

Sales_K-means_clustering

November 8, 2023

```
[4]: import numpy as np
import pandas as pd
```

```
[7]: df = pd.read_csv(r"C:\Users\lalit\OneDrive\Desktop\practicals\machine_
learning\practical4\sales_data_sample.csv")
```

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

```
[8]:  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]
```

```
[9]: df.head(50)
```

```
[9]:
```

	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	
5	10168	36	96.66	1	3479.76	
6	10180	29	86.13	9	2497.77	
7	10188	48	100.00	1	5512.32	
8	10201	22	98.57	2	2168.54	
9	10211	41	100.00	14	4708.44	
10	10223	37	100.00	1	3965.66	
11	10237	23	100.00	7	2333.12	
12	10251	28	100.00	2	3188.64	
13	10263	34	100.00	2	3676.76	
14	10275	45	92.83	1	4177.35	
15	10285	36	100.00	6	4099.68	
16	10299	23	100.00	9	2597.39	
17	10309	41	100.00	5	4394.38	
18	10318	46	94.74	1	4358.04	
19	10329	42	100.00	1	4396.14	
20	10341	41	100.00	9	7737.93	
21	10361	20	72.55	13	1451.00	
22	10375	21	34.91	12	733.11	
23	10388	42	76.36	4	3207.12	
24	10403	24	100.00	7	2434.56	
25	10417	66	100.00	2	7516.08	
26	10103	26	100.00	11	5404.62	
27	10112	29	100.00	1	7209.11	
28	10126	38	100.00	11	7329.06	
29	10140	37	100.00	11	7374.10	
30	10150	45	100.00	8	10993.50	
31	10163	21	100.00	1	4860.24	
32	10174	34	100.00	4	8014.82	
33	10183	23	100.00	8	5372.57	
34	10194	42	100.00	11	7290.36	
35	10206	47	100.00	6	9064.89	
36	10215	35	100.00	3	6075.30	
37	10228	29	100.00	2	6463.23	
38	10245	34	100.00	9	6120.34	
39	10258	32	100.00	6	7680.64	
40	10270	21	100.00	9	4905.39	
41	10280	34	100.00	2	8014.82	
42	10291	37	100.00	11	7136.19	
43	10304	47	100.00	6	10172.70	

44	10312	48	100.00	3	11623.70
45	10322	40	100.00	1	6000.40
46	10333	26	100.00	3	3003.00
47	10347	30	100.00	1	3944.70
48	10357	32	100.00	10	5691.84
49	10369	41	100.00	2	4514.92

	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	...
5	10/28/2003 0:00	Shipped	4	10	2003	...
6	11/11/2003 0:00	Shipped	4	11	2003	...
7	11/18/2003 0:00	Shipped	4	11	2003	...
8	12/1/2003 0:00	Shipped	4	12	2003	...
9	1/15/2004 0:00	Shipped	1	1	2004	...
10	2/20/2004 0:00	Shipped	1	2	2004	...
11	4/5/2004 0:00	Shipped	2	4	2004	...
12	5/18/2004 0:00	Shipped	2	5	2004	...
13	6/28/2004 0:00	Shipped	2	6	2004	...
14	7/23/2004 0:00	Shipped	3	7	2004	...
15	8/27/2004 0:00	Shipped	3	8	2004	...
16	9/30/2004 0:00	Shipped	3	9	2004	...
17	10/15/2004 0:00	Shipped	4	10	2004	...
18	11/2/2004 0:00	Shipped	4	11	2004	...
19	11/15/2004 0:00	Shipped	4	11	2004	...
20	11/24/2004 0:00	Shipped	4	11	2004	...
21	12/17/2004 0:00	Shipped	4	12	2004	...
22	2/3/2005 0:00	Shipped	1	2	2005	...
23	3/3/2005 0:00	Shipped	1	3	2005	...
24	4/8/2005 0:00	Shipped	2	4	2005	...
25	5/13/2005 0:00	Disputed	2	5	2005	...
26	1/29/2003 0:00	Shipped	1	1	2003	...
27	3/24/2003 0:00	Shipped	1	3	2003	...
28	5/28/2003 0:00	Shipped	2	5	2003	...
29	7/24/2003 0:00	Shipped	3	7	2003	...
30	9/19/2003 0:00	Shipped	3	9	2003	...
31	10/20/2003 0:00	Shipped	4	10	2003	...
32	11/6/2003 0:00	Shipped	4	11	2003	...
33	11/13/2003 0:00	Shipped	4	11	2003	...
34	11/25/2003 0:00	Shipped	4	11	2003	...
35	12/5/2003 0:00	Shipped	4	12	2003	...
36	1/29/2004 0:00	Shipped	1	1	2004	...
37	3/10/2004 0:00	Shipped	1	3	2004	...
38	5/4/2004 0:00	Shipped	2	5	2004	...

