

# Earthquake Detection with Deep Learning

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# Goals

- Build Convolutional Neural Network ([CNN](#)) and Long-Short Term Memory ([LSTM](#)) models to predict earthquake class, magnitude, p-wave and s-wave arrival times
- Deploy best classification model to an [AWS Lambda](#) function to predict whether a signal is an earthquake in [real-time](#)

# Use Case

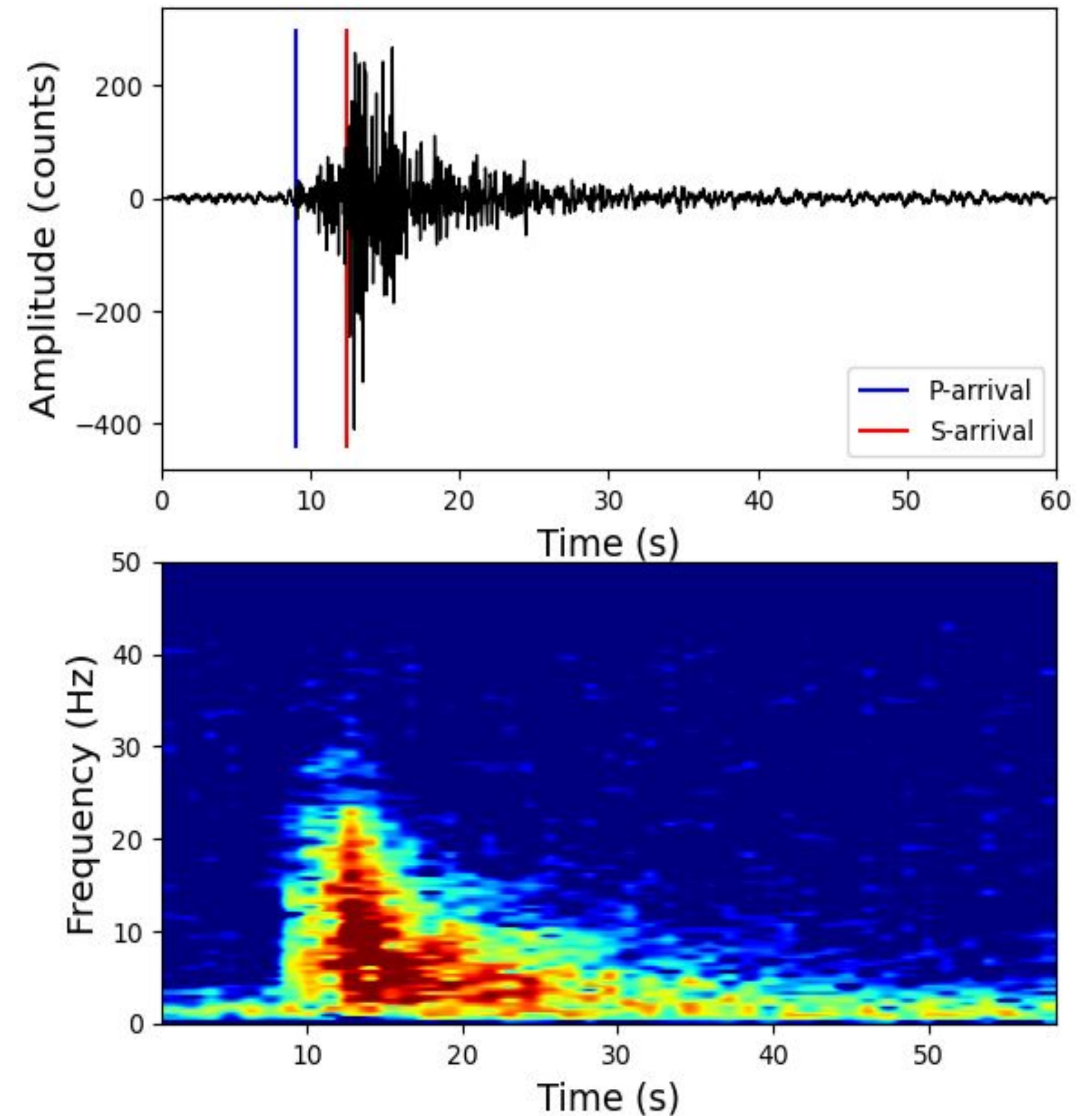
- Potential for faster or more reliable earthquake detection, as current methods typically depend on non-learning algorithms

# Data

100,000 global seismic signals from the STanford EArthquake Dataset (STEAD), plus metadata

- 34 features
- ~40% noise signals
- ~60% earthquake signals
- Each signal is 60 seconds

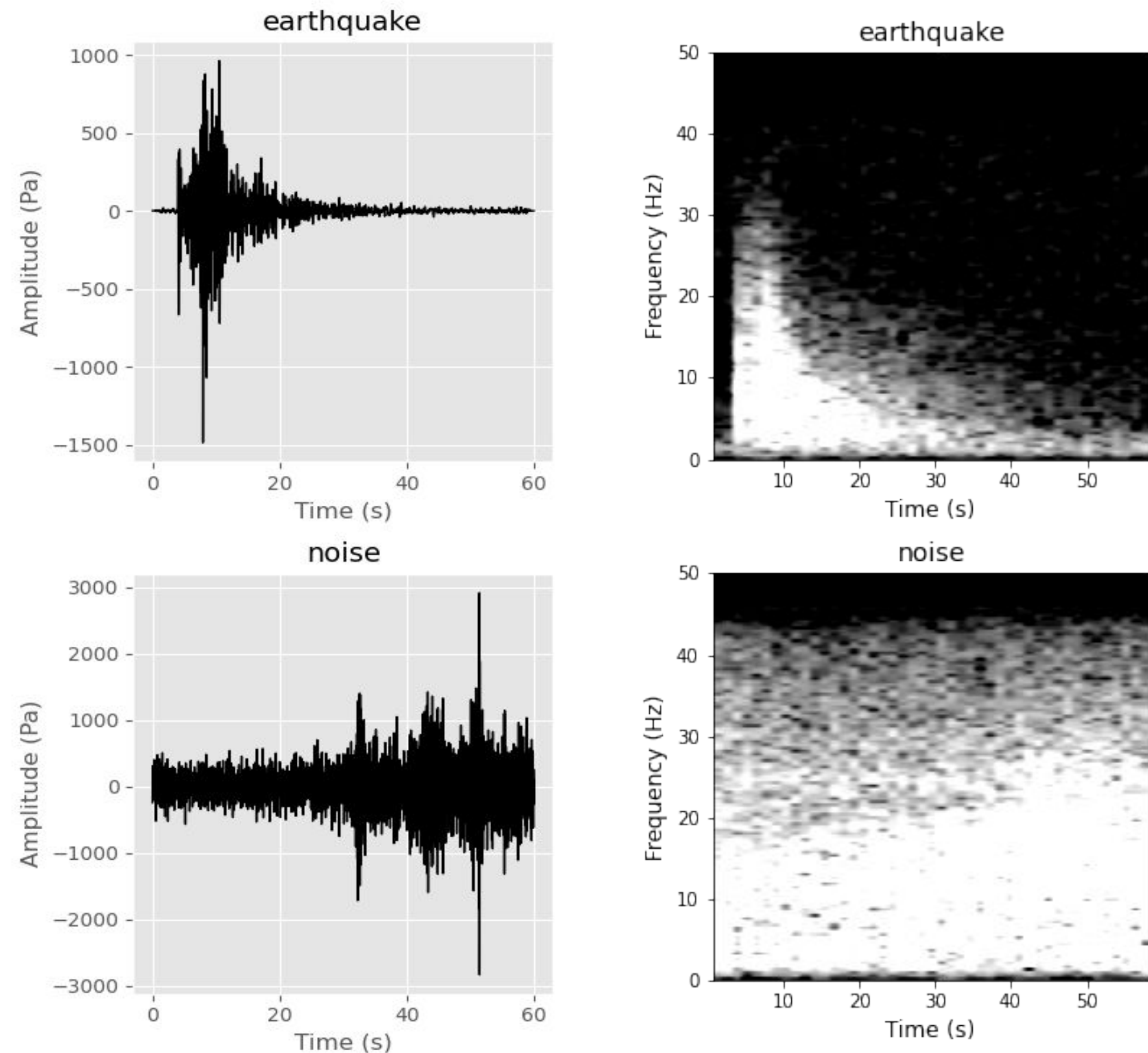
## Example Earthquake Waveform & Spectrogram



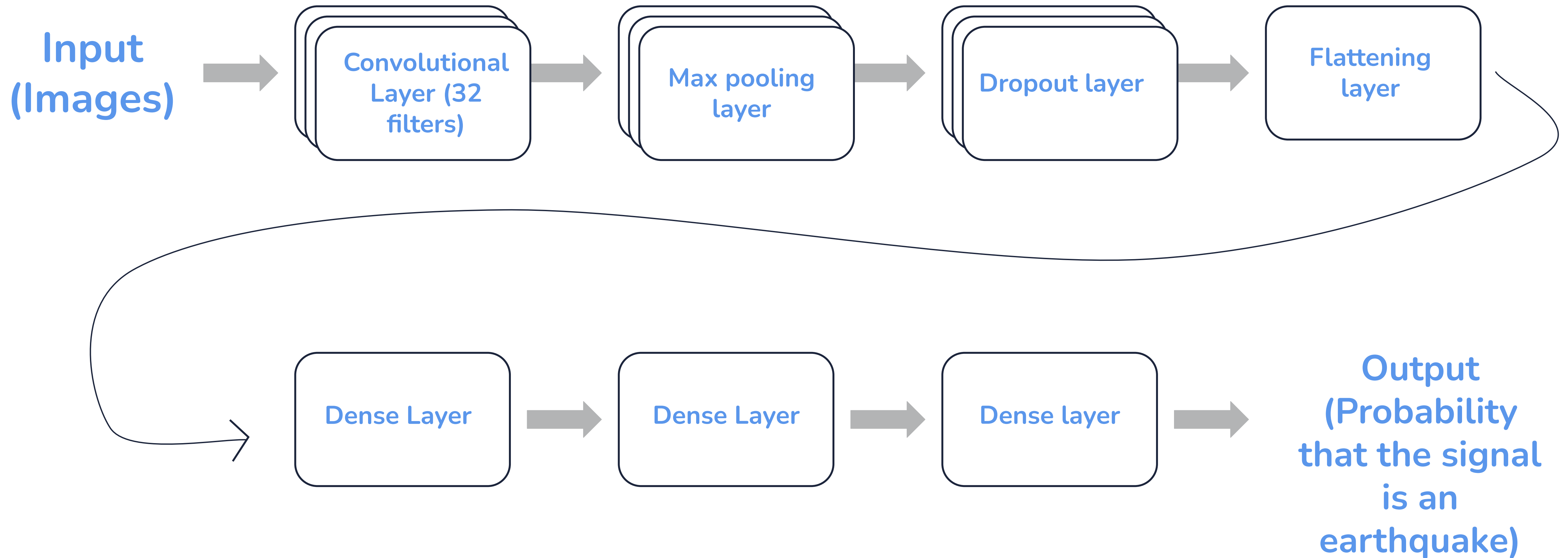
# Methods: CNN

- Models trained/tested on spectrograms or waveform figures depending on target
- Used classification CNN model for predicting earthquakes, regression CNN for other targets, each with one convolutional layer

## Example Waveforms & Spectrograms



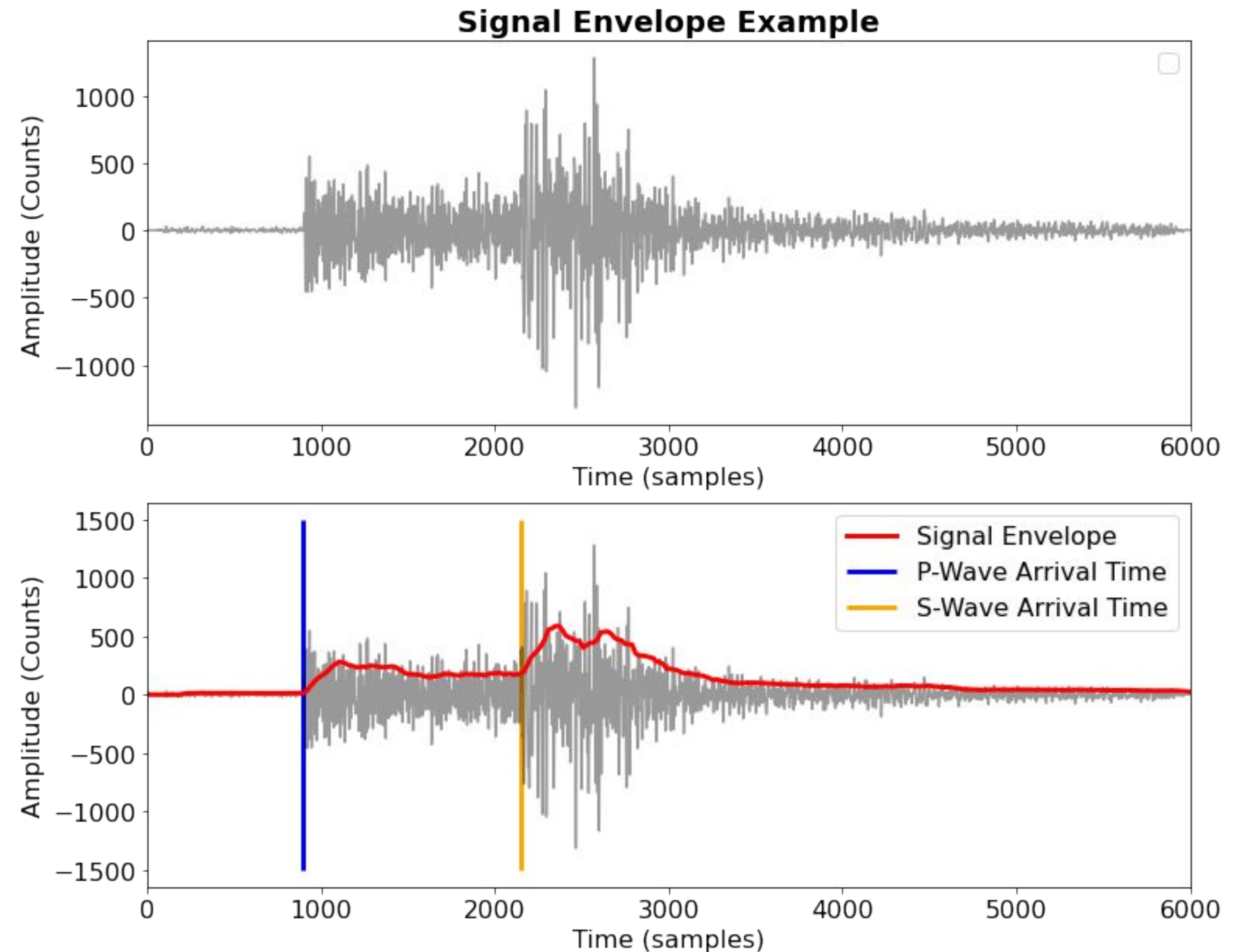
# Classification CNN Architecture



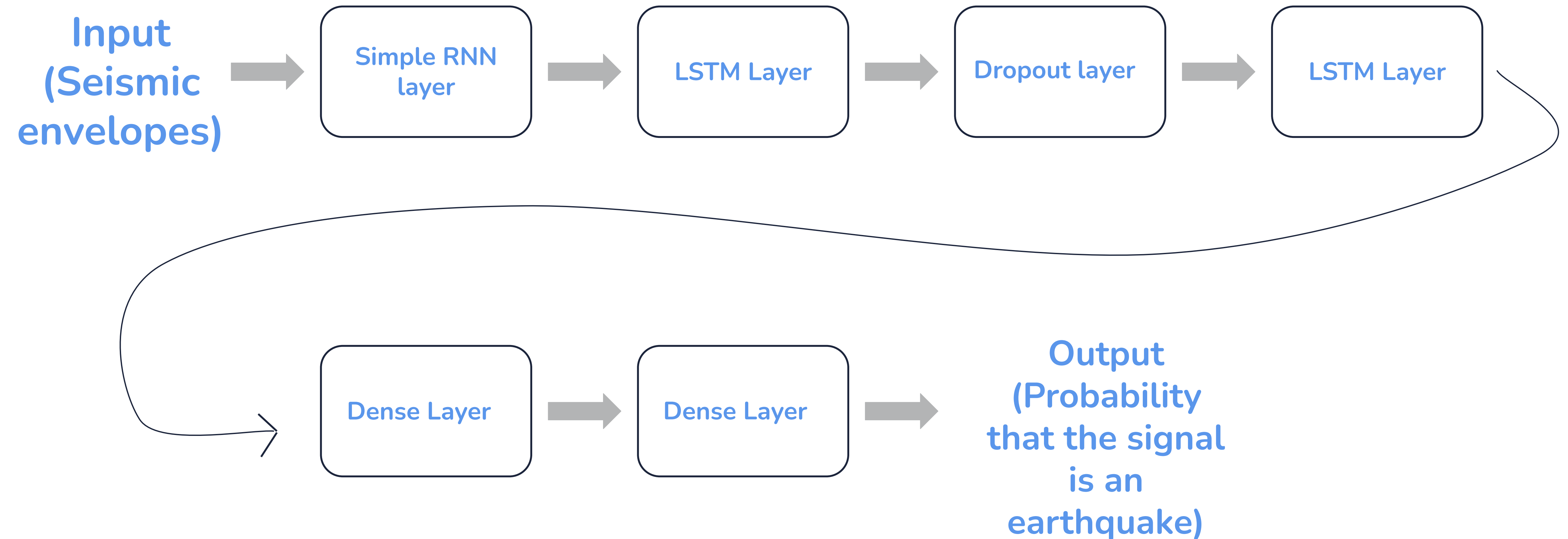


# Methods: LSTM

- Built an LSTM model to use the time-series signal data, to see if I could remove the need to generate images as needed for the CNN
- Models trained/tested on seismic envelopes (time-series), each with 2 LSTM layers

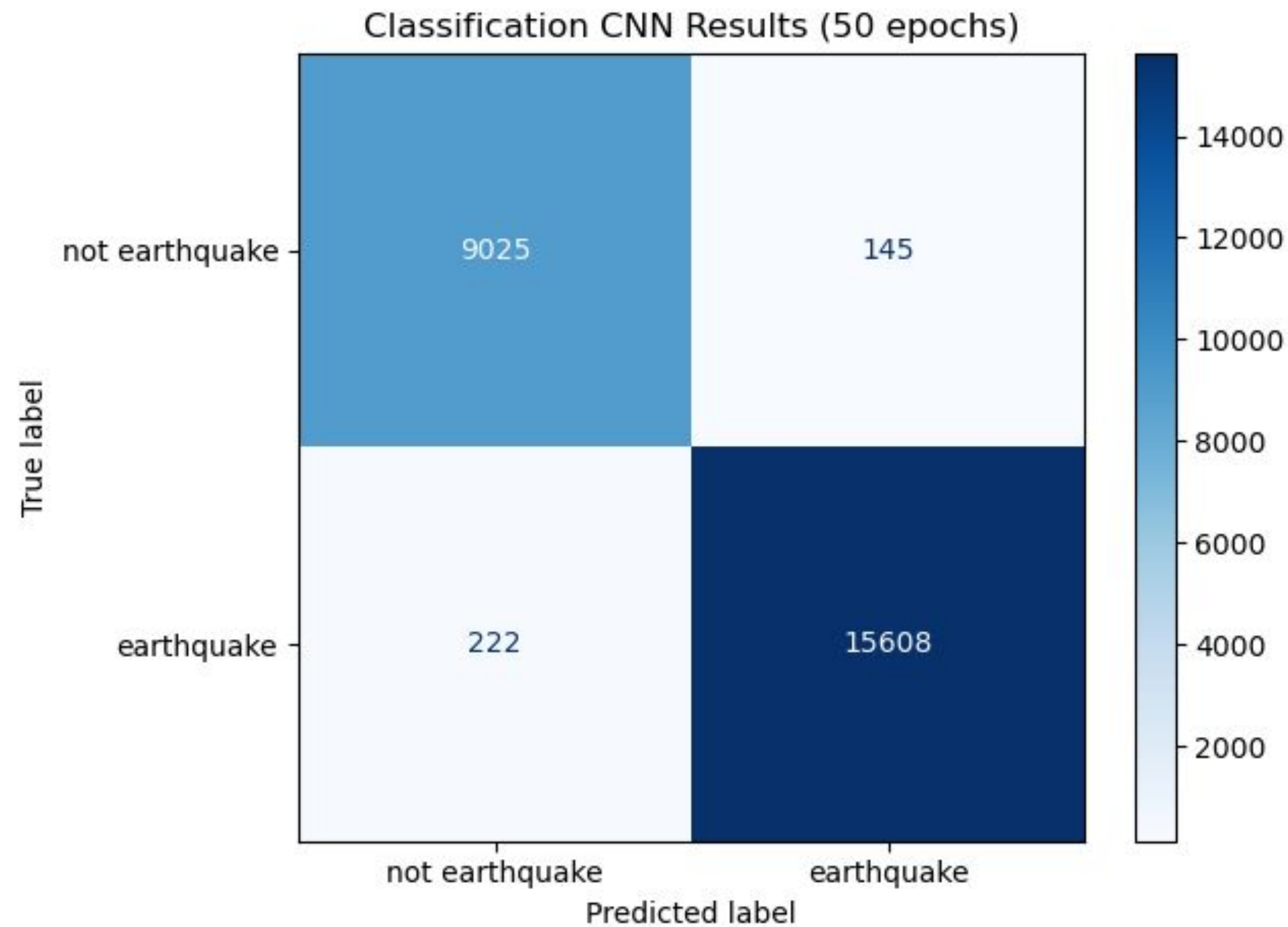


# Classification LSTM Architecture



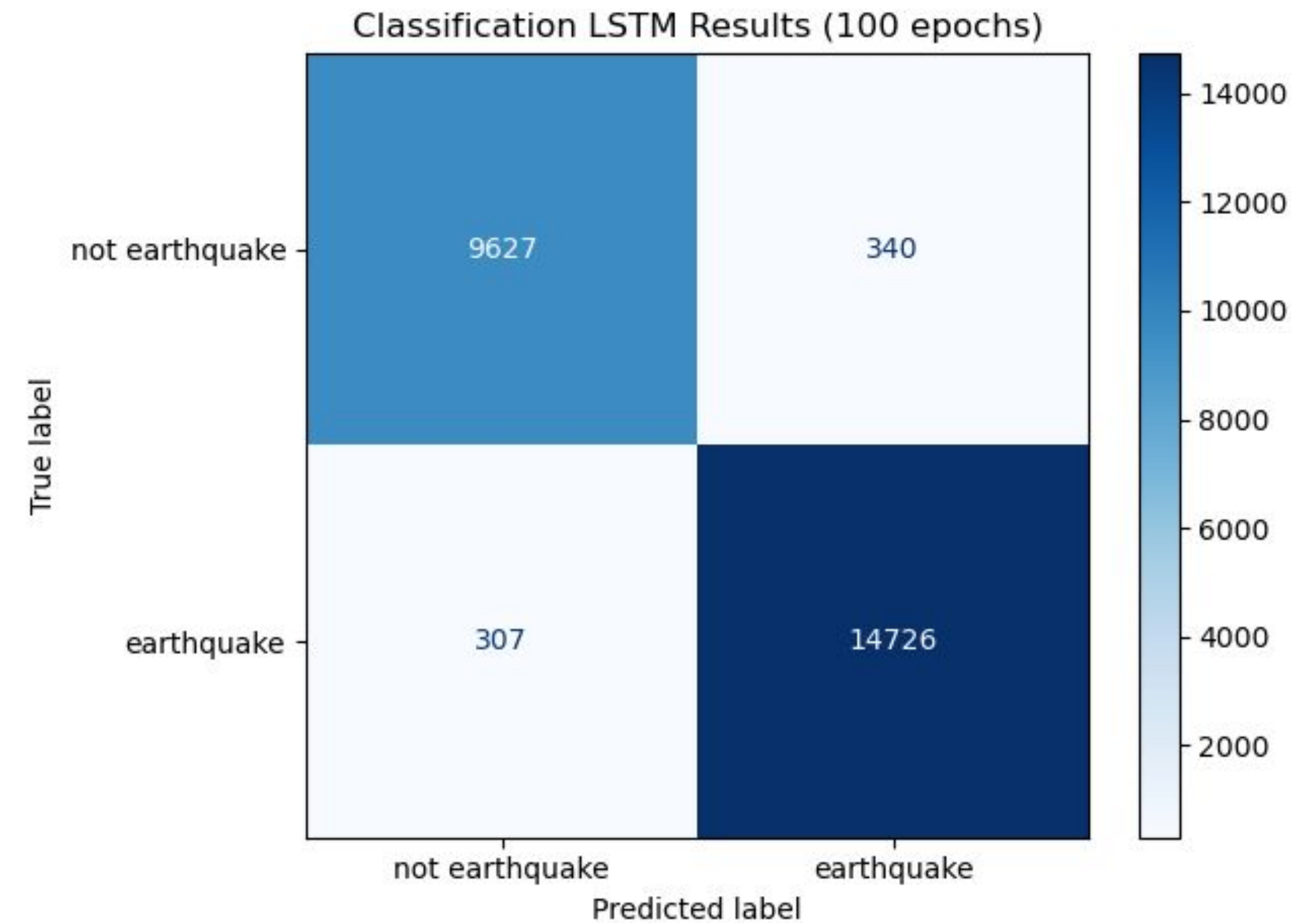
# Results: Classifying Signals as 'Earthquake' or 'noise'

## CNN



Accuracy: 0.98532  
Precision: 0.99080  
Recall: 0.98598

## LSTM

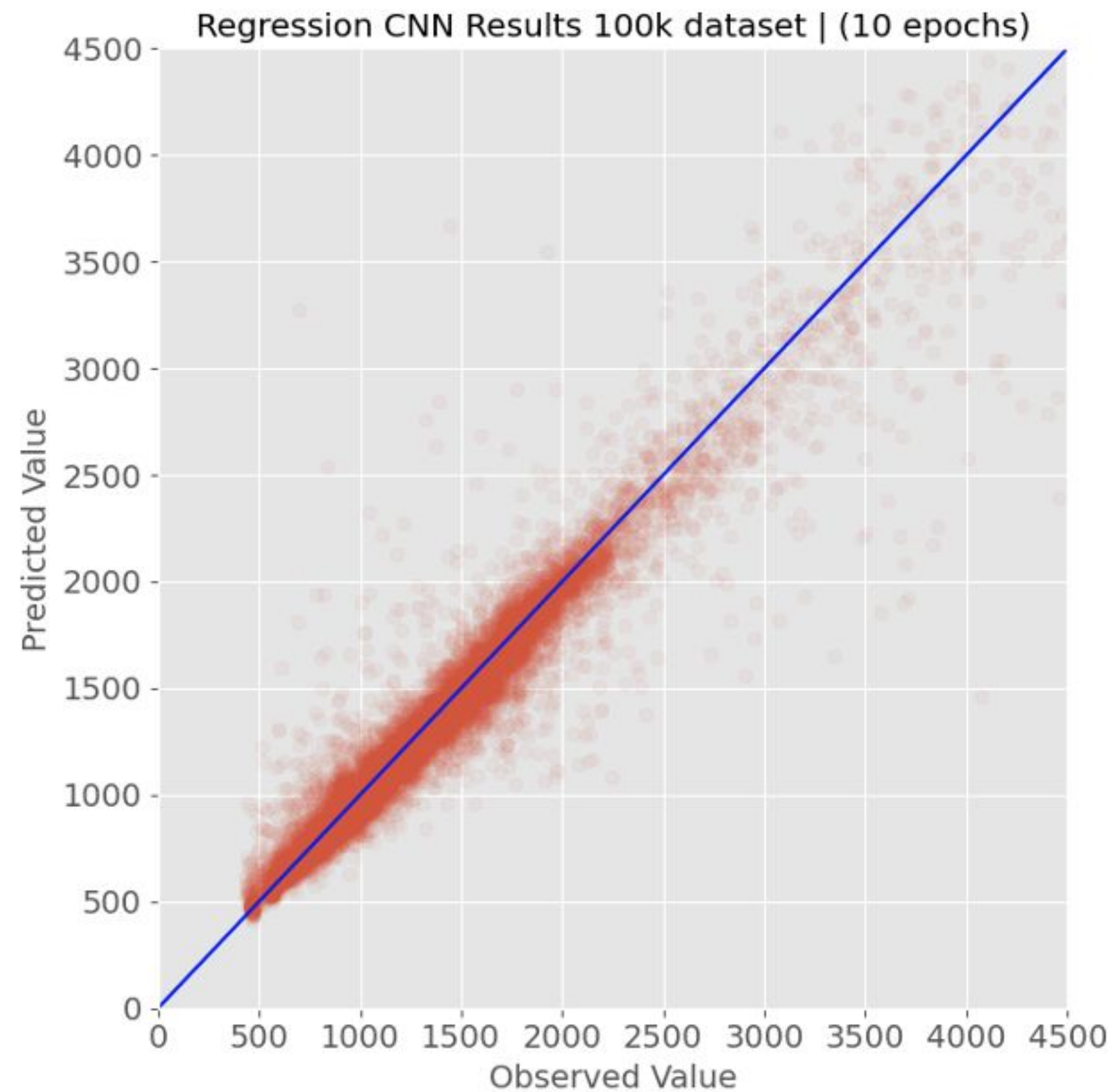


Accuracy: 0.97412  
Precision: 0.97743  
Recall: 0.97958



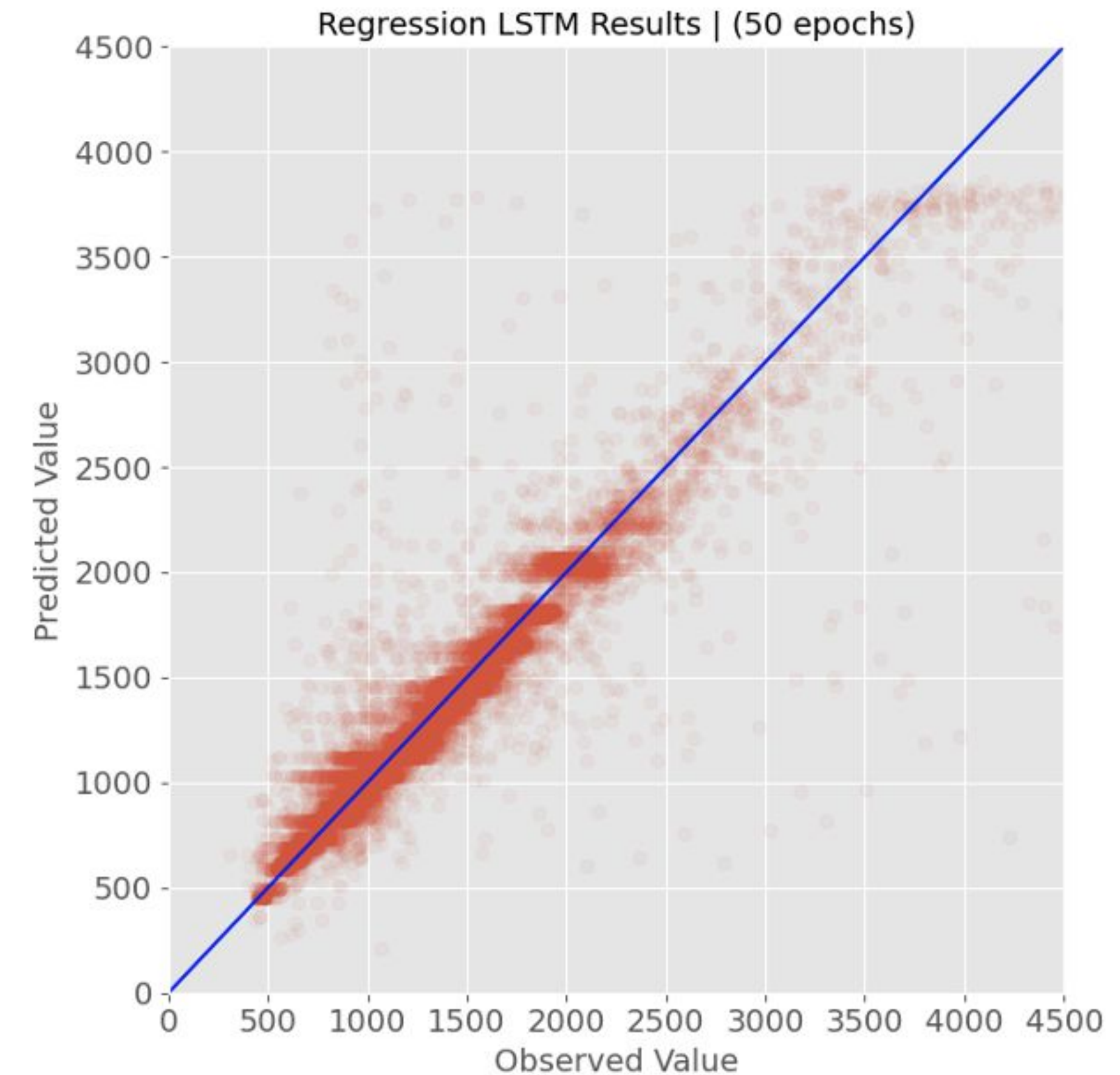
# Results: S-Wave Arrival Time Prediction

