**A Review of Machine Learning Techniques for Crop Yield**

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**Abstract:**

When machine learning (ML) techniques are used to analyses agricultural sensor data acquired by sensor devices, one of the agricultural sector's difficulties, yield forecasting, is resolved. The Internet of Things (IoT) provides different applications in digital agriculture, including crop growth monitoring, crop selection, and irrigation decision assistance. The Internet of Things (IoT) combines physical items transforming data. To assist farmers in making the best decisions, sensors are linked to the Internet of Things (IoT) in the field of agriculture. This is done through sensor networks. The main benefit is that using WSN in Precision Farming (PF) will reduce the need for water fertilizers while increasing crop output.

**Keywords:**

Crop yield, Internet of Things (IoT), Precision Farming (PF), Machine Learning (ML)

**1. Introduction**

In India, agriculture is the primary source of income for the population. A little more than 70% of people work in agriculture. A crucial part of daily life is agriculture. For us, food is a need. We must create an enormous farming industry to provide food to others. India and other developing nations rely on agriculture. So, the growth of the Indian economy is dependent on agriculture. In precision agriculture, we may employ IoT to increase efficiency and production. Sensor nodes in the fields can collect agricultural data, which is then sent to a data server where it is processed and stored. IoT applications in agriculture are crucial in making functions accessible and appropriate. Precision farming, wildlife monitoring, and other uses are made possible by the combination of WSNs with IoT. The weather and environmental factors, such as seasonal temperature, daily temperature ranges, and water cycles between the soil and atmosphere, have a significant impact on crop growth, quality, and yield. Precision farming (PF) may use inputs including temperature, humidity, soil moisture, and pH value to maximise crop growth while using the fewest resources possible. We can anticipate rainfall, crop selection and production predictions, and agricultural disease predictions with the use of machine learning algorithms. Crop management in agriculture involves controlling insecticides, fertilisers, and irrigation for optimum development. For precise agricultural production forecasts, many authors provide a variety of ways using IoT-based WSNs and machine learning algorithms.

***1.1 Overview of IoT***

The Internet of Things (IoT) is the interconnection of physical objects with the capacity to transmit data across a network without the need for human involvement.

Things (items) in the IoT may be anything, including sensors, people, cameras, computers, and phones. Any kind of data may be uploaded to the internet with these devices. IoT has several uses in agriculture, including "smart agriculture," which refers to uses of IoT solutions. The following facts explain why IoT technology is so effective.

1. Device-based global connection.

2. Minimising human effort is possible.

3. More rapid access.

4. Efficiency of time.

**1.2 Role of IoT in Agriculture**

The use of Internet of Things (IoT) devices in the agricultural sector has the potential to improve crop productivity significantly. This technological advancement is often referred to as intelligent farming. IoT devices have been used for the purpose of monitoring the soil condition, temperature, and humidity of crops in agricultural areas. Internet of Things (IoT) sensors can provide instantaneous data pertaining to many agricultural aspects such as crop yields, rainfall patterns, insect infestation levels, and soil nutrition content.

**1.2.1 IoT Application in Agriculture**

There are several types of Internet of Things (IoT) sensors and apps that can be effectively used within the agricultural sector.

• Climate Condition Monitoring

• Intelligent Greenhouse

• The topic of crop water management is of great importance in the field of agriculture.

• Agricultural drones have emerged as a significant technological advancement in the agricultural sector.

• The conservation of fertiliser and chemical crop protection agents.

Precision farming, often known as precision agriculture, is a modern farming approach that utilises advanced technologies and data analysis to optimize agricultural practices.

***1.3 WSN with IoT in Agriculture***

The topic of interest is a wireless sensor network, which refers to a network that operates without the need for physical connections and is composed of sensors. This network has a specific and well-defined topology. A wireless sensor network (WSN) is a network comprised of miniature electronic devices that are equipped with various types of sensors. The purpose of the wireless sensor network is to collect data from the aforementioned sensors. Wireless Sensor Networks (WSN) may be seen as a subset of the Internet of Things (IoT). Wireless Sensor Networks (WSNs) have the capability to establish connections with the Internet of Things (IoT).

