ML_5

November 11, 2025

[1]:
```
import pandas as pd
```

[3]:
```
df = pd.read_csv(r"C:
↪\Users\suraj\OneDrive\Desktop\LP3-master\ML\datasets\sales_data_sample_utf8.
↪csv")
```

[5]:
```
df.head()
```

[5]:

	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	

	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	…	

	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	

	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

[5 rows x 25 columns]

[7]:
```
data = df.select_dtypes(include=['float64','int64'])
data = data.fillna(data.mean())
data.head()
```

[7]:

	ORDERNUMBER	QUANTITYORDERED	PRICEEACH	ORDERLINENUMBER	SALES	QTR_ID	\
0	10107	30	95.70	2	2871.00	1	
1	10121	34	81.35	5	2765.90	2	
2	10134	41	94.74	2	3884.34	3	
3	10145	45	83.26	6	3746.70	3	
4	10159	49	100.00	14	5205.27	4	

	MONTH_ID	YEAR_ID	MSRP
0	2	2003	95
1	5	2003	95
2	7	2003	95
3	8	2003	95
4	10	2003	95

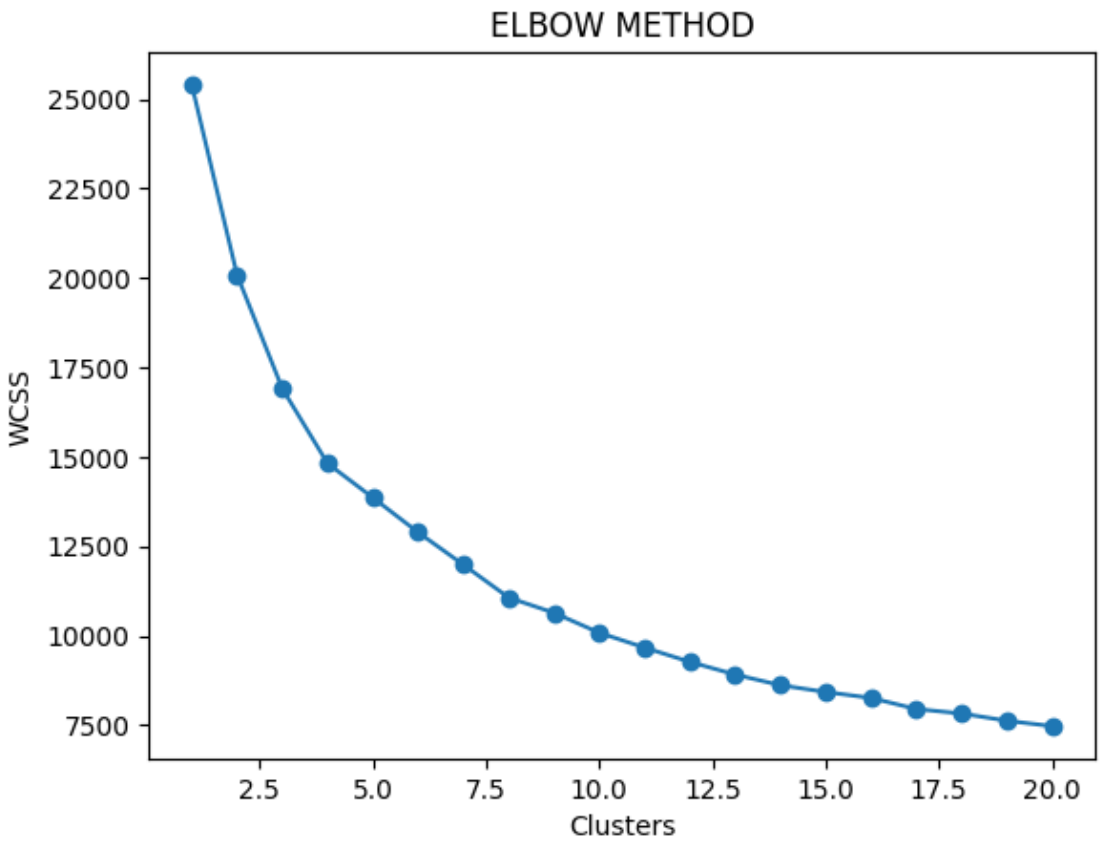
[9]:
```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
scaleddata = sc.fit_transform(data)
```

[17]:
```
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
```

[18]:
```
wcss = []

for i in range(1, 21):
    kmeans = KMeans(n_clusters=i,random_state=42)
    kmeans.fit(scaleddata)
    wcss.append(kmeans.inertia_)
```

[21]:
```
plt.plot(range(1,21),wcss,marker='o')
plt.xlabel("Clusters")
plt.ylabel("WCSS")
plt.title("ELBOW METHOD")
plt.show()
```



[23]:
```
kmeans = KMeans(n_clusters = 5,random_state=42)
df['Clusters']=kmeans.fit_predict(scaleddata)
```

[25]:
```
df['Clusters'].head()
```

[25]:
```
0    2
1    2
2    4
3    4
4    1
Name: Clusters, dtype: int32
```

[27]:
```
df['Clusters'].value_counts()
```

[27]:
```
Clusters
3    647
2    631
1    575
4    508
0    462
```

```
Name: count, dtype: int64
```

[29]:
```
df.head()
```

[29]:
```
   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

          ORDERDATE   STATUS  QTR_ID  MONTH_ID  YEAR_ID  … ADDRESSLINE2  \
0   2/24/2003 0:00  Shipped       1         2     2003  …           NaN
1    5/7/2003 0:00  Shipped       2         5     2003  …           NaN
2    7/1/2003 0:00  Shipped       3         7     2003  …           NaN
3   8/25/2003 0:00  Shipped       3         8     2003  …           NaN
4  10/10/2003 0:00  Shipped       4        10     2003  …           NaN

            CITY STATE POSTALCODE COUNTRY TERRITORY CONTACTLASTNAME  \
0            NYC    NY      10022     USA       NaN              Yu
1          Reims   NaN      51100  France      EMEA         Henriot
2          Paris   NaN      75508  France      EMEA        Da Cunha
3       Pasadena    CA      90003     USA       NaN           Young
4  San Francisco    CA        NaN     USA       NaN           Brown

  CONTACTFIRSTNAME DEALSIZE Clusters
0             Kwai    Small        2
1             Paul    Small        2
2           Daniel   Medium        4
3            Julie   Medium        4
4            Julie   Medium        1

[5 rows x 26 columns]
```

[ ]: