

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

```
In [8]: df=pd.read_csv("sales_data_sample.csv",encoding = 'Latin-1')
## if we don't add "",encoding = 'Latin-1'"" then we get error
```

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

```
Out [9]:
```

| | ORDERNUMBER | QUANTITYORDERED | PRICEEACH | ORDERLINENUMBER | SALES | ORDERDATE | STATUS | QTR_ID | MONTH_ID | YEAR_ID |
|---|-------------|-----------------|-----------|-----------------|---------|-----------------|---------|--------|----------|---------|
| 0 | 10107 | 30 | 95.70 | 2 | 2871.00 | 2/24/2003 0:00 | Shipped | 1 | 2 | 2003 |
| 1 | 10121 | 34 | 81.35 | 5 | 2765.90 | 5/7/2003 0:00 | Shipped | 2 | 5 | 2003 |
| 2 | 10134 | 41 | 94.74 | 2 | 3884.34 | 7/1/2003 0:00 | Shipped | 3 | 7 | 2003 |
| 3 | 10145 | 45 | 83.26 | 6 | 3746.70 | 8/25/2003 0:00 | Shipped | 3 | 8 | 2003 |
| 4 | 10159 | 49 | 100.00 | 14 | 5205.27 | 10/10/2003 0:00 | Shipped | 4 | 10 | 2003 |

5 rows × 25 columns

```
In [11]: 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
```

```
In [13]: df.isnull().sum()
```

```
Out [13]: 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
```

```
In [14]: ## so only two columns are important of the dataset i.e. QuantityOrdered and PriceEach other are irrelevant
data=df[['QUANTITYORDERED','PRICEEACH']]
```

Out [16]:

| | QUANTITYORDERED | PRICEEACH |
|---|-----------------|-----------|
| 0 | 30 | 95.70 |
| 1 | 34 | 81.35 |
| 2 | 41 | 94.74 |
| 3 | 45 | 83.26 |

```
In [18]: ## Do normalization of the data
from sklearn.preprocessing import StandardScaler
# make object of it
scaler=StandardScaler()
normalized_data=scaler.fit_transform(data)
print(normalized_data)
```

```
[ -0.52289086  0.5969775 ]
[ -0.11220131 -0.11445035]
[  0.60650538  0.54938372]
...
[  0.81185016  0.81015797]
[ -0.11220131 -1.06186404]
[  1.2225397  -0.89925195]
```

```
In [19]: ## Using elbow method , determine the best value of k
# wcss= within cluster sum of squares . It's a measure of how close data points are to the centroid

from sklearn.cluster import KMeans

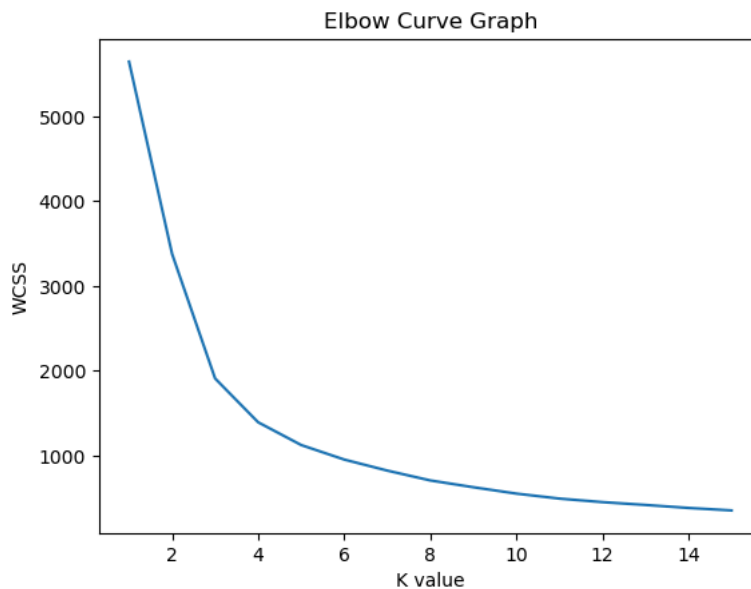
wcss=[]
for i in range(1,16):
    k_means=KMeans(n_clusters=i,init='k-means++',max_iter=300,n_init=10,random_state=10)
    # Here
    # n_clusters specifies the number of clusters you want the algorithm to find in your data.
    # init determines the method for initializing the positions of the cluster centers (centroids).
    # max_iter=300 Sets the maximum number of iterations the algorithm will run for a single initial
    # n_init=10 Specifies the number of times the KMeans algorithm will run with different centroid
    k_means.fit(normalized_data)
    wcss.append(k_means.inertia_)
```

