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# @title Import Required 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
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
from sklearn.ensemble import RandomForestClassifier

# @title Load the Dataset
df = pd.read_csv("IRIS.csv")
df.head()

{"summary": {"\n    \"name\": \"df\", \n    \"rows\": 150, \n    \"fields\": [\n        {\n            \"column\": \"sepal_length\", \n            \"properties\": {\n                \"dtype\": \"number\", \n                \"std\": 0.8280661279778629, \n                \"min\": 4.3, \n                \"max\": 7.9, \n                \"num_unique_values\": 35, \n                \"samples\": [\n                    6.2, \n                    4.5, \n                    5.6\n                ], \n                \"semantic_type\": \"\", \n                \"description\": \"\\n            \"}, \n                {\n                    \"column\": \"sepal_width\", \n                    \"properties\": {\n                        \"dtype\": \"number\", \n                        \"std\": 0.4335943113621737, \n                        \"min\": 2.0, \n                        \"max\": 4.4, \n                        \"num_unique_values\": 23, \n                        \"samples\": [\n                            2.3, \n                            4.0, \n                            3.5\n                        ], \n                        \"semantic_type\": \"\", \n                        \"description\": \"\\n            \"}, \n                    {\n                        \"column\": \"petal_length\", \n                        \"properties\": {\n                            \"dtype\": \"number\", \n                            \"std\": 1.7644204199522617, \n                            \"min\": 1.0, \n                            \"max\": 6.9, \n                            \"num_unique_values\": 43, \n                            \"samples\": [\n                                6.7, \n                                3.8, \n                                3.7\n                            ], \n                            \"semantic_type\": \"\", \n                            \"description\": \"\\n            \"}, \n                        {\n                            \"column\": \"petal_width\", \n                            \"properties\": {\n                                \"dtype\": \"number\", \n                                \"std\": 0.7631607417008414, \n                                \"min\": 0.1, \n                                \"max\": 2.5, \n                                \"num_unique_values\": 22, \n                                \"samples\": [\n                                    1.2, \n                                    0.2, \n                                    1.3\n                                ], \n                                \"semantic_type\": \"\", \n                                \"description\": \"\\n            \"}, \n                            {\n                                \"column\": \"species\", \n                                \"properties\": {\n                                    \"dtype\": \"category\", \n                                    \"num_unique_values\": 3, \n                                    \"samples\": [\n                                        \"Iris-setosa\", \n                                        \"Iris-versicolor\", \n                                        \"Iris-virginica\"\n                                    ], \n                                    \"semantic_type\": \"\", \n                                    \"description\": \"\\n            \"}, \n                                \"type\": \"dataframe\", \n                                \"variable_name\": \"df\"\n                            }\n                        }\n                    }\n                }\n            }\n        }\n    ]\n}, \n    \"type\": \"dataframe\", \n    \"variable_name\": \"df\"\n}

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# @title Understand the Dataset
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   sepal_length    150 non-null   float64
 1   sepal_width     150 non-null   float64
 2   petal_length    150 non-null   float64
 3   petal_width     150 non-null   float64
 4   species        150 non-null   object  
dtypes: float64(4), object(1)
memory usage: 6.0+ KB

df.isnull().sum()

sepal_length    0
sepal_width     0
petal_length    0
petal_width     0
species         0
dtype: int64

df.describe()

