

# Machine Learning-Based Method for Earthquake Analysis in Japan: A Preliminary Study

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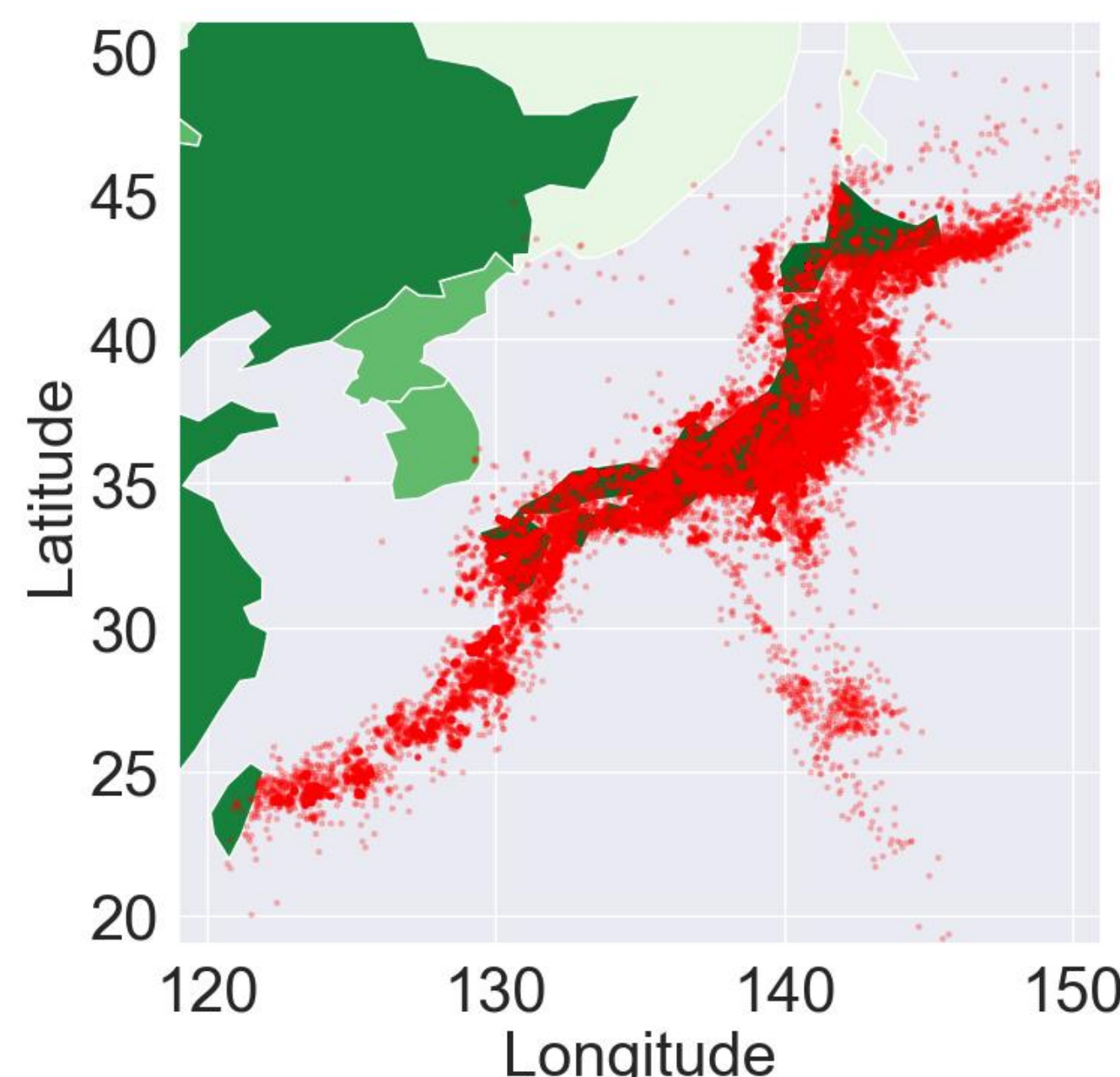
## Earthquake Occurrences in Japan

### Purpose

In Japan, earthquakes frequently occur. Each time causes devastating damages to human lives and infrastructure. The purpose of [this preliminary study](#) is to utilize [machine learning-based method](#) to analyze the earthquake data in Japan.

