

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

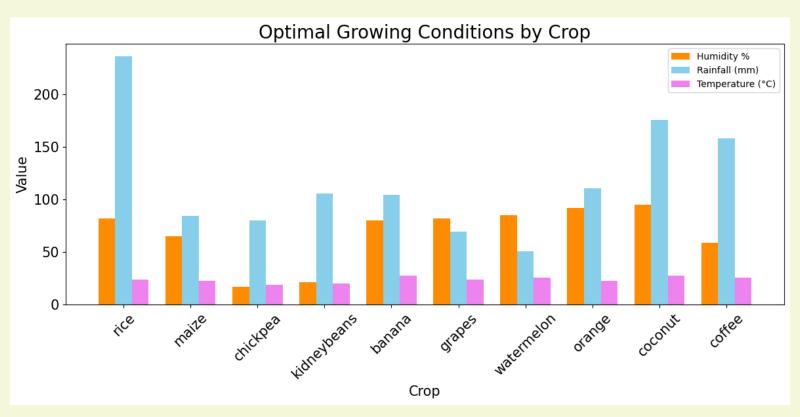
Soil components

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

Environmental factors

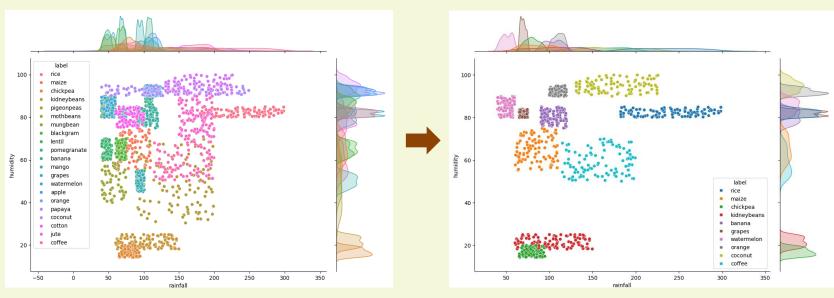
- Temperature
- Humidity
- Rainfall

Data Description



Data Description

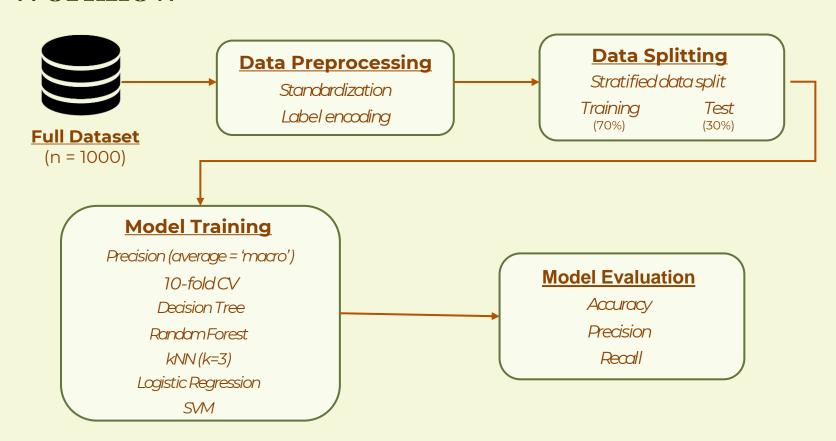
- Output labels (21 -> 10)
- Pairwise euclidean distance



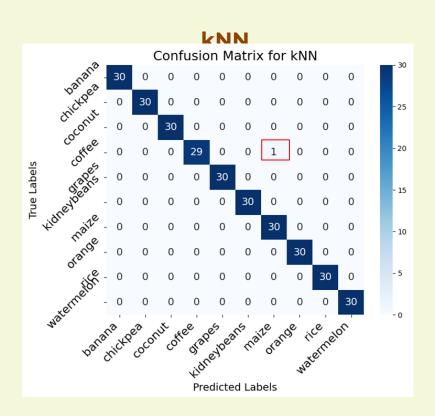
Raw (label numbers = 21)

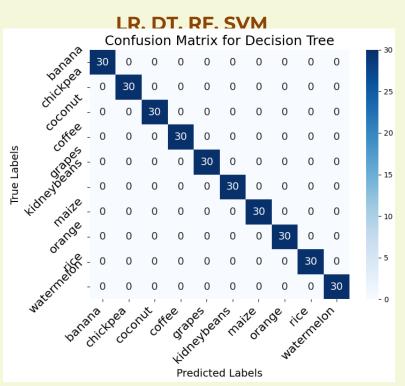
Subset (label numbers = 10)

Workflow



Confusion Matrix





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