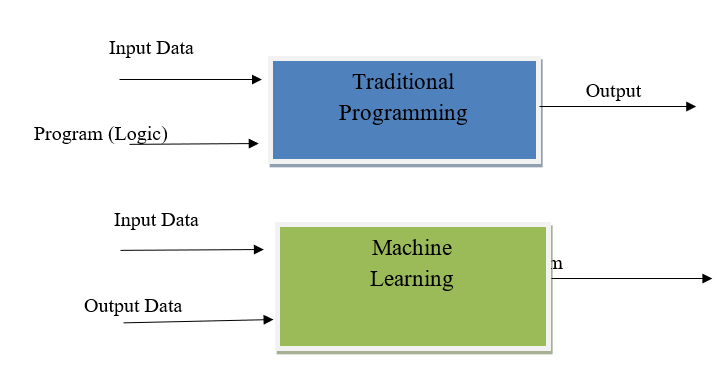
***1.4 Precision Farming***

Precision agricultural (PF) or Precision Agriculture (PA) refers to a highly precise and technologically advanced agricultural approach that involves the use of AI (Artificial Intelligence) processed data for the purpose of planting, cultivating, and harvesting crops. Crop data plays a crucial role in predicting optimal strategies for various agricultural activities. The functionality of PF is contingent upon the integration of hardware, software, and IT services. Precision agriculture (PA) is a farm management method that leverages information technology (IT) services to optimize the production of plants and commodities.

***1.5 Machine Learning***

A computer programme is considered to possess the ability to acquire knowledge from experience E in relation to a certain task T and performance measure P, if its performance on task T, as evaluated by performance measure P, demonstrates improvement as a result of experience E.

Arthur Samuel defines Machine Learning (ML) as a discipline that empowers computers to acquire knowledge and improve performance via learning, without the need for explicit programming. The primary objective is to enable computers to acquire knowledge autonomously, without the need for human intervention.



Traditional

Programming

Figure 1: Procedure Traditional programming and Machine Learning

Machine

Learning

In the context of traditional programming, the input data and programmed logic are provided, afterwards executed on a computer, and provide an output as the resultant outcome.

In the discipline of Machine Learning, input data and corresponding output are provided to the machine during the training phase. The machine then utilizes this data to generate a customized programmed or logic. This programmed may then be evaluated and tested during the testing phase.

***Applications of Machine Learning***

***Image Recognition-*** ML can be used for face recognition in an image.

***Medical diagnosis-*** ML methods and tools that can assistance in the diagnosis of diseases.

***Prediction-*** ML can also be used in the prediction systems.

***Agriculture***- ML applications have been used in Agriculture like Selection of the crop and crop Yield forecasting, water forecasting, smart irrigation, Crop Disease Prediction.

Table 1: Various Applications of Machine Learning Algorithms in Agriculture

|  |  |  |
| --- | --- | --- |
| **S. No** | **ML Applications in Agriculture** | |
| **Field of Study** | **ML Algorithms** |
| 1 | Crop Selection and Yield Prediction | Artificial Neural Network(ANN) |
|  |  | Random Forest Algorithms |
| 2 | Weather Forecasting | Support Vector Machine (SVM) |
| 3 | Smart Irrigation | Depends on data dimension we can use Support Vector Machine (SVM) and KNN (K-Nearest Neighbor) |
| 4 | Crop Disease Prediction | ANN |
|  |  | Regression Trees |
|  |  | Random Forest |

