Assignment5_ML_Sales_data

October 9, 2024

[1]:
```
#Implement K-Means clustering/ hierarchical clustering on sales_data_sample.csv dataset.
#Determine the number of clusters using the elbow method.
#Dataset link : https://www.kaggle.com/datasets/kyanyoga/sample-sales-data
```

[2]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
```

[3]:
```
df =  pd.read_csv("/home/pc13/Downloads/sales_data_sample.csv",encoding='latin')
df
```

[3]:

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	SALES	\
0	10107	30	95.70	2	2871.00	
1	10121	34	81.35	5	2765.90	
2	10134	41	94.74	2	3884.34	
3	10145	45	83.26	6	3746.70	
4	10159	49	100.00	14	5205.27	
…	…	…	…	…	…	
2818	10350	20	100.00	15	2244.40	
2819	10373	29	100.00	1	3978.51	
2820	10386	43	100.00	4	5417.57	
2821	10397	34	62.24	1	2116.16	
2822	10414	47	65.52	9	3079.44	

	ORDERDATE	STATUS	QTR_ID	MONTH_ID	YEAR_ID	…	\
0	2/24/2003 0:00	Shipped	1	2	2003	…	
1	5/7/2003 0:00	Shipped	2	5	2003	…	
2	7/1/2003 0:00	Shipped	3	7	2003	…	
3	8/25/2003 0:00	Shipped	3	8	2003	…	
4	10/10/2003 0:00	Shipped	4	10	2003	…	
…	…	…	…	…	…	…	
2818	12/2/2004 0:00	Shipped	4	12	2004	…	
2819	1/31/2005 0:00	Shipped	1	1	2005	…	
2820	3/1/2005 0:00	Resolved	1	3	2005	…	
2821	3/28/2005 0:00	Shipped	1	3	2005	…	

```
2822    5/6/2005 0:00   On Hold       2         5     2005  …

                      ADDRESSLINE1  ADDRESSLINE2           CITY STATE  \
0            897 Long Airport Avenue           NaN            NYC    NY
1                 59 rue de l'Abbaye           NaN          Reims   NaN
2      27 rue du Colonel Pierre Avia           NaN          Paris   NaN
3                 78934 Hillside Dr.           NaN       Pasadena    CA
4                    7734 Strong St.           NaN  San Francisco    CA
…                                …           …            …   …
2818              C/ Moralzarzal, 86           NaN         Madrid   NaN
2819                    Torikatu 38           NaN           Oulu   NaN
2820              C/ Moralzarzal, 86           NaN         Madrid   NaN
2821           1 rue Alsace-Lorraine           NaN       Toulouse   NaN
2822              8616 Spinnaker Dr.           NaN         Boston    MA

     POSTALCODE  COUNTRY TERRITORY CONTACTLASTNAME CONTACTFIRSTNAME DEALSIZE
0         10022      USA       NaN              Yu              Kwai    Small
1         51100   France      EMEA         Henriot              Paul    Small
2         75508   France      EMEA        Da Cunha             Daniel   Medium
3         90003      USA       NaN           Young             Julie   Medium
4           NaN      USA       NaN           Brown             Julie   Medium
…         …      …        …             …               …      …
2818      28034    Spain      EMEA          Freyre             Diego    Small
2819      90110  Finland      EMEA       Koskitalo             Pirkko   Medium
2820      28034    Spain      EMEA          Freyre             Diego   Medium
2821      31000   France      EMEA          Roulet           Annette    Small
2822      51003      USA       NaN         Yoshido              Juri   Medium

[2823 rows x 25 columns]
```

[4]:
```
df.dtypes
```

[4]:
```
ORDERNUMBER           int64
QUANTITYORDERED       int64
PRICEEACH           float64
ORDERLINENUMBER       int64
SALES               float64
ORDERDATE            object
STATUS               object
QTR_ID                int64
MONTH_ID              int64
YEAR_ID               int64
PRODUCTLINE          object
MSRP                  int64
PRODUCTCODE          object
CUSTOMERNAME         object
PHONE                object
```