39	6/15/2004 0:00	Shipped	2	6	2004	...
40	7/19/2004 0:00	Shipped	3	7	2004	...
41	8/17/2004 0:00	Shipped	3	8	2004	...
42	9/8/2004 0:00	Shipped	3	9	2004	...
43	10/11/2004 0:00	Shipped	4	10	2004	...
44	10/21/2004 0:00	Shipped	4	10	2004	...
45	11/4/2004 0:00	Shipped	4	11	2004	...
46	11/18/2004 0:00	Shipped	4	11	2004	...
47	11/29/2004 0:00	Shipped	4	11	2004	...
48	12/10/2004 0:00	Shipped	4	12	2004	...
49	1/20/2005 0:00	Shipped	1	1	2005	...

	ADDRESSLINE1	ADDRESSLINE2	CITY
0	897 Long Airport Avenue	NaN	NYC \
1	59 rue de l'Abbaye	NaN	Reims
2	27 rue du Colonel Pierre Avia	NaN	Paris
3	78934 Hillside Dr.	NaN	Pasadena
4	7734 Strong St.	NaN	San Francisco
5	9408 Furth Circle	NaN	Burlingame
6	184, chausse de Tournai	NaN	Lille
7	Drammen 121, PR 744 Sentrum	NaN	Bergen
8	5557 North Pendale Street	NaN	San Francisco
9	25, rue Lauriston	NaN	Paris
10	636 St Kilda Road	Level 3	Melbourne
11	2678 Kingston Rd.	Suite 101	NYC
12	7476 Moss Rd.	NaN	Newark
13	25593 South Bay Ln.	NaN	Bridgewater
14	67, rue des Cinquante Otages	NaN	Nantes
15	39323 Spinnaker Dr.	NaN	Cambridge
16	Keskuskatu 45	NaN	Helsinki
17	Erling Skakkas gate 78	NaN	Stavern
18	7586 Pompton St.	NaN	Allentown
19	897 Long Airport Avenue	NaN	NYC
20	Geislweg 14	NaN	Salzburg
21	Monitor Money Building, 815 Pacific Hwy	Level 6	Chatswood
22	67, rue des Cinquante Otages	NaN	Nantes
23	1785 First Street	NaN	New Bedford
24	Berkeley Gardens 12 Brewery	NaN	Liverpool
25	C/ Moralarzal, 86	NaN	Madrid
26	Erling Skakkas gate 78	NaN	Stavern
27	Berguvsv„gen 8	NaN	Lule
28	C/ Araquil, 67	NaN	Madrid
29	9408 Furth Circle	NaN	Burlingame
30	Bronz Sok., Bronz Apt. 3/6 Tesvikiye	NaN	Singapore
31	5905 Pompton St.	Suite 750	NYC
32	31 Duncan St. West End	NaN	South Brisbane
33	782 First Street	NaN	Philadelphia