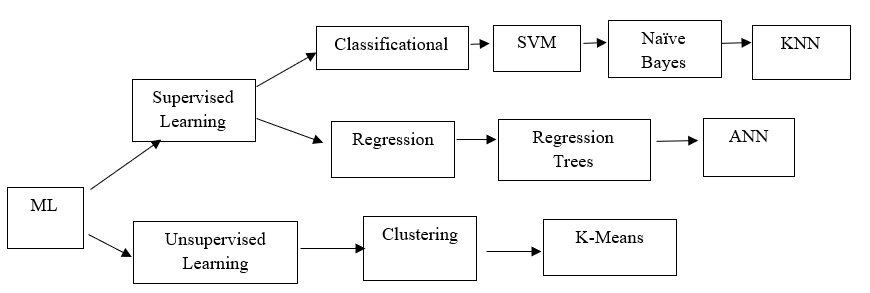
**Types of Machine Learning Algorithms**

Figure 2: Categorical ML

*Supervised Learning*: In Supervised learning, a known data (label) is available for a certain dataset like training data. Means as an input uses the known and label data and it gives accurate and reliable results. Supervised learning is applied in classificational algorithms like Support Vector Machine (SVM), Naïve Bayes and K-Nearest Neighbors (KNN), regression types Algorithms like Regression trees and Artificial Neural Network (ANN).

*Unsupervised Learning*: There is no training will be given to the machine, where only the input data is present and no corresponding output variable is there.

Unsupervised learning classified into Clustering where you want to discover the inherent groupings in the data.

**General procedure for Crop Yield Prediction System**

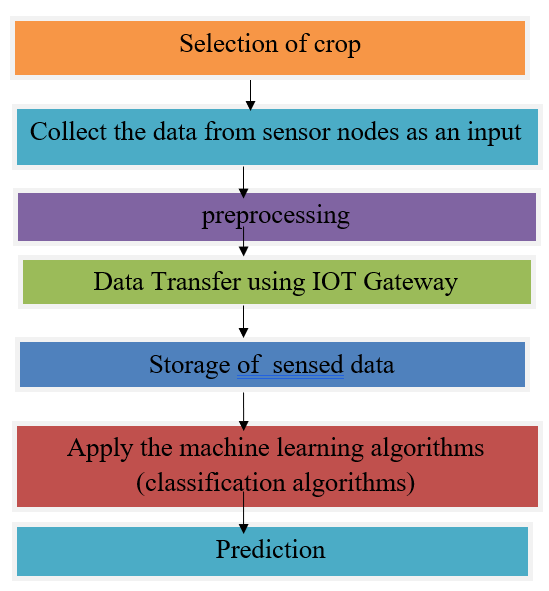
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Figure 3: Flow chart Diagram of Prediction

The methodology for predicting crop production is shown in Figure 3. The processing procedure involves selecting a specific crop and gathering data from various sensor nodes located inside the agricultural field. Data preprocessing refers to the process of preparing data for analysis by addressing issues such as redundancy and inconsistency within the dataset. The elimination of redundant data from sensory data can be achieved through the implementation of filtering techniques such as Kalman filtering. Once the noise has been removed from the sensed data, the resulting high-quality data can be transmitted to cloud or secondary storage devices, such as an SD card, via an IoT Gateway. Subsequently, machine learning algorithms can be applied to the stored data in order to predict crop yield.

Section 2 provides an overview of the existing literature on crop selection and yield prediction, highlighting the many machine-learning techniques proposed by different authors in this domain. Section 3 presents a comparative analysis of the outcomes obtained by several writers who used distinct machine learning methods. In contrast, Section 4 offers conclusive remarks.

**2.Related work:**

**Hetal Patel and Dharmendra Patel (2016) [1**], authors propose different data mining methods especially classification methods such as J48, Simple Cart, and Naïve Bayes algorithms of the decision tree are discovering harvest forecast with accuracy 89.33%,85.66%, and 82.66% respectively.

**D. Ramesh and B. Vardhan (2015**) [2], authors proposed methodology for crop yield forecast using Density-based clustering technique and Multiple Linear Regression (MLR) for certain crops.