CNN



MSE: 25648.9

LSTM



MSE: 42508.6

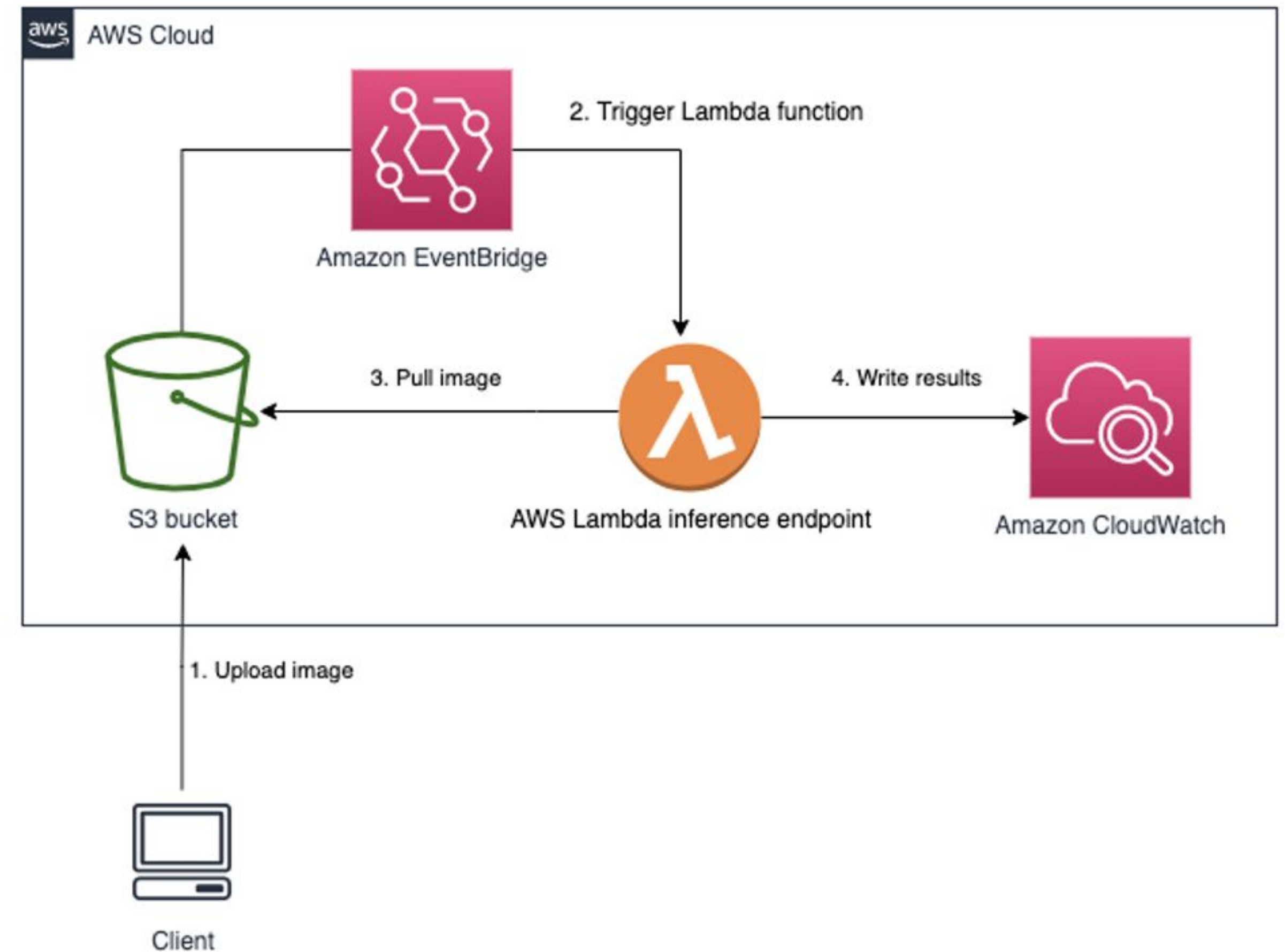
# CNN vs. LSTM Comparison

	Classifier (Accuracy)	Magnitude (MSE)	P-Wave (MSE)	S-Wave (MSE)	Pre-Processing Time
Baseline	0.52164	0.95674	30772.7	366620.9	
Best CNN model	0.98532	0.15895	1216.2	25648.9	0.00761 s
Best LSTM model	0.97412	0.38736	3212.2	42508.5	0.05056 s

# Deploying the Best Model

## Model deployment method:

1. Containerized model using Docker
2. Created s3 bucket to store images
3. Deploy Docker image as AWS Lambda function

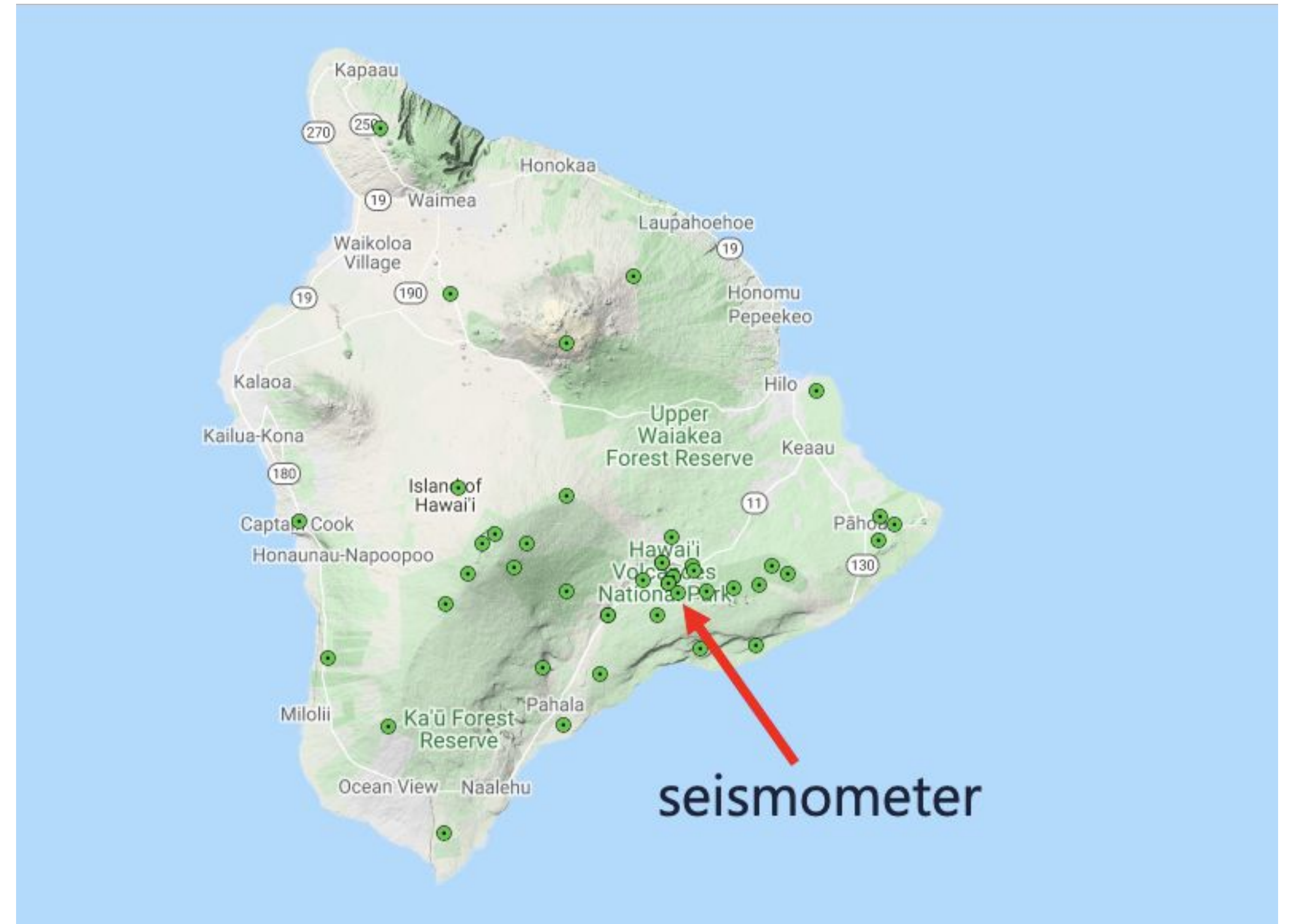




# Real-Time Prediction

To make predictions in real time:

- Built a pipeline to stream live data from a seismometer on Kilauea, Hawaii because the volcano produces a lot of earthquakes
- Create an image every ~15 seconds, send to S3 bucket
- S3 bucket triggers the Lambda function, runs the model and predicts image class
- Compare model predictions to USGS public website

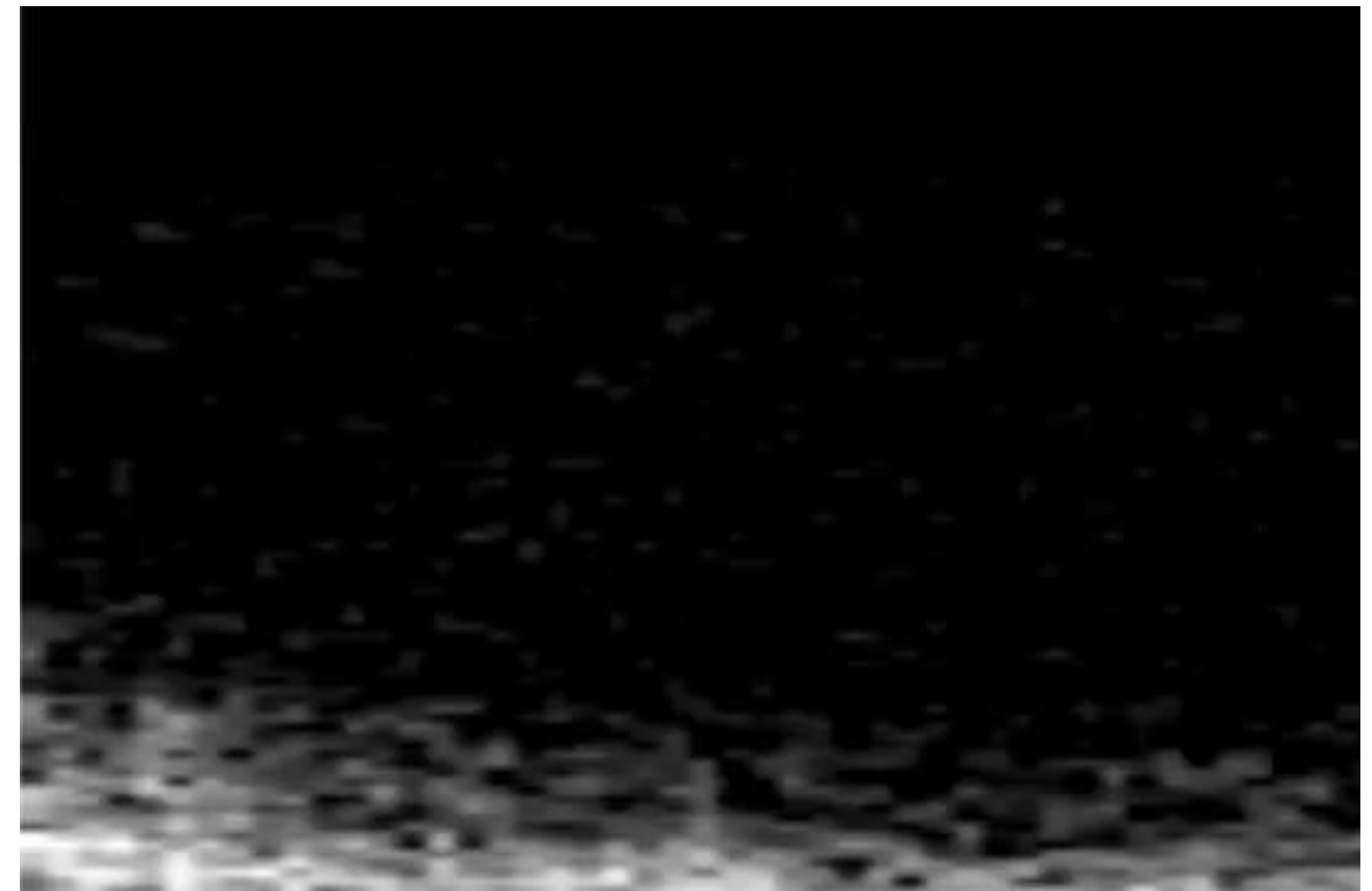


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Network: HV | Station: AHUD | Channel: EHZ





# AWS Lambda Demo





# Thanks!

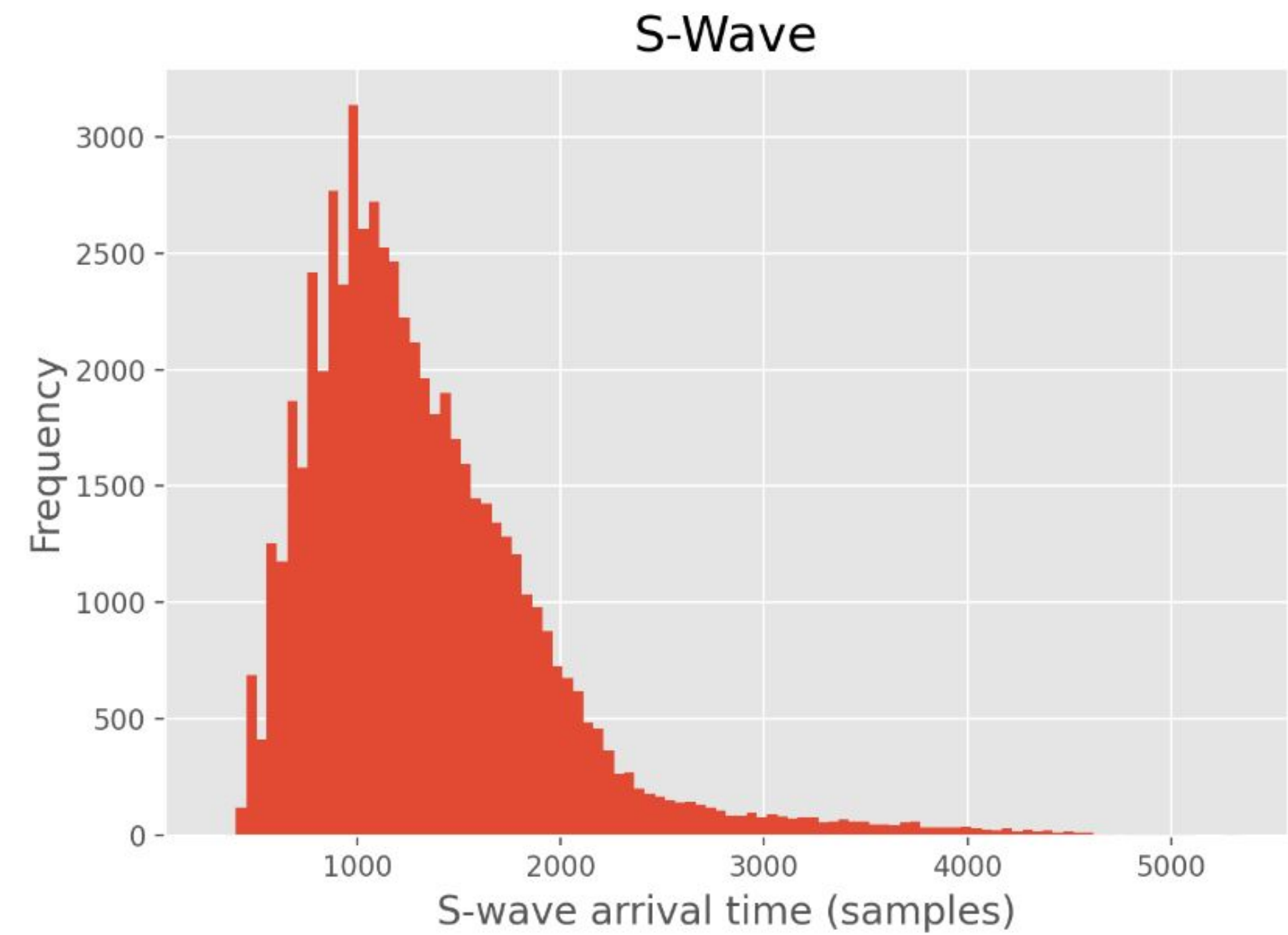
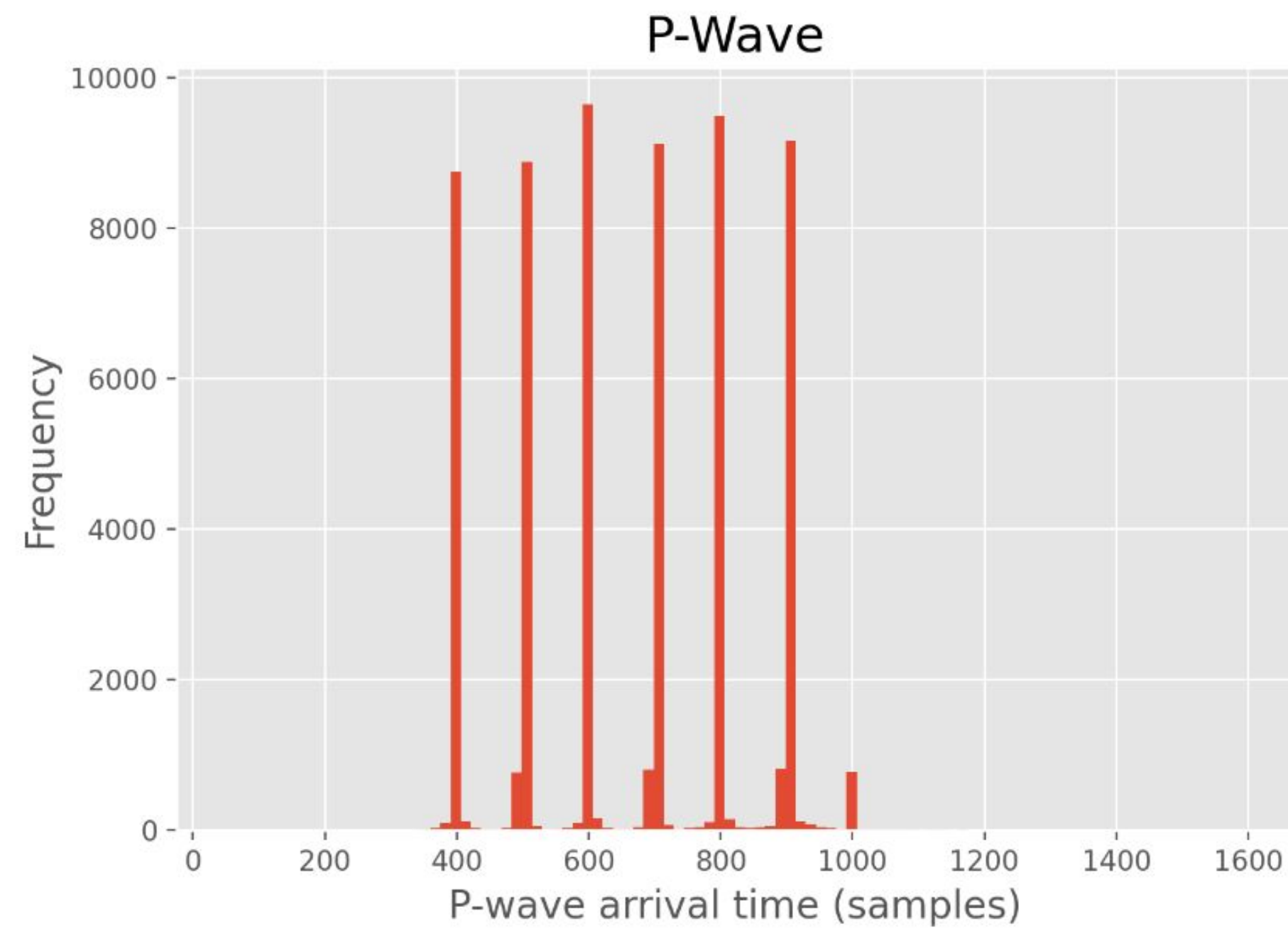
Any questions?





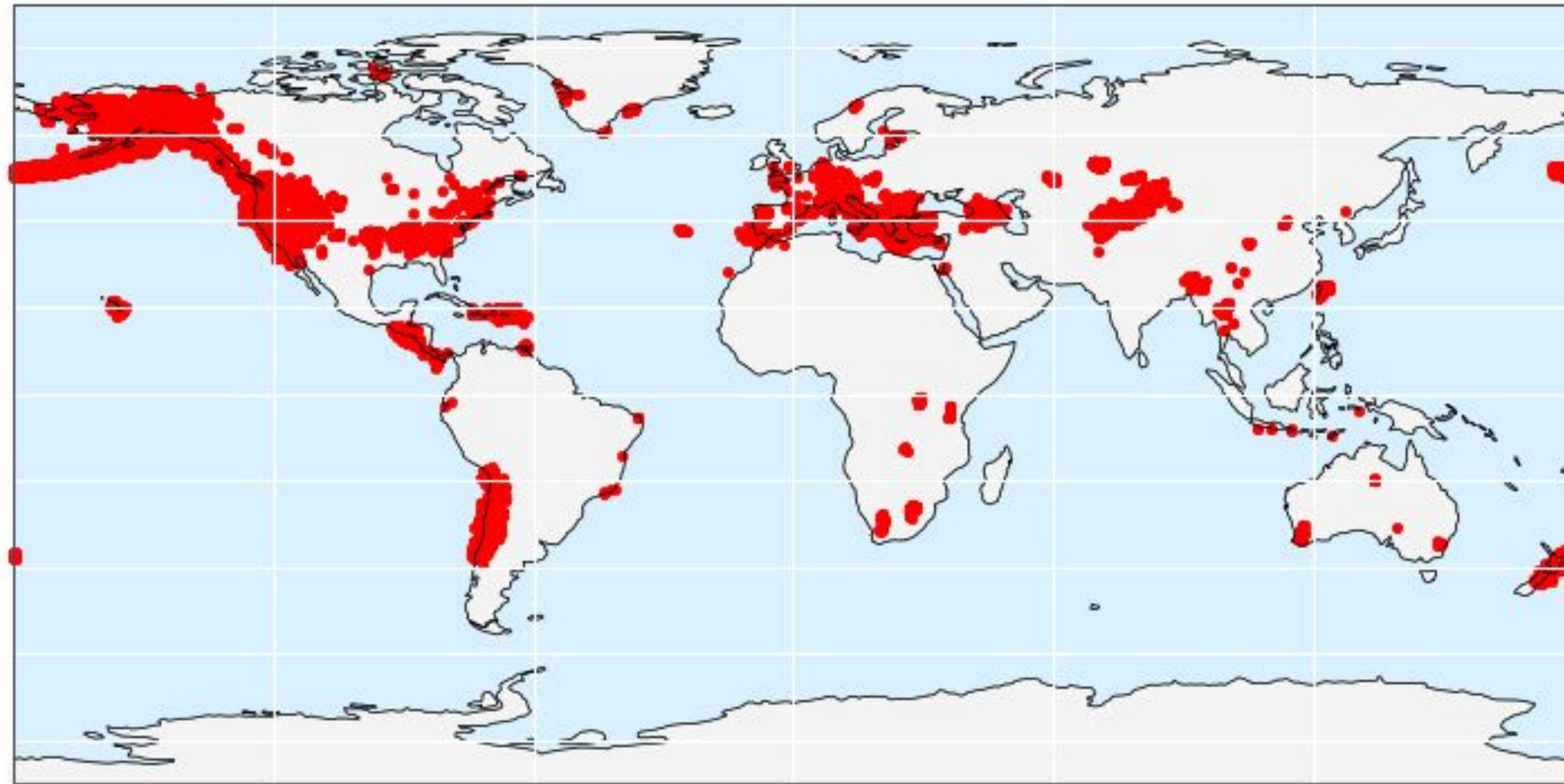
# Appendices

# More EDA





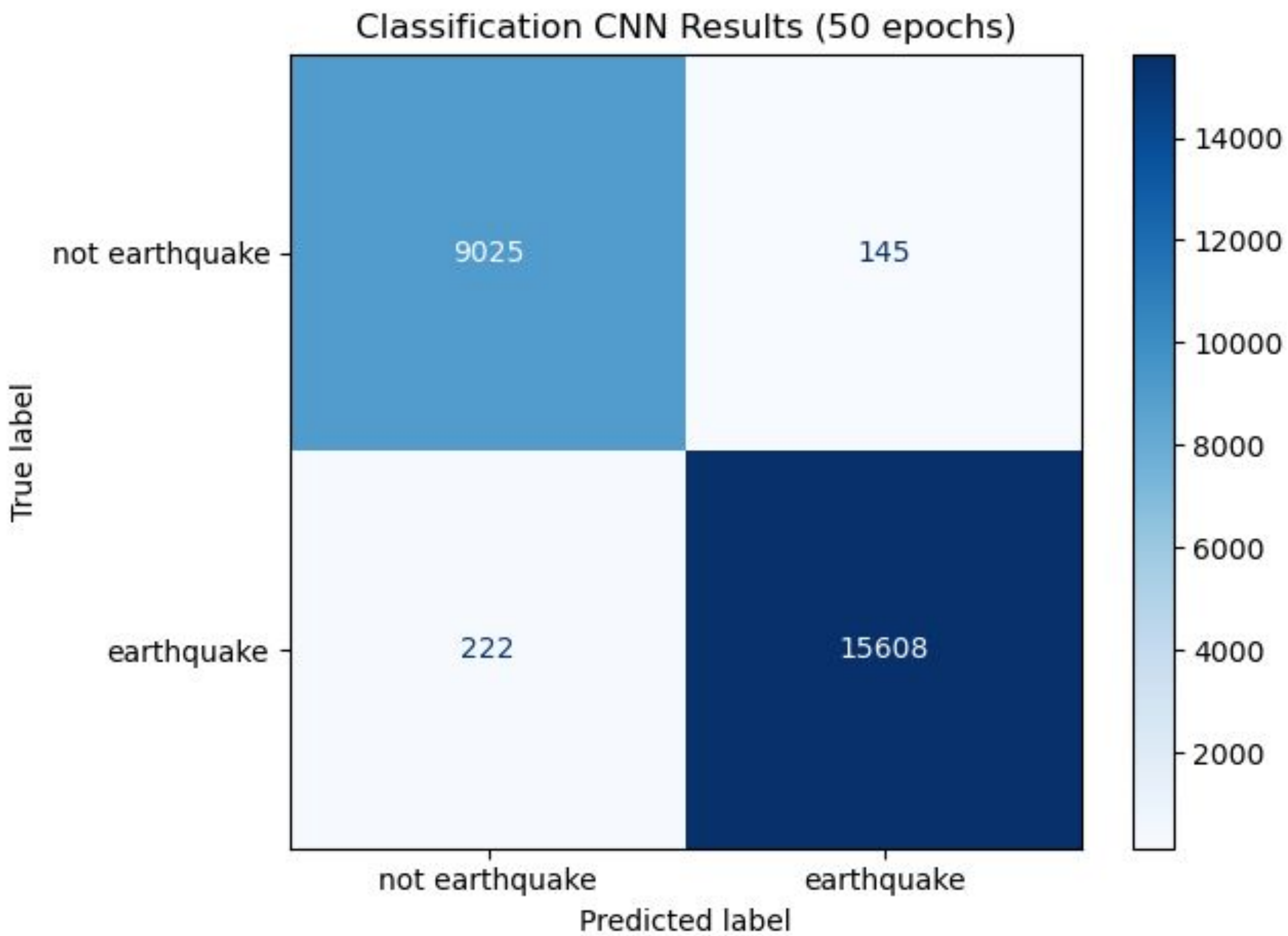
# More EDA



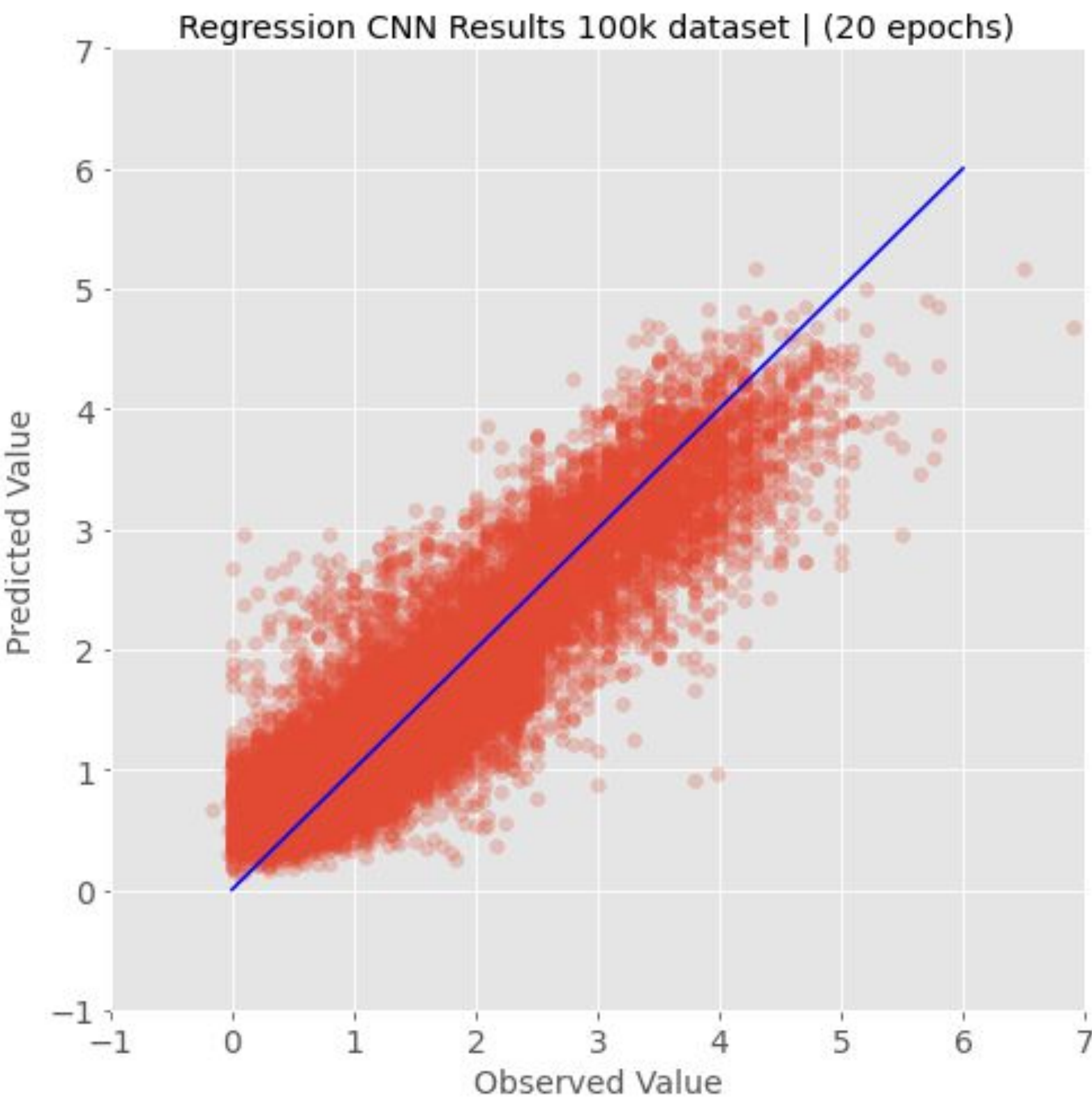
• earthquakes

# CNN Results

## Class Prediction



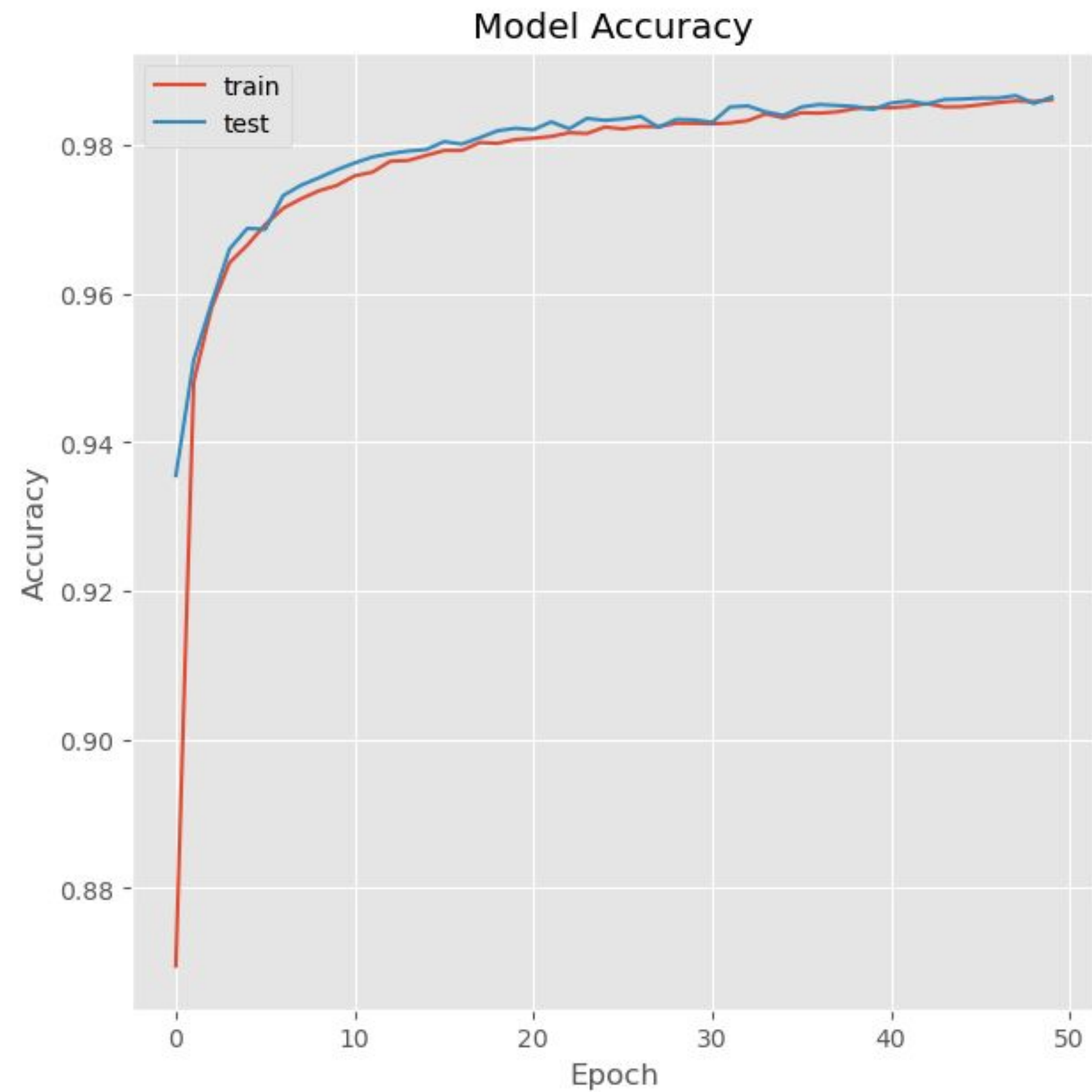
## Magnitude Prediction



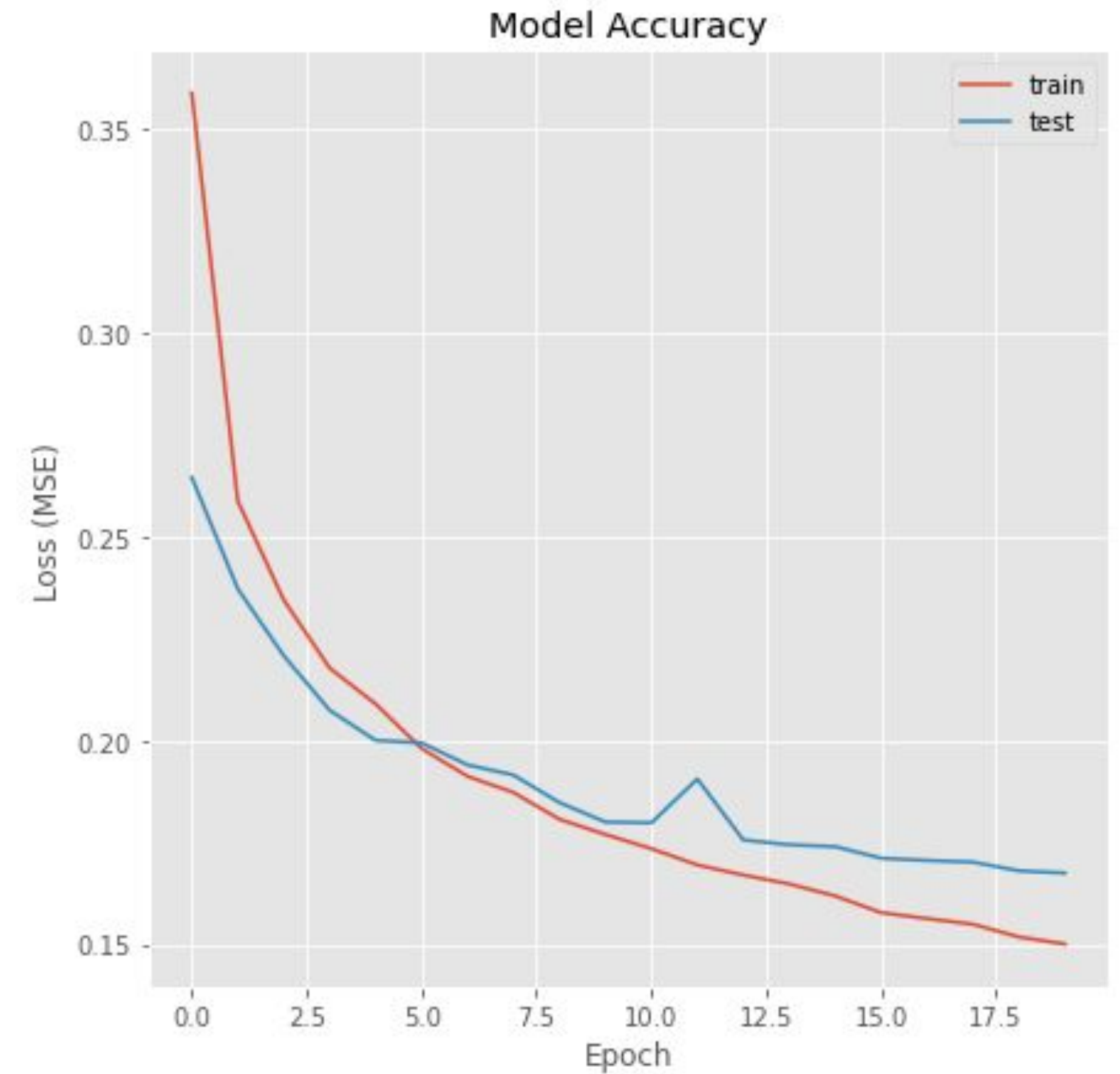


# CNN Results

## Class Prediction



## Magnitude Prediction





# CNN Results

## Class Prediction

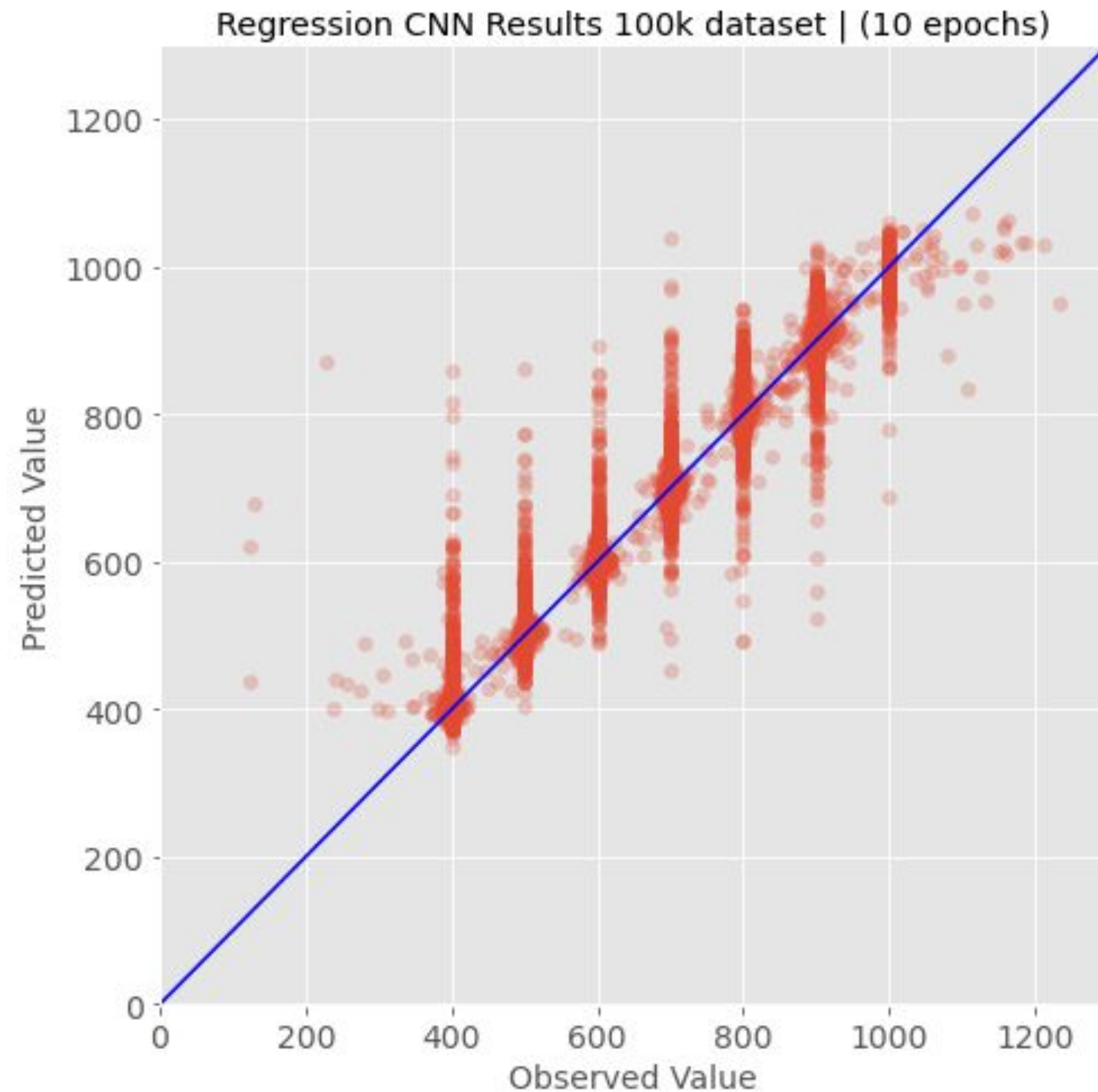
Model: "sequential_1"		
Layer (type)	Output Shape	Param #
=====		
conv2d_1 (Conv2D)	(None, 100, 150, 64)	1664
max_pooling2d_1 (MaxPooling2D)	(None, 50, 75, 64)	0
dropout_1 (Dropout)	(None, 50, 75, 64)	0