34		2, rue du Commerce	NaN	Lyon
35		1900 Oak St.	NaN	Vancouver
36		3675 Furth Circle	NaN	Burbank
37		4658 Baden Av.	NaN	Cambridge
38		567 North Pendale Street	NaN	New Haven
39		2-2-8 Roppongi	NaN	Minato-ku
40	Monitor Money Building,	815 Pacific Hwy	Level 6	Chatswood
41		Via Monte Bianco 34	NaN	Torino
42		?kergatan 24	NaN	Boras
43		67, avenue de l'Europe	NaN	Versailles
44		5677 Strong St.	NaN	San Rafael
45		2304 Long Airport Avenue	NaN	Nashua
46		5557 North Pendale Street	NaN	San Francisco
47		636 St Kilda Road	Level 3	Melbourne
48		5677 Strong St.	NaN	San Rafael
49		7825 Douglas Av.	NaN	Brickhaven

	STATE	POSTALCODE	COUNTRY	TERRITORY	CONTACTLASTNAME
0	NY	10022	USA	NaN	Yu \
1	NaN	51100	France	EMEA	Henriot
2	NaN	75508	France	EMEA	Da Cunha
3	CA	90003	USA	NaN	Young
4	CA	NaN	USA	NaN	Brown
5	CA	94217	USA	NaN	Hirano
6	NaN	59000	France	EMEA	Rance
7	NaN	N 5804	Norway	EMEA	Oeztan
8	CA	NaN	USA	NaN	Murphy
9	NaN	75016	France	EMEA	Perrier
10	Victoria	3004	Australia	APAC	Ferguson
11	NY	10022	USA	NaN	Frick
12	NJ	94019	USA	NaN	Brown
13	CT	97562	USA	NaN	King
14	NaN	44000	France	EMEA	Labrune
15	MA	51247	USA	NaN	Hernandez
16	NaN	21240	Finland	EMEA	Karttunen
17	NaN	4110	Norway	EMEA	Bergulfsen
18	PA	70267	USA	NaN	Yu
19	NY	10022	USA	NaN	Yu
20	NaN	5020	Austria	EMEA	Pipps
21	NSW	2067	Australia	APAC	Huxley
22	NaN	44000	France	EMEA	Labrune
23	MA	50553	USA	NaN	Benitez
24	NaN	WX1 6LT	UK	EMEA	Devon
25	NaN	28034	Spain	EMEA	Freyre
26	NaN	4110	Norway	EMEA	Bergulfsen
27	NaN	S-958 22	Sweden	EMEA	Berglund
28	NaN	28023	Spain	EMEA	Sommer

29	CA	94217	USA	NaN	Hirano
30	NaN	79903	Singapore	Japan	Natividad
31	NY	10022	USA	NaN	Hernandez
32	Queensland	4101	Australia	APAC	Calaghan
33	PA	71270	USA	NaN	Cervantes
34	NaN	69004	France	EMEA	Saveley
35	BC	V3F 2K1	Canada	NaN	Tannamuri
36	CA	94019	USA	NaN	Thompson
37	MA	51247	USA	NaN	Tseng
38	CT	97823	USA	NaN	Murphy
39	Tokyo	106-0032	Japan	Japan	Shimamura
40	NSW	2067	Australia	APAC	Huxley
41	NaN	10100	Italy	EMEA	Accorti
42	NaN	S-844 67	Sweden	EMEA	Larsson
43	NaN	78000	France	EMEA	Tonini
44	CA	97562	USA	NaN	Nelson
45	NH	62005	USA	NaN	Young
46	CA	NaN	USA	NaN	Murphy
47	Victoria	3004	Australia	APAC	Ferguson
48	CA	97562	USA	NaN	Nelson
49	MA	58339	USA	NaN	Nelson

	CONTACTFIRSTNAME	DEALSIZE
0	Kwai	Small
1	Paul	Small
2	Daniel	Medium
3	Julie	Medium
4	Julie	Medium
5	Juri	Medium
6	Martine	Small
7	Veysel	Medium
8	Julie	Small
9	Dominique	Medium
10	Peter	Medium
11	Michael	Small
12	William	Medium
13	Julie	Medium
14	Janine	Medium
15	Marta	Medium
16	Matti	Small
17	Jonas	Medium
18	Kyung	Medium
19	Kwai	Medium
20	Georg	Large
21	Adrian	Small
22	Janine	Small
23	Violeta	Medium

