Crop Yield Classification Using Machine Learning

a. Title Page

Problem Statement:
To build a machine learning classification model that predicts the yield category of a crop based on factors such as soil quality, rainfall, and seed type.

Name: Harsh
Roll No.: 202401100300113
Course: BTech
Institution: KIET Group of Institution

b. Introduction

Agricultural productivity is influenced by multiple factors, and predicting crop yield can help farmers and agronomists make better decisions. In this project, we developed a classification model to predict the yield category based on three main features:
- Soil Quality
- Rainfall
- Seed Type

A dataset containing these attributes was analyzed and used to train a machine learning model to predict yield outcomes.

c. Methodology

1. Data Loading & Cleaning: Loaded the dataset using pandas. Checked for and removed any rows with missing values.
2. Feature Selection: Selected 'soil_quality', 'rainfall', and 'seed_type' as features, and 'yield_category' as the target variable.

3. Label Encoding: Used LabelEncoder to convert categorical text features into numerical format.
4. Splitting Data: Split the dataset into training and test sets using an 80-20 ratio with stratified sampling.
5. Feature Scaling: Standardized the input features using StandardScaler.
6. Model Training: Trained a Random Forest Classifier with 100 trees and a maximum depth of 10.
7. Evaluation: Evaluated the model using a confusion matrix, accuracy, precision, recall, and a classification report.
8. Feature Importance: Computed and displayed the importance of each feature using the model's built-in feature_importances_.

d. Code

```
# Import necessary libraries
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix, classification_report,
accuracy_score, precision_score, recall_score
import os

# Load your CSV file with error handling
file_path = '/crop_yield.csv'
try:
    if not os.path.exists(file_path):
        raise FileNotFoundError(f"File {file_path} not found. Please check
the file path.")
    df = pd.read_csv(file_path)
except Exception as e:
    print(f"Error loading file: {e}")
    exit(1)

# Check for missing values and drop them
missing_values = df.isnull().sum()
print("Missing values per column:\n", missing_values)
if missing_values.sum() > 0:
```

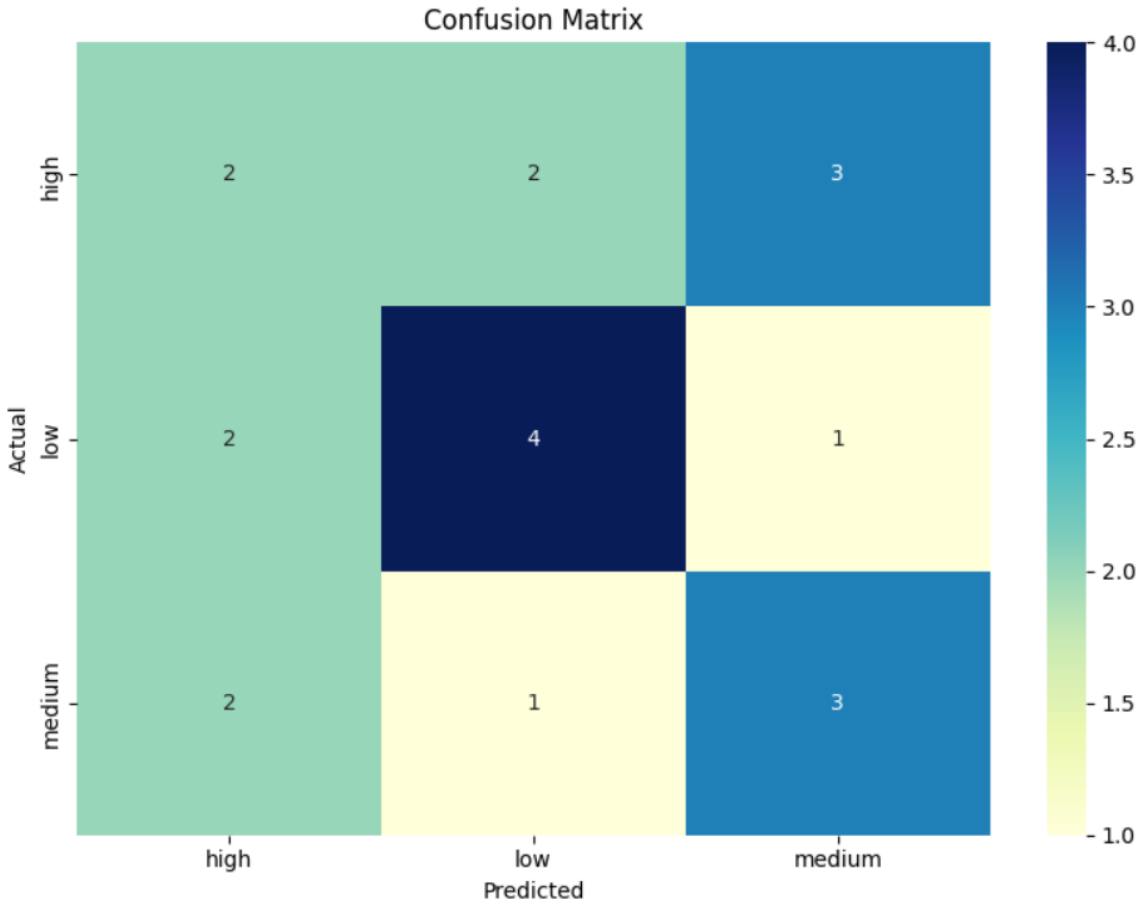
```
    print(f"Dropping {df.isnull().any(axis=1).sum()} rows with missing
values.")
    df = df.dropna()

# Verify column names
print("Columns in dataset:", df.columns.tolist())
feature_cols = ['soil_quality', 'rainfall', 'seed_type']

target_col = 'yield_category'

# Check if specified columns exist
missing_cols = [col for col in feature_cols + [target_col] if col not in
df.columns]
if missing_cols:
    print(f"Error: Columns {missing_cols} not found in dataset.")
    exit(1)

# Encode categorical columns
label_encoders = {}
for col in feature_cols + [target_col]:
    if df[col].dtype == 'object' or df[col].dtype.name == 'category':
        le = LabelEncoder()
        df[col] = le.fit_transform(df[col])
        label_encoders[col] = le
        print(f"Encoded column '{col}' with classes: {le.classes_}")

# Define features and target
X = df[feature_cols]
y = df[target_col]

# Check if target has enough classes
if len(y.unique()) < 2:
    print("Error: Target variable has fewer than 2 classes. Cannot proceed
with classification.")
    exit(1)

# Split into train/test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42, stratify=y)
print(f"Training set size: {X_train.shape[0]} samples")
print(f"Test set size: {X_test.shape[0]} samples")

# Scale the features
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
```

```
X_test = scaler.transform(X_test)

# Train the classifier with tuned hyperparameters
clf = RandomForestClassifier(n_estimators=100, max_depth=10,
random_state=42, n_jobs=-1)
try:
    clf.fit(X_train, y_train)
except Exception as e:
    print(f"Error training model: {e}")
    exit(1)

# Predictions
y_pred = clf.predict(X_test)

# Confusion matrix
cm = confusion_matrix(y_test, y_pred)

# Plot heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt='d', cmap='YlGnBu',
            xticklabels=label_encoders[target_col].classes_,
            yticklabels=label_encoders[target_col].classes_)
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.title('Confusion Matrix')
plt.tight_layout()
plt.show()

# Evaluation metrics
print("\n✅Evaluation Metrics:")
print(f"Accuracy: {accuracy_score(y_test, y_pred):.4f}")
print(f"Precision (macro): {precision_score(y_test, y_pred,
average='macro'):.4f}")
print(f"Recall (macro): {recall_score(y_test, y_pred,
average='macro'):.4f}")

# Detailed classification report
print("\n Detailed Classification Report:")
print(classification_report(y_test, y_pred,
target_names=label_encoders[target_col].classes_))

# Feature importance
feature_importance = pd.DataFrame({
    'Feature': feature_cols,
    'Importance': clf.feature_importances_
```

```
}).sort_values(by='Importance', ascending=False)
print("\n Feature Importance:")
print(feature_importance)
```

e. Output/Result

```
Missing values per column:
 soil_quality      0
rainfall           0
seed_type          0
yield_category     0
dtype: int64
Columns in dataset: ['soil_quality', 'rainfall', 'seed_type', 'yield_category']
Encoded column 'seed_type' with classes: ['A' 'B' 'C']
Encoded column 'yield_category' with classes: ['high' 'low' 'medium']
Training set size: 80 samples
Test set size: 20 samples
```



```
✅ Evaluation Metrics:
Accuracy: 0.4500
Precision (macro): 0.4444
Recall (macro): 0.4524

📋 Detailed Classification Report:
              precision    recall  f1-score   support

        high       0.33      0.29      0.31         7
         low       0.57      0.57      0.57         7
      medium       0.43      0.50      0.46         6

    accuracy                           0.45        20
   macro avg       0.44      0.45      0.45        20
weighted avg       0.45      0.45      0.45        20


🔍 Feature Importance:
        Feature  Importance
0  soil_quality    0.510236
1      rainfall    0.419603
2     seed_type    0.070161
```

f. References/Credits

Dataset Source: Provided as /crop_yield.csv

Libraries Used:

- pandas
- seaborn
- matplotlib
- sklearn (scikit-learn)

IDE/Notebook: Jupyter Notebook / VS Code (specify your environment)