flatten_1 (Flatten)	(None, 240000)	0
dense_3 (Dense)	(None, 16)	3840016
dense_4 (Dense)	(None, 1)	17
=====		
Total params: 3,841,697		
Trainable params: 3,841,697		
Non-trainable params: 0		
=====		

## Magnitude Prediction

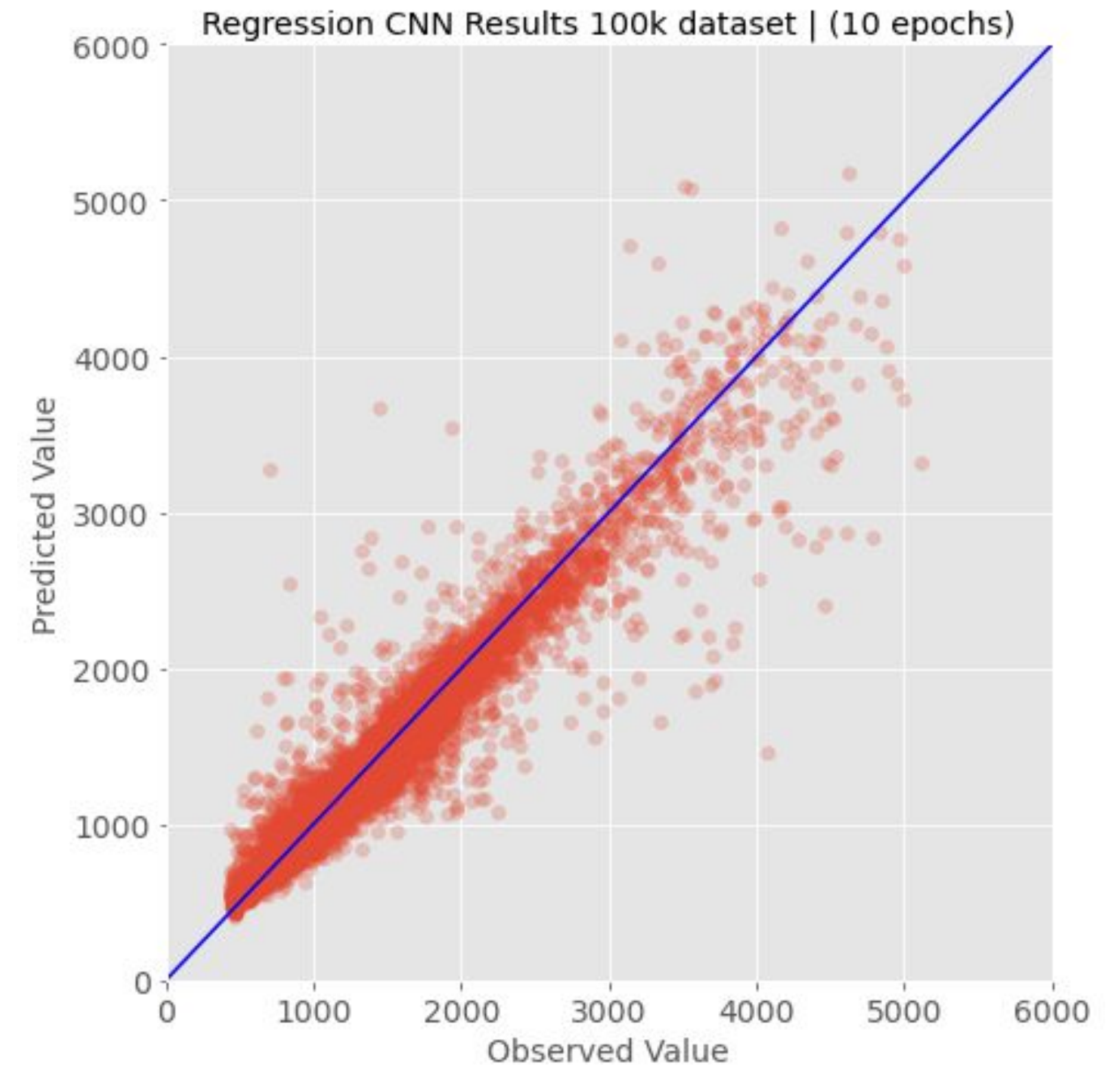
Model: "sequential_19"		
Layer (type)	Output Shape	Param #
=====		
conv2d_6 (Conv2D)	(None, 100, 150, 32)	832
max_pooling2d_6 (MaxPooling2D)	(None, 50, 75, 32)	0
dropout_24 (Dropout)	(None, 50, 75, 32)	0
flatten_6 (Flatten)	(None, 120000)	0
dense_54 (Dense)	(None, 64)	7680064
dense_55 (Dense)	(None, 16)	1040
dense_56 (Dense)	(None, 2)	34
=====		
Total params: 7,681,970		
Trainable params: 7,681,970		
Non-trainable params: 0		
=====		

# CNN Results

## P-Wave Prediction



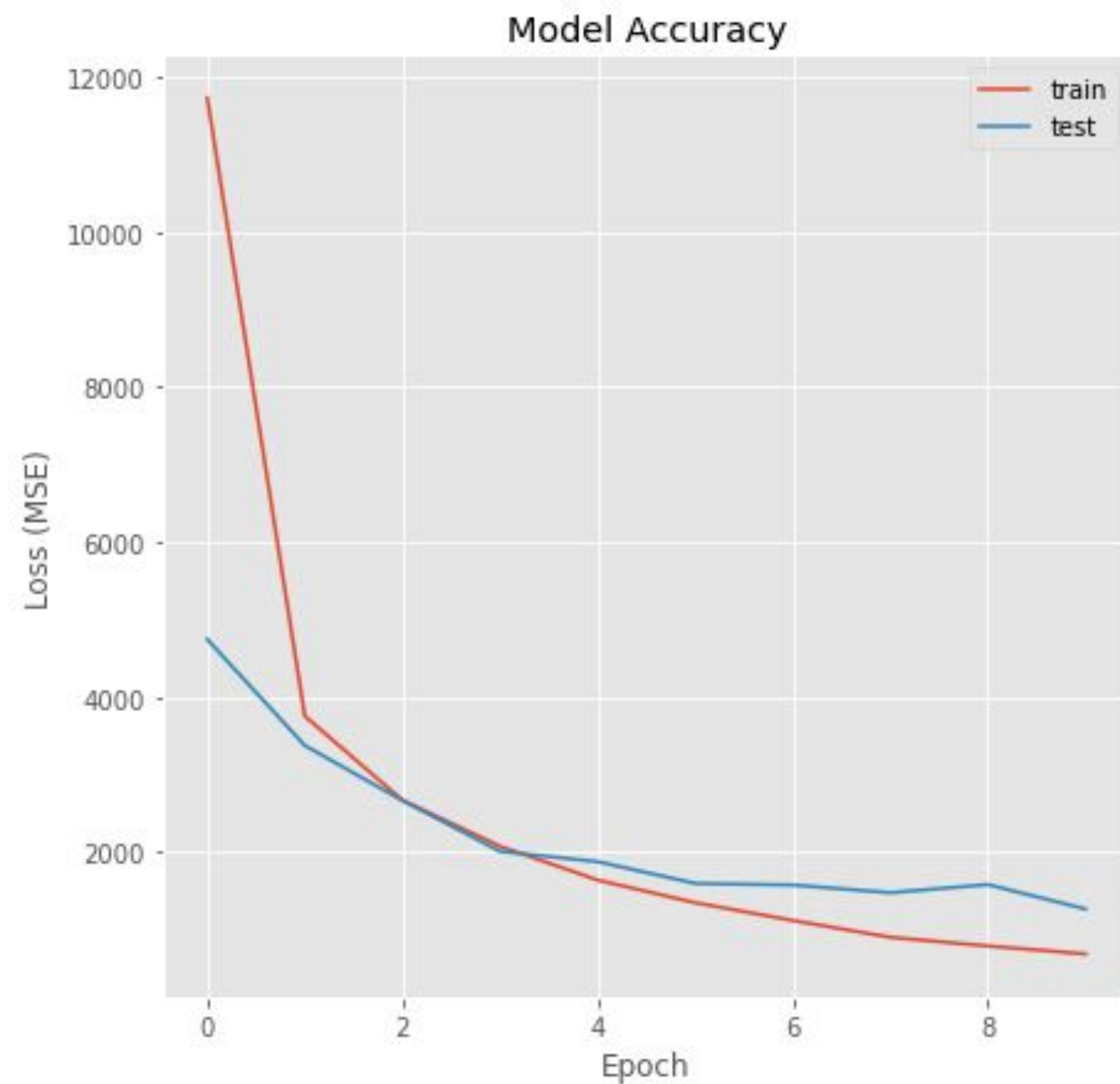
## S-Wave Prediction



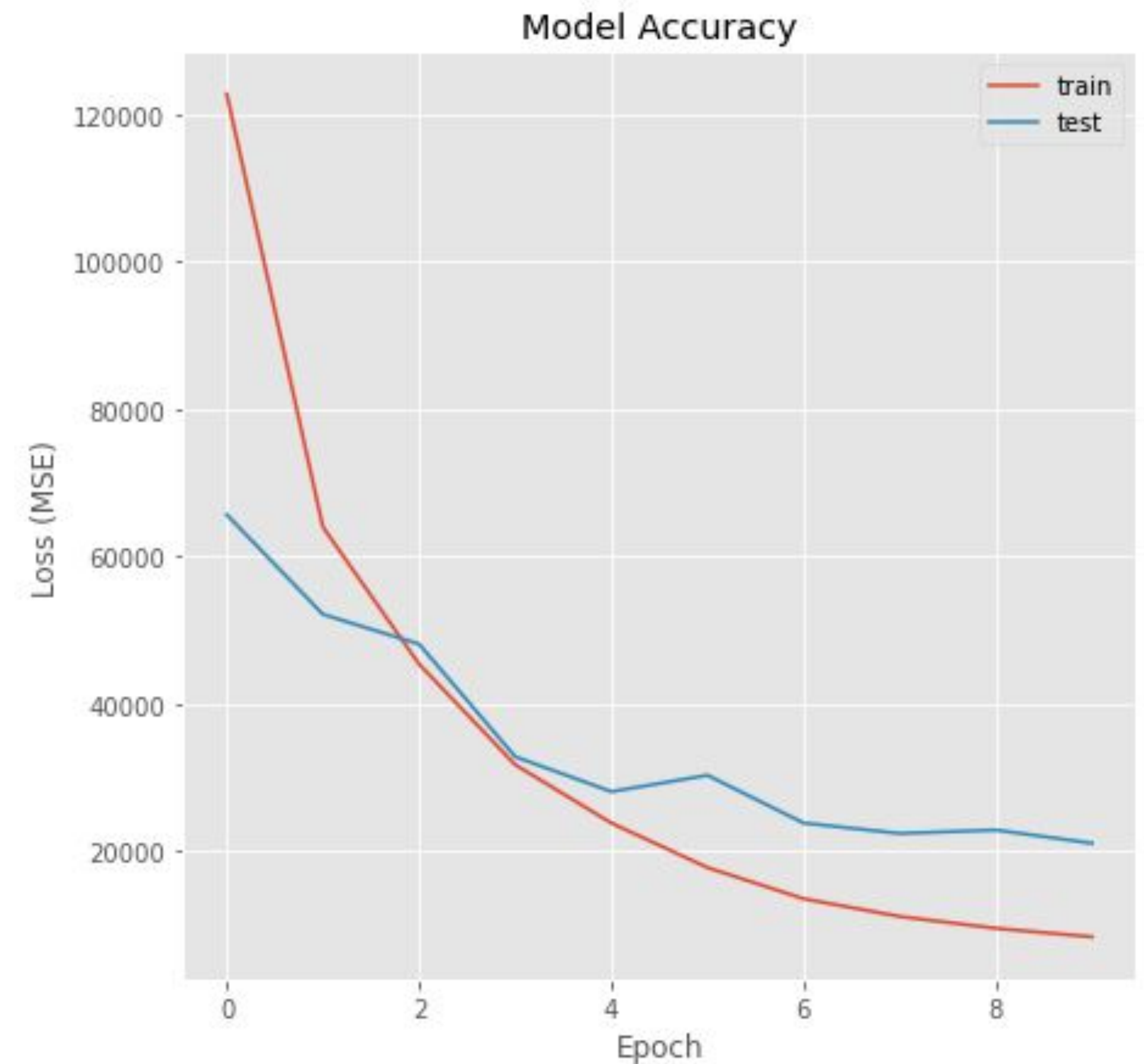


# CNN Results

## P-Wave Prediction



## S-Wave Prediction





# CNN Results

## P-Wave Prediction

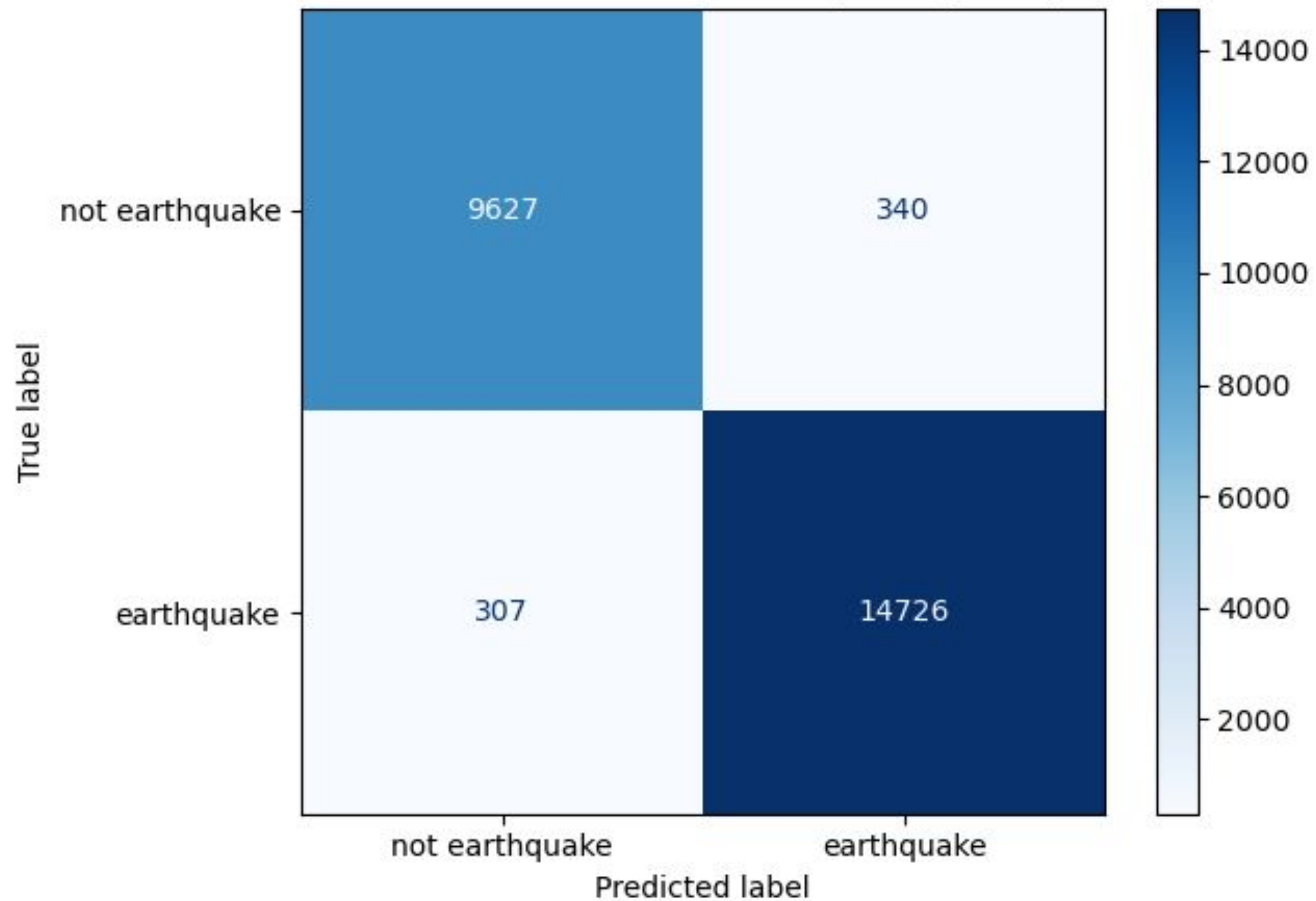
## S-Wave Prediction

Model: "sequential_5"		
Layer (type)	Output Shape	Param #
=====		
conv2d_5 (Conv2D)	(None, 110, 309, 64)	1664
-----		
max_pooling2d_5 (MaxPooling2D)	(None, 55, 154, 64)	0
-----		
dropout_5 (Dropout)	(None, 55, 154, 64)	0
-----		
flatten_5 (Flatten)	(None, 542080)	0
-----		
dense_12 (Dense)	(None, 64)	34693184
-----		
dense_13 (Dense)	(None, 16)	1040
-----		
dense_14 (Dense)	(None, 1)	17
=====		
Total params: 34,695,905		
Trainable params: 34,695,905		
Non-trainable params: 0		
-----		

# LSTM Results

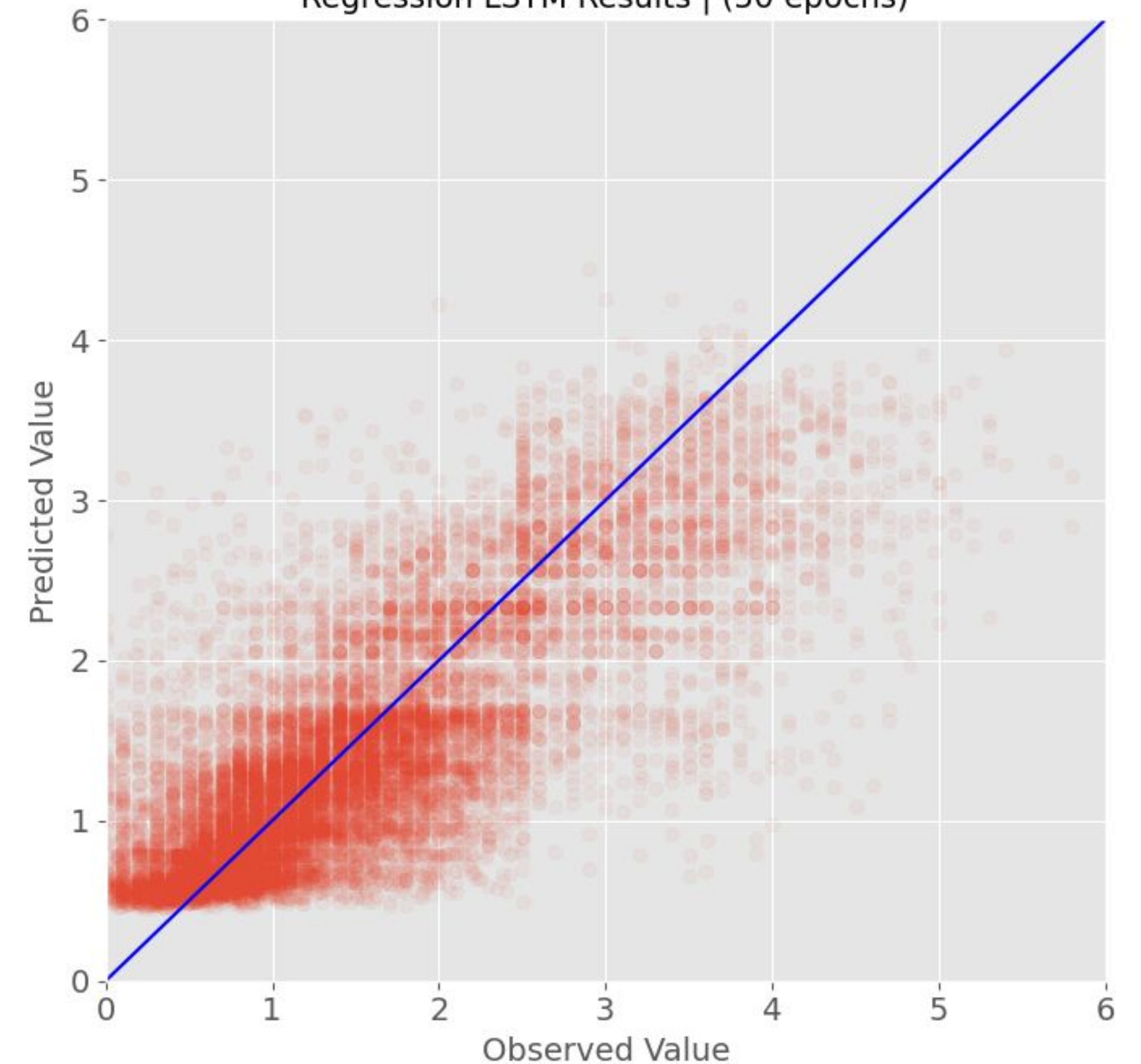
## Class Prediction

Classification LSTM Results (100 epochs)



## Magnitude Prediction

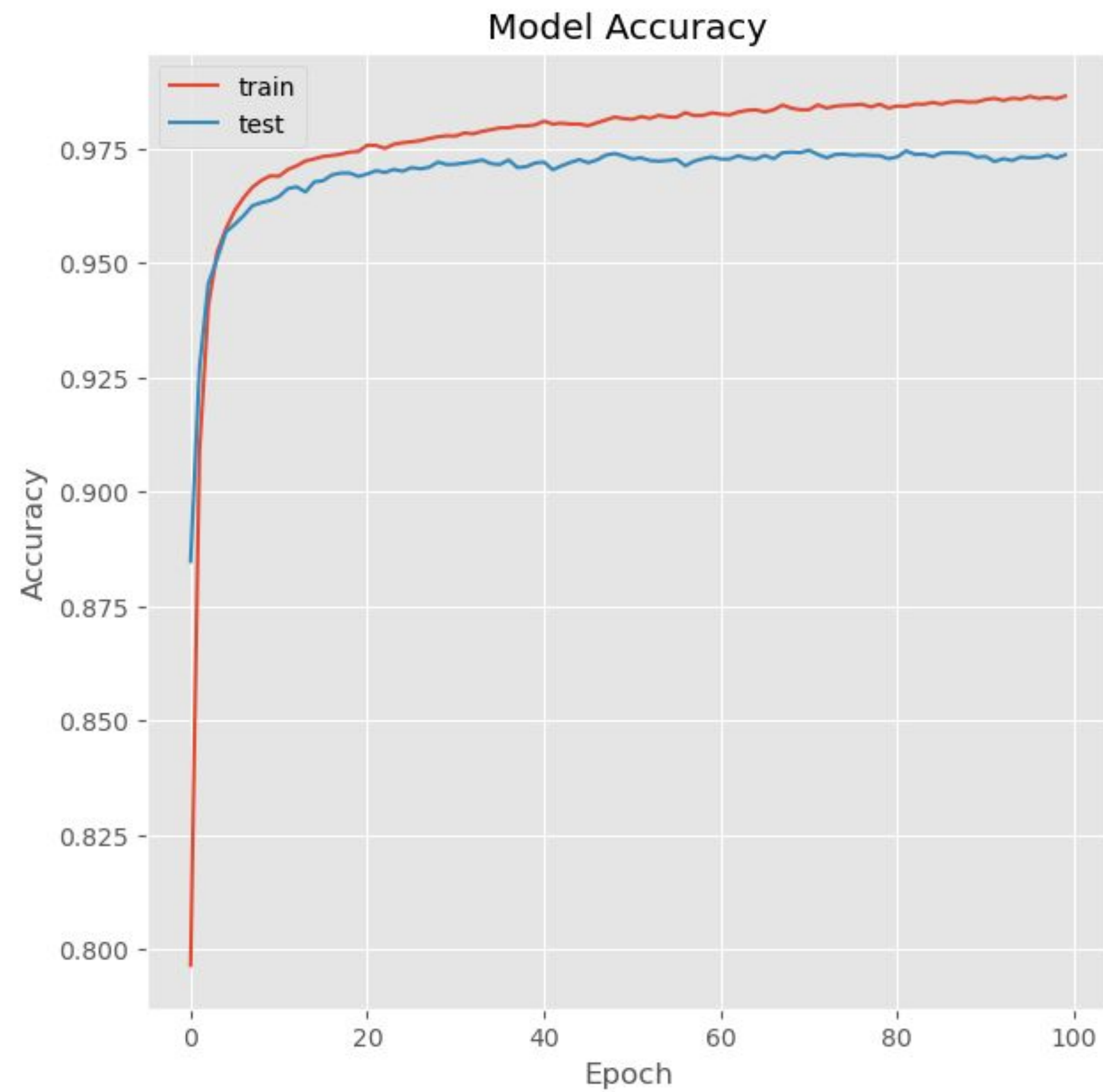
Regression LSTM Results | (50 epochs)



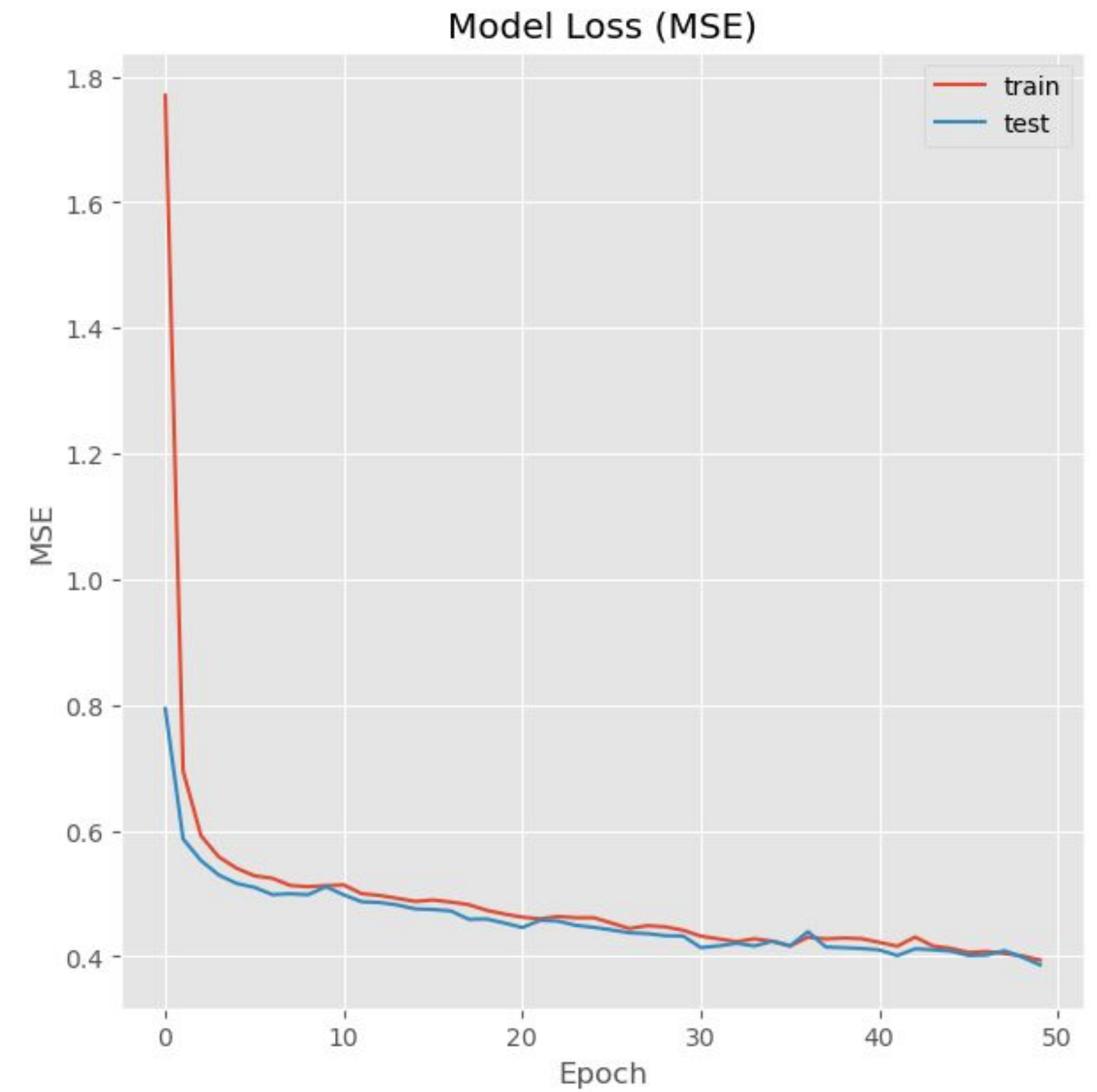


# LSTM Results

## Class Prediction



## Magnitude Prediction



# LSTM Results

## Class Prediction

Model: "sequential_10"		
Layer (type)	Output Shape	Param #
=====		
simple_rnn (SimpleRNN)	(None, 1, 64)	23360
lstm_6 (LSTM)	(None, 1, 64)	33024
dropout_10 (Dropout)	(None, 1, 64)	0
lstm_7 (LSTM)	(None, 32)	12416
dense_27 (Dense)	(None, 16)	528
dense_28 (Dense)	(None, 1)	17
=====		
Total params: 69,345		
Trainable params: 69,345		
Non-trainable params: 0		
=====		

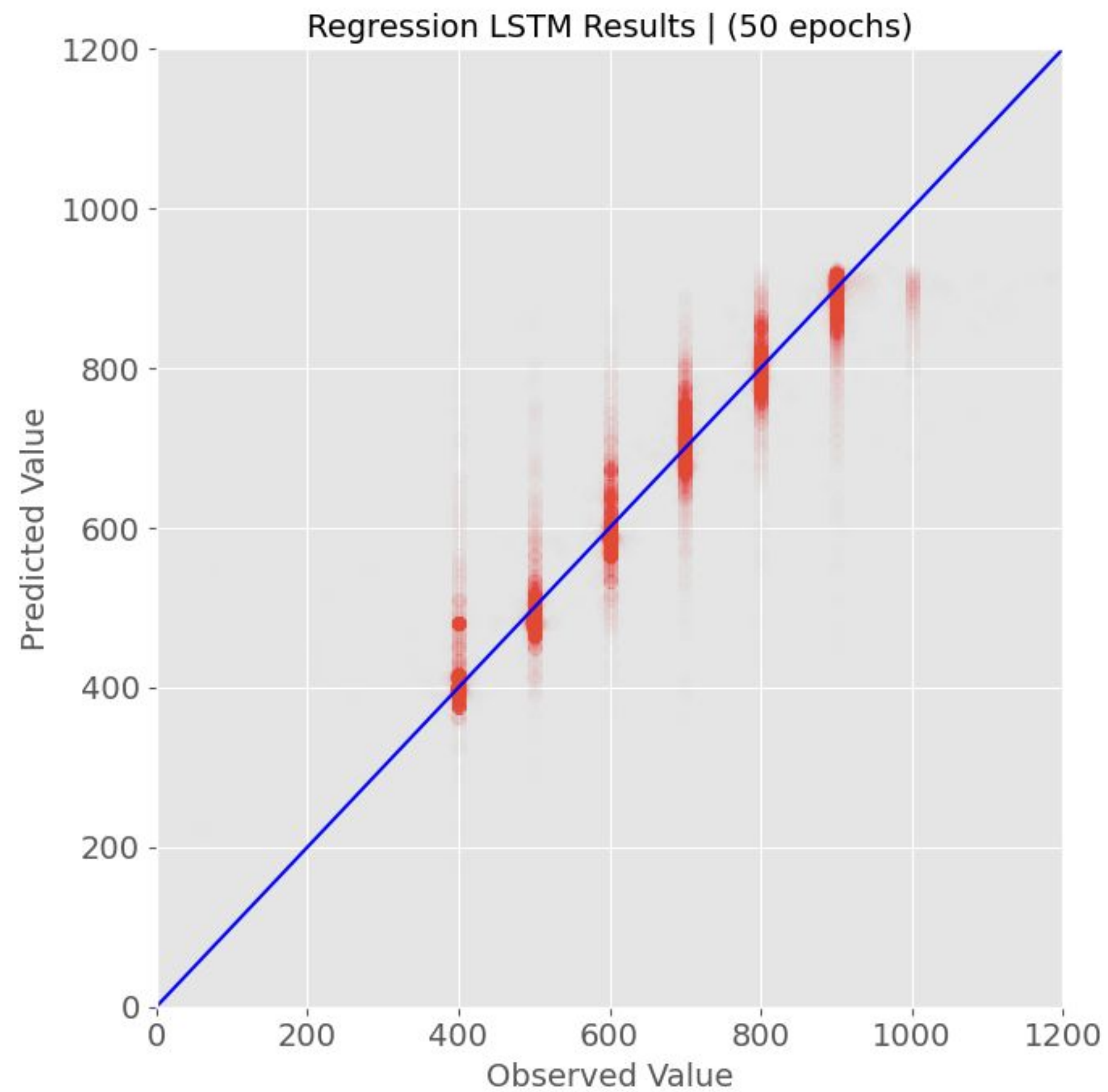
## Magnitude Prediction

Model: "sequential_9"		
Layer (type)	Output Shape	Param #
=====		
lstm_4 (LSTM)	(None, 1, 32)	42624
lstm_5 (LSTM)	(None, 32)	8320
dropout_9 (Dropout)	(None, 32)	0
dense_24 (Dense)	(None, 32)	1056
dense_25 (Dense)	(None, 16)	528
dense_26 (Dense)	(None, 1)	17
=====		
Total params: 52,545		
Trainable params: 52,545		
Non-trainable params: 0		
=====		

