

## PREDICTION OF MAJOR EARTH QUAKE EVENTS USING DIFFERENT MACHINE LEARNING ALGORITHMS

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**ABSTRACT** At least two basic categories of earthquake prediction exist: short-term predictions and forecast ones. Short term earthquake predictions are made hours or days in advance, while forecasts are predicted months to years in advance. The majority of studies are done on forecast, taking into consideration the history of earthquakes in specific countries and areas. In this context, the core idea of this work is to predict whereas an event is classified as negative or positive major earthquake by applying different machine learning algorithms. Eight different algorithms have been applied on a real earthquake dataset, namely: Random Forest, Naive Bayes, Logistic Regression, Multi Layer Perceptron, AdaBoost, K-nearest neighbours, Support Vector Machine, and Classification and Regression Trees. For each selected model, various hyperparameters have been selected, and obtained prediction results have been fairly compared using various metrics, leading to a reliable prediction of major events for 3 of them.

**Keywords:** Random Forest, Naïve Bayes, Logistic Regression, AdaBoost, KNN, Support Vector Machine and Multi-Layer Perceptron Classifier

## **1.INTRODUCTION**

Rapid global urbanization leads to an increase in earthquakes, a major catastrophe which has occupied 60% of deaths overall natural disasters [8]. Obviously, human cannot stop natural disasters, while the application of machine learning is a powerful and invaluable method and technique, used by researchers as new area of study in geology, in order to reduce, as much as they can, the loss of life and billions of dollars in infrastructure and housing costs. Traditional classification methods rely on statistical assumptions for earthquakes that turn out to be unsatisfactory in danger state



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prediction. Therefore, the use of machine learning algorithms is becoming more widespread, that adapt and learn a problem by simulating a biological or natural system. Introducing the machine learning techniques in the field of geology and earthquake beats the traditional and standard methods used in earlier years and provides the scientists a new fresh method for assessing seismic risk and triggering future earthquakes. Data mining applications have achieved great success in geophysics and geology such as ecology, weather prediction, modelling, etc. and finally predicting the earthquakes. Two basic categories of earthquake prediction exist: short-term predictions and forecast ones. Short term earthquake predictions are made hours or days in advance, while forecasts are predicted months to years in advance. Due to the chaotic and complex phenomena of earthquake process, it is considered that short-term predictions are very difficult to address. Thus, the majority of studies are done on forecast, taking into consideration the history of earthquakes in specific countries and areas.

The core idea of this work is to predict whereas an event is classified as negative or positive major earthquake by applying different machine learning algorithms. It is well known that there is no best algorithm or one solution that fits all the problems and datasets for machine learning since the performance of algorithms depends on many factors. Some algorithms are best for small data, while others perform better for a tons of data sample. Some algorithms require categorical inputs, while others need quantitative. Another important criterion while choosing the algorithm is the complexity of the dataset and how many features the model needs to learn and predict. This is why, in this work, eight different algorithms have been applied on an earthquake dataset, namely: Random Forest, Naive Bayes, Logistic Regression, Mult iLayer Perceptron, AdaBoost, K-nearest neighbours, Support Vector Machine, and Classification and Regression Trees. For each selected model, various hyperparameters have been tested, and obtained prediction results have been fairly compared using various metrics, leading to a reliable prediction of major events for 3 of them.

#### 2.LITERATURE SURVEY

[1] KM Asim, F Mart Inez-Alvarez, A Basit, and T Iqbal. Earthquake ´ magnitude prediction in hindukush region using machine learning techniques. Natural Hazards, 85(1):471– 486, 2017. **International Journal For Advanced Research** 



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Earthquake magnitude prediction for Hindukush region has been carried out in this research using the temporal sequence of seismic historic activities in combination with the machine learning classifiers. Prediction has been made on the basis of mathematically calculated seismic indicators eight using the earthquake cata-log of the region. These parameters are based on the well-known geophysical facts of Gutenberg-Richter's inverse law, distribution of characteristic earthquake magnitudes and seismic quiescence. In this research, four machine learning techniques including pattern recognition neural network, recurrent neural network, random forest and linear programming boost ensemble classifier are separately applied to model relationships between calculated seismic parameters and future earthquake The occurrences. problem is formulated as a binary classification task and predictions are made for earthquakes of magnitude greater than or equal to 5.5, for the duration of 1 month. Furthermore, the analysis of earthquake prediction results is carried out for every machine learning classifier in terms of sensitivity, specificity, true and false predictive values. Accuracy is another performance measure considered for analyzing the results. Earthquake

magnitude prediction for the Hindukush using these aforementioned techniques show significant and encouraging results, thus constituting a step forward toward the final robust prediction mechanism which is not available so far.

[2] Louise K Comfort. Self Organization in disaster response: The great hanshin, japan earthquake of january17, 1995. In Self organization in disaster response: The Great Hanshin, Japan earthquake of January 17, 1995. US University of Colorado. Natural Hazards center, 1995.

This report presents findings from a quick response study of the response system that evolved following the Great Hanshin, Japan, Earthquake on January 17, 1995. The study was initially designed to examine the processes of self organization specifically in the medical response to this earthquake. However, in making plans for the study, it became clear that the medical response was inseparable from the larger disaster response system, given the interdependent characteristics of local and intergovernmental disaster management in Japan. Consequently, this study addresses the process of self organization in the broader disaster response system for this earthquake, with particular attention to emergency medical services. Building a





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community capacity for rapid transition in event of disaster is fundamental to effective disaster response. In every disaster environment, evidence exists of spontaneous reallocation of resources and reorganization of action to meet urgent human needs. However, such systems vary significantly in timing, efficiency, effectiveness, and reliability in practice, with associated costs in lives and property. This study explores the characteristics of self organization in order to characterize more accurately the dynamics of this process in a complex, interdependent, densely populated, metropolitan environment

[3] M. Gomes, R. Martins, J. J.
Almeida, P. Henriques and P. Novais,
"Hate Speech Classification in Social edia Using Emotional Analysis," 2018
7th Brazilian Conference on Intelligent Systems (BRACIS), 2018, pp. 61-66, doi: 10.1109/BRACIS.2018.00019.

Remote sensing continues to be an invaluable tool in earthquake damage assessments and emergency response. This study evaluates the effectiveness of multilayer feedforward neural networks, radial basis neural networks, and Random Forests in detecting earthquake damage caused by the 2010 Port-au-Prince, Haiti 7.0 moment magnitude (M<sub>w</sub>) event. Additionally, textural and structural features including entropy, dissimilarity, Laplacian of Gaussian, and rectangular fit are investigated as key variables for high spatial resolution imagery classification. Our findings show that each of the algorithms achieved nearly a 90% kernel density match using the United Nations Operational Satellite Applications Programme (UNITAR/UNOSAT) dataset as validation. The multilayer feedforward network was able to achieve an error rate below 40% in detecting damaged buildings. Spatial features of texture and structure were far more important in algorithmic classification than spectral information, highlighting the potential for future implementation of machine learning algorithms which use panchromatic or pansharpened imagery alone.

#### **3.PROPOSED SYSTEM**

Proposed several machine learning models to classify whether there is a chance of occurring earthquake or not, but none have adequately addressed this misdiagnosis problem. Also, similar studies that have proposed models for evaluation of such performance classification mostly do not consider the heterogeneity and the size of the data Therefore, we propose a Random Forest, Naïve Bayes, Logistic Regression,



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AdaBoost, KNN, Support Vector Machine and Multi-Layer Perceptron Classifier to predict.

## **Input Model:**

The user must provide input values for the certain fields in order to get results.

## View Results:

User view's the generated results from the model.

### View score:

Here user have ability to view the score in %

#### System

### Working on dataset:

System checks for data whether it is available or not and load the data in csv files.

## **Pre-processing:**

Data need to be pre-processed according the models it helps to increase the accuracy of the model and better information about the data.

## Training the data:

After pre-processing the data will split into two parts as train and test data before training with the given algorithms.



#### **Fig 1:Architecture**

#### **3.1 IMPLEMENTATION**

User:

#### View Home page:

Here user view the home page of the Hate Speech Classification web application.

#### View Upload page:

In the about page, users can learn more about the hate speech prediction.



#### Load:

In the load page, users can load the earthquake dataset.

	₩ info@example.com	🌜 +012 345 i	f 🍠 in 💿 🚥						
EARTHQUAKE		Home	Load	View	Preprocess	Model	Prediction		Log Out
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		Detect	ion o	f Eart	hquake				
	With	the he	lp of	Mach	ine Learr	ning			
		Choose Fil	le No file	chosen					
			Up	oload					

#### View:

Here we can see the uploaded data set.

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EARTHQUAKE		Home	Load	View	Preprocess	Model	Prediction	Log Or	ut

#### Detection of Earthquake With the help of Machine Learning

Date	Time	Latitude	Longitude	Туре	Depth	Depth Error	Depth Seismic Stations	Magnitude	Magnitude Type	Magnitude Error	Magnitude Seismic Stations	Azimuthal Gap
01-02-1965	13:44:18	19.246	145.616	Earthquake	131.6	nan	nan	6.0	MW	nan	nan	nan
01-04-1965	11:29:49	1.8630000000000002	127.352	Earthquake	80.0	nan	nan	5.8	MW	nan	nan	nan
01-05-1965	18:05:58	-20.579	-173.972	Earthquake	20.0	nan	nan	6.2	MW	nan	nan	nan
01-08-1965	18:49:43	-59.076	-23.557	Earthquake	15.0	nan	nan	5.8	MW	nan	nan	nan
01-09-1965	13:32:50	11 93799999999999999	126 427	Farthquake	15.0	nan	nan	5.8	N4\A/	nan	nan	nan

#### **Pre-process:**



Here we can prepare our data in such a way that our system should understand i.e., we will make our data noise free.

	对 info@example.com	<b>4</b> +012 345	6789					f 🎔 in 💿 📼
EARTHQUAKE		Home	Load	View	Preprocess	Model	Prediction	Log Out
	Data Prep	rocess	ed ar	nd It S	plits Suc	cessf	ullv	
	Datariop				pinto o de		,	
		Detect	ion o	f Eart	hquake			
	With	the he	lp of	Mach	ine Lear	ning		
		Test Spli	t Size					
		1000 0011	0120					
			Su	ıbmit				

## Model:

Here we can train our data using different algorithm.

S info@example.com	<b>¢</b> +012 345	6789					f 🎔 in 💿 🚥
EARTHQUAKE	Home	Load	View	Preprocess	Model	Prediction	Log Out
The accuracy obtained by	y Rand	om Fo	orest	Classifie	r is 99	.98576309794989%	5
	Detect	ion o	f Eart	hquake			
With	the he	lp of	Mach	ine Learr	ning		
	Choose	an Algori	thm	~			
		Su	bmit				

## **Prediction:**

This page show the detection result of the earthquake prediction data.



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There is a chance of happening Earthquake

With the help of Machine Learning							
Enter Latitude	Enter Longitude						
Enter Depth	Enter Depth Error						
Enter Depth Seismic Stations	Enter Magnitude						
Enter Magnitude Type	✓ Enter Magnitude Error						
Enter Magnitude Seismic Stations	Enter Azimuthal Gap						
Enter Horizontal Distance	Enter Horizontal Error						

## Detection of Earthquake With the help of Machine Learning

Enter Horizontal Distance

## **5.CONCLUSION**

Seven machines learning algorithms have been tested for our work to classify the major earthquake events between negative and positive. The study has been applied to a dataset collected from a center in California, which was recording inputs for 36 vears. Every machine learning technique shows different results from each other. KNN, Random Forest and MLP are the best by producing the least false output (FP) while SVM, KNN, MLP and Random Forest classify the higher number of output correctly. Future work involves working on case studies based on intervention real data and add

consideration to features selection methods.

## REFERENCES

[1] KM Asim, F Mart'ınez-Alvarez, A Basit, and T Iqbal. Earthquake ´ magnitude prediction in hindukush region using machine learning techniques. Natural Hazards, 85(1):471–486, 2017.

[2] Giuseppe Bonaccorso. Machine Learning Algorithms: A Reference Guide to Popular Algorithms for Data Science and Machine Learning. Pack Publishing, 2017.

[3] Leo Breiman. Random forests.Machine learning, 45(1):5–32, 2001.



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[4] Louise K Comfort. Self organization in disaster response: The great Hanshin Japan earthquake of January 17, 1995. In Self organization in disaster response: The Great Hanshin, Japan earthquake of January 17, 1995. US University of Colorado. Natural Hazards Center, 1995.

[5] Austin Cooner, Yang Shao, and JamesCampbell. Detection of urban damageusing remote sensing and machine learningalgorithms: Revisiting the 2010 Haitiearthquake. Remote Sensing, 8(10):868,2016.

[6] Jean-Franoi's Couchot, Christophe Guyeux, and Guillaume Royer. Anonymously forecasting the number and nature of firefighting operations. In Proceedings of the 23rd International Database Applications & Engineering Symposium, pages 30:1–30:8. ACM, June 2019.

[7] Dursun Delen, Cemil Kuzey, and Ali
Uyar. Measuring firm performance using
financial ratios: A decision tree approach.
Expert Systems with Applications,
40(10):3970–3983, 2013.

[8] Laigen Dong and Jie Shan. A comprehensive review of earth quake induced building damage detection with remote sensing techniques. ISPRS Journal of Photogrammetry and Remote Sensing, 84:85–99, 2013.

[9] Wiem Elghazel, Kamal Medjaher, Noureddine Zerhouni, Jacques Bahi, Ahmad Farhat, Christophe Guyeux, and Mourad Hakem. Random forests for industrial device functioning diagnostics using wireless sensor networks. In AERO 2015, 2015 IEEE Aerospace conference, pages 1–9, Big Sky, Montana, USA, March 2015. IEEE.

[10] Wiem Elghazel, Kamal Medjaher, Noureddine Zerhouni, Jacques Bahi, Ahmad Farhat, Christophe Guyeux, and Mourad Hakem. Random forests for industrial device functioning diagnostics using wireless sensor networks. In 2015 IEEE Aerospace Conference, pages 1–9. IEEE, 2015.

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