## **Import Libraries**

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
from sklearn.datasets import load_iris
```

## Load the Dataset

```
In [3]: # Load the dataset
    iris_data_path = "IRIS.csv"
    iris_data = pd.read_csv(iris_data_path )
```

0.2 Iris-setosa

## **Explore and Preprocess the Data**

4.7

3.2

```
In [6]:
         # Display basic information about the dataset
         print(iris data.info())
         # Display the first few rows of the dataset
         print(iris data.head())
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 5 columns):
                          Non-Null Count Dtype
             Column
            sepal length 150 non-null float64
            sepal width 150 non-null float64
             petal length 150 non-null float64
             petal width 150 non-null
                                        float64
             species
                          150 non-null
                                          object
        dtypes: float64(4), object(1)
        memory usage: 6.0+ KB
        None
           sepal length sepal width petal length petal width
                                                                   species
                                              1.4
        0
                    5.1
                                3.5
                                                          0.2 Iris-setosa
        1
                    4.9
                                3.0
                                              1.4
                                                          0.2 Iris-setosa
```

1.3

```
3
                    4.6
                                 3.1
                                              1.5
                                                           0.2 Iris-setosa
                     5.0
                                 3.6
                                               1.4
                                                           0.2 Iris-setosa
In [10]:
          # Display the column names in the dataset
          print(iris_data.columns)
         Index(['sepal length', 'sepal width', 'petal length', 'petal width',
                'species'],
               dtype='object')
In [13]:
          # Separate features (X) and target variable (y)
          X = iris data.drop('species', axis=1)
          y = iris data['species']
        Build and Train the Model
In [15]:
          # Split the data into training and testing sets
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
In [16]:
          # Initialize and train the Random Forest Classifier
          model = RandomForestClassifier(n estimators=100, random state=42)
          model.fit(X_train, y_train)
Out[16]:
                  RandomForestClassifier
         RandomForestClassifier(random state=42)
        Make Predictions and Evaluate the Model
```

```
In [17]: # Make predictions on the test set
y_pred = model.predict(X_test)

# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy:.2f}")

# Display classification report
```

```
print("Classification Report:")
print(classification_report(y_test, y_pred))
Accuracy: 1.00
Classification Report:
                             recall f1-score
                 precision
                                                support
    Iris-setosa
                     1.00
                               1.00
                                          1.00
                                                      10
Iris-versicolor
                     1.00
                               1.00
                                          1.00
                                                       9
Iris-virginica
                     1.00
                               1.00
                                          1.00
                                                      11
       accuracy
                                          1.00
                                                      30
                     1.00
                                1.00
                                          1.00
                                                      30
      macro avg
                               1.00
                                                      30
   weighted avg
                     1.00
                                          1.00
```

## **Feature Importance**

```
In [19]: # Display feature importance (if applicable to the model)
   if hasattr(model, 'feature_importances_'):
        feature_importance = pd.DataFrame({'Feature': X.columns, 'Importance': model.feature_importances_})
        feature_importance = feature_importance.sort_values(by='Importance', ascending=False)
        print("Feature Importance:")
        print(feature_importance)
Feature Importance:
```

Feature Importance
2 petal\_length 0.439994
3 petal\_width 0.421522
0 sepal\_length 0.108098
1 sepal width 0.030387