

**K.I.E.T. Group of Institutions**

**Ghaziabad**



**Project title-Vegetable Classification Based on Nutritional Content**

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## Introduction

This project involves classifying vegetables into categories such as **leafy**, **root**, and **fruit** based on their nutritional attributes including Vitamin A, Vitamin C, and fiber content. Using machine learning classification algorithms, particularly Random Forest, the project aims to evaluate how well a model can distinguish between vegetable types based on nutrition.

## **Methodology**

1. **Dataset:** The data contains nutritional information for various vegetables. The main features used were:

- Vitamin a
- Vitamin c
- fiber
- Target: type (root, leafy, fruit)

2. **Preprocessing:** The data was split into training and testing sets using an 80-20 split.

3. **Model:** A RandomForestClassifier from Scikit-learn was used to train the model.

4. **Evaluation:** The performance was measured using:

- Confusion matrix (visualized as a heatmap)
- Accuracy
- Precision
- Recall
- Classification report

## Code:

```
import pandas as pd

import numpy as np

from sklearn.model_selection import train_test_split

from sklearn.ensemble import RandomForestClassifier

from sklearn.metrics import confusion_matrix, classification_report, accuracy_score, precision_score,
recall_score

import seaborn as sns

import matplotlib.pyplot as plt


# Step 1: Load Dataset

df = pd.read_csv('vegetables.csv')


# Step 2: Preprocess Data

X = df.drop('type', axis=1)

y = df['type']


# Train-test split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)


# Step 3: Train the Model

model = RandomForestClassifier(random_state=42)

model.fit(X_train, y_train)


# Step 4: Predictions

y_pred = model.predict(X_test)


# Step 5: Evaluation

# Confusion Matrix

cm = confusion_matrix(y_test, y_pred)

labels = model.classes_


# Heatmap
```

```
plt.figure(figsize=(8,6))

sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=labels, yticklabels=labels)

plt.xlabel('Predicted')

plt.ylabel('Actual')

plt.title('Confusion Matrix')

plt.show()
```

```
# Metrics
```

```
accuracy = accuracy_score(y_test, y_pred)

precision = precision_score(y_test, y_pred, average='macro') # macro for multiclass

recall = recall_score(y_test, y_pred, average='macro')
```

```
print("Accuracy:", accuracy)

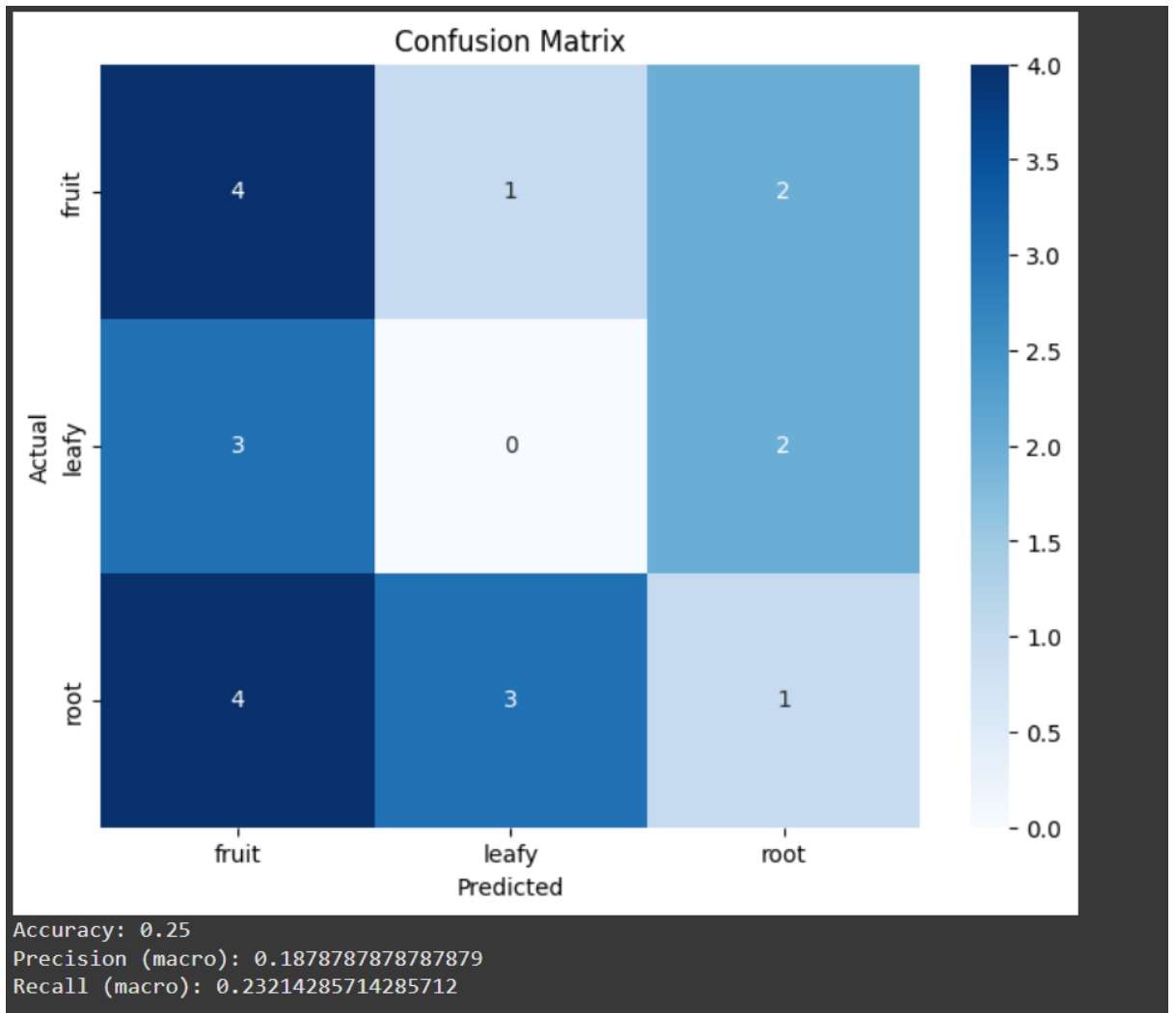
print("Precision (macro):", precision)

print("Recall (macro):", recall)
```

```
# Detailed report
```

```
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

# Output



Classification Report:				
	precision	recall	f1-score	support
fruit	0.36	0.57	0.44	7
leafy	0.00	0.00	0.00	5
root	0.20	0.12	0.15	8
accuracy			0.25	20
macro avg	0.19	0.23	0.20	20
weighted avg	0.21	0.25	0.22	20

## References

- **Dataset:** Provided.
- **Libraries:** Scikit-learn, Pandas, Matplotlib, Seaborn
- **Tools:** Google collab