Classify Plants Based on Water Needs

Problem Statement:
Classify Plants Based on Water Needs — Predict plant categories based on environmental preferences and watering frequency.

Student Details:

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- **Course:** AI - Mid Semester Examination
- **Platform Used:** Google Colab

Introduction

Plants have different water requirements based on their environmental conditions. In this project, the goal is to build a machine learning model that classifies the water needs (e.g., Low, Medium, High) of a plant using input features like sunlight, soil type, temperature, humidity, and watering frequency.

Machine learning techniques are applied to train a classifier and evaluate its performance using metrics like accuracy, precision, recall, and confusion matrix.

Methodology

1. **Dataset**: A dataset named `plants.csv` was provided, containing columns such as Sunlight, Soil Type, Temperature, Humidity, Watering Frequency, etc.
2. **Preprocessing**:
    - Handled missing values (if any).
    - Encoded categorical features using Label Encoding.
    - Standardized features using `StandardScaler`.
3. **Model**: A Random Forest Classifier was trained on the dataset using an 80-20 train-test split.
4. **Evaluation**: The model was evaluated using accuracy, precision, recall, classification report, and confusion matrix (heatmap).
5. **Visualization**: The confusion matrix was visualized using Seaborn's heatmap.

Code

The code was written and executed on Google Colab:

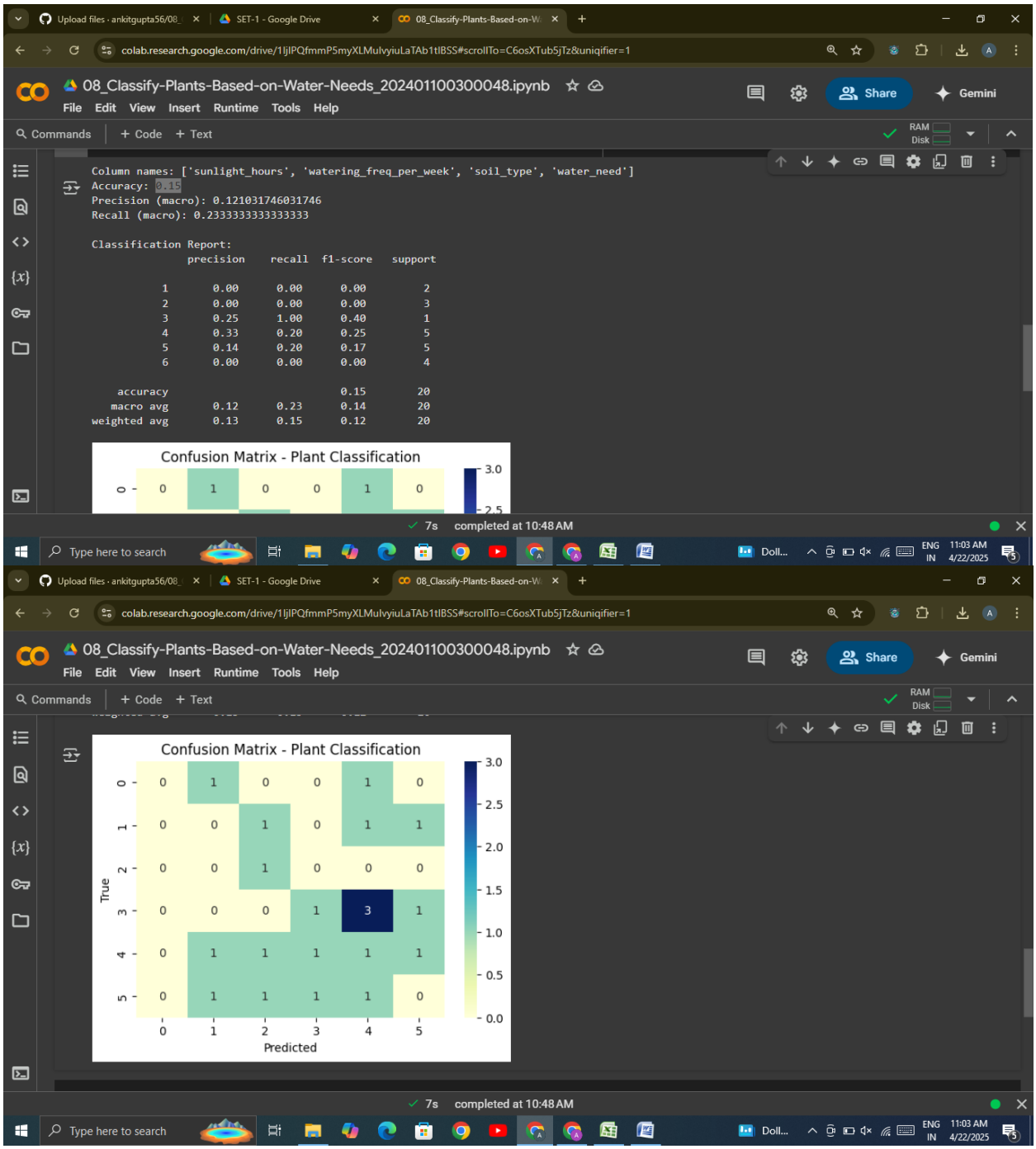
```
# Install required libraries
!pip install -q seaborn scikit-learn matplotlib pandas

# Step 1: Import 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 classification_report, confusion_matrix,
accuracy score, precision score, recall score

# Step 2: Load the uploaded dataset
df = pd.read_csv('/content/plants.csv')

# Step 3: Display column names to verify exact spelling
print("Column names:", df.columns.tolist())

# Step 4: Replace spaces in column names with underscores
# This ensures consistency and avoids issues with spaces within column names
df.columns = df.columns.str.replace(' ', '_')

# Step 5: Encode categorical columns
le = LabelEncoder()
for col in df.select_dtypes(include=['object']).columns:
    df[col] = le.fit transform(df[col])

# Step 6: Define target column correctly after replacing spaces with underscores
# Ensure this matches the exact column name after replacing spaces
target_column = 'watering_freq_per_week'  # Change this to the correct column if
needed - it looks like 'watering frequency' in the data

# Step 7: Define X and y
X = df.drop(target_column, axis=1)
y = df[target_column]

# Step 8: Train-test split and scaling
X train, X test, y train, y test = train test split(X, y, test size=0.2,
random_state=42)

scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
```

```
X_test_scaled = scaler.transform(X_test)

# Step 9: Train classifier
clf = RandomForestClassifier(random_state=42)
clf.fit(X_train_scaled, y_train)
y_pred = clf.predict(X_test_scaled)

# Step 10: Evaluate
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Precision (macro):", precision_score(y_test, y_pred, average='macro',
zero_division=0))
print("Recall (macro):", recall_score(y_test, y_pred, average='macro',
zero_division=0))
print("\nClassification Report:\n", classification_report(y_test, y_pred,
zero_division=0))

# Step 11: Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(6, 4))
sns.heatmap(cm, annot=True, fmt='d', cmap='YlGnBu')
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix - Plant Classification')
plt.show()
```

Output/Result

- **Model Accuracy**: *0.15*
- **Precision**: *0.121031746031746*
- **Recall**: *0.23333333333333*

- **Screenshot** of output:



References / Credits

- Dataset provided for academic purposes.
- Libraries used: `pandas`, `scikit-learn`, `matplotlib`, `seaborn`
- Code executed using Google Colab.
- RandomForestClassifier from `sklearn.ensemble`.