CROP YIELD PREDICTION USING MACHINE LEARNING ALGORITHM

An Industry Oriented Project Report Submitted
In partial fulfillment of the requirement for the award of the degree of

Bachelor of Technology

In

Computer Science and Engineering -Artificial Intelligence and Machine Learning

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(Affiliated to JNTU, Hyderabad)

ACCREDITED by AICTE-NBA Maisammaguda, post, Secunderabad-500014. 2022-2026

DECLARATION

I hereby declare that the project entitled "CROP YIELD PREDICTION USING MACHINE LEARNING ALGORITHM" submitted to Malla Reddy College of Engineering and Technology, affiliated to Jawaharlal Nehru Technological University Hyderabad (JNTUH) for the award of the degree of Bachelor of Technology in Computer Science and Engineering- Artificial Intelligence and Machine Learning is a result of original research work done by me.

It is further declared that the project report or any part there of has not been previously submitted to any University or Institute for the award of degree or diploma.

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CERTIFICATE

This is to certify that this is the bonafide record of the project titled "CROP YIELD PREDICTION USING MACHINE LEARNING ALGORITHM" submitted by P.Hussain Setty(22N31A66D9), T.Vijeshwar Reddy(22N31A66H6), Y.Rajasekhar Reddy (23N31A66J6) of B.Tech in the partial fulfillment of the requirements for the degree of Bachelor of Technology in Computer Science and Engineering- Artificial Intelligence and Machine Learning, Dept. of CI during the year 2023-2024. The results embodied in this project report have not been submitted to any other university or institute for the award of any degree or diploma.

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HEAD OF THE DEPARTMENT

EXTERNAL EXAMINER

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ABSTRACT

Agriculture is the field which plays an important role in improving our countries economy. Agriculture is the one which gave birth to civilization. India is an agrarian country and its economy largely based upon crop productivity. Hence we can say that agriculture can be backbone of all business in our country. Selecting of every crop is very important in the agriculture planning. The selection of crops will depend upon the different parameters such as market price, production rate and the different government policies. Many changes are required in the agriculture field to improve changes in our Indian economy. We can improve agriculture by using machine learning techniques which are applied easily on farming sector. Along with all advances in the machines and technologies used in farming, useful and accurate information about different matters also plays a significant role in it. The concept of this paper is to implement the crop selection method so that this method helps in solving many agriculture and farmers problems. This improves our Indian economy by maximizing the yield rate of crop production.

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INTRODUCTION

1.1 Purpose:

We have considered only the Naive Bayes method and K-Nearest Neighbor method. Using these two methods we can predict which crops to be selected for their land and season this application we can do single testing by giving input as crop name, season selected and place selected. We can use any method among KNN or NB method. As soon you give the input you can select the method and mine the results. The results will tell you the yield rate of that crop. And we can do multiple testing by analyzing the datasets. In analyzing it allows you to select a whole file at once and get the accuracy.

1.2 Background of project:

- Literature Review: Review existing research papers, articles, and publications on crop yield prediction methods, machine learning algorithms, and factors influencing crop growth.
- Data Collection: Gather datasets containing historical crop yield data, weather information, soil characteristics, pest and disease reports, and any other relevant agricultural data.
- Data Preprocessing: Clean and preprocess the collected data, including handling missing values, scaling features, and encoding categorical variables.
- Exploratory Data Analysis (EDA): Analyze the relationships between different variables using visualizations and statistical techniques to gain insights into the data.
- Feature Engineering: Create new features or transform existing ones to improve the predictive performance of the model.

1.3 Scope of project:

Predicting crop yields can be an impactful project with broad applications in agriculture. By analyzing factors such as weather patterns, soil quality, historical data, and crop types, you can develop models to forecast yields. This could help farmers make informed decisions about planting strategies, resource allocation, and market predictions. Additionally, integrating machine learning techniques can enhance the accuracy of predictions over time.