```
ADDRESSLINE1          object
ADDRESSLINE2          object
CITY                  object
STATE                 object
POSTALCODE            object
COUNTRY               object
TERRITORY             object
CONTACTLASTNAME       object
CONTACTFIRSTNAME      object
DEALSIZE              object
dtype: object
```

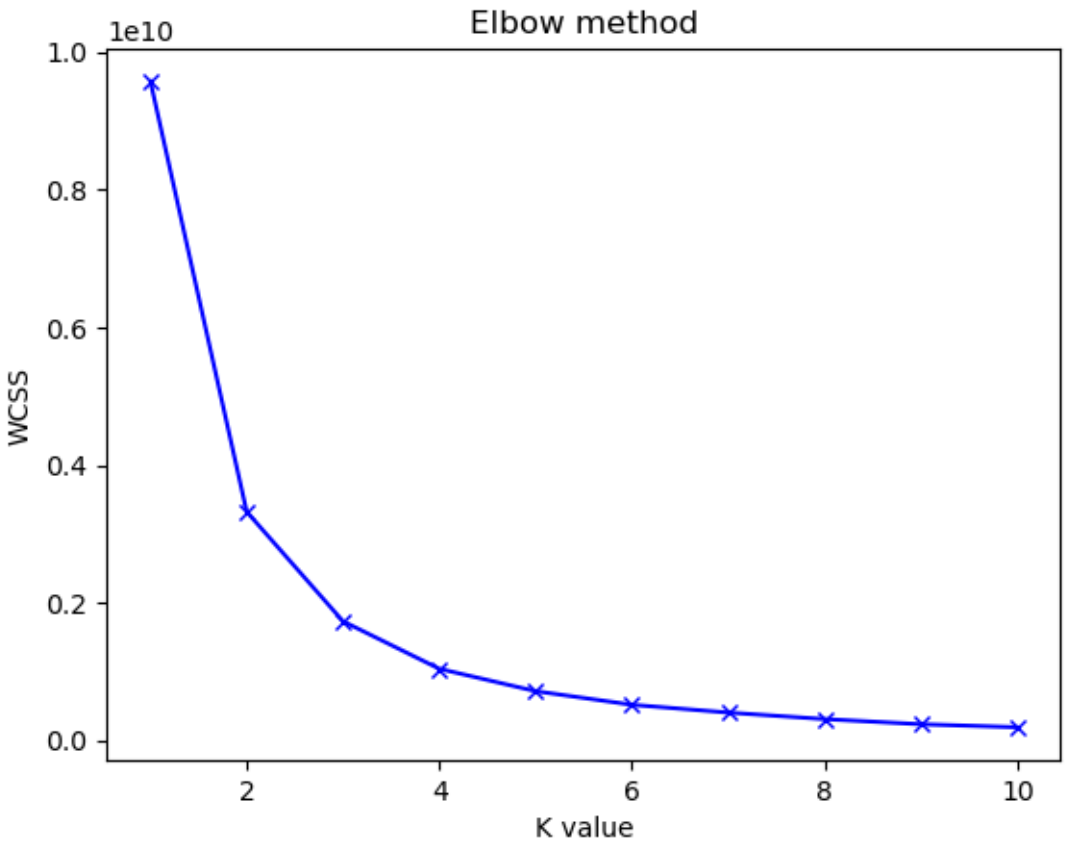
[5]:
```
X = df.iloc[:, [3,4]].values #access rows and columns by their integer index␣
↪positions.
```

[19]:
```
wcss = []   #within cluster sum of square

for i in range(1,11):
    #init argument is the method for initializing the centroid
    kmeans = KMeans(n_clusters=i, init="k-means++", random_state=42)
    kmeans.fit(X)
    #we calculate wcss value for each k value
    wcss.append(kmeans.inertia_)

ks = [1,2,3,4,5,6,7,8,9,10]
plt.plot(ks, wcss, 'bx-')
plt.title("Elbow method")
plt.xlabel("K value")
plt.ylabel("WCSS")
```

[19]:
```
Text(0, 0.5, 'WCSS')
```



[12]:
```
df.describe()
```

[12]:

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	\
count	2823.000000	2823.000000	2823.000000	2823.000000	
mean	10258.725115	35.092809	83.658544	6.466171	
std	92.085478	9.741443	20.174277	4.225841	
min	10100.000000	6.000000	26.880000	1.000000	
25%	10180.000000	27.000000	68.860000	3.000000	
50%	10262.000000	35.000000	95.700000	6.000000	
75%	10333.500000	43.000000	100.000000	9.000000	
max	10425.000000	97.000000	100.000000	18.000000	

	SALES	QTR_ID	MONTH_ID	YEAR_ID	MSRP
count	2823.000000	2823.000000	2823.000000	2823.00000	2823.000000
mean	3553.889072	2.717676	7.092455	2003.81509	100.715551
std	1841.865106	1.203878	3.656633	0.69967	40.187912
min	482.130000	1.000000	1.000000	2003.00000	33.000000
25%	2203.430000	2.000000	4.000000	2003.00000	68.000000
50%	3184.800000	3.000000	8.000000	2004.00000	99.000000
75%	4508.000000	4.000000	11.000000	2004.00000	124.000000

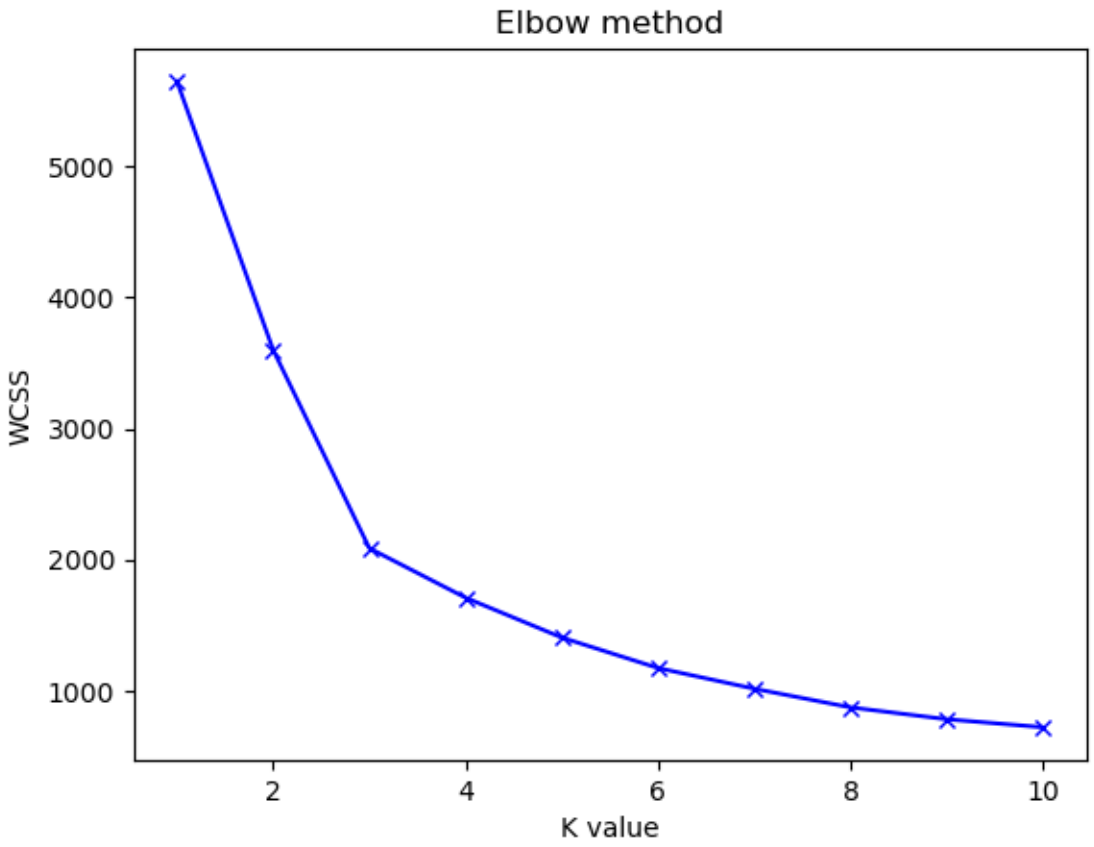
```
max     14082.800000      4.000000     12.000000   2005.00000    214.000000
```

[18]:
```
wcss =[]

for i in range(1,11):
    clustering = KMeans(n_clusters=i, init="k-means++", random_state=42)
    clustering.fit(scaled)
    wcss.append(clustering.inertia_)

ks = [1,2,3,4,5,6,7,8,9,10]
plt.plot(ks, wcss, 'bx-')
plt.title("Elbow method")
plt.xlabel("K value")
plt.ylabel("WCSS")
```

[18]:
```
Text(0, 0.5, 'WCSS')
```



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