```
In [1]: import pandas as pd
        import numpy as np
        from sklearn.preprocessing import StandardScaler
        from sklearn.cluster import KMeans
        from sklearn.metrics import davies bouldin score
        import matplotlib.pyplot as plt
        import seaborn as sns
In [2]: # Load customer and transaction data
        customers = pd.read_csv('Customers.csv')
        transactions = pd.read_csv('Transactions.csv')
        # Merge datasets on 'CustomerID'
        data = pd.merge(transactions, customers, on='CustomerID', how='inner')
In [3]: # Check for missing values
        print(data.isnull().sum())
        # Drop rows with missing values
        data.dropna(inplace=True)
       TransactionID
       CustomerTD
                          0
       ProductID
                          0
                        0
       TransactionDate
       Ouantity
       TotalValue
                        0
       Price
                         0
       CustomerName
                         0
       Region
                          0
       SignupDate
                          0
       dtype: int64
In [4]: #Convert 'TransactionDate' to datetime
        data['TransactionDate'] = pd.to_datetime(data['TransactionDate'])
In [5]: # Define a reference date for recency calculation
        reference date = data['TransactionDate'].max() + pd.Timedelta(days=1)
        # Calculate Recency, Frequency, and Monetary value for each customer
        rfm = data.groupby('CustomerID').agg({
            'TransactionDate': lambda x: (reference_date - x.max()).days,
            'TransactionID': 'nunique',
            'Price': 'sum'
        }).reset index()
        # Rename columns
        rfm.columns = ['CustomerID', 'Recency', 'Frequency', 'Monetary']
In [6]: # Select features for clustering
        features = rfm[['Recency', 'Frequency', 'Monetary']]
        # Log transformation to reduce skewness
        features log = np.log1p(features)
        # Standardize the features
        scaler = StandardScaler()
        features scaled = scaler.fit transform(features log)
In [7]: # Calculate Within-Cluster Sum of Squares (WCSS) for different cluster counts
        wcss = []
        for k in range(2, 11):
            kmeans = KMeans(n clusters=k, random state=42)
            kmeans.fit(features_scaled)
            wcss.append(kmeans.inertia_)
        # Plot the Elbow Method graph
        plt.figure(figsize=(10, 6))
        plt.plot(range(2, 11), wcss, marker='o')
        plt.title('Elbow Method for Optimal k')
        plt.xlabel('Number of Clusters')
        plt.ylabel('WCSS')
        plt.grid(True)
        plt.show()
```

C:\Users\panmo\AppData\Local\Programs\Python\Python312\Lib\site-packages\sklearn\cluster_kmeans.py:1416: Future Warning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning

super(). check params vs input(X, default n init=10)

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super()._check_params_vs_input(X, default_n_init=10)

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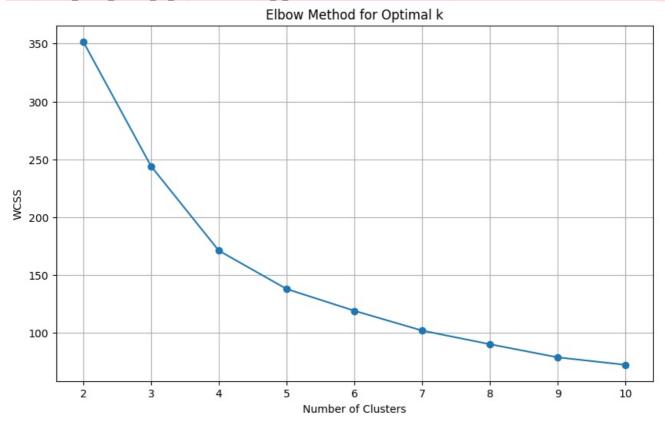
super()._check_params_vs_input(X, default_n_init=10)

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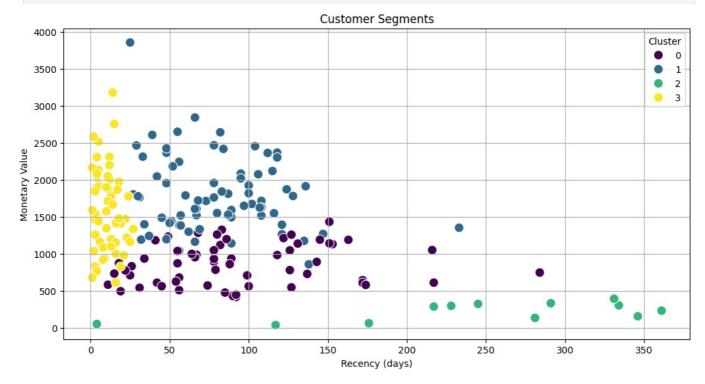
In [8]: # Choose the optimal number of clusters (e.g., k=4) optimal_k = 4 kmeans = KMeans(n_clusters=optimal_k, random_state=42) rfm['Cluster'] = kmeans.fit_predict(features_scaled)

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Warning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitl
y to suppress the warning
super(). check params vs input(X, default n init=10)

```
In [9]: # Calculate Davies-Bouldin Index
db_index = davies_bouldin_score(features_scaled, rfm['Cluster'])
print(f'Davies-Bouldin Index: {db_index:.2f}')
```

Davies-Bouldin Index: 0.85

```
In [10]: # Plot clusters
plt.figure(figsize=(12, 6))
sns.scatterplot(data=rfm, x='Recency', y='Monetary', hue='Cluster', palette='viridis', s=100)
plt.title('Customer Segments')
plt.xlabel('Recency (days)')
plt.ylabel('Monetary Value')
plt.legend(title='Cluster')
plt.grid(True)
plt.show()
```



In []:

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