



INSTITUTE OF EARTH SCIENCES
ACADEMIA SINICA

SUMMER INTERNSHIP PROGRAM

Deep Learning Based Seismic Wave Identification

Author :

Gauss CHANG

Advisor :

Dr. Kuo-Fong MA

Dr. Wen-Tzong LIANG

August 25, 2023

Abstract

Abstract Taiwan is a region frequently hit by earthquakes and has a dense population. This situation exerts significant pressure on structural facilities and the earthquake early warning systems. To achieve more timely earthquake warnings and post-seismic structural integrity analysis, we collaborate with QGIS in this project. Our aim is to promptly collect earthquake signals while avoiding the pitfalls of the traditional STA/LTA method [1], which is prone to false activations [2] [3].

I Introduction

Taiwan is a region characterized by frequent earthquakes and a dense population. These conditions place immense pressure on building structures and earthquake early warning systems. In pursuit of earthquake early warning and post-quake structural integrity analysis, we introduce “QGIS”. Harnessing the power of deep-learning, we present an affordable solution for earthquake detection. This project is meant to be:

1. Cost-Efficient Alerts: Using MEMS technology, we can widely distribute the sensor at low costs.
2. Instantaneous Detection: Real-time alerts for swift precautionary actions.
3. Widespread Coverage: Designed for city-wide deployment to ensure comprehensive safety.
4. Customization: Fine-tuning capabilities to cater to unique building or location requirements.
5. Structural Integrity Monitoring: For different floors in a single building, identify the damaged ones using the waveform.

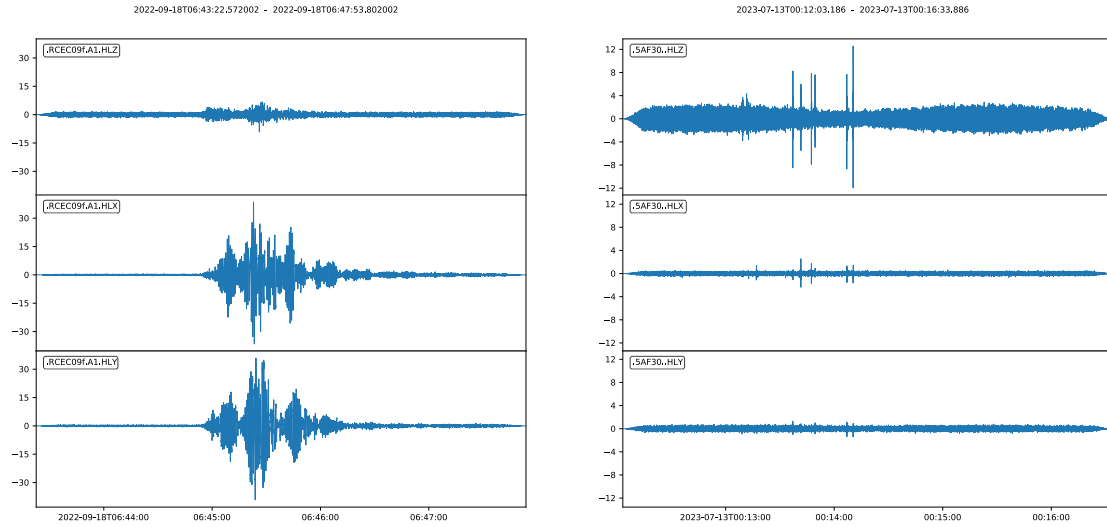


Figure 1: Seismic and Noise Waveform

In Figure 1 shows how the seismic and noise waveform look like.

I.I Importance of this Work

I.II Model Structure

Following is a model structure I have used in this project.

