

## Introduction

- Focus: optimize agricultural productivity
- Align crops with regional climate conditions
- Benefits:
  - Increased crop yield
  - Improved resource utilization
  - Reduced risk of crop failure
  - Sustainable farming practices



## **Our Machine Learning Driven Solution**

Our approach combines environmental data analysis with machine learning to create a reliable crop recommendation system.

- Dataset source: We used a curated dataset from Kaggle containing agricultural data from various hypothetical regions in India
- Data quality: Each crop had 100 entries, ensuring balanced representation
- **Data refinement:** We started with 21 crops and refined to 10 unique crops by removing similar varieties

## **Data Description**

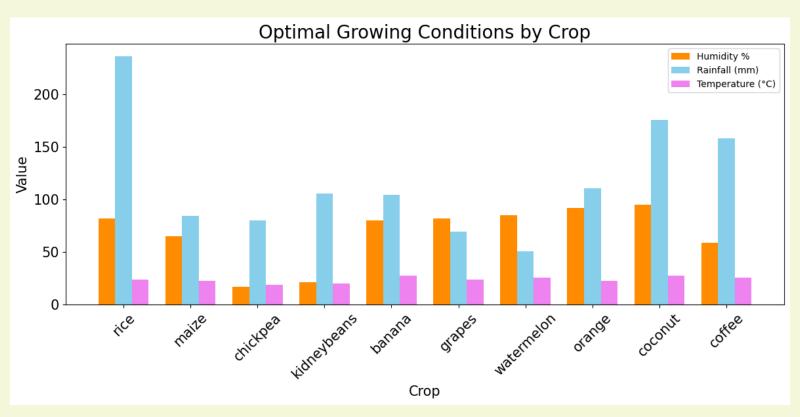
### Soil components

- Nitrogen (N)
- Phosphorus (P)
- Potassium (K)
- o pH

### **Environmental factors**

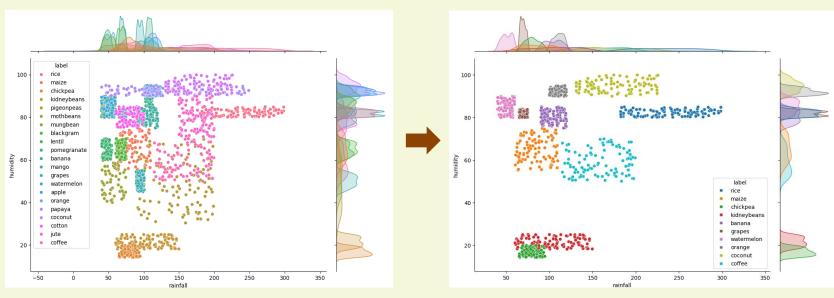
- Temperature
- Humidity
- Rainfall

## **Data Description**



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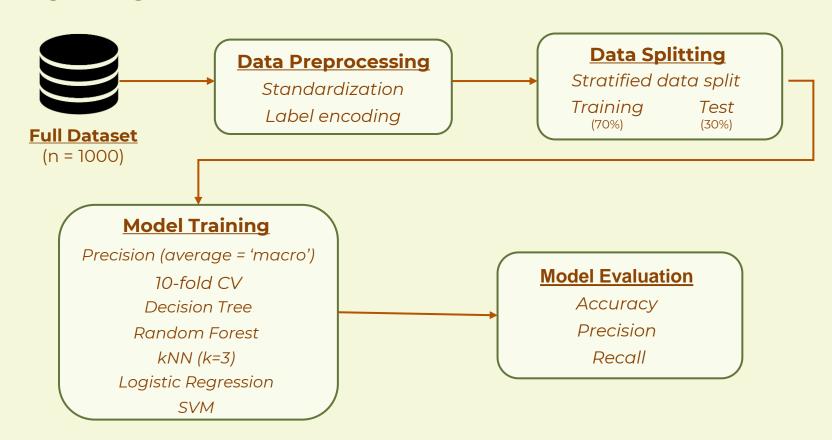
- Output labels (21 -> 10)
- Pairwise euclidean distance



Raw (label numbers = 21)

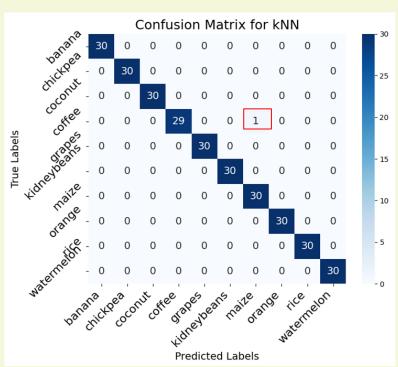
Subset (label numbers = 10)

## Workflow

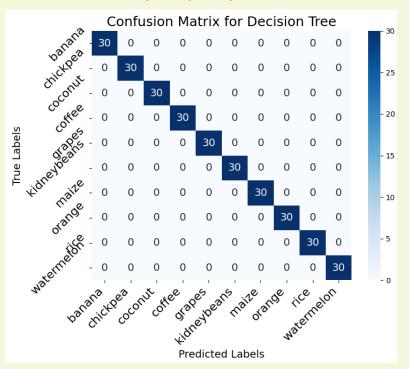


### **Confusion Matrix**

#### **kNN**



#### LR, DT, RF, SVM



# **Model Comparison Result**

Model	Accuracy	Precision	Recall
Decision Tree	1.000	1.000	1.000
Random Forest	1.000	1.000	1.000
Logistic Regression	1.000	1.000	1.000
SVM	1.000	1.000	1.000
knn	0.997	0.997	0.997

# **Crop Recommendation**

Soil nutrients and environmental factors

### Example 1 - hot and humid

Recommended Crop for Input [50, 40, 60, 29.6, 87, 6.5, 260]:

Decision Tree: rice Random Forest: rice

Logistic Regression: rice

SVM: rice

#### Example 2 - warm and moderately humid

Recommended Crop for Input [100, 80, 52, 27, 80, 6.1, 120]:

Decision Tree: banana Random Forest: banana

Logistic Regression: banana

SVM: banana





