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# Machine Learning Approach for Prediction of Crop Yield

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## ABSTRACT

India is a global agriculture powerhouse. The average productivity of many crops in India is quite low and the current situation faced by farmers in India leads to increase in suicide rate over years, due to the impact of climate change in country. Crop productivity can be increased using Machine Learning (ML) methods and climate data. For weather forecasting, Machine Learning approaches like Recurrent Neural Network (RNN) are utilised, while Machine Learning categorization techniques like Decision Tree, Random Forest, and Nave Bayes are used to predict appropriate crops. Therefore, its necessary to build a model which takes into consideration of all the parameters for the better selection of crops which increases the crops yield.

**KEYWORDS:** Agriculture, Recurrent Neural Network (RNN), Decision Tree, Random Forest, Naïve Bayes.

## I. INTRODUCTION

When it comes to the economic prosperity of a country like India, agriculture plays a critical role. Weather and soil parameters have the greatest impact on crop yield. It's critical to comprehend the effects of climate change on crop productivity. Weather has an impact on a plant's development and growth at every stage. Crop choosing using real-time contingencies depending on weather & soil parameters. Techniques that produce the highest yield while using the minimum amount of energy in constrained land areas must be employed. Precision Agriculture (PA) [1] is a popular term for this approach. Precision agriculture aims to improve crop yields and economics while reducing the amount of conventional inputs required to cultivate crops (land, water, fertilizer, herbicides, and insecticides). Several Machine Learning (ML) algorithms have been employed in agricultural in recent times for climate and groundwater quality forecasting and strategic thinking. Unsupervised and Supervised Machine Learning models are the two types of machine learning algorithms.

Supervised learning is a sort of artificial intelligence in which computers are trained with well-labeled training examples and then predict the outcome using that information. The labelled data indicates that some input data has already been tagged with the desired output. K-means grouping, KNN (K-nearest neighbours), and other unsupervised machine learning methods are examples. Supervised machine learning is employed throughout the classification methods used, like Decision tree, Random forest, and Nave Bayes [2-3].

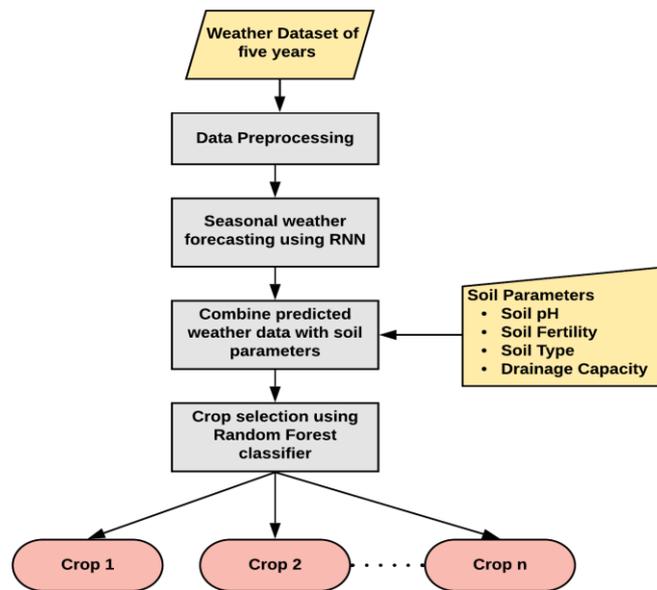
## II. LITERATURE SURVEY

Agriculture production is increased by using supervised Machine Learning technique. It gives information about how the crop should be selected so as to maximize the profit. It includes six major crops consisting of rice species like aus, aman, boro and other crops like potato, wheat and jute. The dataset is collected from Bangladesh Agricultural Research Council according to area. The algorithms used are Decision Tree Learning, ID3 and K-nearest neighbour algorithm. This method is proposed by Md. TahmidShakoor et.al. It is used to provide solution in selection of crop in more profitable manner [4]. Another method is presented by S. Veenadhri et.al designed machine on climatic parameter used in web application. It provides information about climatic condition which would affect the crop yield. The climatic terms that affect the crop yield in Madhya Pradesh was discovered which uses C4.5 Algorithm. The agro input parameters are not considered in this application as they differ with change in space and measure of individual field [5]. Nishit Jain et.al attempted to present a crop selection method based on various environmental factor using machine learning. This method helps the user to choose the crop which will maximize the crop yield by taking into consideration of all the parameters which affects the growth of crop [6-8]. This method also assists in predicting crop sequence as well as providing monetary benefits to farmers. Economic and environmental parameters such as rainfall, temperature and type of soil plays a major role. Based upon all the parameters crop is cultivated based on seasons [9]. Igor Oliveira et al. present an advanced machine learning system for pre-season agricultural yield forecasting in another approach. Agriculture stakeholder department yield prognosis, which can be obtained using machine learning modelling and obtaining additional data from external sources [10-14]. To create a pre-season projection using soil characteristics databases, satellite-derived precipitation, climate patterns forecast data using physical models, and other resources. The yield of soya beans or maize can be predicted without using data from the Normalized Difference Vegetation Index Information (NDVI). It yields useful results by obviating the need for elevated remote control visual representation and allowing farmers to plan for climate changes that affect harvest oscillation. Soya bean and maize

forecasts are based on Brazil and America, which accounted for 44 percent of global food security in 2016 [15-17].

### III. METHODOLOGY

Weather patterns estimation is important using a recurrent neural network (RNN). The random forest categorization technique is used to classify appropriate crops [18].



**Figure-1: Schematic flow of proposed weather and soil forecasting based crop selection method**

**a) Weather Dataset of five years:** At this stage the dataset is collected, The dataset includes minimum and maximum temperature, PH, Humidity, and soil arduino board, and rainfall data is collected from Agri Ministry, IISc, and IIM.

**b) Data Preprocessing :** During this stage raw data is transformed into useful and efficient format. It involves cleaning of data, identifying the missing data, filling of missing values, ignoring the tuples.

**c) Seasonal Weather Forecasting using RNN:** At this stage RNN used in variant of Neural network. This Forecast predicts,

Weather at monthly interval upto 7 months and seasonal crop selection method suggests parameters. These data is collected through one or more crop suitable for a season.

**d) Combine predicted weather data with soil parameters:** During this stage soil parameters and they are as follows,

*Soil pH* : It is a measures of acidity and basicity of a soil. It influences the availability of essential nutrients. pH levels range from 0 to 14.

*Soil fertility*: It increases the fertility soil we can use the method of recycling the organic waste.

*Soil type*: it determines the type of soil suitable for a particular crop such as sandy soil, clay soil etc.

*Soil drainage*: it determines how the water moves out of the soil as a result of force of gravity [19].

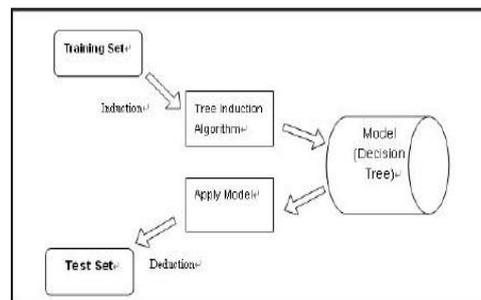
e) **Crop selection using Random Forest Classifier** :At this stage, it is suitable to predict crop by considering all the parameters to have a good yield.

The algorithms used in this model are:

1. Decision Tree
2. Naïve Bayes
3. Random Forest

### 1] Decision Tree

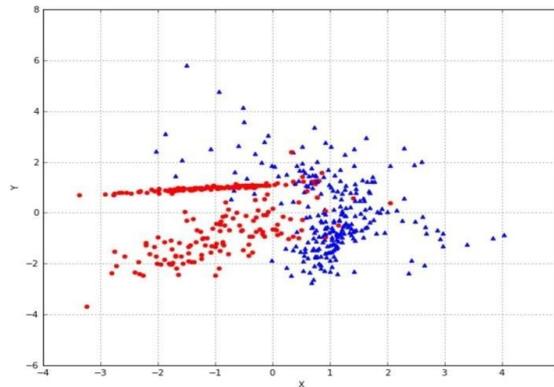
Internal nodes represent properties of the dataset, branches indicate decision rules, and each node represents the result in a decision tree. The process starts with root node in a decision tree for estimating agricultural yield given a data set. The best property in a dataset that can be chosen using attribute selection methods is the root node of a tree [12-13]. The method analyzes the roots attribute to the actual dataset attributes, then continues the branch and goes towards the next node presents a comparison. It's done again and again until the leaf node is found [20-24].



**Figure-2: Architecture of Decision Tree Algorithm**

### 2]. Naives Bayes

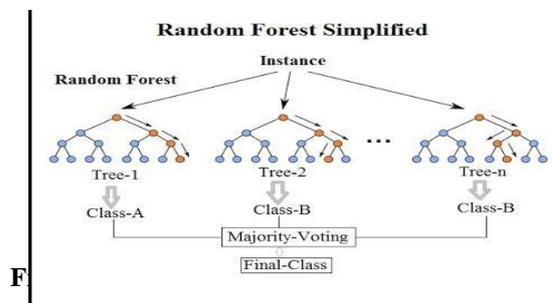
The Nave Bayes classifier makes predictions based on an object's likelihood. It is based on the Bayes theorem premise. With prior knowledge, the Discrete wavelet transform (dwt) is used to compute the likelihood of a hypothesis. The weather dataset is turned into a frequency distribution table utilizing the Nave Bayes classifier, and a probability table is constructed by calculating the probabilities for each parameter utilizing the Bayes theorem [24-27].



**Figure-3: Graph Representation of Naïve Bayes Classification**

### 3] RandomForest

Random forest classification is based on ensemble learning. It is the process of combining multiple classifiers to improve the performance of the model [28]. It has a large number of decision trees on different subsets of the supplied weather dataset and averages them to increase the predicted accuracy of the dataset, which is very useful. The higher the number of trees in the forest, the better the accuracy and the less chance of overfitting.



## IV. MODELLING AND ANALYSIS

The dataset is divided into training and testing data in this study. The training set is a part of the data used to train a model, and the testing set is a subset of the data used to test the trained model [29-31]. The initial stage is to gather data, which includes variables like temperatures, humid, PH, rain, and other soil factors. These are dependent variables where as the predicted crop is an independent variable. Using proven classification methods, the final forecast crop is described as a simpler depiction of the complex link among weather and soil on only one hand, and crop quality (such as growth or yield) on another [32-34].

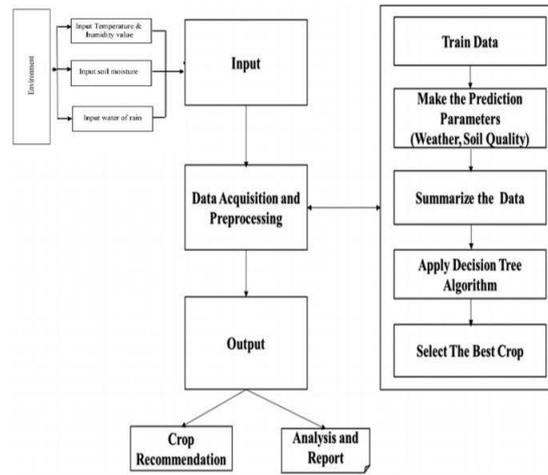


Figure-5: Flowchart of crop yield prediction

### V. RESULT AND DISCUSSION

In this section, all experimental details and trail data sets output have been analysed. All experiments are performed using Arduino boards and windows OS installed on intel Core i5 CPU with 4GB RAM. We followed the guidelines which was mentioned in the Sonal Jain *et al* [14] using some of the data from the training set and some from the testing set and feeding these sets into the classification algorithms [35-40]. Further as mentioned in the Sonal Jain *et al*, we have taken the images of both schematic flow and rain fall graph as shown in fig-1 and andfig-7.

	Temperature	Humidity	pH	soilWater	soilTemp
1	20.87912175	82.30274423	6.50296292	202.9353262	xxx
2	21.77046346	80.12344688	7.02887643	204.4654174	xxx
3	20.44901651	81.12917676	7.44021544	204.4654174	xxx
4	20.49196165	80.15742624	6.48000005	204.8654124	xxx
5	20.19218461	80.40491367	6.24012991	204.1246845	xxx
6	21.03848212	83.37011772	7.07301063	204.0549968	xxx
7	22.70027986	81.87481394	6.70080002	204.1246845	xxx
8	20.27776242	82.84946219	6.738427176	204.8743949	xxx
9	20.12086864	81.61742614	6.402164261	204.8743949	xxx
10	21.23701386	83.01022041	6.330212425	204.2084918	xxx
11	20.52222613	81.47173064	5.700197789	204.614097146	xxx
12	21.37898217	81.45001296	7.52001296	204.08222346	xxx
13	20.48026864	80.80084622	6.50001786	204.614097146	xxx
14	21.04979122	82.05007182	6.80015386	199.2771389	xxx
15	21.44000209	80.80084622	6.40001786	204.58001786	xxx
16	20.82000415	80.30021047	7.042190008	204.08001786	xxx
17	21.10112177	80.17401708	6.070000004	204.0512101	xxx
18	21.01910167	80.48119127	6.070000004	204.0512101	xxx
19	21.8012514	80.13210008	6.501012176	204.55500169	xxx
20	21.79210008	80.13210008	6.501012176	204.55500169	xxx
21	21.47201316	80.61747076	6.442170179	199.4097412	xxx
22	21.17470161	81.12114176	6.070170429	204.3800169	xxx
23	22.84767739	80.97284295	6.02410001	204.35000169	xxx
24	21.05210008	81.12114176	6.24012991	204.1246845	xxx
25	20.48101344	81.12114176	7.170401681	204.0500169	xxx
26	21.07470161	80.12344688	7.700012176	204.0500169	xxx
27	20.19218461	80.40491367	6.24012991	204.1246845	xxx
28	20.19218461	80.40491367	6.24012991	204.1246845	xxx
29	20.19218461	80.40491367	6.24012991	204.1246845	xxx
30	20.19218461	80.40491367	6.24012991	204.1246845	xxx

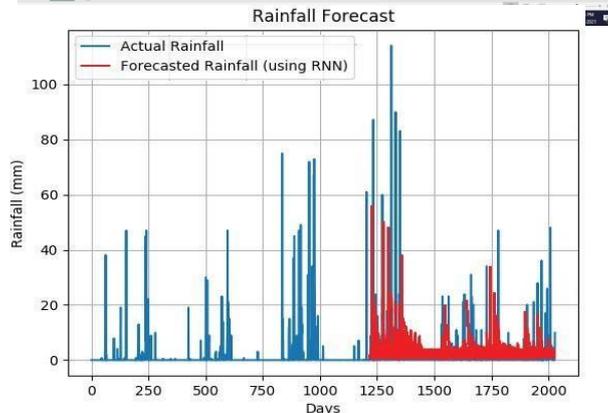


Figure-6: Sample Dataset for training classification model

Figure-7: Rain fall Forecast



**Figure-8:**Predicted Output

## VI. CONCLUSION

A result is the consequence of actions or events expressed qualitatively or quantitatively. Performance analysis is a set of basic quantitative relationship between the performance quantities. Although there are many techniques which shows promising results in crop yield detection. As some times there might be sudden change in weather so prediction, so weather plays a prominent role. Therefore, in this paper we prepare a profitable model for crop prediction. Hence, our model is the best situated for the modern crop yield detection system when compared to the traditional and unconventional techniques.

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