

Intelligent Crop Recommendation System using Machine Learning

Sanjivani Kadam¹, Tejal Survase², Siddhi Velapure³, Mayuri Swami⁴, Ratan Kore⁵, Prapti Atkale⁶

Assistant Professor, Department of Computer Science and Engineering¹

Students, Department of Computer Science and Engineering^{2,3,4,5,6}

SVERI's College of Engineering, Pandharpur, Maharashtra, India

Punyashlok Ahilyadevi Holkar Solapur University, Gopalpur, Pandharpur, Maharashtra, India

Abstract: *Intelligent crop recommendation systems have gained significant attention in modern agriculture for their potential to optimize crop selection and enhance agricultural productivity. This report aims to provide a comprehensive analysis of existing intelligent crop recommendation systems and propose a novel framework for future system development. The report explores the current state-of-the-art in the field, identifies the key components and functionalities of existing systems, and evaluates their strengths and limitations. Building upon this analysis, a proposed system framework is presented, encompassing data acquisition, preprocessing, machine learning algorithms, recommendation generation, and user interface. The proposed system addresses the limitations of existing systems and leverages emerging technologies for improved accuracy, scalability, and sustainability. The report concludes with a discussion on the potential impact of the proposed system on agricultural practices and highlights future research directions*

Keywords: *Agriculture, Maximum Crop Yield, Fertilizer Suggestion, Environmental Factor, Economic Factor, Machine Learning(ML), Plant Disease Classification*

I. INTRODUCTION

In recent years, there has been a growing need for efficient and sustainable agricultural practices to meet the increasing demand for food production. One critical aspect of successful farming is the selection of suitable crops for specific environmental conditions, soil characteristics, and market demands. However, this task can be challenging for farmers due to the complexity and variability of these factors. To address this challenge, intelligent crop recommendation systems have emerged as valuable tools for optimizing crop selection. These systems leverage advanced technologies such as machine learning, data analytics, and artificial intelligence to provide data-driven recommendations to farmers. By analyzing large volumes of agricultural data, including historical crop yields, climate data, soil information, and market trends, these systems can offer personalized and accurate crop recommendations. The fundamental goal of an intelligent crop recommendation system is to enhance agricultural productivity and profitability while minimizing environmental impact. By suggesting the most suitable crops based on the specific conditions of a farm, these systems can help farmers make informed decisions and maximize their yields. Additionally, intelligent crop recommendation systems can aid in resource management by optimizing the use of fertilizers, pesticides, and water, thus reducing waste and potential harm to the environment.

The success of intelligent crop recommendation systems relies heavily on the quality and diversity of the data used for analysis. Integration of various data sources, such as satellite imagery, weather stations, soil sensors, and market data, provides a comprehensive understanding of the agricultural landscape. With access to real-time and historical data, these systems can continuously learn and adapt, improving the accuracy of their recommendations over time.

Furthermore, intelligent crop recommendation systems can consider multiple objectives and constraints, such as crop rotation, disease resistance, and market demand, to provide holistic and tailored recommendations to farmers. By considering a wide range of factors, these systems help optimize crop selection for long-term sustainability and economic viability.

II. LITERATURE REVIEW

[1] S.Pudumalar*, E.Ramanujam*, R.HarineRajashreeñ, C.Kavyań, T.Kiruthikań, J.Nishań. :: In a country like India in which agriculture plays a dominant role. The prosperity of nation increases because of farmers prosperity. Thus our work would help farmers in sowing the right seed based on soil requirements to enhance productivity and gain profit out of such a technique. Thus the farmers can plant the right crop improving their income and also increasing the gross productivity of the state. Our future work is focused on an upgraded data set with a large number of attributes and also implements yield prediction

[2] Shikha Ujjainia, Pratima Gautam, S. Veenadhari :: From the above analysis, it has been concluded that technology has achieved that level of competency by which agriculture can easily predict crop yielding production. Machine learning technology proposed the model integrated form of a concept. By estimating the different parameters of biosystems has been recognized that the technology used for making the crop yielding prediction device is very much variegated. The parameters of biosystems vary concerning changing location and a single concept of algorithm is not sufficient to fulfill the requirement of crop prediction.