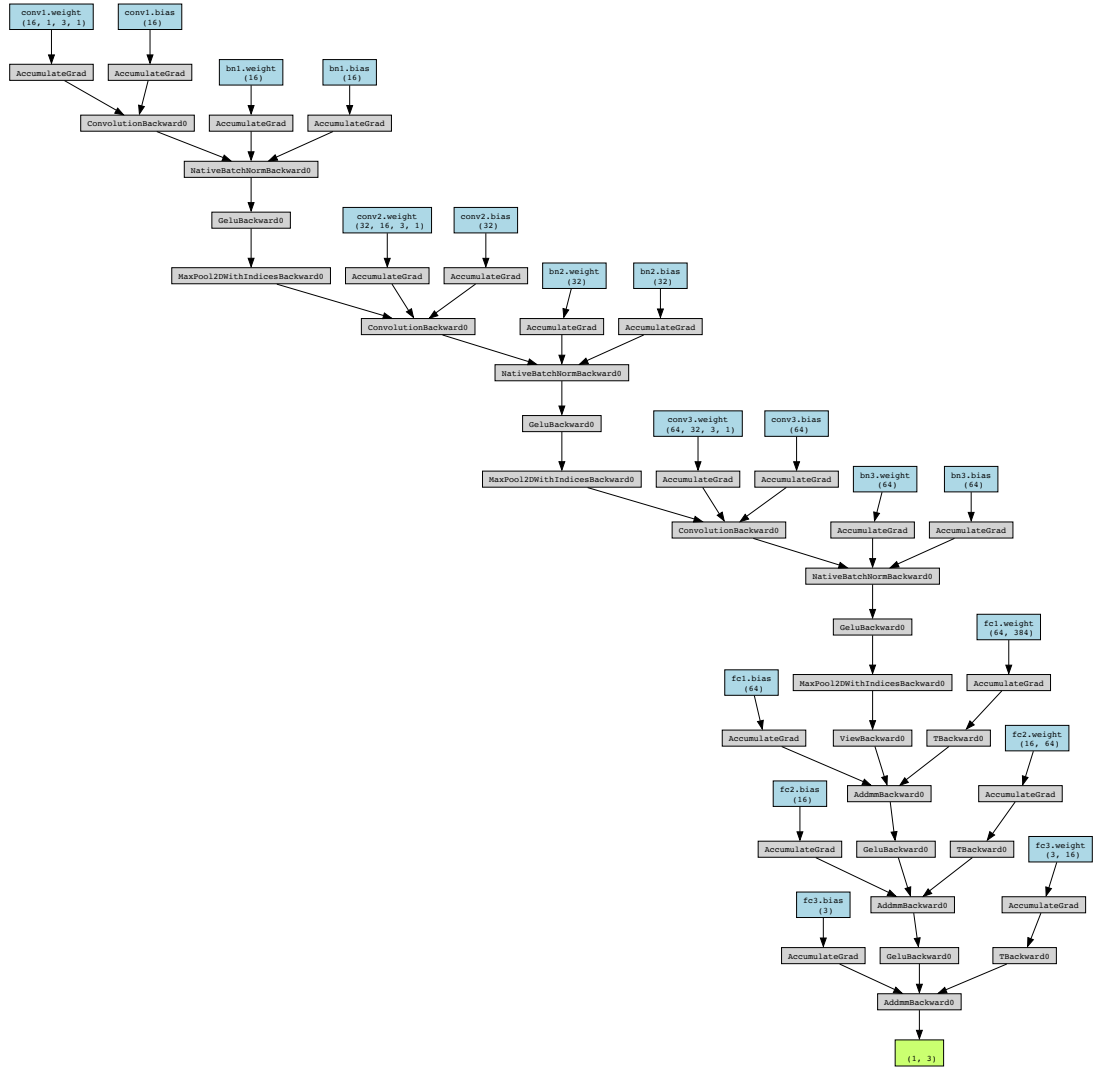


Figure 2: Model Structure

II Training Data Structure

Given that the QGIS project is still in its early stages, its earthquake information is somewhat lacking compared to more established databases. Therefore, for this initiative, we utilized two databases for training. Initially, we pre-trained our model using data from the STanford EArthquake Dataset (STEAD) [4].

The STEAD database contains 73,582 usable noise and event recordings, divided approximately into 40% Noise, 30% P-waves, and 30% S-waves.

Building on this, we fine-tuned the model using the QGIS dataset, which comprises 241 usable events and 725 noise recordings. The distribution within the QGIS dataset is approximately 60% Noise, 20% P-waves, and 20% S-waves.

Figures 3 showcase the spectrogram from the QGIS dataset for two different buildings: IES and RCEC. As observed from the diagrams, the earthquake data recorded by IES predominantly has wave durations around 4 Hz, whereas the data recorded by RCEC mostly centers around 1.3 Hz. This suggests that earthquake signals within buildings are highly building- dependent.

