Experimental Analysis of Machine Learning Algorithms Based on Agricultural Dataset for Improving Crop Yield Prediction

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Abstract: Agriculture is the primary research study area in India as agriculture is the main source of income for various communities. In classification algorithm for agricultural dataset according to production, area, crop and seasons. Here, four classification algorithms are used with the help of WEKA tool. These algorithms are namely the present scenario, there is a call to renovate the enormous agriculture data into diverse technologies and make them accessible to the farmer for improved decision making. The endeavor of this work is to find out the finest Random Tree, J48, Bayes Net and KStar etc. The captured results revealed that Random tree algorithm performed well in terms of error rate and provides slightly better performance than KStar, Bayes Net and J48 classifiers. In this paper, our objective is to apply machine learning techniques to mine constructive information from the agricultural dataset to improve the crop yield prediction for major crops in Nashik district of Maharashtra.

Keywords: WEKA tool, J48, Bayes Net, KStar and Random Tree, Machine learning and Crop Yield Prediction

I. INTRODUCTION

In India population is increasing rapidly as compare to the world population. Hence the demand of food will increase exponentially with limited resources and land. According to the Government data approximately seventy percent of the Indian population is living in villages. In villages, agriculture is a kind of employment for farmers and their growth depend on the agricultural output. But there are many issues untouched in the agricultural domain. In these issues crop yield prediction is the major research area. Early crop yield predictions can help the farmer’s to avoid the losses and maximize their crop yield.In broad sense we can say this is the issue originally related to the agriculture planning [8]. Here our primary objective is to experimentally analyze the various machine learning algorithms and suggest the best one to get the higher crop yield accuracy results. Crop yield prediction generally depends on the weather conditions, soil and geography [15]. Secondly, we want to equip the farmers with better technologies in order to make the profit from their agricultural output. In concrete terms machine learning algorithms are classified as supervised and unsupervised machine learning algorithms. In both the algorithms raw data is given in order to predict or find the various data trends to find out the most accurate crop yield prediction. This paper is organized in a sequential manner. Section II is about the related work and Section III is about the core of this research i.e. proposed approach. Section IV describes the Methods and Materials required for the study. Dataset description discussed on Section V. Section VI was discussed with Results and analysis. Section VII concludes the work with possible future enhancement.

II. RELATED WORK

In the field of crop yield prediction various researchers have contributed. Here, in this section we have tried to communicate the earlier research as mentioned below: Diepeveen and Armstrong [6] discuss about various key attributes which will decide the performance of crop in different areas depending on the season, sowing and soil type etc. This paper mainly concentrates on different data mining techniques to enlighten the crop yield performance. Monali Paul et al. [10] In order to predict the yielding of the crops, the crops are analyzed and based on the categorized analysis. This categorization is done based on data mining algorithms. This paper gives insight into various classification rules like Naïve Bayes, K-Nearest Neighbor. Murynin et al. [7] study the dependency between the prediction and the accuracy of the forecast. Linear model is used for yield prediction. This model is extended with non-linear attributes in order to improve the accuracy of the prediction. The accuracy of the model has been expected based on the time period between the moment of the forecast formation and the time of harvest. Hemageetha [8] mainly focuses on using the soil parameters like pH, Nitrogen, moisture etc. for crop yield prediction. Naïve Bayes algorithm is used to classify the soil and a77% accuracy is achieved. Apriori algorithm is used to associate the soil with the crops that could provide maximum yield in them. A comparison of accuracy achieved during classification using Naïve Bayes, J48 and JRIP is also presented.

III. RESEARCH METHODS

The main components which contributes to this research are sample study area, agricultural datasets and firm methodology.

A. Study Area

For this particular research, we have selected Nashik district of Maharashtra state as a sample study area. In Maharashtra state there are total thirty six districts, out of these all Nashik is agriculture oriented and catering the agricultural demands of the nearby states. In Nashik, farmers are growing various crops and out of the various crops arhar, bajra, caster seed, cotton, groundnut, maize,
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moong, niger seed, ragi, rice, soyabean, sunflower, safflower, gram, jowar, wheat and sugarcane produced in this district we have selected etc.

B. Data Set
For this experiment, dataset is extracted from link https://data.gov.in. In this government website agricultural datasets are shared containing data of past few years. This dataset contains various columns in Microsoft Excel like production area, crop, Crop year and season etc. This CSV file has the hundred records of eight districts of Maharashtra state of India. Moreover this file contains data used for twenty different crops.