**S.Nagini** [3], This paper explains data analysis and considers designing of the various predictive techniques. In this paper, relative learning of several data analytics method were gained. This helps us to refer what type of technique was finest suitable for our proposed system.

**R. Kalpana** [4], This study focus to search out proper data processing techniques to understand more precision and predict potential. Lastly, exploitation data processing systems in agriculture could be an up to date technique to find out the proper solution over the normal and traditional systems.

**Sellam,** et al [5], describe several environmental variables such as Area Under Cultivation (AUC), Annual rainfall (AR) and Food price index (FPI) which were work on the crop harvest and the correlation among these variables were analyzed.

**Monali Paul**, [6], This classification is done based on data mining technique. In this article, we examined the cataloging instructions and recognized which will be suitable for data set.

**N. Hema Geetha** [7], Authors explained about several methods like Market-based Analysis, Association rule mining, Decision Trees (DT).

**AwanitKumar** [8], Authoer explain about K-Means algorithm and how it will be applied for analysing the data for maximizing the crop yield based on soil and weather data.

[**M. Gunasundari Ananthara**](https://ieeexplore.ieee.org/search/searchresult.jsp?searchWithin=%22Authors%22:.QT.M.%20Gunasundari%20Ananthara.QT.&newsearch=true) [9], This paper explains the Bee Hive algorithm for forecasting harvest yield. This system deals with the huge data file.

**Thomas Truong** [10], applying the support vector machine regression (SVMr) technique on raw data and process a raw data and forecast the result. SVM gives result but which is less accurate than other algorithms.

**Giritharan Ravichandran [11]**, the author explains in this paper about Artificial Neural Network which is a successful tool and applied for modeling and forecasting.

**Jagielska et al. [12],** authors defined Yield forecasting which is an important problem in agriculture.

**Veenadhari**, [13], The Decision Tree analysis explains the paddy crop yield. The guidelines have been formed from the Decision Tree and which are applied to discovering context planned for maximum or minimum harvest efficiency.

**Shalvi D [14],** The Bayesian System is a crucial tool and extensively applied in data file of agriculture. The method succeeds for applications of agriculture which has been based on the Bayesian Network Technique. consequence display that Bayesian Networks are achievable and valuable.

**Charles L. Hornbaker [15]**, have built a spatial method of maize yields in the US Corn Belt that uses the Bayesian prior estimation model for every state in the belt region which influence spatial smoothness among the regression coefficients to mitigate the effects of noisy data across regions and to develop yield prediction. This helps in formulating an in-season prediction mode2.

The combination of Information Technology (IT) and agriculture assist in prediction harvest yield. It is compulsory to create an appropriate technique will have some merits over the conservative prediction mechanism. A comprehensive study of different ML algorithms applied for different applications in agriculture as shown in below table 2.