[3] Mahendra Choudhary, Rohit Sartandel, Anish Arun, LeenaLadge SIES Graduate School of Technology, Maharashtra, India Corresponding author: Rohit Sartandel, Email:rohit.sartandel18@siesgst.ac.i :: At present, our farmers are not using technology and analytics productively, so there is a probability of fallacious crop selection for cultivation, which will reduce their income. To fend off such dropping, we have developed a farmer-friendly system with a graphical user interface (GUI) that will predict which crop would be the best fit for a specific plot of land. This system will also furnish details on the essential nutrients to add to the soil and help them to pick out crop diseases. As a result, farmers are more likely to make the proper decisions for crop selection, and farmers will benefit from earning more profits. With the help of more composite algorithms, the efficiency of the model can be increased. To boost the efficiency of crop selection and yield, incorporate the crop recommendation system with yield prediction. By increasing the dataset's size and by including more crop information and pictures of infected plants, the efficiency of the model can be increased.

II. PROPOSED SYSTEM

The proposed system harnesses the power of data analysis technology to actively monitor and update crop yield rates. Its primary objective is to introduce a robust crop selection method that effectively tackles the challenges faced by farmers and the agricultural sector. By maximizing crop yield rates, this approach seeks to make a substantial contribution to the overall improvement of the Indian economy.

One of the key strengths of the system lies in its ability to consider diverse land conditions and accurately identify the quality of crops. Through this process, the system distinguishes high-quality crops, enabling farmers to make informed decisions about their cultivation choices. To enhance the precision of crop predictions, the system leverages an ensemble of classifiers, such as Decision Tree and Random Forest classifiers. By combining the outputs of multiple classifiers, the system enhances decision-making capabilities, leading to more reliable predictions and improved efficiency.

The integration of ensemble classifiers into the decision-making process further strengthens the overall system performance. By incorporating the insights and predictions from multiple classifiers, the system attains higher accuracy and better adaptability. This integration ensures that the proposed system maximizes the potential of data analysis technology and effectively addresses the challenges faced by farmers and the agricultural sector. In summary, the proposed system capitalizes on data analysis technology to continually monitor and update crop yield rates. Its core objective is to introduce a comprehensive crop selection method that overcomes existing agricultural challenges. By prioritizing the maximization of crop yield rates, the system aims to contribute significantly to the advancement of the Indian economy. Through its consideration of diverse land conditions, accurate crop quality identification, and integration of ensemble classifiers, the system enhances prediction accuracy, decision-making capabilities, and overall system efficiency.

2.1 Being System

The proposed project introduces a novel model aimed at addressing existing agricultural challenges. The system's key focus is to guide farmers in maximizing crop yield and selecting the most profitable crops for specific regions.

By considering economic and environmental conditions, the proposed model offers crop selection recommendations that prioritize both maximizing yield and meeting the country's increasing food demand. The model predicts crop yield based on factors such as state, district, area, and season, while also assisting in determining the optimal timing for fertilizer application.

For production-related inputs, the user provides information on the state, district, season, crop, and area. For crop recommendation, the user provides details on the state, district, season, and area. Based on these inputs, the model predicts the crop yield for a specific crop and suggests the most profitable crop option, along with the ideal time for fertilizer application.

The primary objective of the system is to offer a diverse range of crops suitable for each season. By providing guidance in crop selection, the proposed system aims to minimize farmers' challenges and maximize overall yield.

Overall, the proposed system presents a valuable solution that assists farmers in selecting crops, maximizing yield, and overcoming difficulties in decision-making.