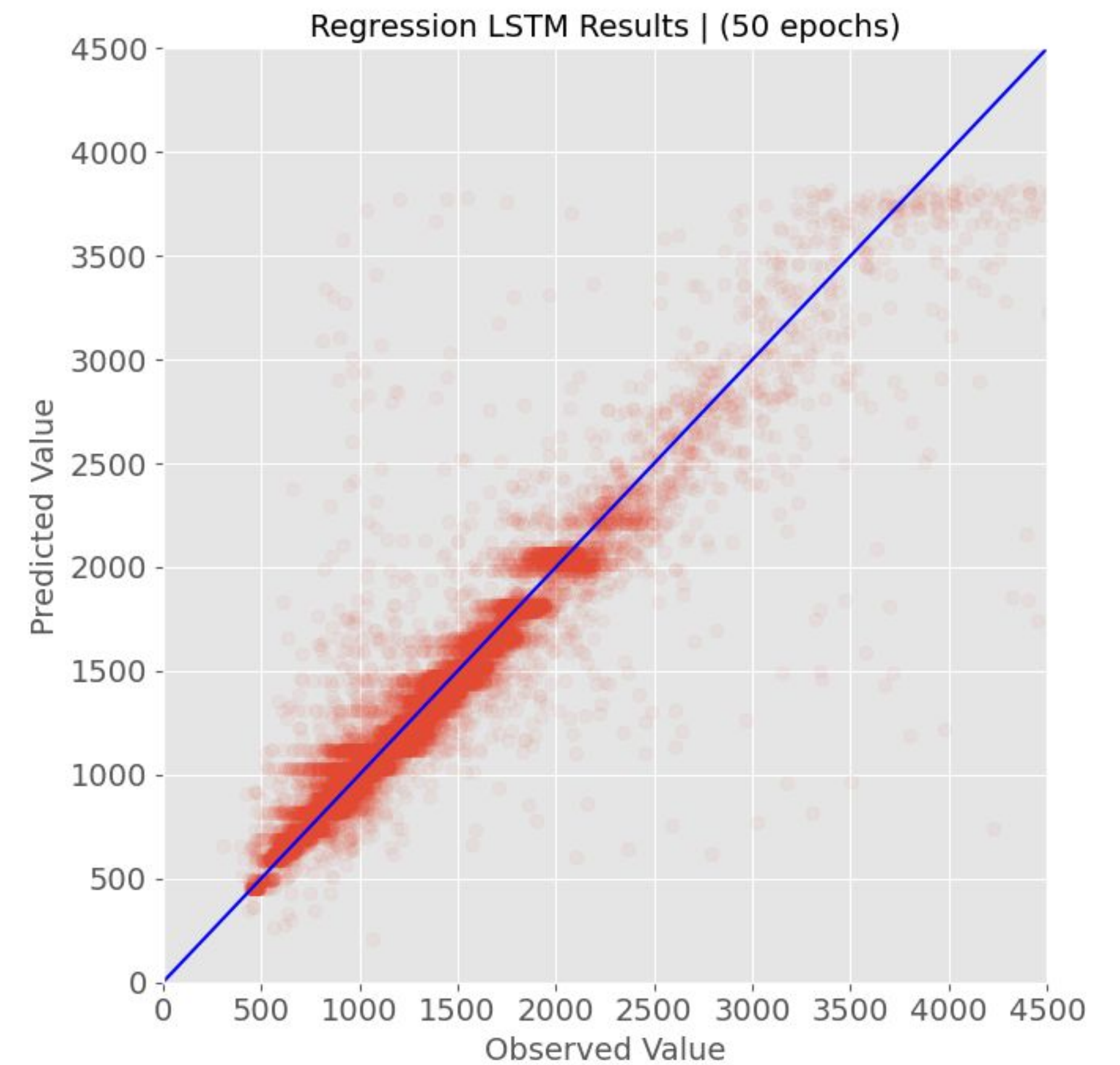


# LSTM Results

## P-Wave Prediction

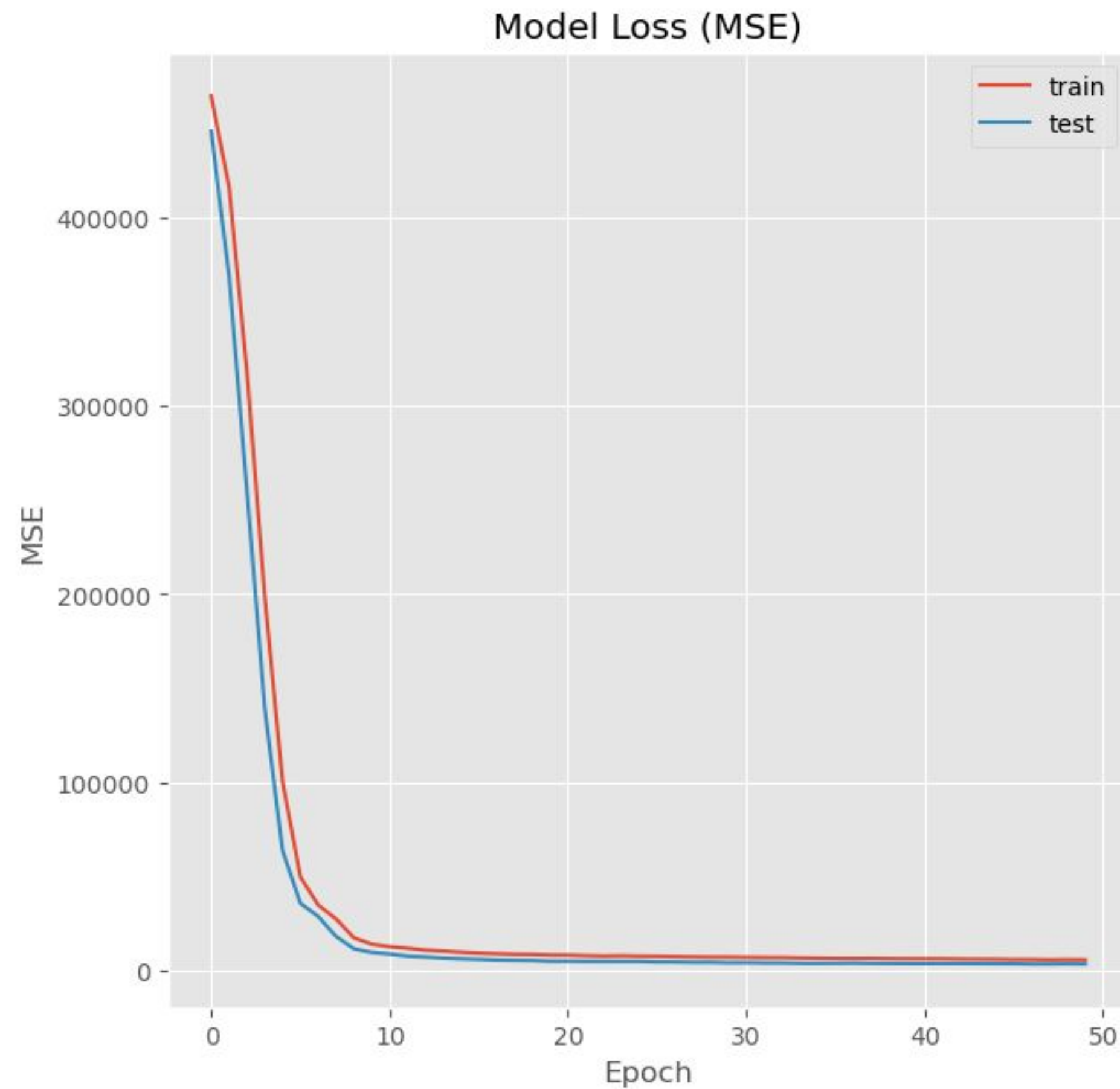


## S-Wave Prediction

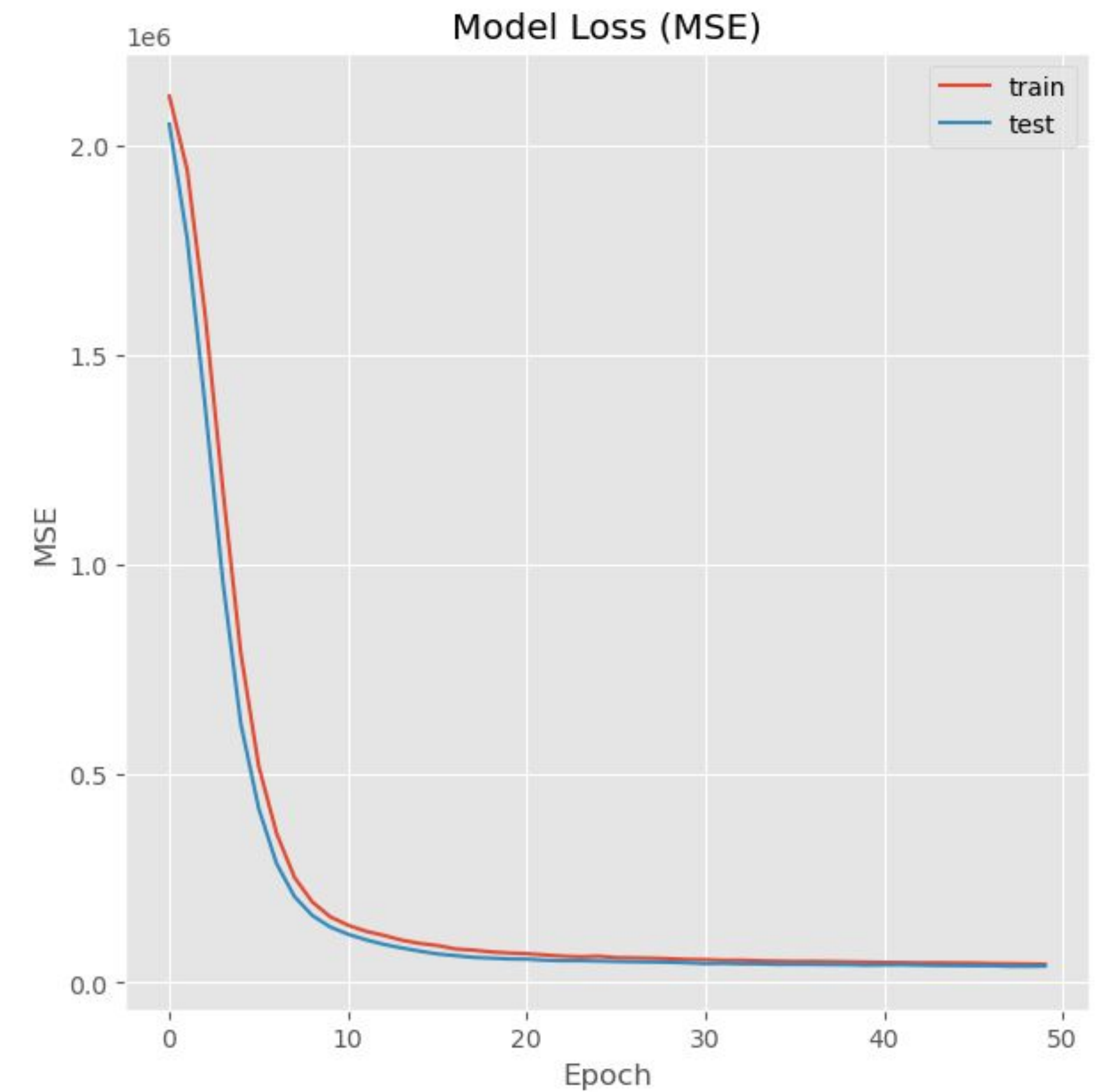


# LSTM Results

## P-Wave Prediction



## S-Wave Prediction





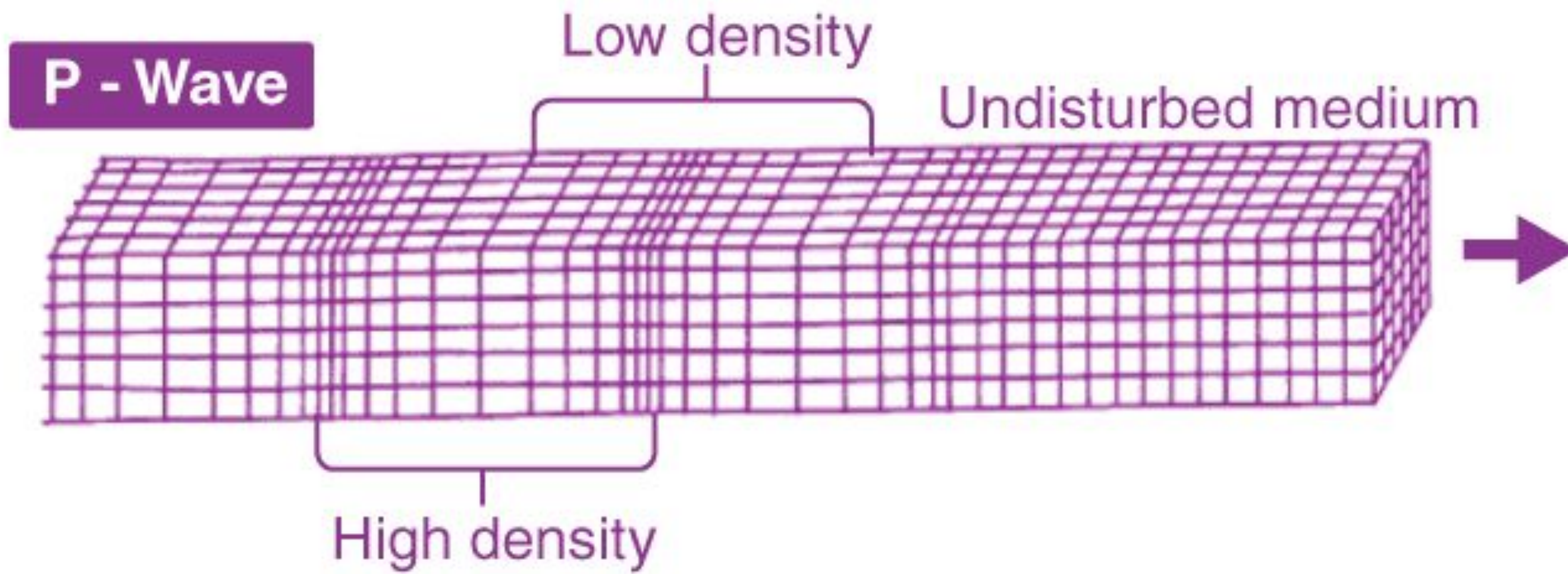
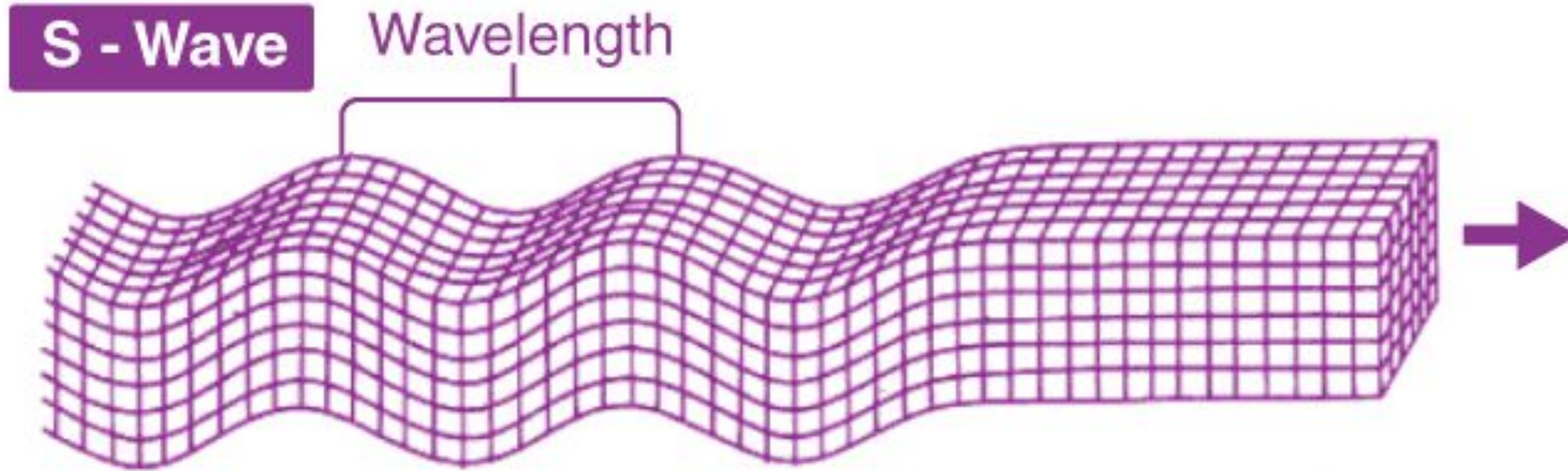
# LSTM Results

## P-Wave Prediction

## S-Wave Prediction

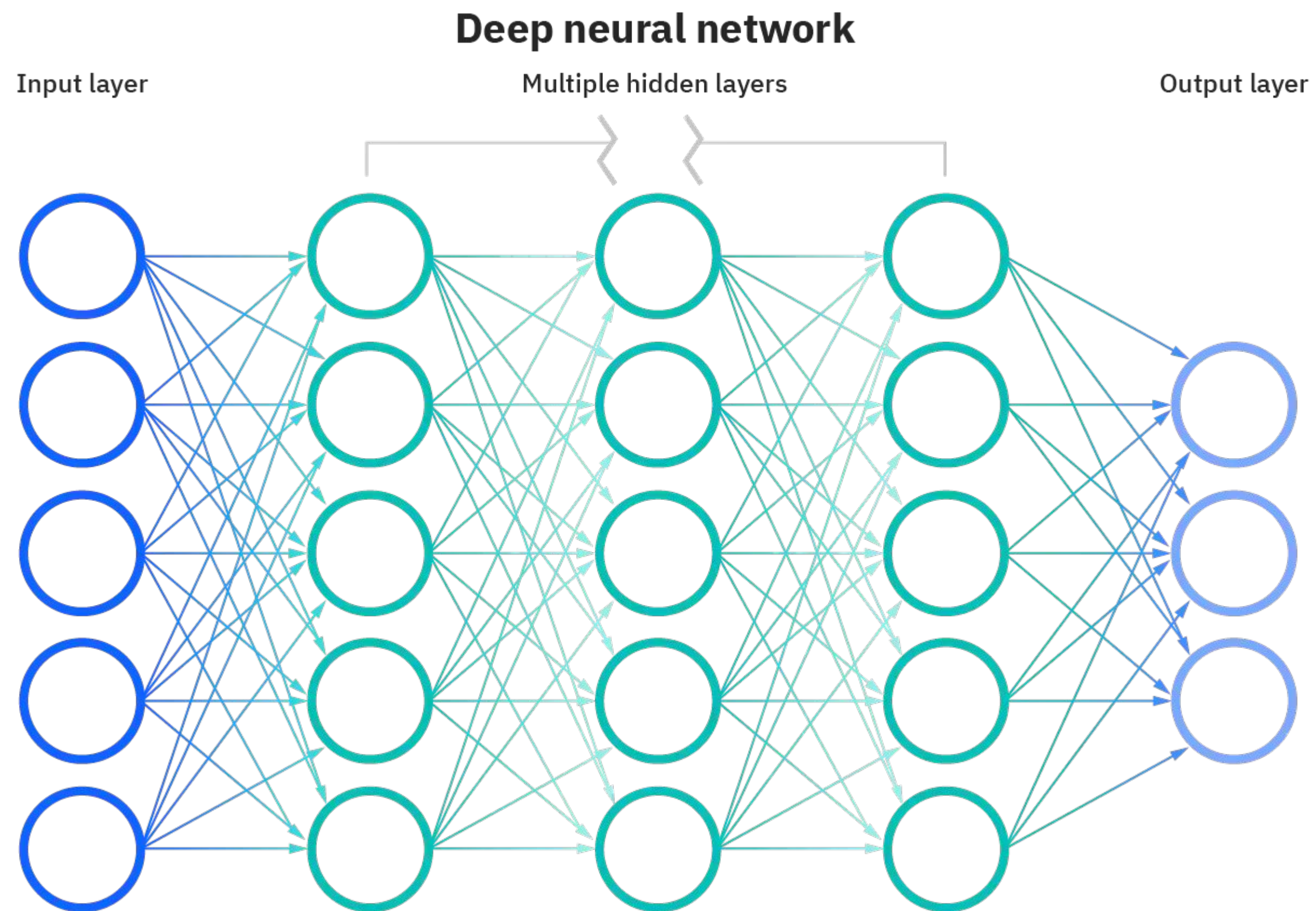
Model: "sequential_9"		
Layer (type)	Output Shape	Param #
=====		
lstm_4 (LSTM)	(None, 1, 32)	42624
lstm_5 (LSTM)	(None, 32)	8320
dropout_9 (Dropout)	(None, 32)	0
dense_24 (Dense)	(None, 32)	1056
dense_25 (Dense)	(None, 16)	528
dense_26 (Dense)	(None, 1)	17
=====		
Total params: 52,545		
Trainable params: 52,545		
Non-trainable params: 0		
=====		



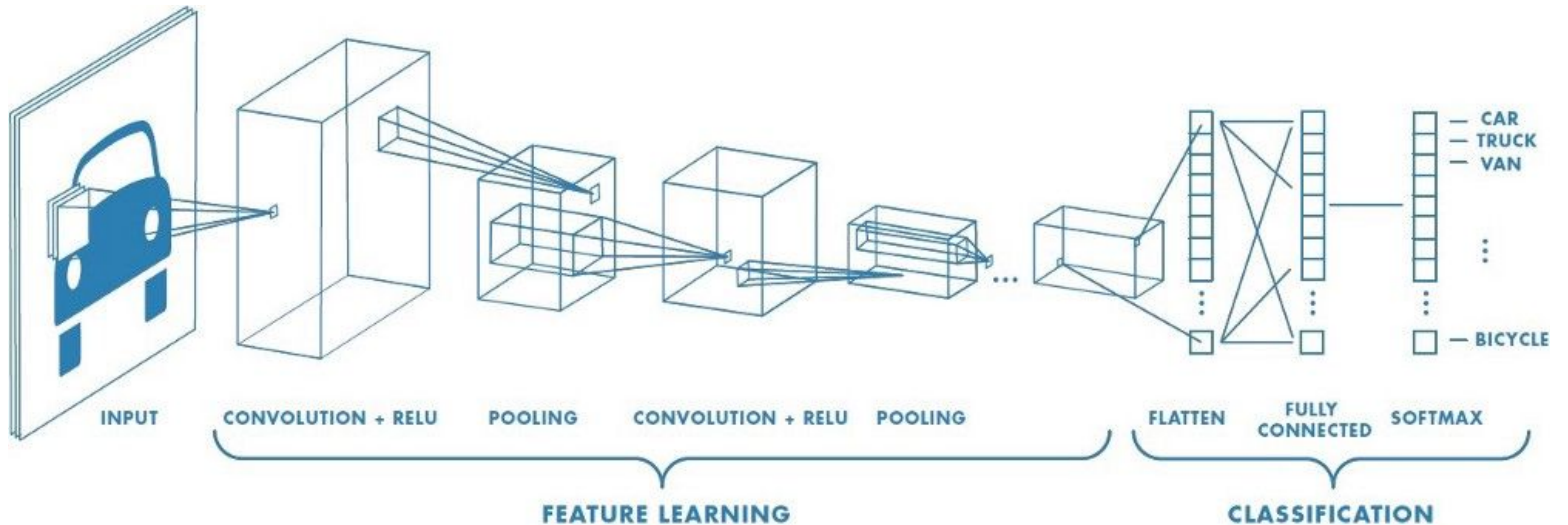




# Neural Networks

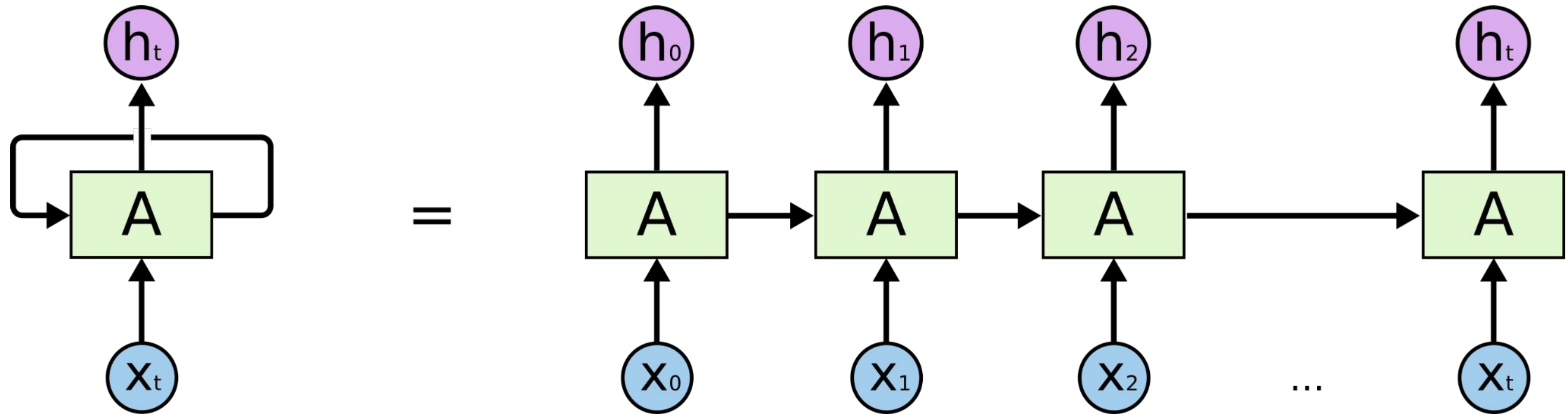


# Convolutional Neural Networks





# Recurrent Neural Networks



# LSTMs

