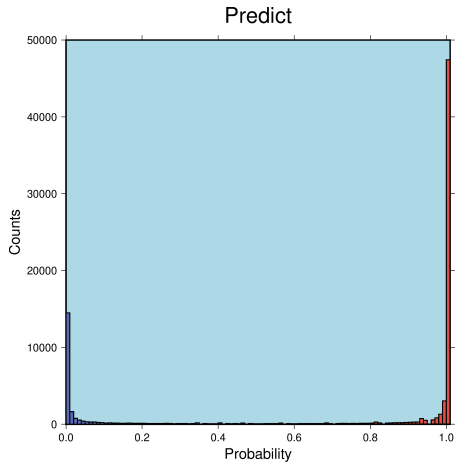
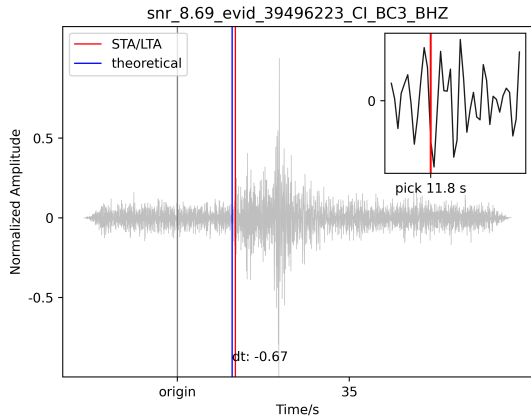
Distribution of quake



- CNN can achieve automatic discrimination without predetermined features.
- CNN can discriminate between earthquakes and quarry blasts with 93% accuracy, comparable to that of manual discrimination.
- The accuracy of CNN prediction is influenced by the the signal to noise ratio.

Train history





- Performance on Kentucky Data (Directly Applied)

	Predicted Quake		Predicted Blast	
True Quake	119		42	
True Blast	1,419		2,516	

	Precision	Recall	F1-score	Support
Quake	0.08	0.74	0.14	161
Blast	0.98	0.64	0.77	3,935

- Performance on Kentucky Data (Retrained)

	Predicted Quake		Predicted Blast	
True Quake	153		8	
True Blast	110		3,825	
	Precision	Recall	F1-score	Support
Quake	0.58	0.95	0.72	161
Blast	1.00	0.97	0.98	3,935

- Performance on Kentucky Data (Pretrained with 29404 Samples)

	Predicted Quake		Predicted Blast	
True Quake	153		8	
True Blast	51		3,884	
	Precision	Recall	F1-score	Support
Quake	0.75	0.95	0.84	161
Blast	1.00	0.99	0.99	3,935

- Performance on Kentucky Data (Pretrained with 15000 Samples)

	Predicted Quake		Predicted Blast	
True Quake	157		4	
True Blast	86		3,849	

	Precision	Recall	F1-score	Support
Quake	0.65	0.98	0.78	161
Blast	1.00	0.98	0.99	3,935