III. PROPOSED WORKFLOW

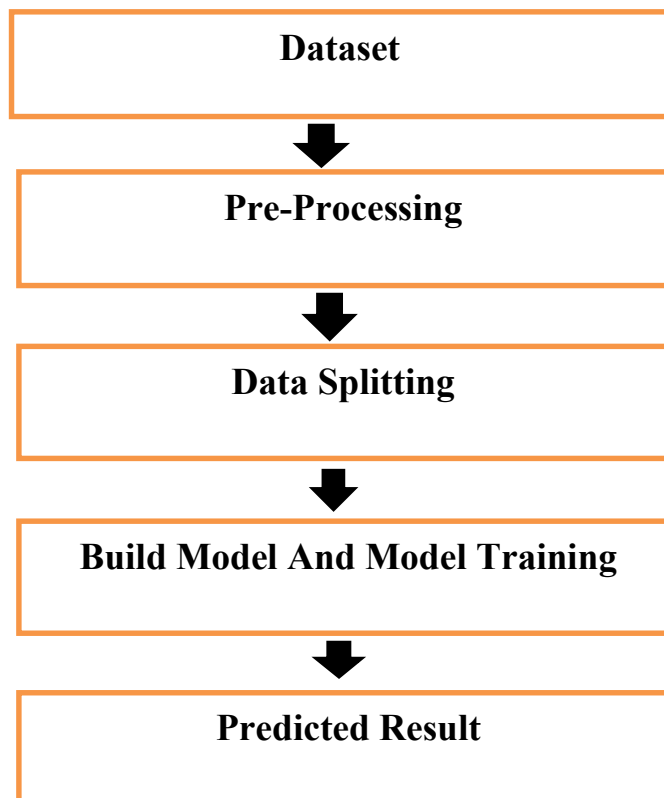
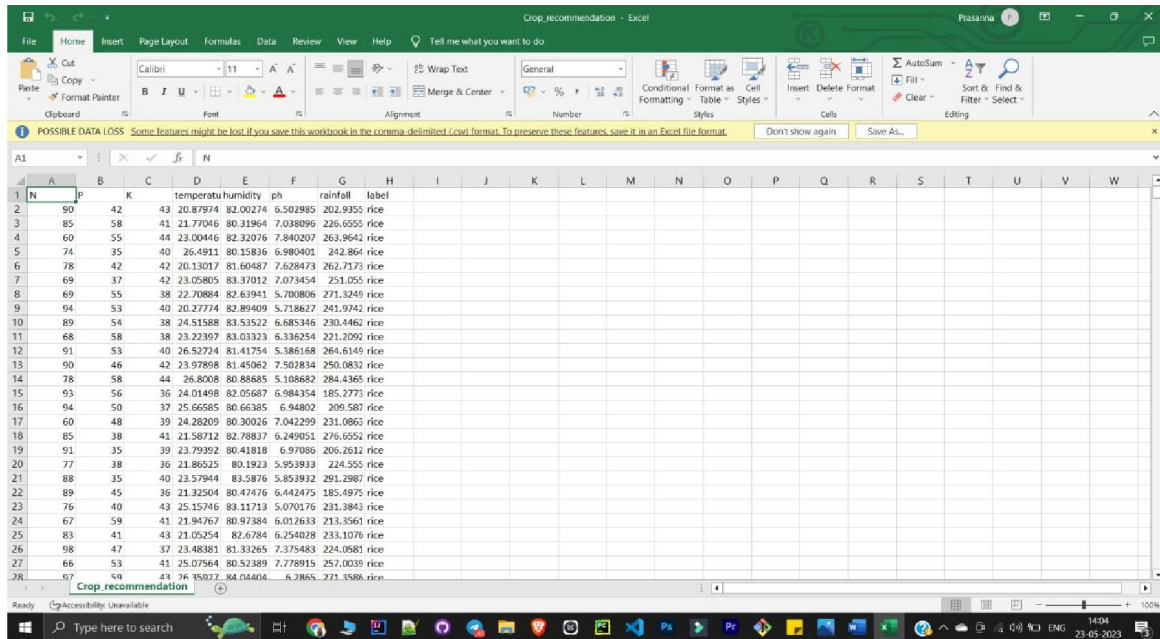


Fig: Proposed Workflow

IV. IMPLEMENTATION

4.1 Input Dataset

The dataset provided to the system as inputs which include parameters like soil pH, humidity, N, P, K, temperature rainfall and the labels of the crops which are going to be given by a system as output.



| | A | B | C | D | E | F | G | H | I | J | K | L | M | N | O | P | Q | R | S | T | U | V | W |
|----|----|----|----|-----------|----------|----------|----------|-------|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| 1 | N | P | K | temperatu | humidity | ph | rainfall | label | | | | | | | | | | | | | | | |
| 2 | 90 | 42 | 43 | 20.87974 | 82.00274 | 6.502985 | 202.9355 | rice | | | | | | | | | | | | | | | |
| 3 | 85 | 58 | 41 | 21.77046 | 80.31964 | 7.038096 | 226.6555 | rice | | | | | | | | | | | | | | | |
| 4 | 60 | 55 | 44 | 23.00446 | 82.32076 | 7.840207 | 263.9642 | rice | | | | | | | | | | | | | | | |
| 5 | 74 | 35 | 40 | 26.9311 | 80.15836 | 6.980401 | 242.864 | rice | | | | | | | | | | | | | | | |
| 6 | 78 | 42 | 42 | 20.13017 | 81.60487 | 7.628473 | 262.7173 | rice | | | | | | | | | | | | | | | |
| 7 | 69 | 37 | 42 | 23.05805 | 83.37012 | 7.073454 | 251.055 | rice | | | | | | | | | | | | | | | |
| 8 | 69 | 55 | 38 | 22.70884 | 82.63941 | 5.700806 | 271.3249 | rice | | | | | | | | | | | | | | | |
| 9 | 94 | 53 | 40 | 20.27774 | 82.89409 | 5.718627 | 241.9742 | rice | | | | | | | | | | | | | | | |
| 10 | 89 | 54 | 38 | 24.51588 | 83.53522 | 6.685346 | 230.4462 | rice | | | | | | | | | | | | | | | |
| 11 | 68 | 58 | 38 | 23.22397 | 83.03323 | 6.336254 | 221.2092 | rice | | | | | | | | | | | | | | | |
| 12 | 91 | 53 | 40 | 26.52724 | 81.41754 | 5.386168 | 264.6149 | rice | | | | | | | | | | | | | | | |
| 13 | 90 | 46 | 42 | 23.97898 | 81.45062 | 7.502834 | 250.0831 | rice | | | | | | | | | | | | | | | |
| 14 | 78 | 58 | 44 | 26.80208 | 80.88685 | 5.108682 | 284.4365 | rice | | | | | | | | | | | | | | | |
| 15 | 93 | 56 | 36 | 24.01498 | 82.05687 | 6.984354 | 185.2773 | rice | | | | | | | | | | | | | | | |
| 16 | 94 | 50 | 37 | 25.66585 | 80.66385 | 6.94802 | 209.587 | rice | | | | | | | | | | | | | | | |
| 17 | 60 | 48 | 39 | 24.28209 | 80.30026 | 7.042299 | 231.0863 | rice | | | | | | | | | | | | | | | |
| 18 | 85 | 38 | 41 | 21.58712 | 82.78837 | 6.249051 | 276.9552 | rice | | | | | | | | | | | | | | | |
| 19 | 91 | 35 | 39 | 23.79392 | 80.41818 | 6.97086 | 206.2612 | rice | | | | | | | | | | | | | | | |
| 20 | 77 | 38 | 36 | 21.86525 | 80.1923 | 5.953933 | 224.555 | rice | | | | | | | | | | | | | | | |
| 21 | 88 | 35 | 40 | 23.57944 | 83.5876 | 5.853932 | 291.2987 | rice | | | | | | | | | | | | | | | |
| 22 | 89 | 45 | 36 | 21.32504 | 80.47476 | 6.442475 | 185.4975 | rice | | | | | | | | | | | | | | | |
| 23 | 76 | 40 | 43 | 25.15746 | 83.11719 | 5.070176 | 231.3843 | rice | | | | | | | | | | | | | | | |
| 24 | 67 | 59 | 41 | 21.94767 | 80.97384 | 6.012633 | 213.3561 | rice | | | | | | | | | | | | | | | |
| 25 | 83 | 41 | 43 | 21.05254 | 82.6784 | 6.254028 | 233.1076 | rice | | | | | | | | | | | | | | | |
| 26 | 98 | 47 | 37 | 23.48381 | 81.33265 | 7.375483 | 224.0981 | rice | | | | | | | | | | | | | | | |
| 27 | 66 | 53 | 41 | 25.07564 | 80.52389 | 7.778915 | 257.0039 | rice | | | | | | | | | | | | | | | |
| 28 | 07 | 60 | 43 | 26.50377 | 84.04044 | 6.2865 | 271.3588 | rice | | | | | | | | | | | | | | | |

Fig: Input dataset

```

import IPython
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from IPython import get_ipython
import warnings
warnings.filterwarnings("ignore")

data = pd.read_csv("D:\pycharmProject\Crop Recommendation System\Dataset\Crop_recommendation.csv")

```

Fig : Dataset connector

Code Generation :

The generation of source code by using different libraries such as streamlit, Numpy, Image, pickle, requests, st_lottie. Here we connect the above dataset and write the program which display frontend of the web.

```

import streamlit as st
import numpy as np
from PIL import Image
import pickle
import requests
from streamlit_lottie import st_lottie

st.set_page_config(page_title="Crop Recommender system", layout="wide")

# --- Important Functions ---
def load_url(url):
    r = requests.get(url) # to access the animation link
    if r.status_code != 200:
        return None
    return r.json()

# Animation Assets
lottie = load_url("https://assets9.lottiefiles.com/packages/lf20_m9qplun.json")
farmer = load_url("https://assets1.lottiefiles.com/packages/lf20_fggnferrn.json")

recommender = pickle.load(open('croprecommender (2).sav', 'rb'))

def crop_output(input_data):
    input_array = np.array(input_data)
    final_input = input_array.reshape(1, 1)
    prediction = recommender.predict(final_input)
    output = prediction[0]
    return ("Field conditions are most suitable for " + output)

def main():
    with st.container():
        c1, c2 = st.columns((1, 1.5))
        with c1:
            st.title("Crop Recommendation System")
            st.write("")
            st.write("Algorithm used: K-Nearest Neighbors")

```

Fig: Source code

Web Front View :

After running source code we will get webpage as :

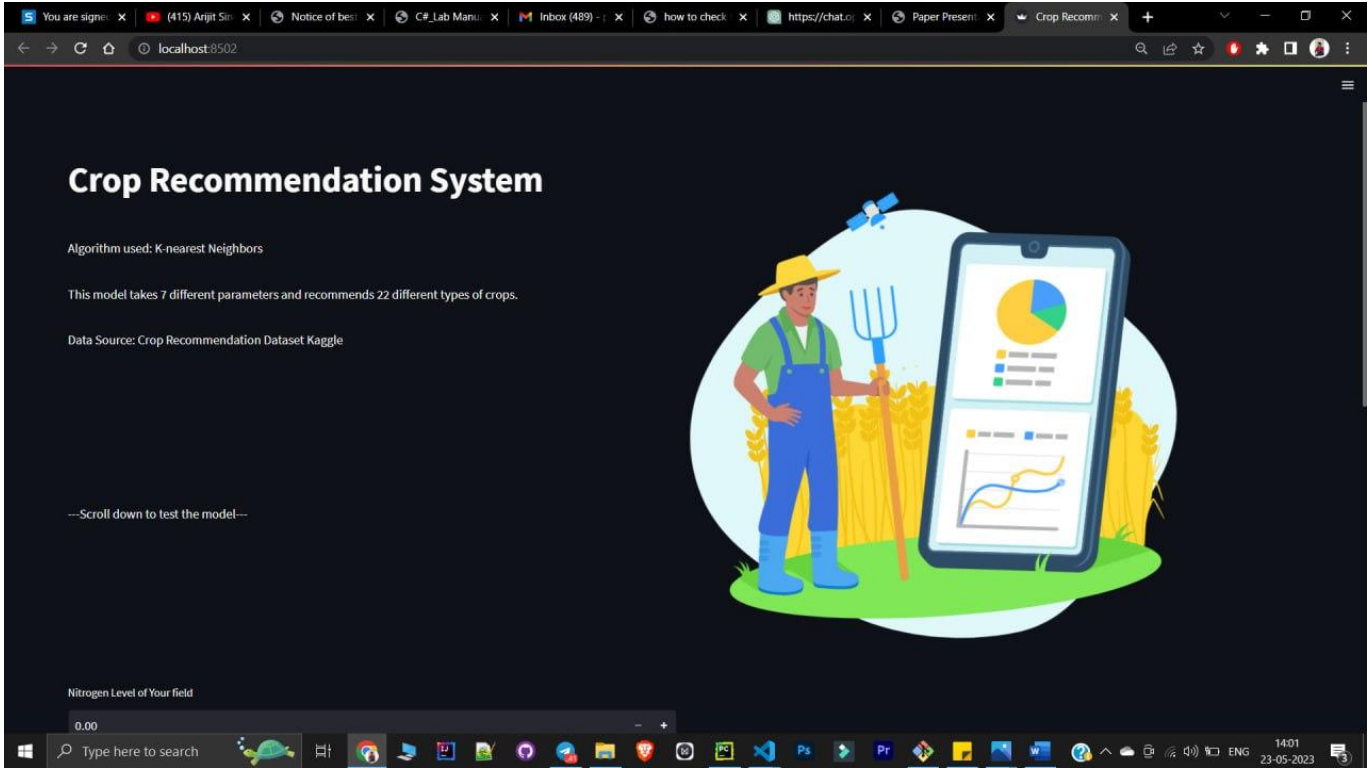


Fig : WebView

Final Output :

It shows final output that is the recommended crop .