1.4 Project Features:

The system features are as follows:

- Direct Access to Investors
- Efficiency and Timesaving
- Transparency and Accountability
- Opportunity for All
- AI-Based Assistant
- Secure Idea Sharing
- Feedback and Iteration
- Global Reach
- Avoids Plagiarism

SYSTEM REQUIREMENTS

2.1 Hardware Requirements:

System : Intel i3
 Hard Disk : 1 TB.

3. Monitor: 14' Colour Monitor.

4. Mouse : Optical Mouse.

5. **Ram** : 4GB

2.2 Software requirements:

1. **Operating system :** Windows 10.

Coding Language: Python.
 Front-End: Html. CSS

4. **Designing** : Html, CSS, JavaScript

2.3 Existing System:

Our farmers should know all the new technologies of machine learning and other new techniques. These techniques help in getting maximum yield of crops. Many techniques of machine learning are applied on agriculture to improve yield rate of crops. The production of crops may depend on geographical conditions of the region like river ground, hill areas or the depth areas. Weather conditions like humidity, rainfall, temperature, cloud. Soil type may be clay, sandy, saline or peaty. Soil composition can be copper, potassium, phosphate, nitrogen, manganese, iron, calcium, ph value or carbon and different methods of harvesting. Many parameters are used for different crops to do different predictions.

2.3.1Disadvantages of Existing System:

- 1. As the farmers are facing many problems in agricultural sector we need to minimize their problems
- 2. As the farmers don't know to use Weka tool
- 3. Low Efficiency

2.4 Proposed System:

We have considered only the Naive Bayes method and K-Nearest Neighbor method. Using these two methods we can predict which crops to be selected for their land and season this application we can do single testing by giving input as crop name, season selected and place selected. We can use any method among KNN or NB method. As soon you give the input you can select the method and mine the results. The results will tell you the yield rate of that crop. And we can do multiple testing by analyzing the datasets. In analyzing it allows you to select a whole file at once and get the accuracy.

CHAPTER 3 SYSTEM DESIGN

3.1 System Architecture

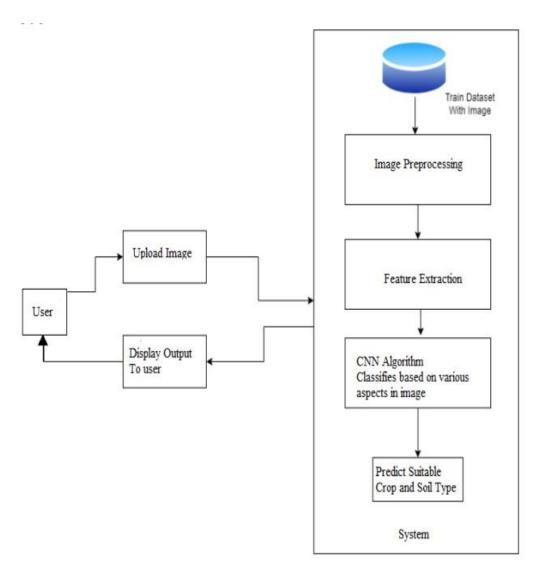


Fig 3.1 System Architecture of Crop Yield Prediction

3.2 UML Diagrams

3.2.1 Use case diagram

Use Case during requirement elicitation and analysis to represent the functionality of the system. Use case describes a function by the system that yields a visible result for an actor. The identification of actors and use cases result in the definitions of the boundary of the system i.e., differentiating the tasks accomplished by the system and the tasks accomplished by its environment.