[illegible]

```
warnings.warn(
C:\Users\Ashvini Mahajan\Anaconda\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: UserWarning: KMeans is known to have
a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the
environment variable OMP_NUM_THREADS=12.
warnings.warn(
C:\Users\Ashvini Mahajan\Anaconda\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: UserWarning: KMeans is known to have
a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the
environment variable OMP_NUM_THREADS=12.
warnings.warn(
```

```
In [21]: ## Elbow Graph
import matplotlib.pyplot as plt
plt.plot(range(1,16),wcss)
plt.xlabel("K value")
plt.ylabel("WCSS")
plt.title("Elbow Curve Graph")
```

Out [21]: Text(0.5, 1.0, 'Elbow Curve Graph')

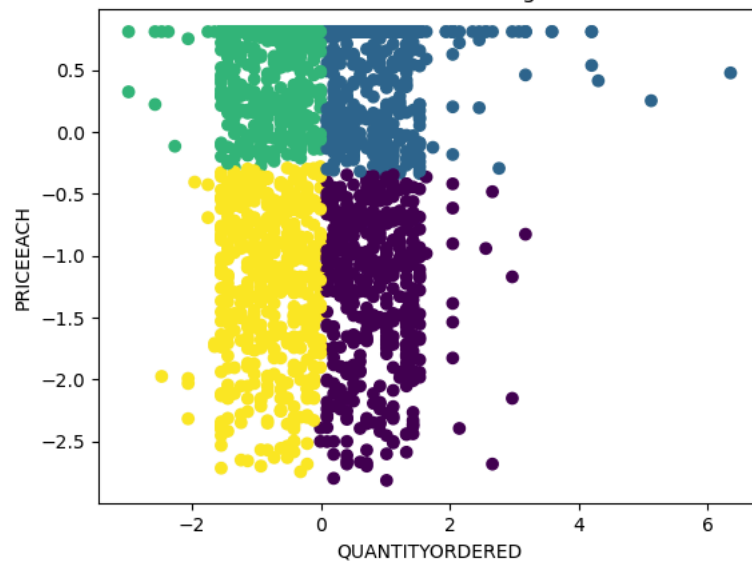


```
In [27]: # from graph we can see that for k=4 is the optimal value , so train the model
k_means=KMeans(n_clusters=4,init='k-means++',max_iter=300,n_init=10,random_state=10)
clusters=k_means.fit_predict(normalized_data)
# The fit_predict method combines the operations of fitting the model and predicting
```

```
C:\Users\Ashvini Mahajan\Anaconda\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: UserWarning: KMeans is known to have
a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the
environment variable OMP_NUM_THREADS=12.
warnings.warn(
```

```
In [29]: ## Visualization of the clusters
plt.scatter(normalized_data[:, 0], normalized_data[:, 1], c=clusters, cmap='viridis')
# x axis numbers =normalized_data[:, 0]
# y axis numbers =normalized_data[:, 1]
# The c parameter specifies the color of the markers (data points) in the scatter plot
# The cmap parameter stands for "colormap." It defines the colormap used to map numerical data to color.
plt.xlabel('QUANTITYORDERED')
plt.ylabel('PRICEEACH')
plt.title('K-Means Clustering')
plt.show()
```

K-Means Clustering



In []: