EXPRIMENT:-6

To perform dimensionality reduction operation using PCA for Houses Data Set

Solution:

```
from sklearn.datasets import fetch california housing
from sklearn.decomposition import PCA
from sklearn.preprocessing import StandardScaler
import pandas as pd
# Load California Housing Dataset
housing = fetch california housing(as frame=True)
data = housing.data
# Standardize the data
scaled_data = StandardScaler().fit_transform(data)
# Apply PCA to retain 95% variance
pca = PCA(n_components=0.95)
principal_components = pca.fit_transform(scaled_data)
# Explained variance ratio
print("Explained Variance Ratio:", pca.explained_variance_ratio_)
print("Number of Components:", pca.n_components_)
```

Output:

```
Explained Variance Ratio: [0.25336868 0.23516245 0.15888635 0.12887971 0.12538195 0.0824225 ]
Number of Components: 6
```

EXPRIMENT:-7

To perform Simple Linear Regression with R.

Solution:

```
# Load dataset
2
   data(mtcars) # Built-in dataset in R
3
  # View the first few rows
4
5 head(mtcars)
6
7 # Perform Simple Linear Regression
   # Predicting mpg (miles per gallon) based on wt (weight of the car)
8
   model <- lm(mpg \sim wt, data = mtcars)
9
10
   # Summary of the model
11
12
   summary(model)
13
14
   # Plot the data and regression line
   plot(mtcars$wt, mtcars$mpg,
15
        main = "Simple Linear Regression: MPG vs Weight",
16
        xlab = "Car Weight (1000 lbs)",
17
        ylab = "Miles Per Gallon",
18
        pch = 19, col = "blue")
19
20
   # Add regression line
21
22
   abline(model, col = "red", lwd = 2)
23
```

Output:

```
mpg cyl disp hp drat wt qsec vs am gear carb
Mazda RX4
                 21.0
                       6 160 110 3.90 2.620 16.46
                                                  0 1
                                                               4
                       6 160 110 3.90 2.875 17.02 0 1
Mazda RX4 Wag
                21.0
                                                          4
                                                               4
                                                          4
Datsun 710
                 22.8
                       4 108 93 3.85 2.320 18.61 1 1
                                                               1
Hornet 4 Drive
                21.4
                       6 258 110 3.08 3.215 19.44 1
                                                     0
                                                          3
                                                               1
Hornet Sportabout 18.7 8 360 175 3.15 3.440 17.02 0 0
                                                          3
                                                               2
Valiant
                 18.1 6 225 105 2.76 3.460 20.22 1
                                                          3
                                                               1
```

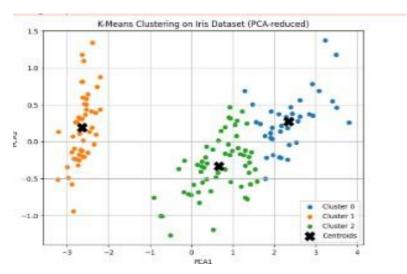
EXPRIMENT: -8:

To perform K-Means clustering operation and visualize for iris data set

Solution:

```
import pandas as pd
from sklearn.cluster import KMeans
from sklearn.datasets import load_iris
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
# Load the Iris dataset
iris = load_iris()
data = pd.DataFrame(iris.data, columns=iris.feature_names)
# Apply K-Means clustering
kmeans = KMeans(n_clusters=3, random_state=42)
data['Cluster'] = kmeans.fit_predict(data)
# Reduce dimensions for visualization using PCA
pca = PCA(n_components=2)
reduced_data = pca.fit_transform(data.iloc[:, :-1])
# Add PCA results to the dataframe
data['PCA1'] = reduced_data[:, 0]
data['PCA2'] = reduced_data[:, 1]
# Visualize clusters
plt.figure(figsize=(8, 6))
for cluster in range(3):
    cluster_data = data[data['Cluster'] == cluster]
    plt.scatter(cluster_data['PCA1'], cluster_data['PCA2'], label=f'Cluster {cluster}')
# Add cluster centers (PCA-reduced)
centers = pca.transform(kmeans.cluster_centers_)
plt.scatter(centers[:, 0], centers[:, 1], s=200, c='black', marker='X', label='Centroids')
plt.title('K-Means Clustering on Iris Dataset (PCA-reduced)')
plt.xlabel('PCA1')
plt.ylabel('PCA2')
plt.legend()
plt.grid()
plt.show()
```

Output:



EXPRIMENT: 9:

Learn how to collect data via web-scraping, APIs and data connectors from suitable sources as specified by the instructor

Solution:

1)-Web Scrapping:

```
import requests
from bs4 import BeautifulSoup

# URL of the website to scrape
url = 'https://example.com'

# Send a GET request to fetch the HTML content
response = requests.get(url)

# Parse the HTML content
soup = BeautifulSoup(response.content, 'html.parser')

# Extract specific data (e.g., all paragraph texts)
paragraphs = soup.find_all('p')
for p in paragraphs:
    print(p.text)
```

2)-APIs:

```
import requests

# TMDB API Key
api_key = 'your_api_key'
url = f'https://api.themoviedb.org/3/movie/popular?api_key={api_key}'

# Send a GET request to fetch the data
response = requests.get(url)
data = response.json()

# Print the movie titles
for movie in data['results']:
    print(movie['title'])
```

3)-Data Connectors:

```
import mysql.connector
# Connect to the database
connection = mysql.connector.connect(
   host='localhost',
   user='your_username',
   password='your_password',
    database='your_database'
# Query the data
cursor = connection.cursor()
cursor.execute('SELECT * FROM your_table')
# Fetch and display results
for row in cursor.fetchall():
   print(row)
# Close connection
cursor.close()
connection.close()
```