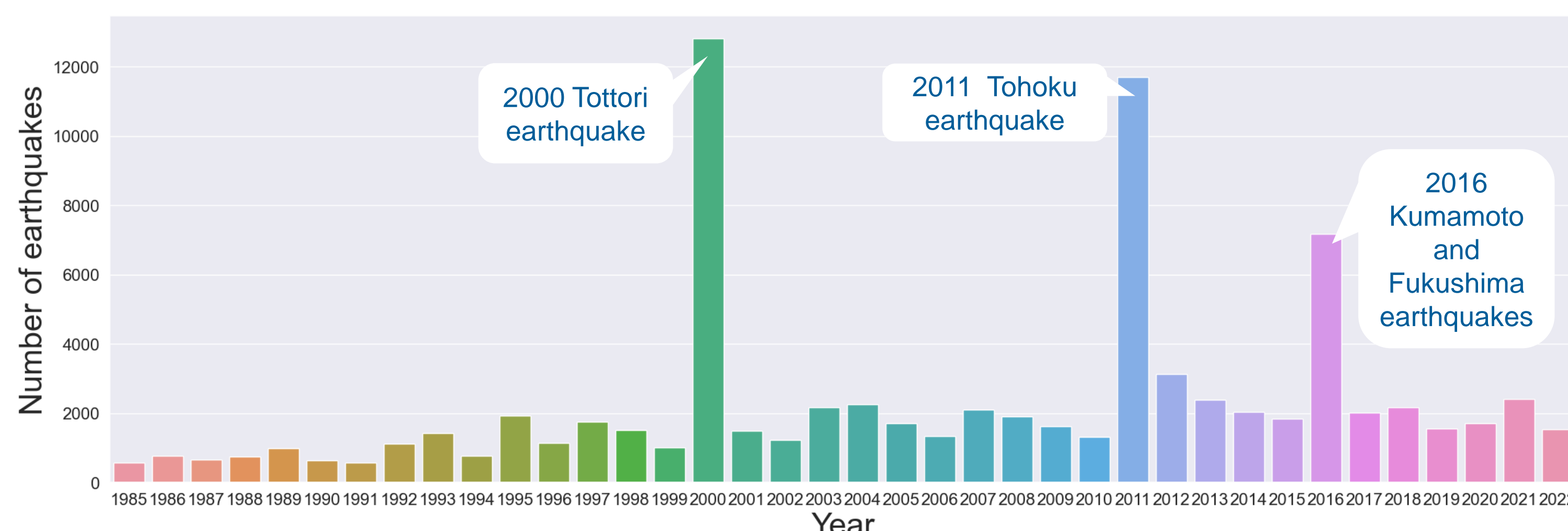
### Data Format and Data Visualization

- We use earthquake data during 1985 to 2022 from Japan Meteorological Agency.
- Data contains time, locations, magnitude and depth.



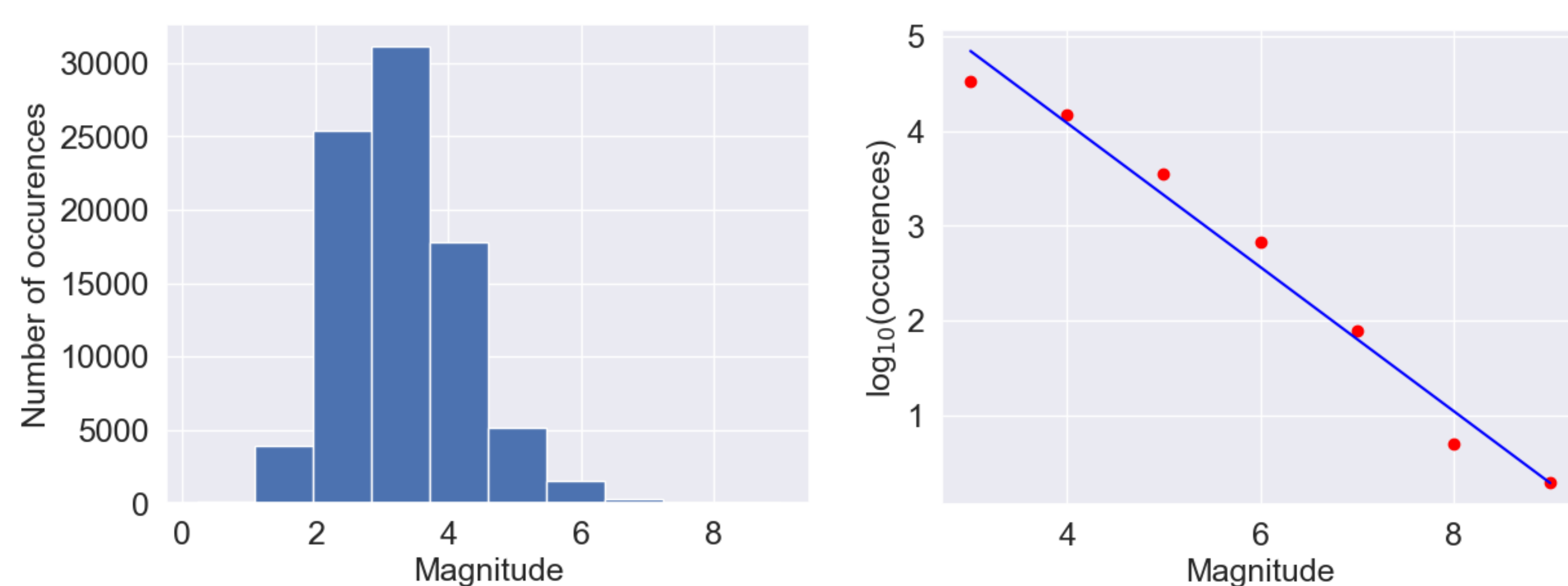
#### Locations of earthquake occurrences

In Japan, earthquakes are more frequent on the Pacific side because it is in the Ring of Fire.



#### Relationship between the number of occurrences and time

Every earthquake accumulates energy before occurring. From the graph above, it can be concluded that in the past, the year with high rate of earthquakes was in 2000, 2011, and 2016.



#### Relationship between the number of occurrences and magnitude (Gutenberg-Richter Law)

$$N = 10^{7.115 - 0.759M}$$

$N$ : number of earthquake occurrences

$M$ : magnitude

#### Train-Test Split

- For relative intensity model, we divided 70% of data to train set and 30% of data to test set.
- For machine learning, we normalized data. Then, we separated data to X(timestamp, latitude, longitude) and Y(magnitude, depth). After that, we divided 80% of data to train set, 10% of data to test set, and 10% of data to validation set.

### Relative Intensity (Conventional)

#### RI Score Calculation

- Study area is defined as point with even spacing ranges from 30°E 128°N to 46°E 146°N
- RI Score is min-max normalization of Cumulative Benioff Strain of a point in study area.
- Cumulative Benioff Strain of a point is the sum of square root of seismic energy from M5.0 or larger earthquake around the point, which can be written mathematically as following.

$$B_{xy}(t) = \sum_{i=1}^{N_{xy}(t)} \sqrt{E_{xy}^{(i)}}$$

- Seismic energy is estimated by magnitude using the following formula.

$$E_{xy}^{(i)} = e^{5.24 + 1.44M}$$

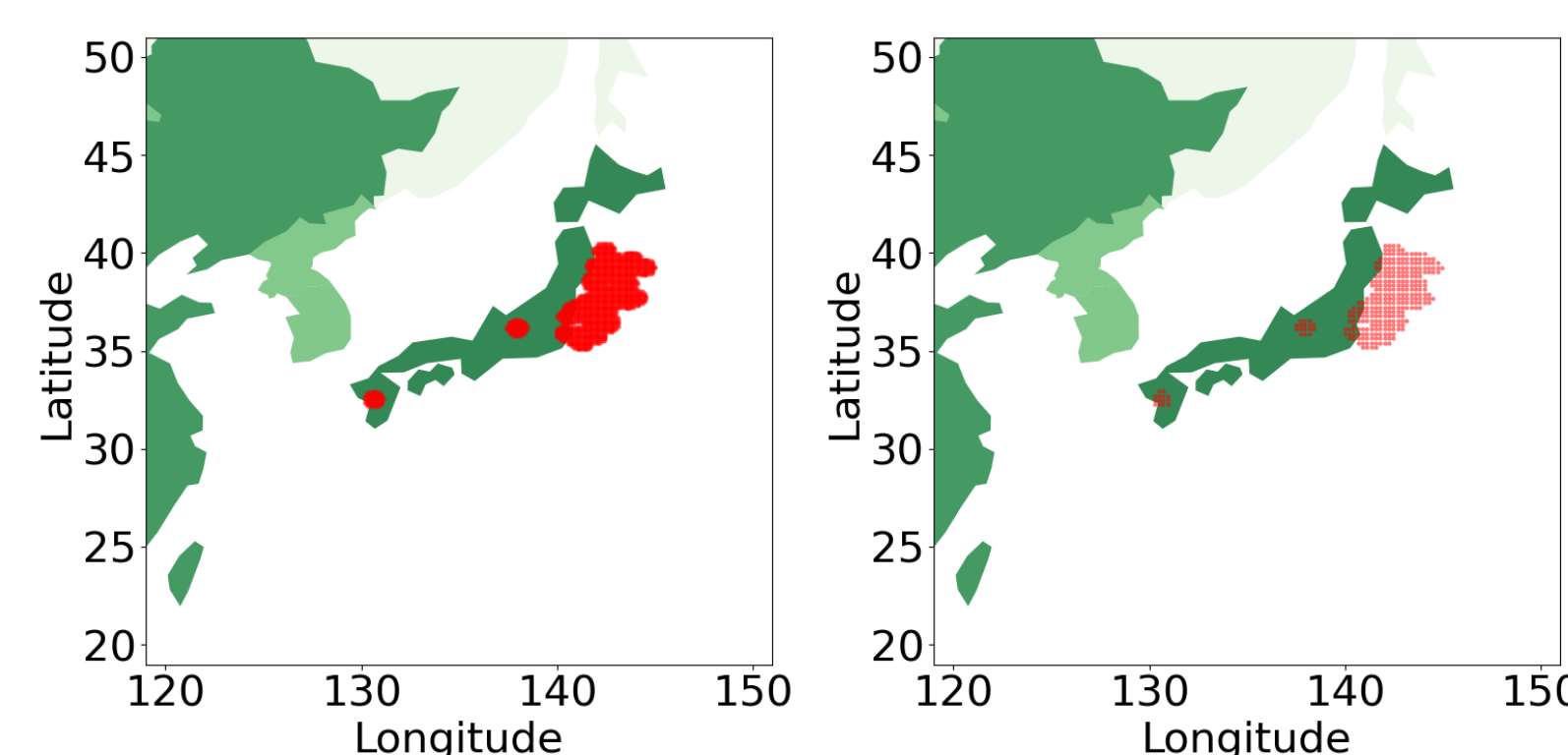
#### Performance evaluation using Pierce's Skill Score

- Pierce's Skill Score is widely use to evaluate performance of the binary classification models.
- Formula for Pierce's Skill Score is as following, with a, b, c, d being element of the confusion matrix, respectively.

$$\text{Pierce's Skill Score} = \frac{(ad - bc)}{[(a + c) \cdot (b + d)]}$$

- Pierce's Skill Score is then further use to find optimal settings (Study Area Spacings, Radius) for the RI model.

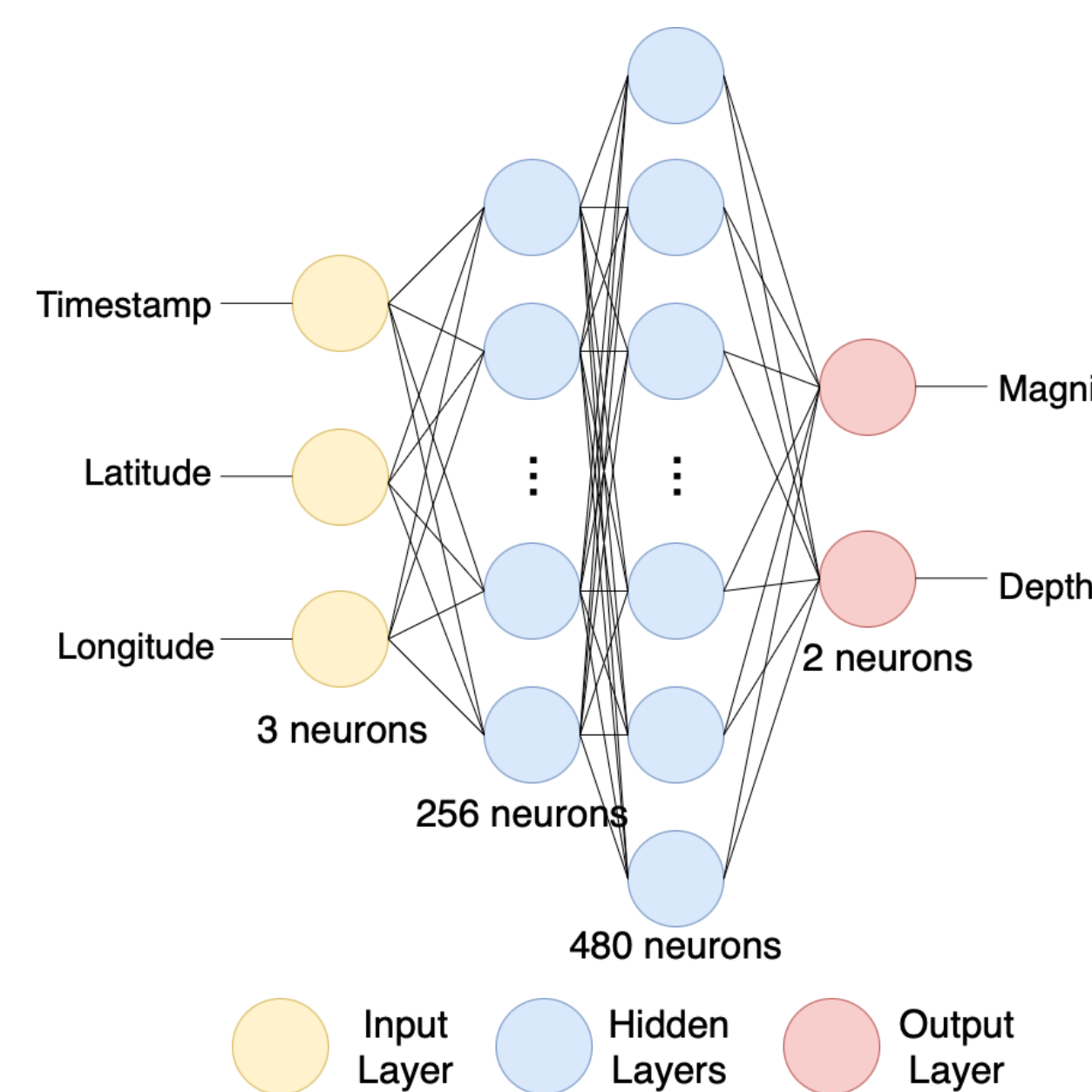
#### Results



Study Area Spacing 10km (left) and 25km (right)

Study Area Spacing	Radius	RI value Threshold	Pierce's Skill Score
10 km.	50 km.	0.101	0.694
25 km.	50 km.	0.107	0.688

### Machine Learning-Based Model (Proposed)



#### Architecture

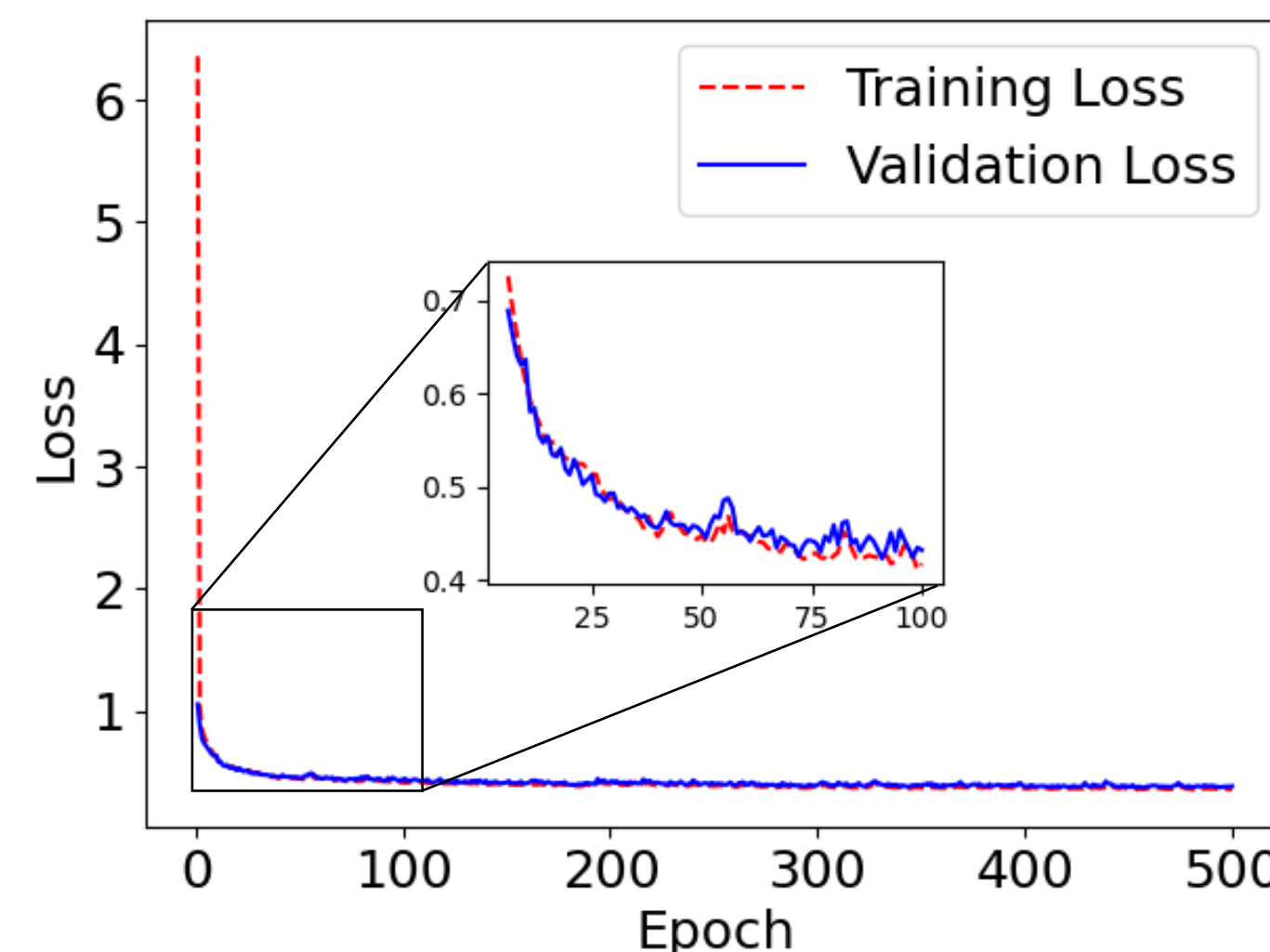
- We built the neural network with the architecture in the left side.
- We used Rectified Linear Unit (ReLU) as the activation function in hidden layers and linear as the activation function in output layer.

#### Training

We trained this model with

- Optimizer: Adam (learning rate=0.05)
- Loss: Mean Squared Error (MSE)
- Batch size: 10,000
- Epochs: 2,000

We plotted loss function of train set and validation set as graph below.



#### Testing and Evaluation

- We used test set to evaluate this model, we got MSE=0.394 as the result.
- From training, we got MSE=0.349 for train set and MSE=0.377 for validation set.

### Conclusion

From the above results from both models, we can conclude that earthquake in Japan has pattern. From this preliminary study, we can use relative intensity to find the areas that have high risk of earthquake. For machine learning-based model, we can find the relationship between time, locations, magnitude, and depth.

## References

### Acknowledgement

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