|  |  |  |  |
| --- | --- | --- | --- |
| **Ref .No** | **Author** | **ML**  **Methodologies** | **Applications** |
| 16 | Chen C,  Mcnairn H | Neural Network | Rice Crop Monitoring |
| 17 | Co HC, Boosarawongse R | Artificial Neural Networks (ANNs) | Forecasting Rice Export |
| 18 | Monisha Kaul M, Robert L | Artificial Neural Networks (ANNs) | Corn and Soybean Yield Forecast |
| 19 | Prasad PR, Begum SA | Regression and Neural Networks Models19 | Forecast of Crop Yield. |
| 20 | Dahikar MSS, Rode SV | Artificial Neural Network Approach20 | Agricultural Crop Yield Forecast |
| 21 | Stathakis D, Savin I, Negre T | Neuro-Fuzzy Modeling | Crop Yield Forecast |
| 22 | Papageorgiou EI, Aggelopoulou KD, Gemtos TA | Fuzzy Cognitive Map learning approach | Yield Forecast in Apples. |
| 23 | Petridis V, Kaburlasos VG. FINk NN | Finken: a fuzzy interval number k-Nearest Neighbor classifier | forecast of Sugar production from populations of samples |
| 24 | Salleh MNM | A Fuzzy Modelling of Decision Support System24 | for Crop Selection |
| 25 | Papageorgiou EI, Aggelopoulou KD | Fuzzy Cognitive Map learning approach | Yield Forecast in apples |
| 26 | Veenadhari S, Mishra B, Singh CD | Decision Tree Algorithms | Soybean Productivity Modelling |
| 27 | Kumar AVTV, Rajini Kanth R | Data mining with the climate variable | Jowar Crop Yield in India |
| 28 | Priya SRK, Suresh KK | Regression | sugarcane yield using climatic variables |
| 29 | Shibayama M | Linear Regression | Forecasting Grain Yield of Maturing Rice |
| 30 | House CC | Nonlinear Regression | Forecasting Corn Yields |
| 31 | Matis JH, Birkett T, Boudreaux D. | Markov Chain Approach | Predicting Cotton Yields |
| 32 | Yiqun Gu Y, James W, McNicol M | Belief Networks | Future Crop Production |
| 33 | Jain RC, Rama subramalliall V | Second Order Markov Chains33 | Forecasting of Crop Yields |
| 34 | Hong-Ying L, Yan-Lin H, Yong-Juan Y, Hui-Ming Z | Time Series Techniques34 | Crop Yield Forecast |
| 35 | Utkarsha P, Narkhede N, Adhiya KP | Modified K-Means Clustering35 | Crop Prediction |

**Table2: Machine Learning methodologies used in agriculture**

**3. Comparison of Results for Different Machine Learning Algorithms applied in Agriculture**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Ref.NO** | **Name of the Author** | **Machine Learning Algorithms** | **Crop Type** | **Accuracy** |
| 1 | Hetal Patel and  Dharmendra Patel | The Classification Techniques  like J48 |  | 89.33% |
|  |  | Simple Cart |  | 85.66% |
|  |  | Naive Bayes |  | 82.66% |
| 36 | D Ramesh, B Vishnu Vardhan | Multiple Linear  Regression | Rice Yield | 90%-95% |
| 37 | Sudhanshu Sekhar Panda,  Daniel P. Ames , and  Suranjan Panigrahi, | Neural Networks | Corn yield | 95% |
| 38 | S.Veenadhari,Dr. Bharat Misra ,Dr. CD Singh | C4.5 Technique and  Decision Tree | Soyabean,  paddy,  maize | For Soyabean=87%  For Paddy=85%  For Maize=76% |
| 39 | Jefferson Lobato Fernandes; Jansle Vieira Rocha | Harmonic Analysis  of NDVI Time Series Algorithm | sugarcane | 86.5% |
| 40 | A.T.M Shakil Ahamed,  Navid Tanzeem Mahmood. | K-Means Algorithm  for clustering  And Classification Linear  Regression, K-NN,  ANN Method. | Wheat ,  Potato | 90% - 95% |
| 41 | D Ramesh,  B.Vishnu Vardhan | Multiple Linear Regression (MLR) | Rice Yield | 90-95% |
|  |  | K-Means algorithm | Rice Yield | 96% |

**Table 3: Comparison of various ML Techniques with Accuracy**.

**CONCLUSION**

This review article examines the use of IoT based WSN and Machine Learning methods for the analysis of Agricultural Data in the context of precision Farming. Various authors have put forth a range of machine learning algorithms, such as Artificial Neural Networks (ANN), Support Vector Machines (SVM), J48, and K-Nearest Neighbours (KNN), to enhance the accuracy and reliability of crop yield prediction in diverse agricultural domains including paddy, sugarcane, wheat, potato, and soybean. Additionally, a comparative analysis of the outcomes obtained from applying different machine learning algorithms to agricultural data is conducted.

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