C. Major Factors Used
The major factors used in this sample dataset are described as below:

Year: This will contains in which year how much production takes place. This is further classified for every crop in various districts of the state.

Season: In India the crop seasons are basically classified in two types. These are scheduled as below:
- Kharif (July- Oct.)
- Rabi (Oct.-March) _Winter crops and (March-June) _Summer crops

Area: It contains cultivation (Hectare) area according to the crop and district wise.

Production: It contains total production (Tons) for according to the crop and district wise.

IV. METHODS AND METHODOLOGY

In this paper, we are focusing to preprocess the data and using weka tool to analyze the simulation results obtained after applying the machine learning classifiers. Comprehensive depiction of weka tool is described as below:

WEKA Tool: The acronym WEKA can be elaborated as Waikato Environment for Knowledge Analysis. This tool was come into existence at University of Waikato, New Zealand. Here in this tool a variety of machine learning algorithms realized that can be implemented directly to agricultural data set. Graphical User Interface is well designed to compare the various experimental results. Machine learning supports the artificial intelligence to learn or train the data without any external assistance. Weka is a Preprocess: This step contains the preprocessing of data set in various ways.
- Classify: The training and testing of data set occurs which performs the classification and regression and also evaluate them.
- Cluster: The clustering of dataset takes place.
- Associate: Association rules are applied for the data set and evaluation takes place.
- Select attributes: Selection of the most relevant attributes of the given dataset.
- Visualize: Visualization of data set occurs in the different two-dimensional plot and interact with them.

The main features of weka are as listed below:
- Platform independent
- Open source and free
- Different machine learning algorithms
- Easy to use
- Data prepossessing tools
- Flexibility for scripting experiments
- Graphical user interface

Weka tool consists of four keys which are described as below:
- Explorer: It is the location where you can explore the data
- Experiment: It is the location where you can execute the experiments and can bear statistical tests between learning outlines.
- Knowledge flow: It is the location which upkeep essentially the similar functions as the EXPLORER nevertheless with a drag and drop interface.
- Simply CLI: It offers an easy command-line interface that permits implementation of WEKA commands for operating systems and does not permit their own command line interface.

| State_Name | District_Name | Crop_Year | Season | Crop | Area | Production |
|------------|---------------|-----------|--------|------|------|------------|
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 105000 | 8300 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 179100 | 17900 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 44400 | 4000 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 47000 | 94500 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 27900 | 28300 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 59000 | 7900 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 173800 | 104000 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 88800 | 7900 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 136000 | 5000 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 121000 | 6700 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 40000 | 34100 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 42500 | 77000 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 100 | 100 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 58000 | 81500 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 100 | 50 |
| Maharashtra | NASHIK | 2011 | Kharif | Ahar | 14900 | 14500 |
| Maharashtra | NASHIK | 2011 | Rabi | Ahar | 17900 | 27000 |
| Maharashtra | NASHIK | 2011 | Rabi | Ahar | 7000 | 5000 |

Fig.1 Sample Agricultural Dataset (Nashik)
A. Data Set and Attribute Selection

We have collected the crop data set which contains 95 instances and 7 attributes. The data file has to be in either ‘CSV’ format or ‘ARFF’ format. In order to explore dataset the following steps need to be integrated.

- Load data
- Pre-process data
- Analyze attributes

The pre-process section enables you to load data into weka tool. The dataset may be imported from a CSV file as well as from numerous data formats like Binary, ARFF and C4.5. Moreover, Weka tool provides the feature to read the data from a database server and from a web address. Sample ARFF file is shown in Fig 2. After importing the data the preprocessing tools are applied to get the output in terms of discretization, re-sampling, normalization, attribute selection etc. The pictorial representation of preprocessing tool is shown in Fig 3. and Fig 4.

B. Filters

The preprocessing tools in weka are called filters. As per our requirement, one can select or remove the attributes and also apply filter on our dataset. Weka filters can be used to modify the data sets, so they are called data preprocessing tools.

C. Explore: Building classifiers:

In WEKA tool, classifiers are the models used for classification and regression.

- Choosing a classifier: Once you have loaded your data set in Weka tool you have to click on the classifier button, and choose the appropriate classifier for your data set and one can start
analyzing the data by making the use of provided algorithms. Here, I have analyzed my data with the Random Tree algorithm.

To describe the performance of used various classifiers; a confusion matrix is formed containing the values inside it. Confusion matrix depicts the performance of the classifiers.

VI. PERFORMANCE EVALUATION

The basic evaluation measures from confusion matrix are described in this section. Confusion matrix is a combination of four outcomes such as accuracy, sensitivity, specificity and precision. This also contains the ROC and precision-recall etc. Four outcomes of classification:

- True positive (TP): As the classifier name appears, TP is just the representation of number of correct classifiers reserved to positive class.
- True negative (TN): As the classifier name appears, TN is just the representation of number of correct classifiers reserved to negative class.
- False positive (FP): As the classifier name appears, FP is just the representation of number of classifiers reserved to positive class but in realism fit in to negative class.
- False negative (FN): As the classifier name appears, FN is just the representation of number of classifiers reserved to negative class but in realism fit in to positive class.

Sensitivity (REC/TPR): This is also called as true positive rate and recall. It can be determined as the no. of correct positive predictions divided by the total no. of positives.

Specificity (SP/TNR): This is also called as true negative rate. It can be determined as the no. of correct negative predictions divided by the total no. of negative.

Precision (PPV): This is also called positive predictive value. It can be determined as the no. of correct positive predictions divided by the total no. of positive.

Accuracy: It can be determined as the no. of all correct predictions divided by the total no. of dataset.

RAE (Relative absolute error): Relative absolute
error is just the replica of relative squared error.

**RRSE (Relative root mean square error):** It is basically applied to calculate the differences between the different values estimated by a model, estimator and the observed values. Sometimes it is also termed as root mean square deviation (RMSD).

**MAE (Mean absolute error):** The difference between the two continuous variables is called as mean absolute error. For the mean absolute error subtract mean value from each term and calculate the mean of modulus i.e. positive of those values.

The formulas used to calculate the performances are described in Table I, as shown below:

| S.No. | Performance Evaluation | Formula used |
|------|------------------------|--------------|
| 1    | Sensitivity            | \[ SN = \frac{TP}{TP + FN} \] |
| 2    | Specificity            | \[ SP = \frac{TN}{TN + FP} \] |
| 3    | Precision              | \[ PREC = \frac{TP}{TP + FP} \] |
| 4    | Accuracy               | \[ Accuracy = \frac{TN + TP}{(TN + TP + FN + FP)} \] |

**VII. EXPERIMENTAL RESULTS AND DISCUSSIONS**

The experimental results are shown with 10 fold cross validation to avoid overlapping. Four classifiers are studied here namely Random Tree, KStar, Bayes Net and J48 etc. The performances of the classifiers are analyzed in Table II.

**A. Performance error of classification algorithms**

Accuracy of different classification algorithms are also shown in Table II according to area harvested production, season and crop. From Table it is observed that Random tree and KStar higher accuracy than Bayes Net and J48 classification algorithms.

| S.No | Classifiers | RAE  | RRSE | MAE  | Accuracy |
|------|-------------|------|------|------|----------|
| 1    | Random Tree | 5.8224 | 24.31 | 0.015 | 97.89    |
| 2    | KStar       | 12.8054 | 33.81 | 0.034 | 95.7     |
| 3    | Bayes net   | 68.4517 | 84.60 | 0.1856 | 70.5     |
| 4    | J48         | 59.2913 | 77.60 | 0.1608 | 78.9     |

The results are captured from the WEKA tool and it is observed that Random tree classifier has higher values of accuracy and less error rate. These results are purely depends on the dataset.

**VIII. CONCLUSION – MACHINE LEARNING VS. STATISTICS**

The recent growing population draws attention to meet the requirements of the people to assure the food security and protect the environment. In this connection crop yield production and its early prediction is very crucial. Crop yield prediction using intelligent machine learning techniques may improve the crop planning decisions. Here in this proposed research many comparisons between the various machine learning algorithms like Random tree, KStar, Bayes Net and J48 using the dataset are performed to get the most accurate technique for crop yield prediction.

In our research, it is predicted that the experimental comparisons of various algorithms have diverse accuracy results which further may be instrumental to the farmers. The essence of this result throws light that choosing accurate algorithm and multidimensional dataset can entrust farmers with early crop yield predictions and recommendations. In future we will apply these algorithms for large datasets to suggest one seasonal decision support system using weather and non-weather information for Indian farmers.

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