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import matplotlib.pyplot as plt
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier, export_graphviz
from sklearn.metrics import accuracy_score
from IPython.display import Image, display
import pydotplus
# Load the Iris dataset
iris = load_iris()
X = iris.data
y = iris.target
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
# Create and train a Decision Tree classifier
clf = DecisionTreeClassifier(random_state=42)
clf.fit(X_train, y_train)
# Make predictions
y_pred = clf.predict(X_test)
# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy * 100:.2f}%')
# Visualize the Decision Tree
def visualize_tree(tree, feature_names):
```

display(visualize_tree(clf, iris.feature_names))

Explanation:

- 1. **Import Libraries**: We import the necessary libraries including matplotlib for plotting, sklearn for machine learning tasks, and pydotplus and graphviz for visualizing the decision tree.
- 2. **Load Dataset**: The Iris dataset is loaded from **sklearn**.
- 3. **Split Dataset**: The dataset is split into training and testing sets using train_test_split.
- 4. **Train Classifier**: A **DecisionTreeClassifier** is created and trained on the training data.
- 5. **Make Predictions**: The trained model is used to make predictions on the test data.
- 6. **Evaluate Model**: The accuracy of the model is calculated using accuracy_score.
- 7. **Visualize Tree**: The decision tree is visualized using **export_graphviz** to create a DOT file, which is then converted to a PNG image using **pydotplus**. The image is displayed using IPython's **display**.