

🌸 Dimensionality Reduction using PCA on Iris Dataset

Step 1: Import required libraries

```
from sklearn import datasets
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
import seaborn as sns
import matplotlib.pyplot as plt
```

Step 2: Load the Iris dataset

```
iris = datasets.load_iris()
df = pd.DataFrame(iris['data'], columns = iris['feature_names'])
print("Original Iris Dataset (first 5 rows):")
print(df.head())
```

Step 3: Standardize the data

```
scalar = StandardScaler()
scaled_data = pd.DataFrame(scalar.fit_transform(df), columns=df.columns)
print("\nScaled Data (first 5 rows):")
print(scaled_data.head())
```

Step 4: Correlation heatmap before PCA

```
plt.figure(figsize=(6,4))
sns.heatmap(scaled_data.corr(), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap (Before PCA)")
plt.show()
```

Step 5: Apply PCA to reduce dimensions to 3 components

```
pca = PCA(n_components=3)
pca.fit(scaled_data)
```

```
data_pca = pca.transform(scaled_data)
data_pca = pd.DataFrame(data_pca, columns=['PC1','PC2','PC3'])
print("\nData after PCA (first 5 rows):")
print(data_pca.head())
```

Step 6: Correlation heatmap after PCA

```
plt.figure(figsize=(6,4))
sns.heatmap(data_pca.corr(), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap (After PCA)")
plt.show()
```

```

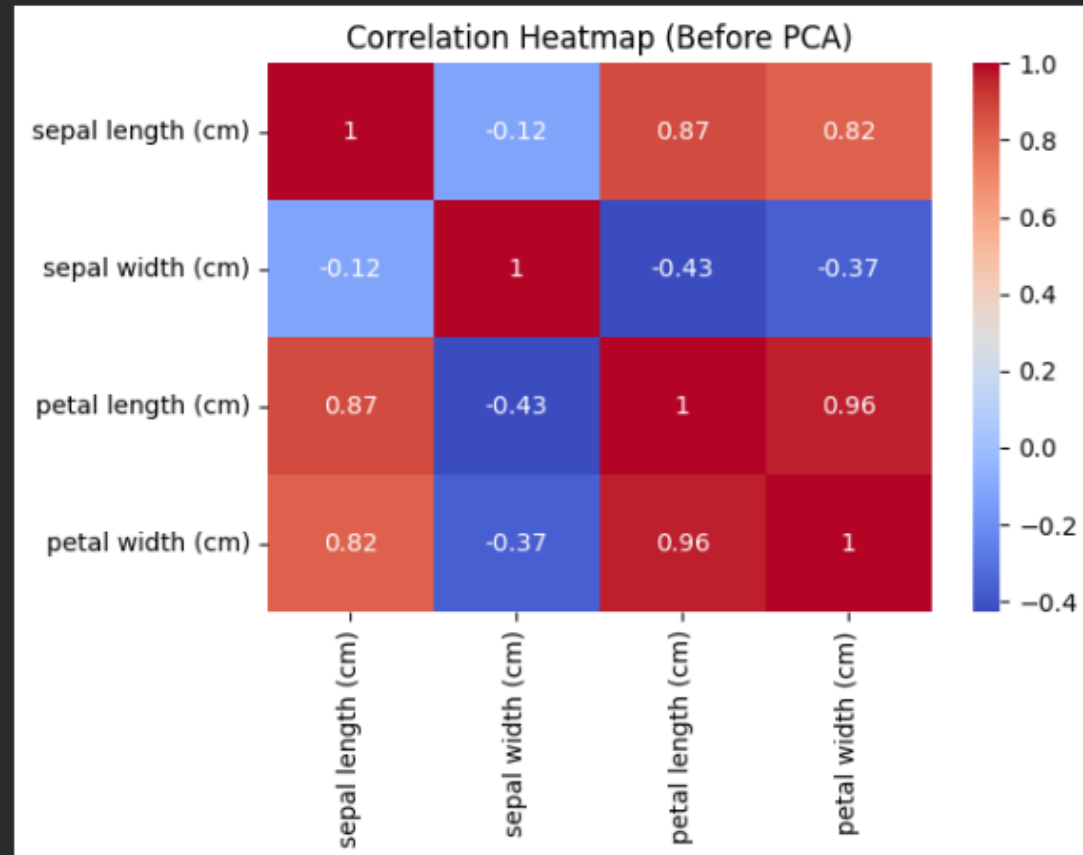
... Original Iris Dataset (first 5 rows):
      sepal length (cm)  sepal width (cm)  petal length (cm)  petal width (cm)
0          5.1           3.5           1.4           0.2
1          4.9           3.0           1.4           0.2
2          4.7           3.2           1.3           0.2
3          4.6           3.1           1.5           0.2
4          5.0           3.6           1.4           0.2

```

```

Scaled Data (first 5 rows):
      sepal length (cm)  sepal width (cm)  petal length (cm)  petal width (cm)
0      -0.900681       1.019004       -1.340227       -1.315444
1      -1.143017       -0.131979       -1.340227       -1.315444
2      -1.385353        0.328414       -1.397064       -1.315444
3      -1.506521        0.098217       -1.283389       -1.315444
4      -1.021849       1.249201       -1.340227       -1.315444

```



Data after PCA (first 5 rows):

	PC1	PC2	PC3
0	-2.264703	0.480027	0.127706
1	-2.080961	-0.674134	0.234609
2	-2.364229	-0.341908	-0.044201
3	-2.299384	-0.597395	-0.091290
4	-2.389842	0.646835	-0.015738