Web FrontView :

After running source code we will get webpage as :

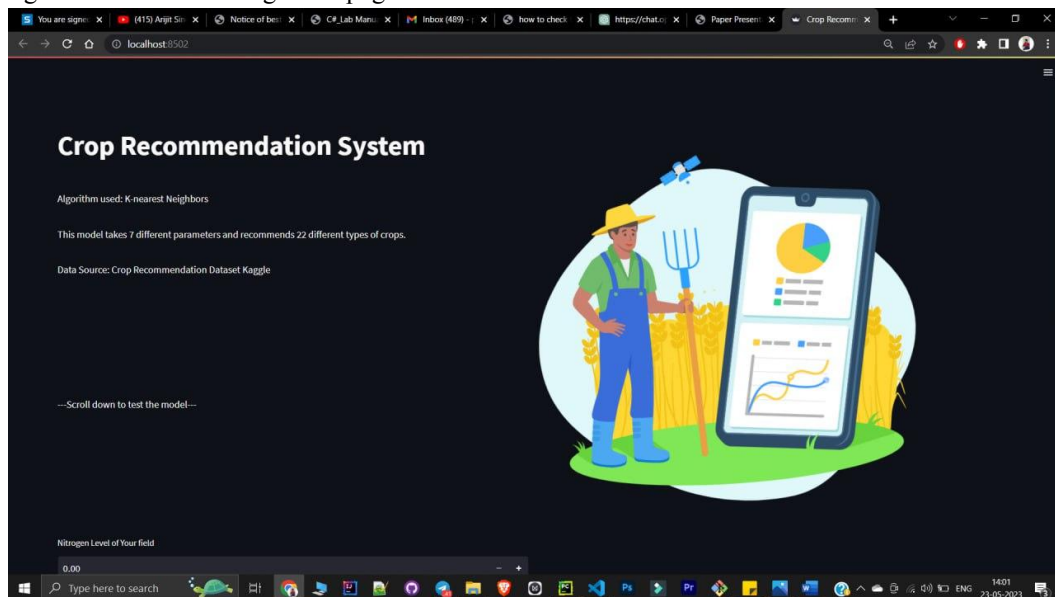


Fig : WebView

Final Output :

It shows final output that is the recommended crop .

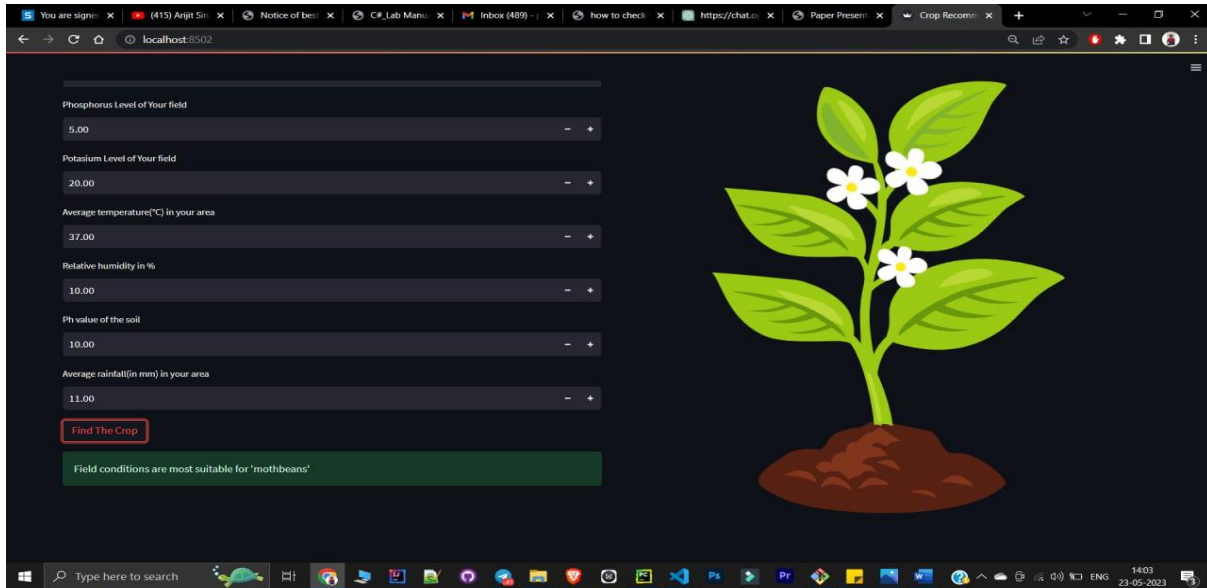


fig: Final output

V. TOOLS AND TECHNOLOGIES USED

- Python: Python is an abundance of powerful tools ready for scientific computing Packages. The packages like NumPy, Pandas and SciPy, Streamlite-lottie, Pillow. Pickle_ming.
- NUMPY: NumPy is a powerful Python library which provides a fundamental infrastructure for numerical computing. It stands for "Numerical Python" and provides efficient data structures, mathematical functions, and tools for working with large arrays and matrices. NumPy forms the foundation for many scientific and data analysis libraries in Python.
- SCIKIT-LEARN : It features numerous classification, clustering and regression algorithms like random forests, SVM and it furthermore supports Python scientific and numerical libraries like SciPy and NumPy.
- TENSORFLOW: In the TensorFlow [22] has an open source software library for numerical computation using data flow graphs.
- IDE used: Jupyter Notebook, PyCharm
- Dataset used: Kaggle Crop Dataset

VI. FUTURE SCOPE

The future scope of intelligent crop recommendation systems includes:

- Integration of IoT and sensor technologies for real-time data collection.
- Harnessing big data and advanced analytics techniques for improved recommendations.
- Utilizing remote sensing and satellite imagery for precise and proactive decision-making.
- Incorporating crop genetics and genomics to recommend resilient and high-yielding crop varieties.
- Developing user-friendly mobile applications and decision support tools for accessibility.
- Promoting collaboration and knowledge sharing among farmers and experts.

These advancements aim to enhance the accuracy, efficiency, and sustainability of crop recommendations, contributing to improved agricultural practices and global food security

VII. CONCLUSION

The proposed system framework for an intelligent crop recommendation system harnesses emerging technologies, and advanced machine learning algorithms, including ensemble methods and deep learning, enhance the accuracy and robustness of the recommendations by capturing complex patterns and relationships in the data. The proposed system framework represents a significant step forward in intelligent crop recommendation systems, offering a comprehensive and technologically advanced approach to assist farmers in making optimal crop decisions. By leveraging the power of data, analytics, and user-friendly interfaces, the system has the potential to revolutionize agriculture and drive sustainable farming practices in the future.

REFERENCES

- [1]. Shilpa Mangesh Pande, Prem Kumar Ramesh, Anmol, B R Aishwarya, Karuna Rohilla and Kumar Shaurya, Crop Recommendation using Machine Learning Approach, Proceedings of the Fifth International Conference on Computing Methodologies and Communication (ICCMC 2021).
- [2]. https://www.researchgate.net/publication/331426761_Crop_Recommendation_System_to_Maximize_Crop_Yield_in_Ramtek_region_using_Machine_Learning
- [3]. Haedong Lee and Aekyung Moon, "Development of Yield Prediction System Based on Real-time Agricultural Meteorological Information", 16th International Conference on Advanced Communication Technology, 2014
- [4]. Pudumalar, S., E. Ramanujam, R. Harine Rajashree, C. Kavya, T. Kiruthika, and J. Nisha. "Crop recommendation system for precision agriculture." In 2016 Eighth International Conference on Advanced Computing (ICoAC), pp. 32-36. IEEE, 2017.
- [5]. T.R. Lekhaa, "Efficient Crop Yield and Pesticide Prediction for Improving Agricultural Economy using Data Mining Techniques", International Journal of Modern Trends in Engineering and Science (IJMTES), 2016, Volume 03, Issue 10
- [6]. https://www.researchgate.net/publication/335799151_Crop_Recommendation_System_for_Precision_Agriculture