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# =====
# 1. Import Libraries
# =====

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score

from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
```

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# =====
# 2. Load Dataset
# =====

df = pd.read_csv("crop_yield_dataset_50000.csv")

print(df.head())
print(df.info())
```

```
Region  Crop_Type  Rainfall_mm  Temperature_C  Humidity_%  Soil_pH \
0      West       Maize        432.52       24.47       59.29      7.11
1   Central      Maize        757.60       32.87       62.20      4.82
2      East      Maize       1567.35       37.62       40.71      5.62
3   Central     Cotton        597.73       30.68       36.53      8.23
4   Central  Sugarcane       698.53       29.32       52.22      5.45
```

```
Fertilizer_kg_per_hectare  Pesticide_kg_per_hectare  Area_hectare \
0                      132.31                  2.98          2.04
1                      159.85                  9.67          12.26
2                      142.69                  3.29          4.87
3                      184.17                  8.04          2.52
```

```
4                      89.29                  7.27      0.89
                                         Crop_Yield_ton_per_hectare
0                           19.22
1                           30.88
2                           33.47
3                           28.28
4                           23.04
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50000 entries, 0 to 49999
Data columns (total 10 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Region          50000 non-null   object  
 1   Crop_Type        50000 non-null   object  
 2   Rainfall_mm       50000 non-null   float64 
 3   Temperature_C    50000 non-null   float64 
 4   Humidity_%        50000 non-null   float64 
 5   Soil_pH          50000 non-null   float64 
 6   Fertilizer_kg_per_hectare  50000 non-null   float64 
 7   Pesticide_kg_per_hectare  50000 non-null   float64 
 8   Area_hectare     50000 non-null   float64 
 9   Crop_Yield_ton_per_hectare 50000 non-null   float64 
dtypes: float64(8), object(2)
memory usage: 3.8+ MB
None
```

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# =====
# 3. Data Preprocessing
# =====

# Check missing values
print(df.isnull().sum())

# Encode categorical variables
le = LabelEncoder()

df["Region"] = le.fit_transform(df["Region"])
df["Crop_Type"] = le.fit_transform(df["Crop_Type"])

# Feature & Target split
```

```
X = df.drop("Crop_Yield_ton_per_hectare", axis=1)
y = df["Crop_Yield_ton_per_hectare"]

# Train-Test Split
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

# Feature Scaling
scaler = StandardScaler()

X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

```
Region          0
Crop_Type       0
Rainfall_mm     0
Temperature_C   0
Humidity_%      0
Soil_pH         0
Fertilizer_kg_per_hectare 0
Pesticide_kg_per_hectare 0
Area_hectare    0
Crop_Yield_ton_per_hectare 0
dtype: int64
```

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# =====
# 4. Model Training
# =====

# Linear Regression
lr = LinearRegression()
lr.fit(X_train, y_train)

# Random Forest
rf = RandomForestRegressor(n_estimators=100, random_state=42)
rf.fit(X_train, y_train)
```

▼ RandomForestRegressor [i](#) [?](#)
RandomForestRegressor(random_state=42)

```
# =====
# 5. Model Evaluation
# =====

def evaluate(model, X_test, y_test, name):

    y_pred = model.predict(X_test)

    rmse = np.sqrt(mean_squared_error(y_test, y_pred))
    mae = mean_absolute_error(y_test, y_pred)
    r2 = r2_score(y_test, y_pred)

    print(f"--- {name} ---")
    print("RMSE:", rmse)
    print("MAE :", mae)
    print("R2  :", r2)
    print()

    return y_pred

lr_pred = evaluate(lr, X_test, y_test, "Linear Regression")
rf_pred = evaluate(rf, X_test, y_test, "Random Forest")
```

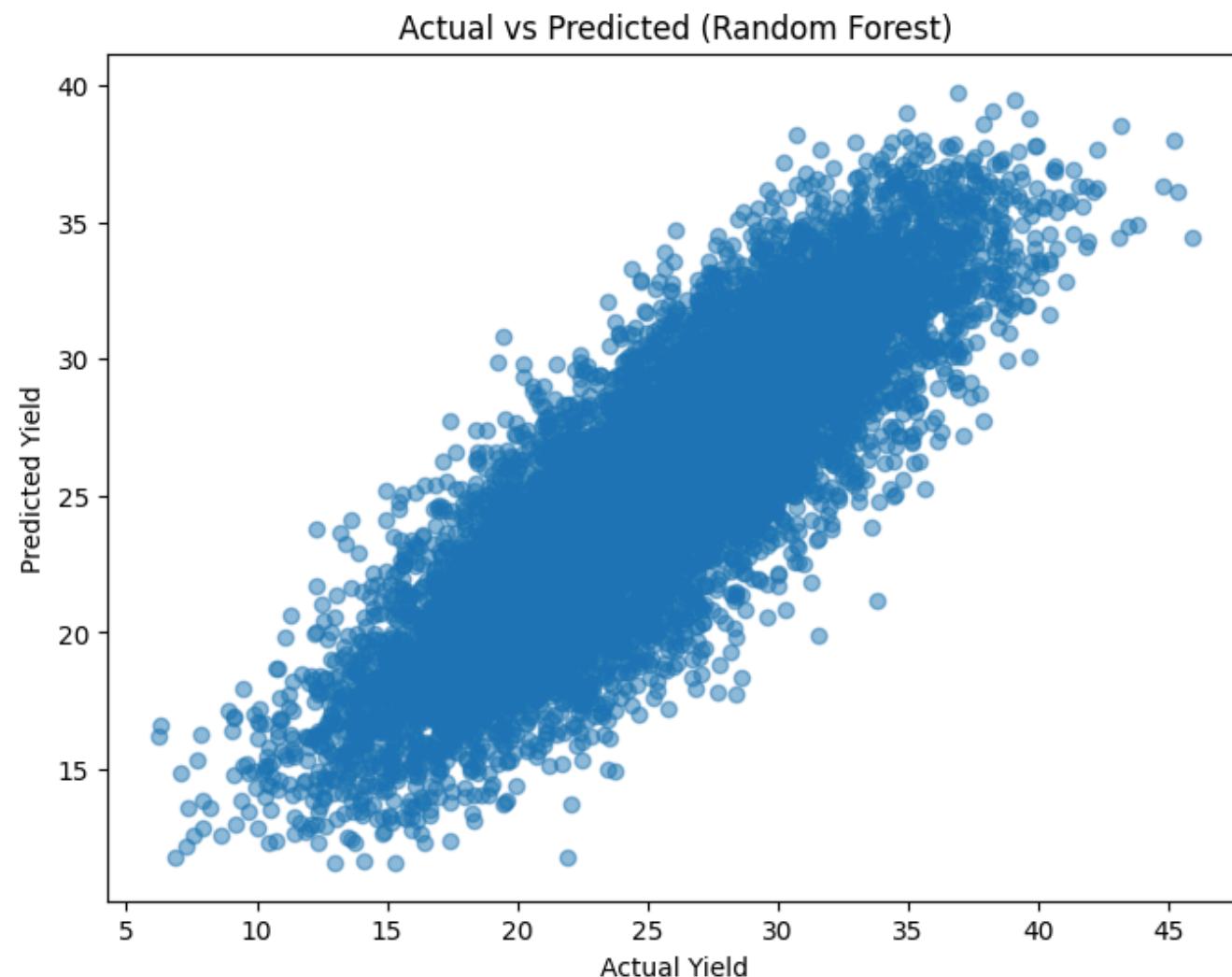
```
--- Linear Regression ---
RMSE: 2.9971372600077117
MAE : 2.378988679568162
R2  : 0.7441756874001002
```

```
--- Random Forest ---
RMSE: 3.1037788921975413
MAE : 2.46370429
```

```
R2 : 0.725646755290246
```

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# =====
# 6. Visualization
# =====

plt.figure(figsize=(8,6))
plt.scatter(y_test, rf_pred, alpha=0.5)
plt.xlabel("Actual Yield")
plt.ylabel("Predicted Yield")
plt.title("Actual vs Predicted (Random Forest)")
plt.show()
```



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# =====
# 7. Feature Importance
# =====

importance = rf.feature_importances_

features = X.columns
```

```
imp_df = pd.DataFrame({  
    "Feature": features,  
    "Importance": importance  
}).sort_values(by="Importance", ascending=False)  
  
print(imp_df)  
  
# Plot  
plt.figure(figsize=(10,6))  
sns.barplot(x="Importance", y="Feature", data=imp_df)  
plt.title("Feature Importance")  
plt.show()
```

	Feature	Importance
3	Temperature_C	0.554693
6	Fertilizer_kg_per_hectare	0.166173
2	Rainfall_mm	0.129629
5	Soil_pH	0.032482
4	Humidity_%	0.032369
8	Area_hectare	0.031640
7	Pesticide_kg_per_hectare	0.031463
1	Crop_Type	0.010856
0	Region	0.010694

Feature Importance



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# =====
# 8. Sample Prediction
# =====

sample = X.iloc[0:1]
sample = scaler.transform(sample)

predicted_yield = rf.predict(sample)
```