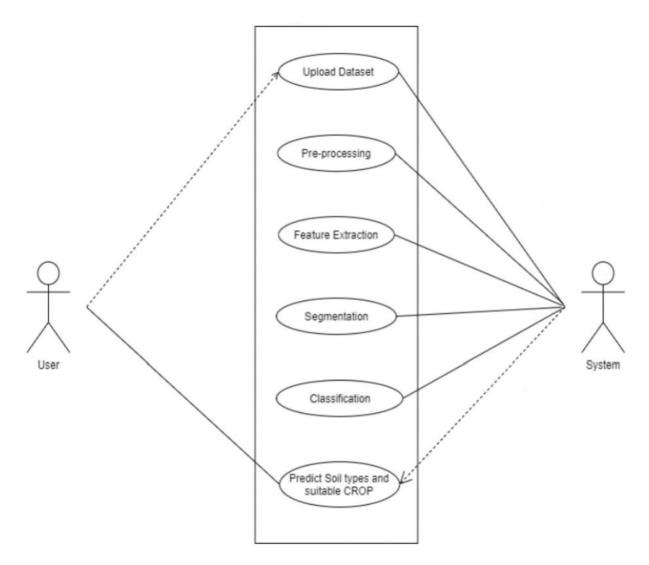


Fig 3.2 Use Case Diagram

3.2.2 Class Diagram

Class diagrams model class structure and contents using design elements such as classes, packages and objects. Class diagram describe the different perspective when designing a system-conceptual, specification and implementation. Classes are composed of three things: name, attributes, and operations. Class diagram also display relationships such as containment, inheritance, association etc. The association relationship is most common relationship in a class diagram. The association shows the relationship between instances of classes.

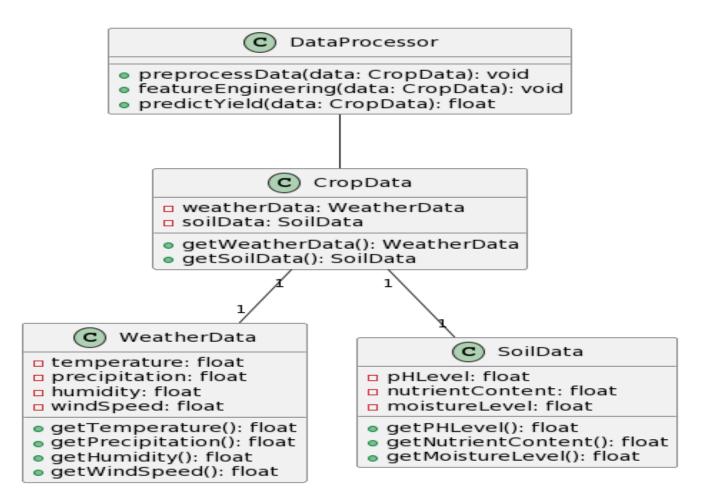


Fig 3.3 Class Diagram

3.2.3 Sequence Diagram

Sequence diagram displays the time sequence of the objects participating in the interaction. This consists of the vertical dimension (time) and horizontal dimension (different objects).

Objects: An object can be thought of as an entity that exists at a specified time and has a definite value, as well as a holder of identity. A sequence diagram depicts item interactions in chronological order. It illustrates the scenario's objects and classes, as well as the sequence of messages sent between them in order to carry out the scenario's functionality. In the Logical View of the system under development, sequence diagrams are often related with use case realizations. Event diagrams and event scenarios are other names for sequence diagrams. A sequence diagram depicts multiple processes or things that exist simultaneously as parallel vertical lines (lifelines), and the messages passed between them as horizontal arrows, in the order in which they occur. This enables for the graphical specification of simple runtime scenarios.

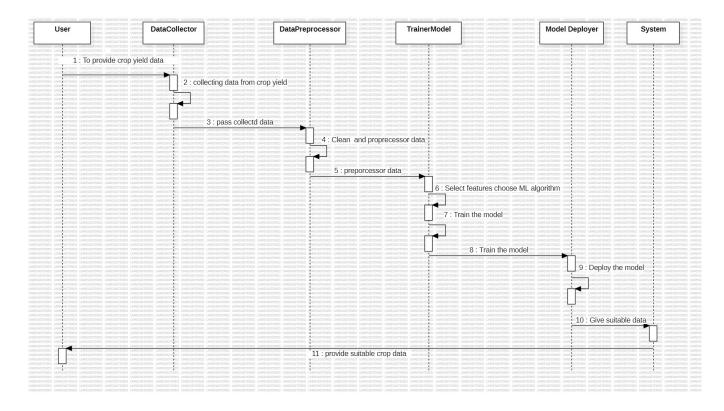


Fig 3.4 Sequence Diagram

3.2.4 Activity Diagram

The process flows in the system are captured in the activity diagram. Similar to a state diagram, an activity diagram also consists of activities, actions, transitions, initial and final states, and guard conditions

