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

# Load the dataset
def load_data(file_path):
    try:
        data = pd.read_csv('/content/rental-price-indexes-september-2023.csv')
        print(f"\nData loaded successfully from {file_path}")
        print("\nDataset Overview:")
        print(data.info())
        return data
    except FileNotFoundError:
        print(f"File not found: {file_path}")
        return None

# Preprocess the data: Scale the features and drop non-numeric columns if any
def preprocess_data(data):
    # Select only numeric columns for PCA
    numeric_data = data.select_dtypes(include=[np.number])

    # Scale the features
    scaler = StandardScaler()
    scaled_data = scaler.fit_transform(numeric_data)

    return scaled_data

# Apply PCA
def apply_pca(scaled_data, n_components):
    # Apply PCA transformation
    pca = PCA(n_components=n_components)
    principal_components = pca.fit_transform(scaled_data)

    # Variance explained by each component
    explained_variance = pca.explained_variance_ratio_
    print(f"\nExplained variance by each component: {explained_variance}")

    return principal_components, explained_variance

# Plot the explained variance
def plot_explained_variance(explained_variance):
    plt.figure(figsize=(10, 6))
    plt.plot(np.cumsum(explained_variance), marker='o')
    plt.xlabel("Number of Principal Components")

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plt.ylabel("Cumulative Explained Variance")
plt.title("Explained Variance by Principal Components")
plt.grid()
plt.show()

# Main function
def main():
    file_path = input("Enter the path to your dataset CSV file: ")
    data = load_data(file_path)
    if data is None:
        return

    # Preprocess the data
    scaled_data = preprocess_data(data)

    # Ask the user to specify the number of components
    max_components = scaled_data.shape[1]
    n_components = int(input(f"Enter the number of components to keep
(1 to {max_components}): "))

    if n_components < 1 or n_components > max_components:
        print(f"Invalid number of components. Please enter a value
between 1 and {max_components}.")
        return

    # Apply PCA
    principal_components, explained_variance = apply_pca(scaled_data,
n_components)

    # Convert the principal components to a DataFrame
    pca_df = pd.DataFrame(data=principal_components,
columns=[f'PC{i+1}' for i in range(n_components)])
    print("\nFirst five rows of the principal components:")
    print(pca_df.head())

    # Plot cumulative explained variance
    plot_explained_variance(explained_variance)

    # Plot the first two principal components if n_components >= 2
    if n_components >= 2:
        plt.figure(figsize=(10, 6))
        sns.scatterplot(x=pca_df['PC1'], y=pca_df['PC2'])
        plt.xlabel("Principal Component 1")
        plt.ylabel("Principal Component 2")
        plt.title("Data in Principal Component Space")
        plt.grid()
        plt.show()

# Execute the main function

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```
if __name__ == "__main__":  
    main()
```

Enter the path to your dataset CSV file: /content/rental-price-indexes-september-2023.csv

Data loaded successfully from /content/rental-price-indexes-september-2023.csv

Dataset Overview:

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 1420 entries, 0 to 1419

Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype
0	SER_REF	1420 non-null	object
1	TIME_REF	1420 non-null	float64
2	DATA_VAL	1420 non-null	int64
3	STATUS	1420 non-null	object
4	UNITS	1420 non-null	object
5	Subject	1420 non-null	object
6	Group	1420 non-null	object
7	Series_title_1	1420 non-null	object
8	Series_title_2	1420 non-null	object
9	Series_title_3	1420 non-null	object

dtypes: float64(1), int64(1), object(8)

memory usage: 111.1+ KB

None

Enter the number of components to keep (1 to 2): 2

Explained variance by each component: [0.97639127 0.02360873]

First five rows of the principal components:

	PC1	PC2
0	-2.297608	0.236238
1	-2.312438	0.218514
2	-2.144569	0.128775
3	-2.126844	0.143605
4	-2.099354	0.168201

