

project-5

The transactions made by a UK-based, registered, non-store online retailer between December 1, 2010, and December 9, 2011, are all included in the transnational data set known as online retail. The company primarily offers one-of-a-kind gifts for every occasion. The company has a large number of wholesalers as clients. Company Objective Using the global online retail dataset, we will design a clustering model and select the ideal group of clients for the business to target.

```
In [1]: import pandas as pd
        from matplotlib import pyplot as plt
        %matplotlib inline
```

```
In [3]: df=pd.read_csv(r"C:\Users\Mastan Reddy\Downloads\onlineretail.csv")
df
```

Out[3]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Cour
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	6	01-12-2010 08:26	2.55	17850.0	Uni Kingd
1	536365	71053	WHITE METAL LANTERN	6	01-12-2010 08:26	3.39	17850.0	Uni Kingd
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	01-12-2010 08:26	2.75	17850.0	Uni Kingd
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	01-12-2010 08:26	3.39	17850.0	Uni Kingd
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	01-12-2010 08:26	3.39	17850.0	Uni Kingd
...
541904	581587	22613	PACK OF 20 SPACEBOY NAPKINS	12	09-12-2011 12:50	0.85	12680.0	Frai
541905	581587	22899	CHILDREN'S APRON DOLLY GIRL	6	09-12-2011 12:50	2.10	12680.0	Frai
541906	581587	23254	CHILDRENS CUTLERY DOLLY GIRL	4	09-12-2011 12:50	4.15	12680.0	Frai
541907	581587	23255	CHILDRENS CUTLERY CIRCUS PARADE	4	09-12-2011 12:50	4.15	12680.0	Frai
541908	581587	22138	BAKING SET 9 PIECE RETROSPOT	3	09-12-2011 12:50	4.95	12680.0	Frai

541909 rows × 8 columns



In [4]: `df.head()`

Out[4]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	6	01-12-2010 08:26	2.55	17850.0	United Kingdom
1	536365	71053	WHITE METAL LANTERN	6	01-12-2010 08:26	3.39	17850.0	United Kingdom
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	01-12-2010 08:26	2.75	17850.0	United Kingdom
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	01-12-2010 08:26	3.39	17850.0	United Kingdom
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	01-12-2010 08:26	3.39	17850.0	United Kingdom

In [5]: `df.tail()`

Out[5]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Coun
541904	581587	22613	PACK OF 20 SPACEBOY NAPKINS	12	09-12-2011 12:50	0.85	12680.0	Frar
541905	581587	22899	CHILDREN'S APRON DOLLY GIRL	6	09-12-2011 12:50	2.10	12680.0	Frar
541906	581587	23254	CHILDRENS CUTLERY DOLLY GIRL	4	09-12-2011 12:50	4.15	12680.0	Frar
541907	581587	23255	CHILDRENS CUTLERY CIRCUS PARADE	4	09-12-2011 12:50	4.15	12680.0	Frar
541908	581587	22138	BAKING SET 9 PIECE RETROSPOT	3	09-12-2011 12:50	4.95	12680.0	Frar



```
In [6]: df['InvoiceNo'].value_counts()
```

```
Out[6]: 573585      1114
        581219       749
        581492       731
        580729       721
        558475       705
        ...
        569873        1
        C572556        1
        546309        1
        558449        1
        562278        1
        Name: InvoiceNo, Length: 25900, dtype: int64
```

```
In [7]: df['CustomerID'].value_counts()
```