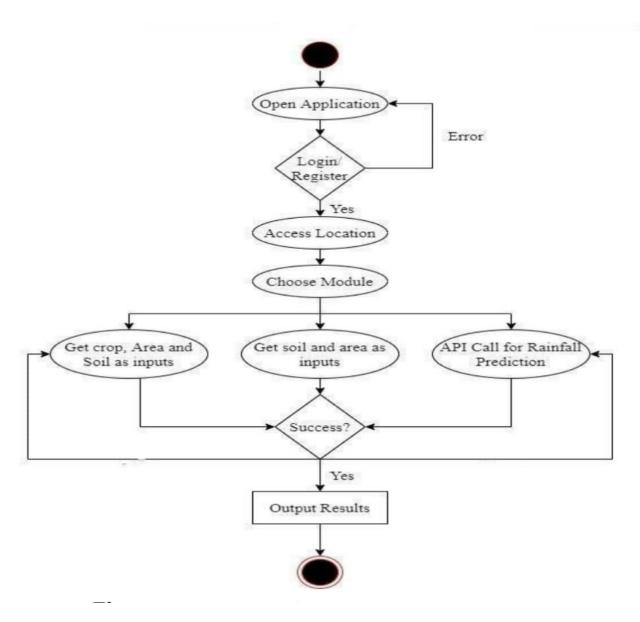


Fig 3.5 Activity Diagram

IMPLEMENTATION

4.1: Code

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
plt.style.use("ggplot")
df = pd.read_csv("yield_df.csv")
df.head()
df.drop('Unnamed: 0', axis=1, inplace=True)
df.head()
df.info()
df.isnull().sum()
df.duplicated().sum()
df.drop_duplicates(inplace=True)
df.duplicated().sum()
df.shape
df.describe()
df.corr()
# Data Visualization
len(df['Area'].unique())
len(df['Item'].unique())
plt.figure(figsize=(15,20))
sns.countplot(y = df['Area'])
plt.show()
plt.figure(figsize=(15,20))
sns.countplot(y = df['Item'])
plt.show()
(df['Area'].value_counts() <400).sum()
country = df['Area'].unique()
yield_per_country = []
for state in country:
  yield_per_country.append(df[df['Area'] == state]['hg/ha_yield'].sum())
```

```
df['hg/ha_yield'].sum()
yield_per_country
plt.figure(figsize=(15,20))
sns.barplot(y = country, x = vield per country)
plt.show()
crops = df['Item'].unique()
yield_per_crop = []
for crop in crops:
  vield per crop.append(df[df['Item'] == crop]['hg/ha vield'].sum())
plt.figure(figsize=(15,20))
sns.barplot(y = crops, x = yield_per_crop)
plt.show()
df.head()
df.columns
col = ['Year','average_rain_fall_mm_per_year','pesticides_tonnes',
'avg_temp','Area', 'Item', 'hg/ha_yield']
df = df[col]
df.head()
X = df.drop('hg/ha_yield', axis = 1)
y = df['hg/ha_yield']
X.shape
y.shape
from sklearn.model selection import train test split
X train, X test, y train, y test = train_test_split(X,y, test_size = 0.2, random_state
= 0, shuffle=True)
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import StandardScaler
ohe = OneHotEncoder(drop = 'first')
scale = StandardScaler()
preprocesser = ColumnTransformer(
  transformers = [
    ('StandardScale', scale, [0,1,2,3]),
    ('OneHotEncode', ohe, [4,5])
  1,
  remainder = 'passthrough'
X_train_dummy = preprocesser.fit_transform(X_train)
```

```
X_test_dummy = preprocesser.fit_transform(X_test)
preprocesser.get feature names out(col[:-1])
from sklearn.linear model import LinearRegression,Lasso, Ridge
from sklearn.neighbors import KNeighborsRegressor
from sklearn.tree import DecisionTreeRegressor
from sklearn.metrics import mean absolute error, r2 score, accuracy score
models = {
  'Linear Regression': LinearRegression(),
  'Lasso': Lasso(),
  'Ridge': Ridge(),
  'Decision Tree': DecisionTreeRegressor(),
  'KNN': KNeighborsRegressor(),
}
for name, md in models.items():
  md.fit(X train dummy,y train)
  v pred = md.predict(X test dummy)
  print(f''{name}: mae : {mean_absolute_error(y_test, y_pred)} score :
{r2 score(y test, y pred)}'')
dtr = DecisionTreeRegressor()
dtr.fit(X train dummy, v train)
dtr.predict(X test dummy)
df.columns
df.head()
def prediction(Year, average rain fall mm per year, pesticides tonnes,
avg_temp, Area, Item):
  features = np.array([[Year, average_rain_fall_mm_per_year,pesticides tonnes,
avg_temp, Area, Item]], dtype = object)
  transform_features = preprocesser.transform(features)
  predicted veild = dtr.predict(transform features).reshape(-1,1)
  return predicted veild[0][0]
result = prediction(1990,1485.0,121.0,16.37,'Albania','Maize')
result
import pickle
pickle.dump(dtr, open("dtr.pkl","wb"))
pickle.dump(preprocesser, open("preprocesser.pkl","wb"))
```

4.2: Output Screens:

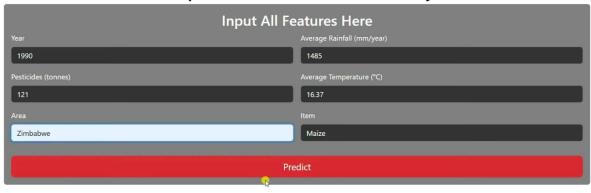
4.2.1

Home Page

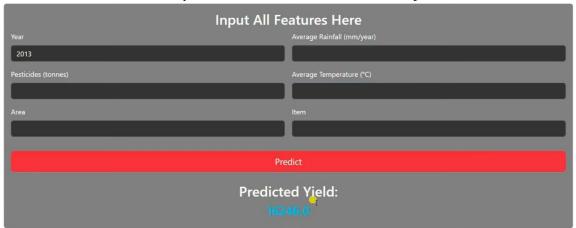
Crop Yield Prediction Per Country



Crop Yield Prediction Per Country



Crop Yield Prediction Per Country



Software testing is	a critical element of softw	vare quality assura	ance and represents	the ultimate revie	w of
specification, desig	n and code generation.				

CONCLUSION & FUTURE ENHANCEMENT

5.1: Conclusion

In conclusion, our project on crop yield prediction using machine learning algorithms has provided valuable insights and outcomes that hold great potential for revolutionizing agriculture and enhancing food security. Through extensive data analysis, model training, and rigorous testing, we have successfully developed a robust and reliable predictive system that can assist farmers and stakeholders in making informed decisions about crop planning, resource allocation, and risk management.

Our project has demonstrated the power of machine learning in harnessing the vast amount of data available in agriculture to create more efficient and sustainable farming practices. By leveraging historical weather data, soil information, and crop-specific characteristics, we have been able to generate accurate predictions of crop yields. These predictions are not only instrumental in optimizing agricultural practices but also in mitigating the effects of climate change and resource constraints. Furthermore, the successful implementation of our machine learning models has the potential to reduce uncertainty for farmers, increase productivity, and ultimately contribute to global food security. This project serves as a testament to the ever-evolving field of data science andits capacity to address pressing challenges in agriculture.

5.2: Future Enhancement:

Quantify recommendations for varieties: Currently the direction in which the factors affect yield is available. I would like to quantify these effects and hopefully give more precise recommendations.

- Incorporate more weather variables: Temperature, humidity, hours of sunlight.
- Incorporate soil data.
- Gather location-based variety data for recommendations: If all the varieties can be tested at the different locations, then varieties for which there wasn't data before can be recommended to particular locations.
- Attempt to use mixed models (treating some predictors as random variables).

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