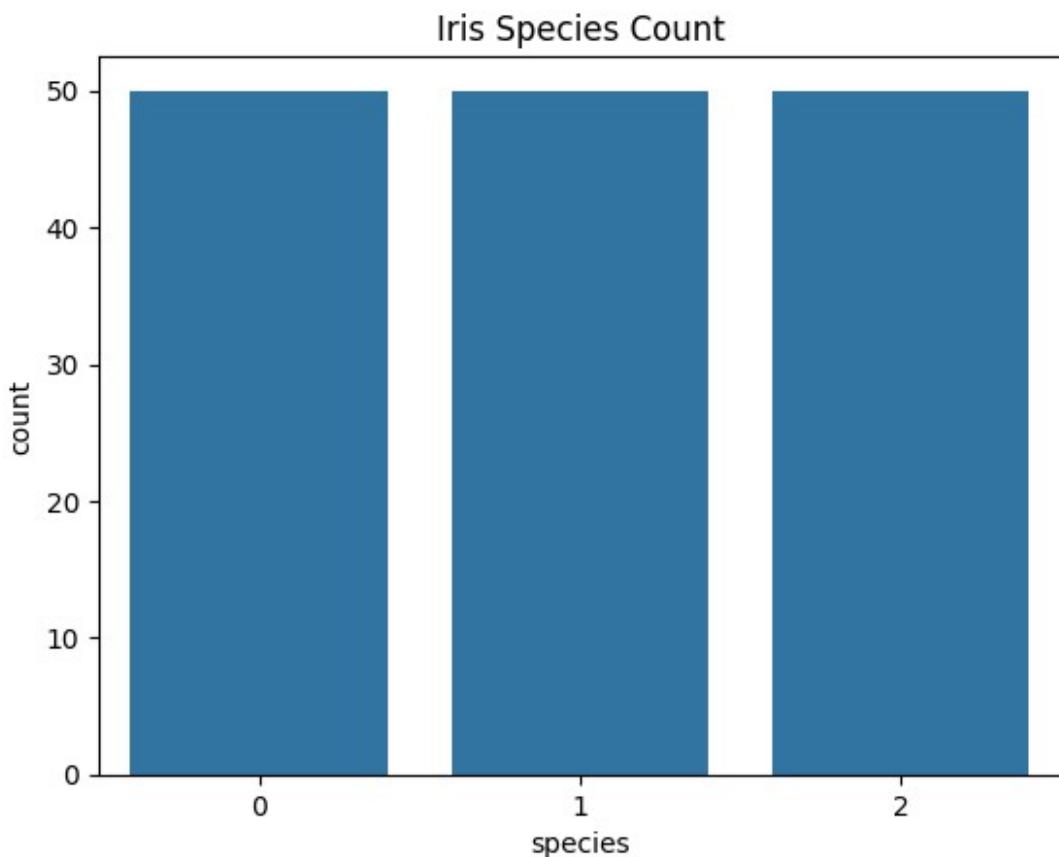
{"summary": "{\n  \"name\": \"df\", \n  \"rows\": 150, \n  \"fields\": [\n    {\n      \"column\": \"sepal_length\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 51.24711349471842, \n        \"min\": 0.8280661279778629, \n        \"max\": 150.0, \n        \"num_unique_values\": 8, \n        \"samples\": [5.8, 150.0]\n      }, \n      \"semantic_type\": \"\", \n      \"description\": \"\"\n    }, \n    {\n      \"column\": \"sepal_width\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 52.08647211421483, \n        \"min\": 0.4335943113621737, \n        \"max\": 150.0, \n        \"num_unique_values\": 8, \n        \"samples\": [3.0, 3.0]\n      }, \n      \"semantic_type\": \"\", \n      \"description\": \"\"\n    }, \n    {\n      \"column\": \"petal_length\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 51.835227940958106, \n        \"min\": 1.0, \n        \"max\": 150.0, \n        \"num_unique_values\": 8, \n        \"samples\": [4.35, 4.35]\n      }, \n      \"semantic_type\": \"\", \n      \"description\": \"\"\n    }, \n    {\n      \"column\": \"petal_width\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 52.63663424340991, \n        \"min\": 0.1, \n        \"max\": 150.0, \n        \"num_unique_values\": 8\n      }\n    }\n  ]\n}
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    \\"samples\": [\n      1.1986666666666668,\n      1.3,\n      150.0\n    ],\n    \\"semantic_type\\": \"\",\n    \\"description\\": \"\"\n  },\n  {\n    \\"column\\":\n      \\"species\\",\n      \\"properties\\": {\n        \\"dtype\\": \"number\",\n        \\"std\\": 52.69404575122032,\n        \\"min\\": 0.0,\n        \\"max\\": 150.0,\n        \\"num_unique_values\\": 5,\n        \\"samples\\": [\n          1.0,\n          2.0,\n          0.8192319205190405\n        ],\n        \\"semantic_type\\": \"\"\n      }\n    }\n  ]\n},\n\"type\\": \"dataframe\"}