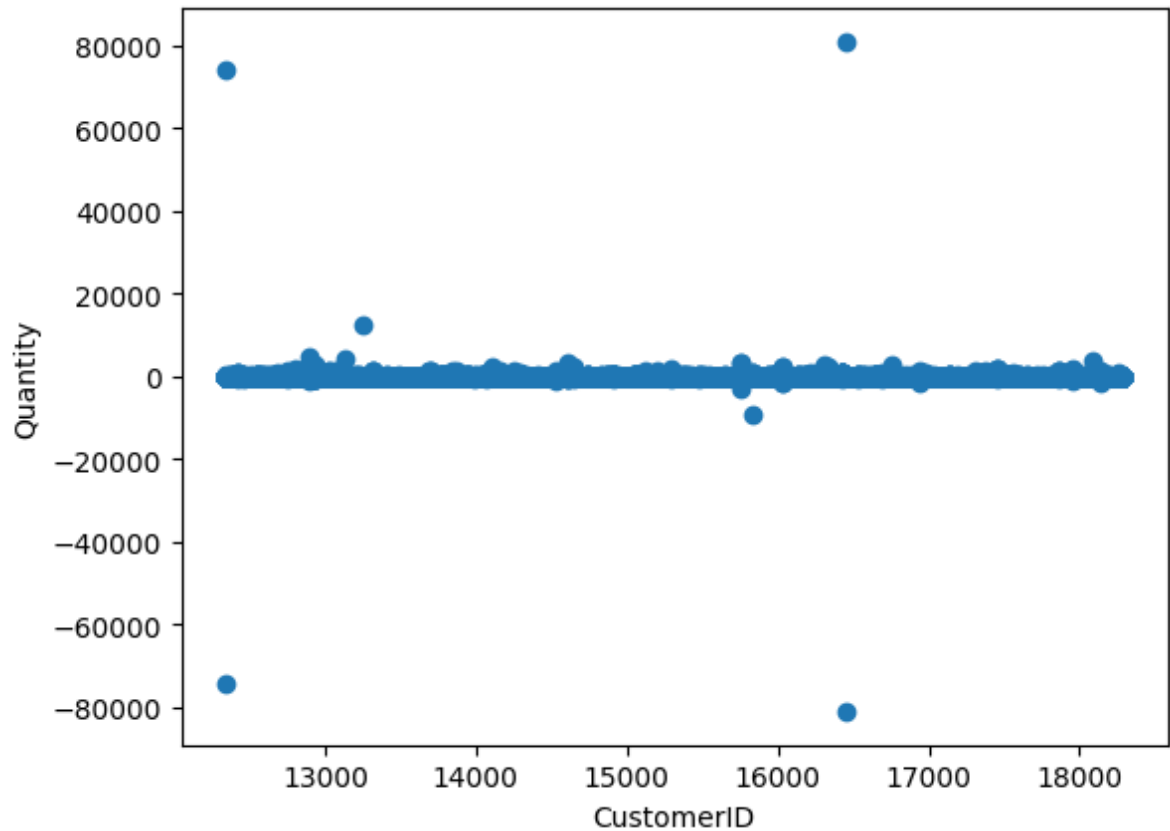
```
Out[7]: 17841.0      7983
        14911.0      5903
        14096.0      5128
        12748.0      4642
        14606.0      2782
        ...
        14351.0        1
        13256.0        1
        12943.0        1
        17923.0        1
        13747.0        1
        Name: CustomerID, Length: 4372, dtype: int64
```

```
In [8]: df['Quantity'].value_counts()
```

```
Out[8]: 1      148227
        2      81829
        12     61063
        6      40868
        4      38484
        ...
        -1681      1
        -657       1
        199        1
        -313       1
        512        1
        Name: Quantity, Length: 722, dtype: int64
```

```
In [9]: plt.scatter(df["CustomerID"],df["Quantity"])
plt.xlabel("CustomerID")
plt.ylabel("Quantity")
```

```
Out[9]: Text(0, 0.5, 'Quantity')
```



```
In [10]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 541909 entries, 0 to 541908
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype  
---  -
0   InvoiceNo        541909 non-null object  
1   StockCode       541909 non-null object  
2   Description     540455 non-null object  
3   Quantity        541909 non-null int64  
4   InvoiceDate     541909 non-null object  
5   UnitPrice       541909 non-null float64 
6   CustomerID     406829 non-null float64 
7   Country        541909 non-null object  
dtypes: float64(2), int64(1), object(5)
memory usage: 33.1+ MB
```

```
In [11]: df.isnull().sum()
```

```
Out[11]: InvoiceNo      0
         StockCode     0
         Description  1454
         Quantity     0
         InvoiceDate    0
         UnitPrice     0
         CustomerID   135080
         Country       0
         dtype: int64
```

```
In [12]: df.fillna(method='ffill',inplace=True)
```

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

```
Out[13]: InvoiceNo      0
         StockCode     0
         Description    0
         Quantity     0
         InvoiceDate    0
         UnitPrice     0
         CustomerID    0
         Country       0
         dtype: int64
```

```
In [14]: from sklearn.cluster import KMeans
         km=KMeans()
         km
```

```
Out[14]: KMeans()
```

```
In [15]: y_predicted=km.fit_predict(df[["CustomerID","Quantity"]])
         y_predicted
```

```
Out[15]: array([3, 3, 3, ..., 4, 4, 4])
```

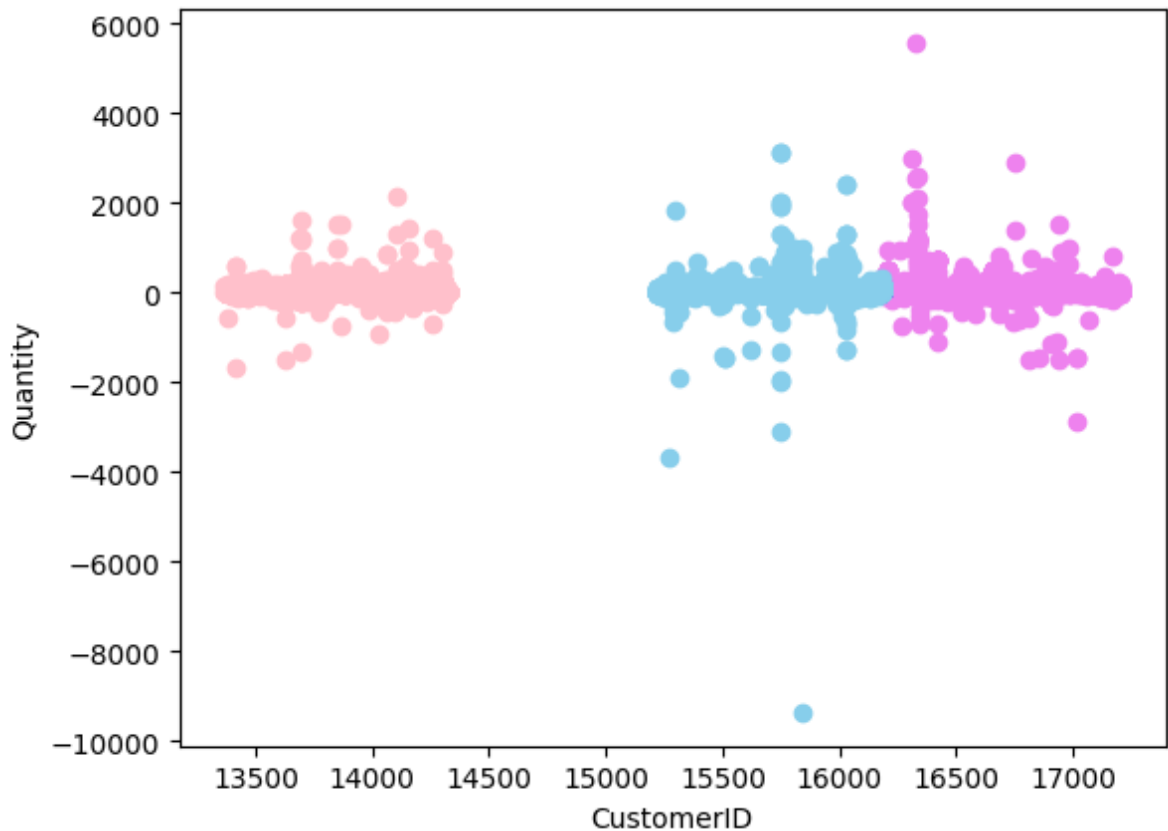
```
In [16]: df["cluster"]=y_predicted  
df.head()
```

Out[16]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	cl
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	6	01-12-2010 08:26	2.55	17850.0	United Kingdom	
1	536365	71053	WHITE METAL LANTERN	6	01-12-2010 08:26	3.39	17850.0	United Kingdom	
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	01-12-2010 08:26	2.75	17850.0	United Kingdom	
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	01-12-2010 08:26	3.39	17850.0	United Kingdom	
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	01-12-2010 08:26	3.39	17850.0	United Kingdom	

```
In [17]: df1=df[df.cluster==0]
df2=df[df.cluster==1]
df3=df[df.cluster==2]
plt.scatter(df1["CustomerID"],df1["Quantity"],color="violet")
plt.scatter(df2["CustomerID"],df2["Quantity"],color="pink")
plt.scatter(df3["CustomerID"],df3["Quantity"],color="skyblue")
plt.xlabel("CustomerID")
plt.ylabel("Quantity")
```

Out[17]: Text(0, 0.5, 'Quantity')




```
In [18]: from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
scaler.fit(df[["Quantity"]])
df["Quantity"]=scaler.transform(df[["Quantity"]])
df.head()
```

Out[18]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	cl
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	0.500037	01-12-2010 08:26	2.55	17850.0	United Kingdom	
1	536365	71053	WHITE METAL LANTERN	0.500037	01-12-2010 08:26	3.39	17850.0	United Kingdom	
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	0.500049	01-12-2010 08:26	2.75	17850.0	United Kingdom	
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	0.500037	01-12-2010 08:26	3.39	17850.0	United Kingdom	
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	0.500037	01-12-2010 08:26	3.39	17850.0	United Kingdom	

```
In [19]: scaler.fit(df[["CustomerID"]])
df["CustomerID"]=scaler.transform(df[["CustomerID"]])
df.head()
```

Out[19]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	cl
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	0.500037	01-12-2010 08:26	2.55	0.926443	United Kingdom	
1	536365	71053	WHITE METAL LANTERN	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	0.500049	01-12-2010 08:26	2.75	0.926443	United Kingdom	
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	

K-MEANS CLUSTURING

```
In [20]: km=KMeans()
```

```
In [21]: y_predicted=km.fit_predict(df[["CustomerID","Quantity"]])
y_predicted
```

Out[21]: array([3, 3, 3, ..., 5, 5, 5])

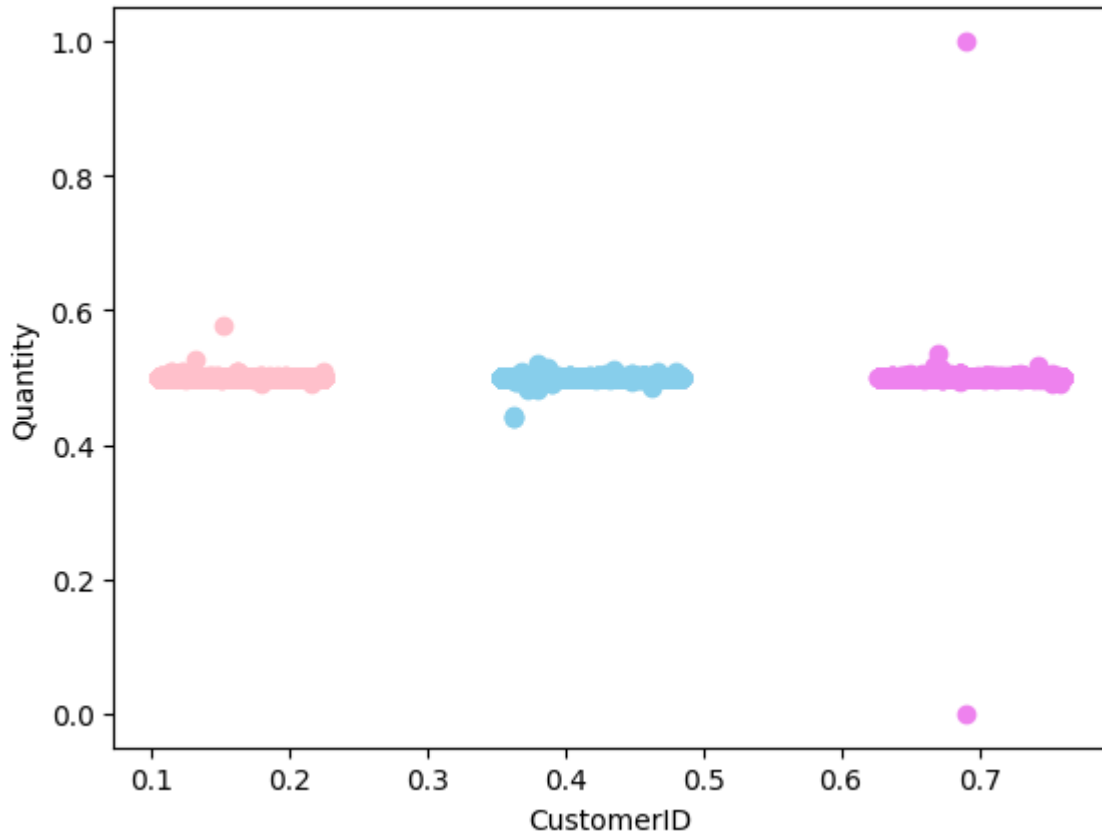
```
In [22]: df["New Cluster"]=y_predicted
df.head()
```

Out[22]:

	InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country	cl
0	536365	85123A	WHITE HANGING HEART T- LIGHT HOLDER	0.500037	01-12-2010 08:26	2.55	0.926443	United Kingdom	
1	536365	71053	WHITE METAL LANTERN	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	0.500049	01-12-2010 08:26	2.75	0.926443	United Kingdom	
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	0.500037	01-12-2010 08:26	3.39	0.926443	United Kingdom	

```
In [23]: df1=df[df["New Cluster"]==0]
df2=df[df["New Cluster"]==1]
df3=df[df["New Cluster"]==2]
plt.scatter(df1["CustomerID"],df1["Quantity"],color="violet")
plt.scatter(df2["CustomerID"],df2["Quantity"],color="pink")
plt.scatter(df3["CustomerID"],df3["Quantity"],color="skyblue")
plt.xlabel("CustomerID")
plt.ylabel("Quantity")
```

Out[23]: Text(0, 0.5, 'Quantity')



```
In [24]: km.cluster_centers_
```

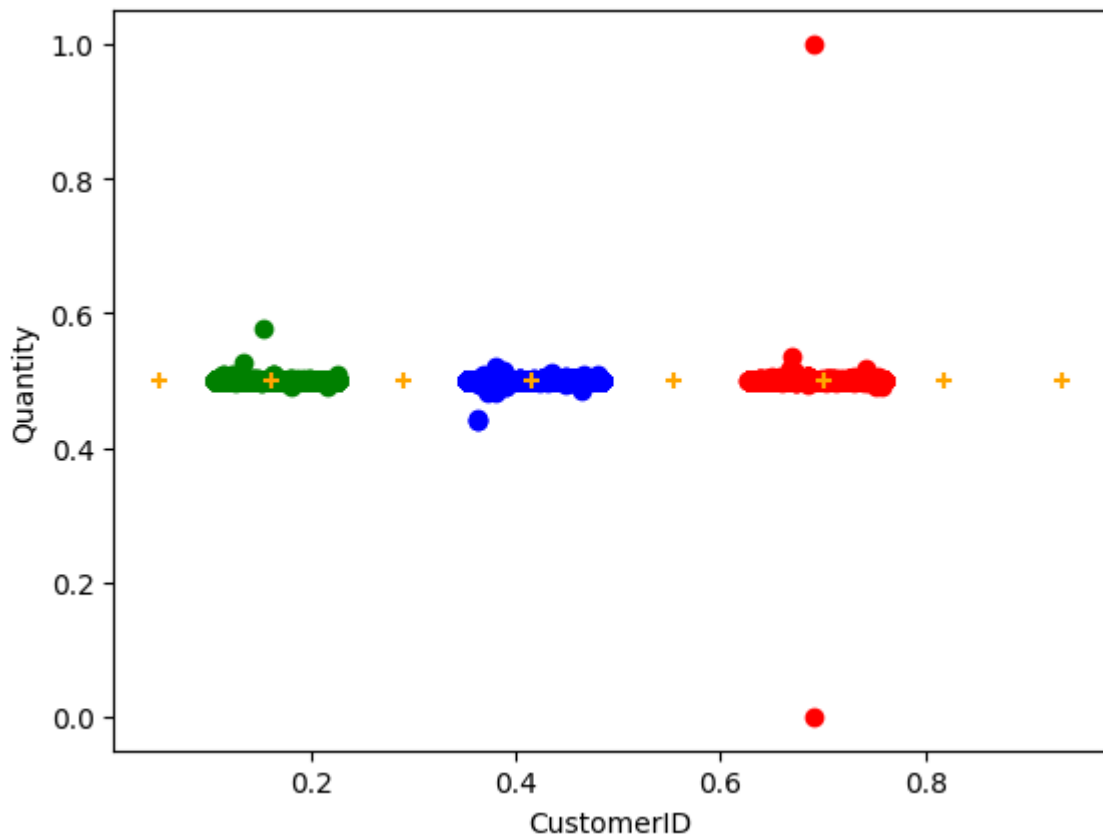
```
Out[24]: array([[0.70077245, 0.50005781],
 [0.16030912, 0.50005651],
 [0.41539441, 0.50005966],
 [0.93308721, 0.50005101],
 [0.55361958, 0.50005385],
 [0.05118569, 0.5000668 ],
 [0.81857738, 0.50006025],
 [0.29096734, 0.50006607]])
```

```

In [25]: df1=df[df["New Cluster"]==0]
df2=df[df["New Cluster"]==1]
df3=df[df["New Cluster"]==2]
plt.scatter(df1["CustomerID"],df1["Quantity"],color="red")
plt.scatter(df2["CustomerID"],df2["Quantity"],color="green")
plt.scatter(df3["CustomerID"],df3["Quantity"],color="blue")
plt.scatter(km.cluster_centers_[0],km.cluster_centers_[1],color="orange",marker='x')
plt.xlabel("CustomerID")
plt.ylabel("Quantity")

```

Out[25]: Text(0, 0.5, 'Quantity')



```

In [ ]: k_rng=range(1,10)
sse=[]

```

For the given dataset we use K-means Clustering and done the grouping based on the given data. In the above dataset we will take customer id and quantity based on that we make the clusters. When the K-value is low error rate is more and the K-value is high error rate is very high. So, finally we can Conclude the above dataset is bestfit for K-Means.

In []: