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PREDICTIVE ANALYSIS OF WIND POWER GENERATION USING MACHINE LEARNING

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Abstract — The proposed aim of project predictive analysis of wind power generation using Machine learning is to reduce the cost, provide cleaner and eco friendly source of energy and increase the efficiency of existing systems to greater extent by calculative analysis [1]. As the world is progressing towards renewable and clean energy it is an interdisciplinary research model that is designed to find an optimal way of harnessing wind energy via machine learning models. It is a viable option for Governments and Companies around the world which will help them to set up Wind Energy Plants at appropriate locations for maximizing the output power generation by prediction from the Machine Learning model.

The experience obtained while working in the simulation of predictive analysis of wind power generation using machine learning based wind conversion can be successfully implemented in real life scenario helping in increasing renewable energy production.

The main objective is to maximize the performance and

efficiency of Wind Mills. This project will help in automation-. The control strategy used in this project is for maximizing the performance and efficiency of a Wind Mills.

I. INTRODUCTION

An impending energy catastrophe is coming speeding towards us and our future generations. It is a well known fact that we are exhausting the nearly exhausted non renewable energy sources like coal, natural gas, petroleum etc. on an unprecedented scale. So it becomes a compulsion for mankind to find ways of efficient extraction and distribution of renewable energy sources like tidal power, geothermal energy, wind energy and others. Amongst these the extraction of wind energy is a viable option as the world has a lot of wind power potential in countries like China, US, European states like Germany, Spain and France. In the context of our country, nine states have humongous wind power potential like Gujarat, Karnataka, Madhya Pradesh, Maharashtra, Rajasthan and Tamil Nadu. The production and distribution costs to be incurred in wind

The data is divided into training data and testing data for training the model and testing the model. 70 % of space is given to training data and 30% for testing purposes.

We applied two classification and one regression technique on our model to determine which particular method provides better accuracy and prediction for our model.

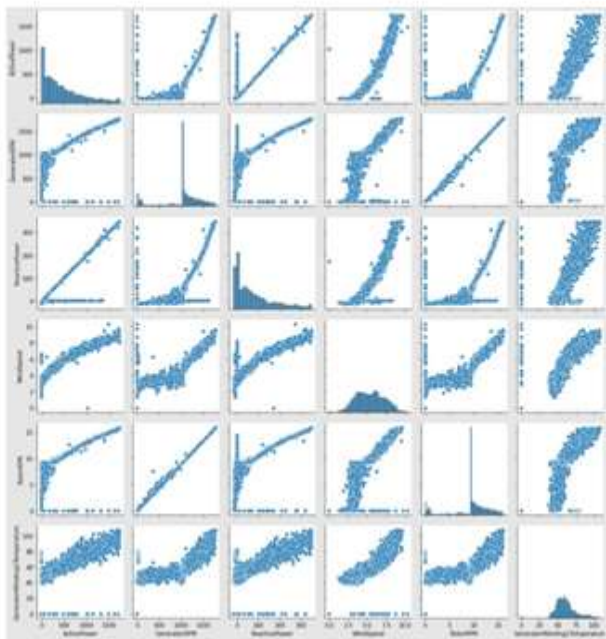


Fig 2 (Scatterplots)

DIFFERENT MACHINE LEARNING ALGORITHMS USED

- 1) **Linear Regression-** It is a machine learning approach in which relation between two variables is established on a graph and shown by a line. For a good linear regression model perpendicular distance of scatter points must be minimum from the line plotted [7].
- 2) **Random Forest-** The random forest classifier is a supervised learning algorithm. It consists of multiple decision trees just as a forest has many trees. It uses randomness to enhance its accuracy and combat over fitting. These algorithms make decision trees based on a random selection of data samples and get predictions from every tree, after that, they select the best viable solution through votes. A tree has several nodes and the model searches for the best solution by going through all nodes [8].
- 3) **XG Boost-** It is an implementation of gradient boosted decision trees. In this algorithm decision trees are

created in sequential form [9]. Weights play an important role in XGBoost. Weights are assigned to all the independent variables which are then fed into a decision tree which predicts results [10].

Error Metrics

i).R2 Score: It is the amount of variation in output dependent attribute which is predictable from input independent variables [11].

$$1 - \frac{\sum_{t=1}^N (y_t - \hat{y}_t)^2}{\sum_{t=1}^N (y_t - \bar{y})^2}$$

ii) MAE: Mean absolute error is calculated as the sum of absolute errors divided by the sample size.

$$\frac{\sum_{i=1}^n |y_i - x_i|}{n}$$

iii) RMSE: Root mean square error is the square root of the difference between the values predicted by the model and the values observed [12].

$$\sqrt{\frac{1}{N} \sum_{t=1}^N (y_t - \hat{y}_t)^2}$$

V. MODEL RESULTS

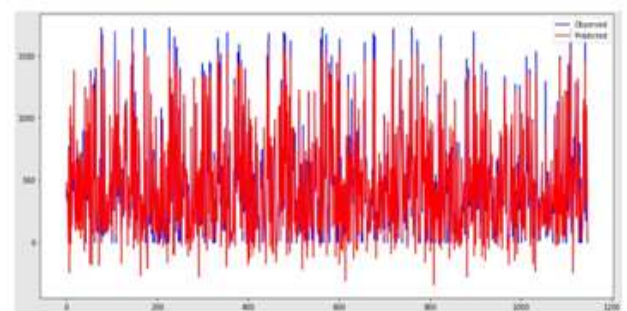


Fig 3 (Linear Regression)

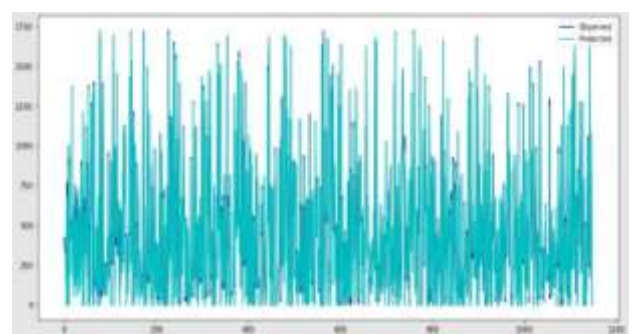


Fig 4 (XG Boost)

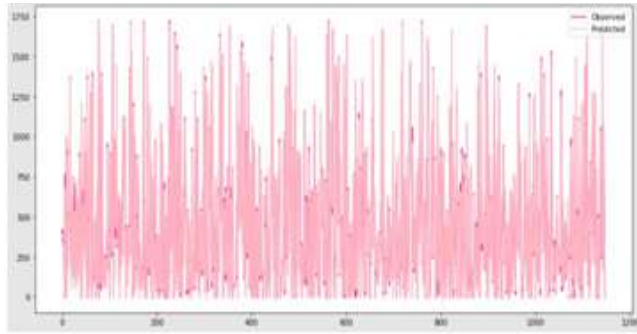


Fig 5 (Random Forest)

Model Name	R2 Score	MAE	RMSE
Linear Regression	0.96	64.86	85.62
XG Boost	0.99	15.58	23.99
Random Forest	0.99	8.55	16.39

Table 1: Performance Score

VI. CONCLUSION

We have taken 3 different types of error metric for comparing our model performance on the dataset. From the table we see that linear regression models have highest mean absolute error and root mean square error whereas random forest have least mean absolute error and root mean square error. So for this data set random forest outperform other model.

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