

Background and Motivation

- ❖ Manual identification of the arrival times of P and S waves from seismometer data offers great precision but involves human bias and effort
- ❖ Deep-learning models can effectively automate this process, but they do not incorporate scientific knowledge, nor do they offer any understanding of the classification process
- ❖ By pre-processing the data into scientifically meaningful features and using them as input to deep-learning models, we hypothesize improved performance and understanding

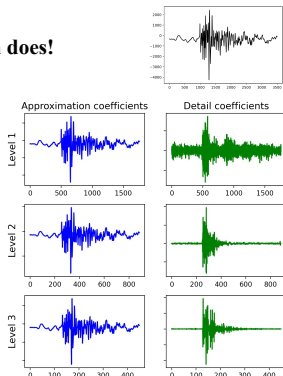
Why Wavelet?

- ❖ Seismic events like earthquakes possess characteristic temporal signatures
- ❖ Templates [1] have been used to capture these signatures.
- ❖ Wavelets [2] are a good way to operationalize this for the purposes of feature engineering for ML methods.

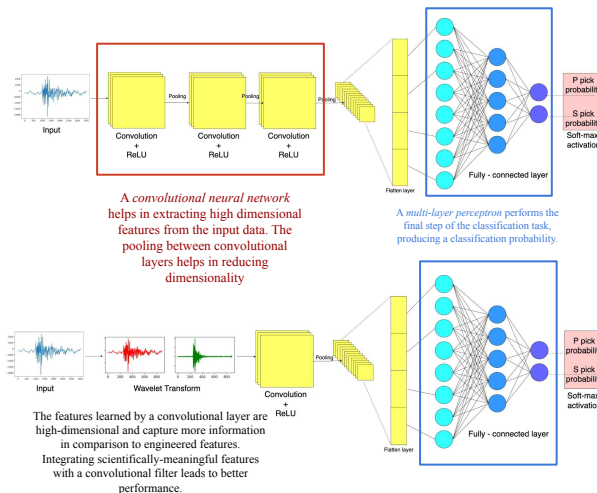
What a wavelet transform does!



The effect of this preprocessing step is not only to remove noise, but also more clearly expose the behavior of interest.



A Comparative Analysis



Details: One year of seismic data from the southern Taupo Volcanic Zone, New Zealand collected using a short-period seismometer with three component sensor aligned to North. Split into training and testing sets in a 70:30 ratio.

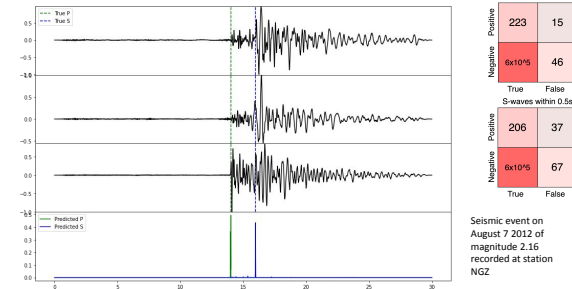
Future Work

- ❖ Explore more feature engineering mechanisms for the seismic waves such as continuous wavelet transform, fingerprints.
- ❖ Explore a neural network model that can retain memory such as Long Short Term Memory (LSTM).
- ❖ Build a model with the ability to classify the various kinds of earthquake events such as tremor and low-frequency earthquakes.

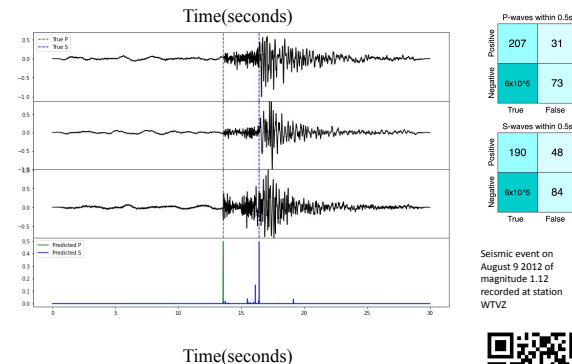
Results

- ❖ Model with featurized input exhibits better performance in terms of precision and proximity to the analyst-picked arrival times.
- ❖ Model with raw input is able to detect similar number of events but with lesser precision.

Wavelet transformed model prediction



Raw input model prediction



Time(seconds)

Scan the QR for more details!