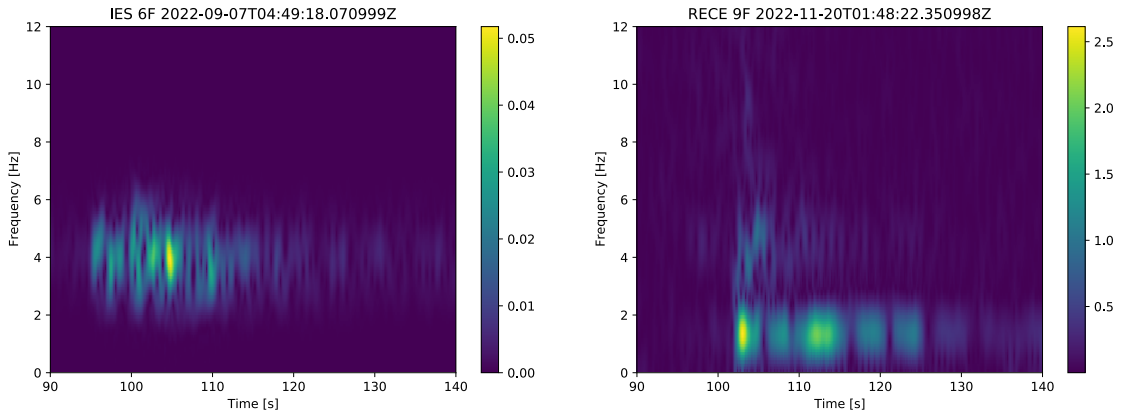


Figure 3: Spectrogram of Training Data

III Implementation

IV Result

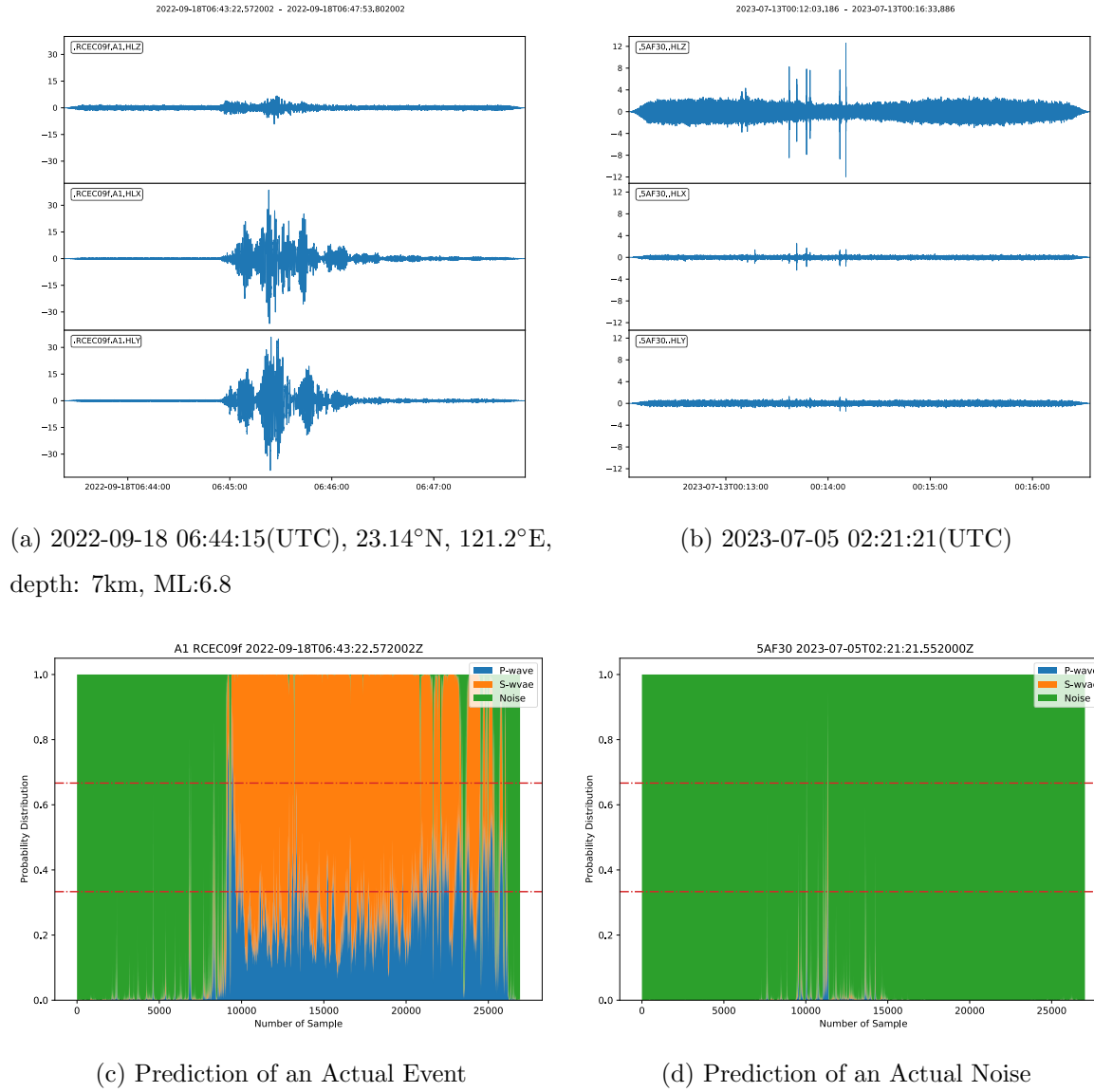


Figure 4: Prediction Results

V Conclusion

References

- [1] Jian Li, Mengmin He, Gaofeng Cui, Xiaoming Wang, Weidong Wang, and Juan Wang. A novel method of seismic signal detection using waveform features. *Applied Sciences*, 10(8), 2020.
- [2] Timothy Clements. Earthquake Detection with tinyML. *Seismological Research Letters*, 94(4):2030–2039, 04 2023.
- [3] Zachary E. Ross, Men-Andrin Meier, Egill Hauksson, and Thomas H. Heaton. Generalized Seismic Phase Detection with Deep Learning. *Bulletin of the Seismological Society of America*, 108(5A):2894–2901, 08 2018.
- [4] S. Mostafa Mousavi, Yixiao Sheng, Weiqiang Zhu, and Gregory C. Beroza. Stanford earthquake dataset (stead): A global data set of seismic signals for ai. *IEEE Access*, 7:179464–179476, 2019.