24	Elizabeth	Small
25	Diego	Large
26	Jonas	Medium
27	Christina	Large
28	Martijn	Large
29	Juri	Large
30	Eric	Large
31	Maria	Medium
32	Tony	Large
33	Francisca	Medium
34	Mary	Large
35	Yoshi	Large
36	Steve	Medium
37	Kyung	Medium
38	Leslie	Medium
39	Akiko	Large
40	Adrian	Medium
41	Paolo	Large
42	Maria	Large
43	Daniel	Large
44	Valarie	Large
45	Valarie	Medium
46	Julie	Medium
47	Peter	Medium
48	Valarie	Medium
49	Allen	Medium

[50 rows x 25 columns]

[10]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2823 entries, 0 to 2822
Data columns (total 25 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ORDERNUMBER           2823 non-null   int64
1   QUANTITYORDERED       2823 non-null   int64
2   PRICEEACH             2823 non-null   float64
3   ORDERLINENUMBER       2823 non-null   int64
4   SALES                 2823 non-null   float64
5   ORDERDATE             2823 non-null   object
6   STATUS                2823 non-null   object
7   QTR_ID               2823 non-null   int64
8   MONTH_ID             2823 non-null   int64
9   YEAR_ID              2823 non-null   int64
10  PRODUCTLINE           2823 non-null   object
11  MSRP                 2823 non-null   int64
```

```

12  PRODUCTCODE      2823 non-null  object
13  CUSTOMERNAME     2823 non-null  object
14  PHONE            2823 non-null  object
15  ADDRESSLINE1     2823 non-null  object
16  ADDRESSLINE2     302 non-null   object
17  CITY             2823 non-null  object
18  STATE            1337 non-null  object
19  POSTALCODE       2747 non-null  object
20  COUNTRY          2823 non-null  object
21  TERRITORY        1749 non-null  object
22  CONTACTLASTNAME  2823 non-null  object
23  CONTACTFIRSTNAME 2823 non-null  object
24  DEALSIZE         2823 non-null  object
dtypes: float64(2), int64(7), object(16)
memory usage: 551.5+ KB

```

