

Implement K-Means clustering/ hierarchical clustering on sales_data_sample.csv dataset. Determine the number of clusters using the elbow method.

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In [ ]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

In [ ]: df = pd.read_csv("sales_data_sample.csv", encoding = 'latin-1')
df

In [ ]: df.isnull().sum()

In [ ]: df.shape

In [ ]: df.drop(['ADDRESSLINE1', "ADDRESSLINE2", "STATE", "TERRITORY", "CONTACTFIRST

In [ ]: df.drop(['PHONE', 'YEAR_ID', 'ORDERDATE'], axis = 1, inplace = True)
df.info()

In [ ]: df.describe()

In [ ]: df.drop('ORDERNUMBER', inplace = True, axis = 1)

In [ ]: df.describe()

In [ ]: df.shape

In [ ]: cat_list = df.select_dtypes(include=object).columns.to_list()

In [ ]: cat_list

In [ ]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()

In [ ]: df[cat_list].nunique()

In [ ]: for i in cat_list:
    df[i] = le.fit_transform(df[i])

In [ ]: X = df[['SALES', 'PRODUCTCODE']]

In [ ]: X.head()
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In [ ]: import warnings

warnings.filterwarnings('ignore', category=FutureWarning, module='sklearn
from yellowbrick.cluster import KElbowVisualizer
from sklearn.cluster import KMeans

visualizer = KElbowVisualizer(KMeans(), k = (1,12)).fit(X)
visualizer.show()
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In [ ]: kmeans_model = KMeans(n_clusters = 3, init = "k-means++").fit(X)
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In [ ]: kmeans_model.labels_
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In [ ]: kmeans_model.cluster_centers_
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In [ ]: from collections import Counter
Counter(kmeans_model.labels_)
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In [ ]: sns.scatterplot(data = X, x = 'SALES', y = 'PRODUCTCODE', hue = kmeans_mod
plt.scatter(kmeans_model.cluster_centers_[0], kmeans_model.cluster_cent
plt.legend()
plt.show()
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In [ ]:
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