# @title Exploratory Data Analysis (EDA)
sns.countplot(x='species', data=df)
plt.title("Iris Species Count")
plt.show()

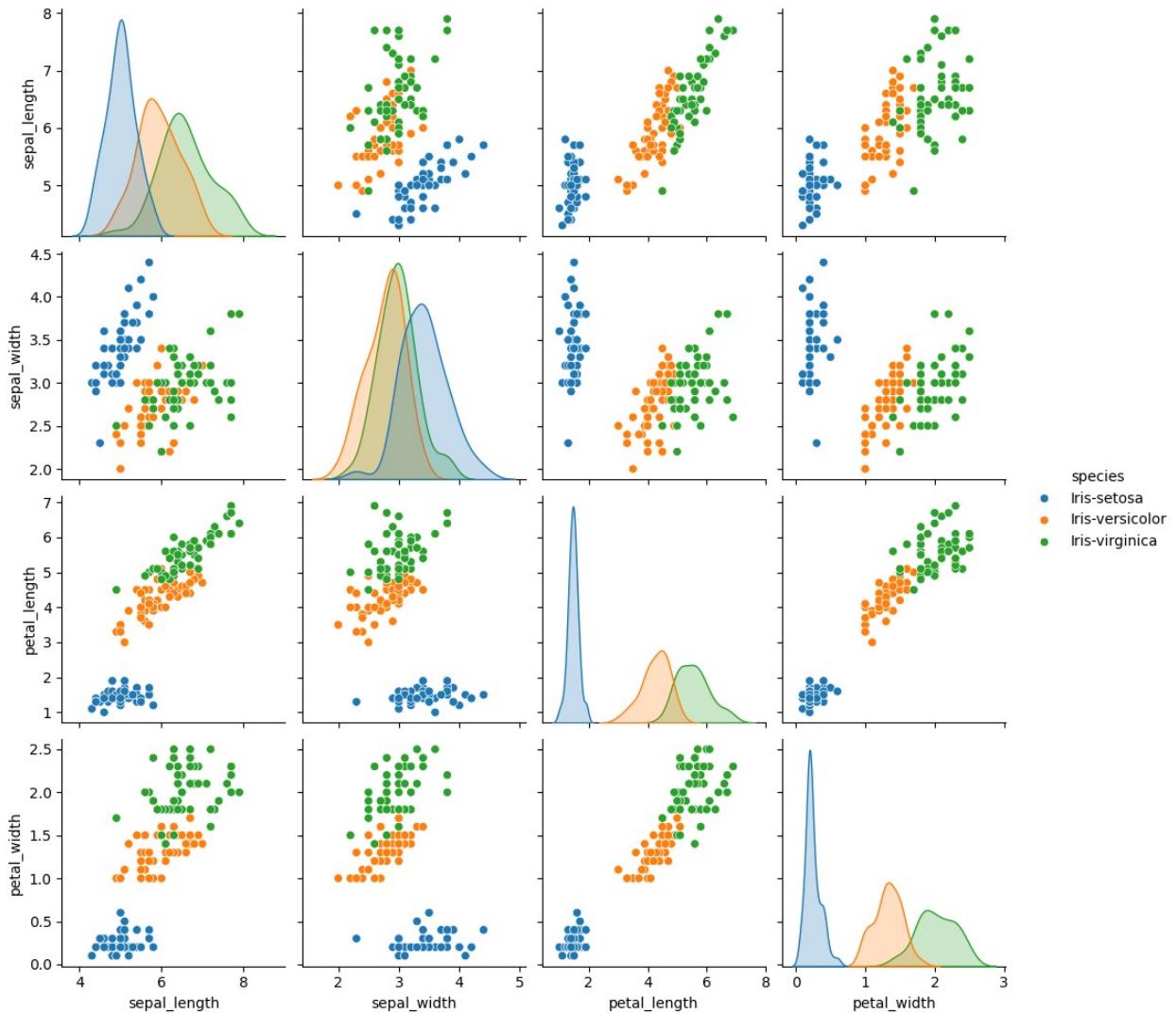
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sns.pairplot(df, hue='species')
plt.show()

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# @title Encode Target Variable
le = LabelEncoder()
df['species'] = le.fit_transform(df['species'])

# @title Feature Selection
X = df.drop(columns=['species'])
y = df['species']

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

# @title Build the Model
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)

RandomForestClassifier(random_state=42)

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# @title Make Predictions
y_pred = model.predict(X_test)

# @title Model Evaluation
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)

Accuracy: 1.0

print(classification_report(y_test, y_pred))

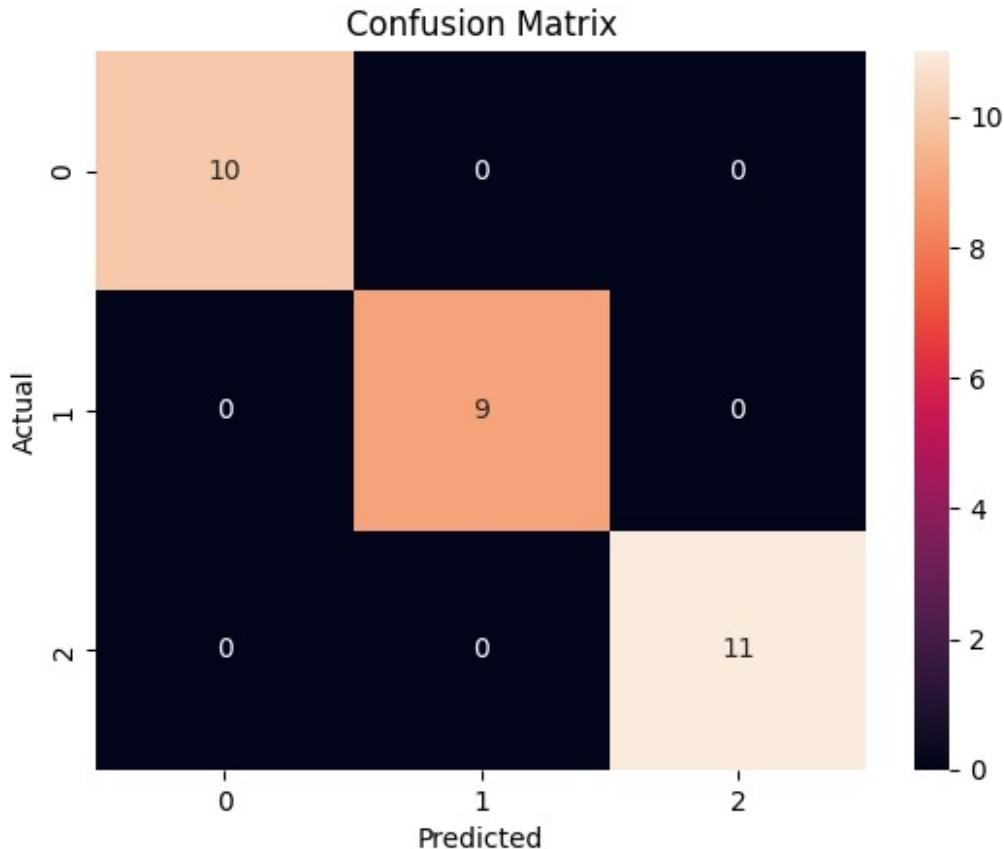
      precision    recall  f1-score   support

          0       1.00     1.00     1.00      10
          1       1.00     1.00     1.00       9
          2       1.00     1.00     1.00      11

   accuracy                           1.00      30
  macro avg       1.00     1.00     1.00      30
weighted avg       1.00     1.00     1.00      30

# @title Model Evaluation
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt='d')
plt.title("Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()

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# @title Feature Importance
feature_importance = pd.DataFrame({
    'Feature': X.columns,
    'Importance': model.feature_importances_
}).sort_values(by='Importance', ascending=False)

feature_importance

{"summary": {"name": "feature_importance", "rows": 4, "fields": [{"column": "Feature", "properties": {"dtype": "string", "num_unique_values": 4, "samples": [{"petal_width": 1, "sepal_width": 5, "semantic_type": "\\", "description": "\n"}, {"petal_length": 2, "semantic_type": "\\", "description": "\n"}, {"Importance": 0.030386812473242528, "min": 0.030386812473242528, "max": 0.43999397414456937, "std": 0.2112530882478962, "number": 0.4215215887397244, "samples": [{"petal_width": 0.4215215887397244, "sepal_width": 1, "semantic_type": "\\", "description": "\n"}, {"Importance": 0.43999397414456937, "min": 0.030386812473242528, "max": 0.43999397414456937}], "column": "Importance", "properties": {"dtype": "number", "std": 0.2112530882478962, "min": 0.030386812473242528, "max": 0.43999397414456937, "num_unique_values": 4, "samples": [{"petal_width": 0.4215215887397244, "sepal_width": 1, "semantic_type": "\\", "description": "\n"}, {"Importance": 0.43999397414456937}], "semantic_type": "\\", "description": "\n"}], "type": "dataframe", "variable_name": "feature_importance"}}
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# @title Conclusion
print("Iris Flower Classification model built successfully.")
print("Petal length and petal width are the most important features.")

Iris Flower Classification model built successfully.
Petal length and petal width are the most important features.
```