```
[11]: df.shape
```

```
[11]: (2823, 25)
```

```
[12]: df.isnull().sum()
```

```

[12]: ORDERNUMBER      0
      QUANTITYORDERED  0
      PRICEEACH        0
      ORDERLINENUMBER  0
      SALES             0
      ORDERDATE        0
      STATUS           0
      QTR_ID           0
      MONTH_ID         0
      YEAR_ID          0
      PRODUCTLINE      0
      MSRP             0
      PRODUCTCODE      0
      CUSTOMERNAME     0
      PHONE            0
      ADDRESSLINE1     0
      ADDRESSLINE2     2521
      CITY             0
      STATE            1486
      POSTALCODE       76
      COUNTRY          0
      TERRITORY        1074
      CONTACTLASTNAME  0
      CONTACTFIRSTNAME 0
      DEALSIZE         0
dtype: int64

```



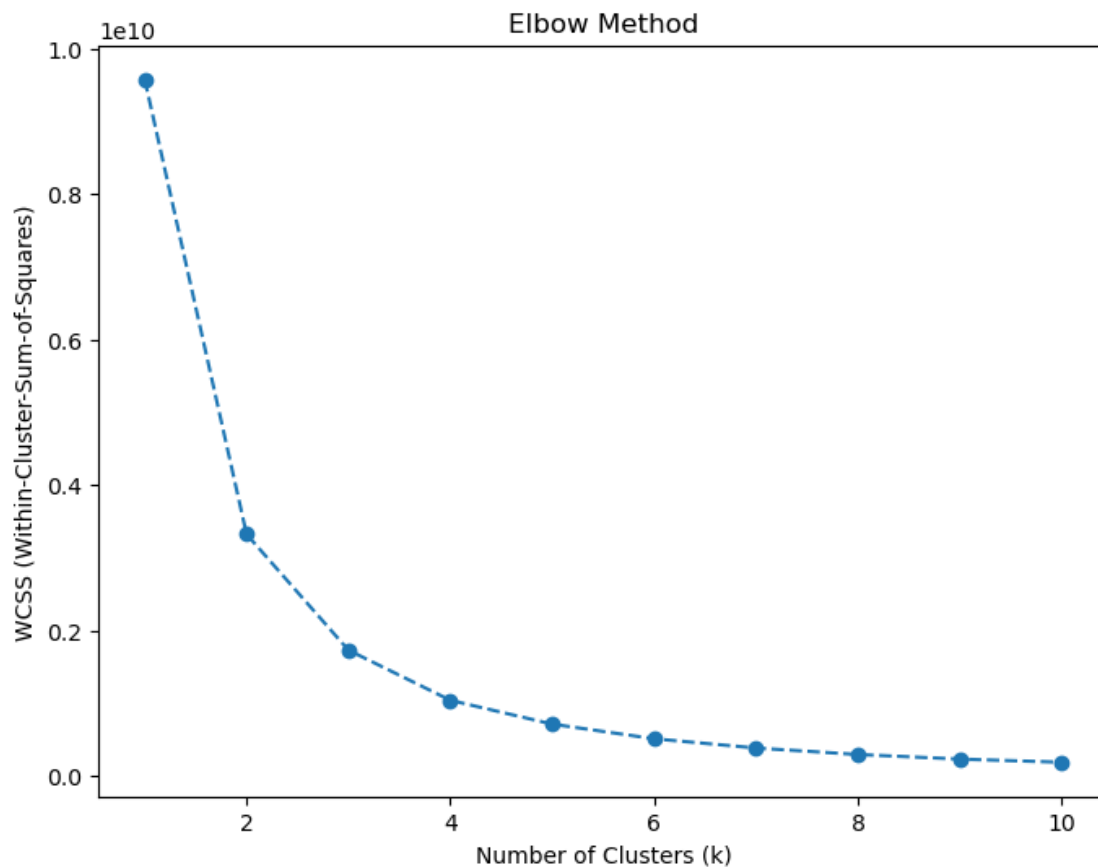
```
[13]: from sklearn.cluster import KMeans
import matplotlib.pyplot as plt

[21]: # Select relevant numerical columns for clustering
numerical_columns = df[['QUANTITYORDERED', 'PRICEEACH', 'SALES']]

[23]: # Determine the optimal number of clusters using the elbow method
wcss = [] # Within-Cluster-Sum-of-Squares

[24]: # Iterate over a range of values for k (number of clusters)
for k in range(1, 11):
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(numerical_columns) # Fit K-Means to the data
    wcss.append(kmeans.inertia_) # Append the inertia value to the list

[25]: # Plot the elbow graph
plt.figure(figsize=(8, 6))
plt.plot(range(1, 11), wcss, marker='o', linestyle='--')
plt.title('Elbow Method')
plt.xlabel('Number of Clusters (k)')
plt.ylabel('WCSS (Within-Cluster-Sum-of-Squares)')
plt.show()
```



```
[26]: # Based on the elbow method, let's say the optimal number of clusters is 3
      optimal_k = 3
```

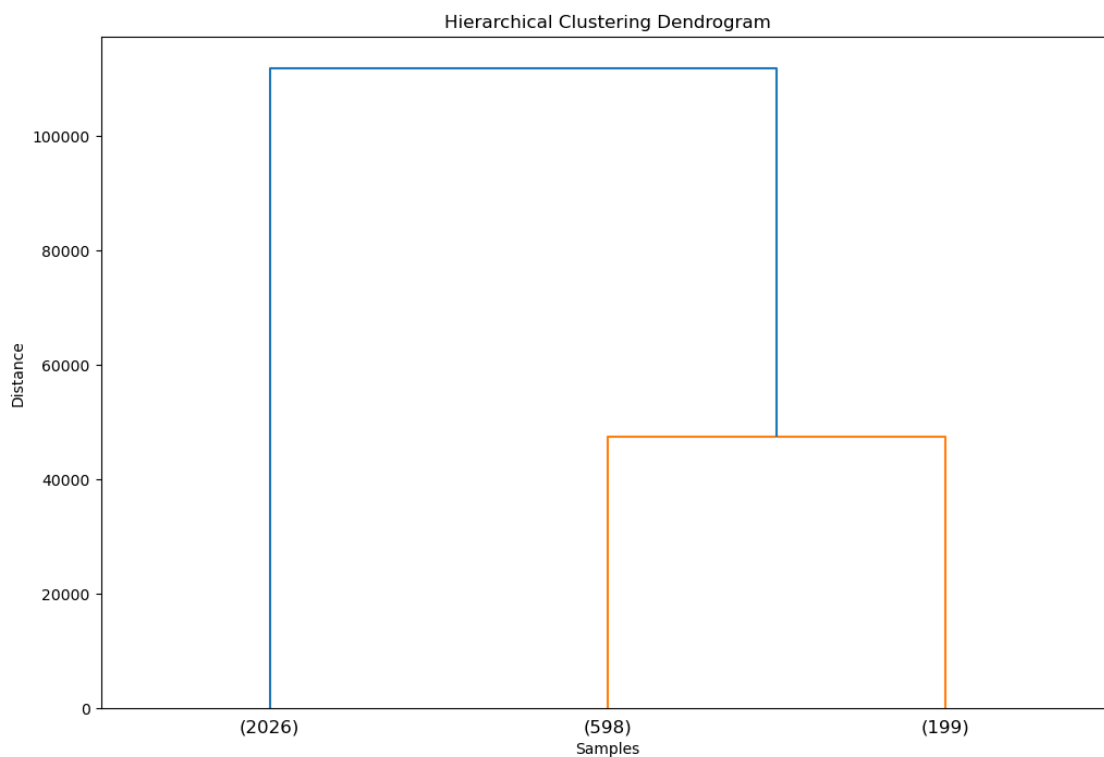
```
[29]: # Perform K-Means clustering with the chosen number of clusters
      kmeans = KMeans(n_clusters=optimal_k, random_state=42)
      df['cluster'] = kmeans.fit_predict(numerical_columns)
```

```
[37]: from scipy.cluster.hierarchy import dendrogram, linkage
      import matplotlib.pyplot as plt

      numerical_columns = df[['QUANTITYORDERED', 'PRICEEACH', 'SALES']]
```

```
[38]: # Create a linkage matrix using the Ward method
      Z = linkage(numerical_columns, method='ward')
```

```
[39]: # Plot the hierarchical clustering dendrogram
      plt.figure(figsize=(12, 8))
      dendrogram(Z, p=optimal_k, truncate_mode='lastp')
      plt.title('Hierarchical Clustering Dendrogram')
      plt.xlabel('Samples')
      plt.ylabel('Distance')
      plt.show()
```



```
[40]: from scipy.cluster.hierarchy import fcluster
```

```
[41]: # Based on the dendrogram, let's say the optimal number of clusters is 3
      optimal_k = 3
```

```
[42]: # Cut the dendrogram to obtain cluster assignments
      cluster_assignments = fcluster(Z, t=optimal_k, criterion='maxclust')
```

```
[44]: # Add the cluster assignments to your original dataset
      df['cluster'] = cluster_assignments
```

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

```
[47]:  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 cluster
0             Kwai     Small        1
1             Paul     Small        1
2          Daniel   Medium        1
3             Julie   Medium        1
4             Julie   Medium        2
```

```
[5 rows x 26 columns]
```

```
[ ]:
```