```
In [12]:
```

```
from sklearn.datasets import load_iris
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
```

In [2]:

```
# Load the Iris dataset
iris = load_iris()
```

In [5]:

```
iris.data.shape
```

Out[5]:

(150, 4)

In [6]:

```
iris.target.shape
```

Out[6]:

(150,)

In [3]:

```
# Split the data into features and target
X = iris.data
y = iris.target
```

In [7]:

```
# Standardize the features
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
```

In [8]:

```
# Perform PCA with two components
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)
```

In [11]:

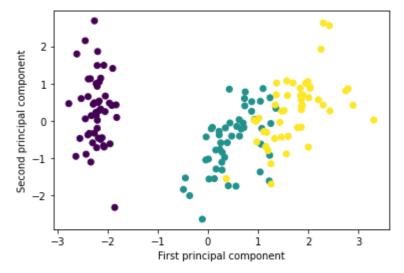
```
X_pca.shape
```

Out[11]:

(150, 2)

In [9]:

```
# Create a scatter plot of the first two principal components
plt.scatter(X_pca[:, 0], X_pca[:, 1], c=y)
plt.xlabel('First principal component')
plt.ylabel('Second principal component')
plt.show()
```



Before PCA

```
In [16]:
```

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.33, random_

In [20]:

y_train.shape

Out[20]:

(100,)

In [14]:

from sklearn.linear_model import LinearRegression

In [15]:

```
lr = LinearRegression()
```

In [21]:

```
lr.fit(X_train, y_train)
```

Out[21]:

LinearRegression()

```
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                                                 PCA - Jupyter Notebook
  In [22]:
 y_pred = lr.predict(X_test)
  In [26]:
  from sklearn.metrics import mean_squared_error, r2_score, accuracy_score
  In [28]:
  # Calculate the mean squared error and R-squared value
 mse = mean_squared_error(y_test, y_pred)
  rmse = mean_squared_error(y_test, y_pred, squared=False)
  r2 = r2_score(y_test, y_pred)
  In [25]:
  # Print the results
  print("Mean squared error: {:.2f}".format(mse))
  print("Root mean squared error: {:.2f}".format(rmse))
  print("R-squared value: {:.2f}".format(r2))
 Mean squared error: 0.04
  Root mean squared error: 0.21
  R-squared value: 0.94
  After PCA
  In [29]:
 X_train_2, X_test_2, y_train_2, y_test_2 = train_test_split(X_pca, y, test_size=0.33, ra
  In [31]:
 X_train_2.shape
  Out[31]:
  (100, 2)
  In [32]:
  lr.fit(X_train_2, y_train_2)
  Out[32]:
  LinearRegression()
```

y_pred_2 = lr.predict(X_test_2)

In [33]:

In [34]:

```
mse = mean_squared_error(y_test_2, y_pred_2)
rmse = mean_squared_error(y_test_2, y_pred_2, squared=False)
r2 = r2_score(y_test_2, y_pred_2)
```

In [35]:

```
print("Mean squared error: {:.2f}".format(mse))
print("Root mean squared error: {:.2f}".format(rmse))
print("R-squared value: {:.2f}".format(r2))
```

Mean squared error: 0.07 Root mean squared error: 0.26

R-squared value: 0.90

In []: