Project - 5 (DATASET: Online Retail) 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 ObjectiveUsing 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\sruth\Downloads\OnlineRetail1.csv")
df

Out[3]:

| | InvoiceNo | StockCode | Description | Quantity | InvoiceDate | UnitPrice | CustomerID |
|----------|-------------|-----------|---|----------|---------------------|-----------|------------|
| 0 | 536365 | 85123A | WHITE HANGING HEART T- LIGHT HOLDER | 6 | 01-12-2010 08:26 | 2.55 | 17850.0 |
| 1 | 536365 | 71053 | WHITE METAL LANTERN | 6 | 01-12-2010 08:26 | 3.39 | 17850.0 |
| 2 | 536365 | 84406B | CREAM CUPID HEARTS COAT HANGER | 8 | 01-12-2010 08:26 | 2.75 | 17850.0 |
| 3 | 536365 | 84029G | KNITTED UNION FLAG HOT WATER BOTTLE | 6 | 01-12-2010 08:26 | 3.39 | 17850.0 |
| 4 | 536365 | 84029E | RED WOOLLY HOTTIE WHITE HEART. | 6 | 01-12-2010 08:26 | 3.39 | 17850.0 |
| | | | | | | | |
| 541904 | 581587 | 22613 | PACK OF 20 SPACEBOY NAPKINS | 12 | 09-12-2011 12:50 | 0.85 | 12680.0 |
| 541905 | 581587 | 22899 | CHILDREN'S APRON DOLLY GIRL | 6 | 09-12-2011 12:50 | 2.10 | 12680.0 |
| 541906 | 581587 | 23254 | CHILDRENS CUTLERY DOLLY GIRL | 4 | 09-12-2011 12:50 | 4.15 | 12680.0 |
| 541907 | 581587 | 23255 | CHILDRENS CUTLERY CIRCUS PARADE | 4 | 09-12-2011 12:50 | 4.15 | 12680.0 |
| 541908 | 581587 | 22138 | BAKING SET 9 PIECE RETROSPOT | 3 | 09-12-2011 12:50 | 4.95 | 12680.0 |
| 541909 i | rows × 8 co | lumns | | | | | |
| 4 | | | | | | | • |

In [4]:

df.head()

Out[4]:

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

In [5]:

df.tail()

Out[5]:

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

```
In [6]:
```

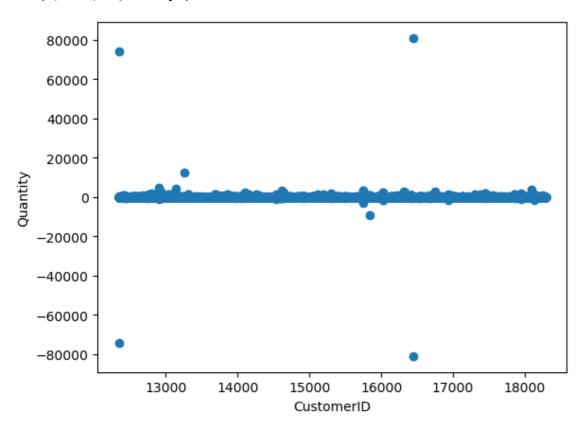
```
df['InvoiceNo'].value_counts()
Out[6]:
573585
           1114
581219
            749
581492
            731
580729
            721
558475
            705
           ...
554023
              1
554022
              1
554021
              1
               1
554020
C558901
              1
Name: InvoiceNo, 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
15070.0
              1
15753.0
               1
17065.0
16881.0
16995.0
               1
Name: CustomerID, Length: 4372, dtype: int64
In [8]:
df['Quantity'].value_counts()
Out[8]:
 1
          148227
 2
           81829
 12
           61063
 6
           40868
           38484
-472
               1
               1
-161
-1206
               1
                1
-272
-80995
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):

| Duca | CO_U | a | | | | | | | |
|------------------------|--|-----------------|---------|--|--|--|--|--|--|
| # | 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 | | | | | | |
| dtype | <pre>dtypes: float64(2), int64(1), object(5)</pre> | | | | | | | | |
| 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
KMeans()
```

In [15]:

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

C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(

Out[15]:

array([0, 0, 0, ..., 2, 2, 2])

In [17]:

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

Out[17]:

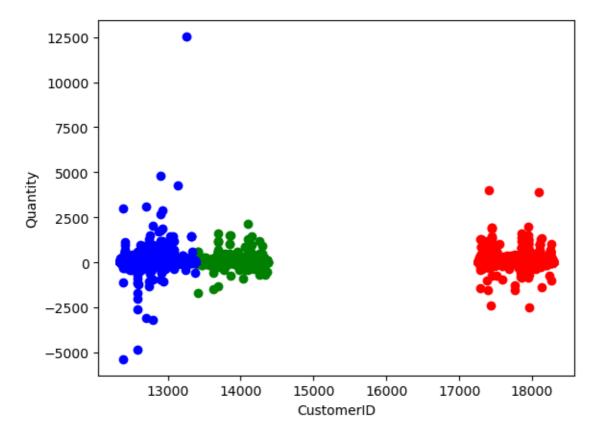
| | InvoiceNo | StockCode | Description | Quantity | InvoiceDate | UnitPrice | CustomerID | Countr |
|---|-----------|-----------|---|----------|---------------------|-----------|------------|------------------|
| 0 | 536365 | 85123A | WHITE HANGING HEART T- LIGHT HOLDER | 6 | 01-12-2010 08:26 | 2.55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

In [18]:

```
df1=df[df.cluster==0]
df2=df[df.cluster==1]
df3=df[df.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.xlabel("CustomerID")
plt.ylabel("Quantity")
```

Out[18]:

Text(0, 0.5, 'Quantity')



In [19]:

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

Out[19]:

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

```
In [20]:
```

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

Out[20]:

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

K-MeansClustering

```
In [22]:
```

```
km=KMeans()
```

```
In [23]:
```

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

C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(

```
Out[23]:
```

```
array([1, 1, 1, ..., 6, 6, 6])
```

In [24]:

df["New Cluster"]=y_predicted
df.head()

Out[24]:

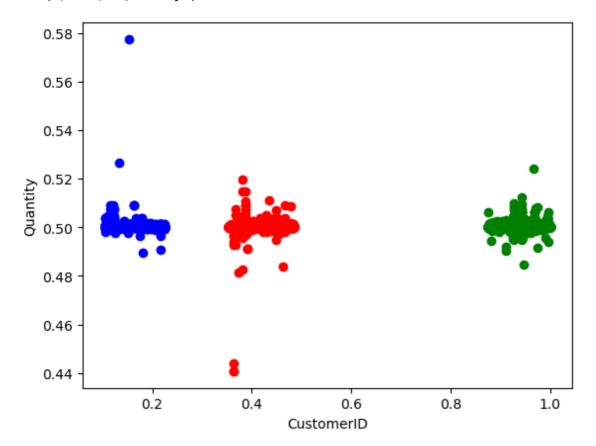
| | InvoiceNo | StockCode | Description | Quantity | InvoiceDate | UnitPrice | CustomerID | Countr |
|---|-----------|-----------|---|----------|---------------------|-----------|------------|------------------|
| 0 | 536365 | 85123A | WHITE HANGING HEART T- LIGHT HOLDER | 0.500037 | 01-12-2010 08:26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

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.xlabel("CustomerID")
plt.ylabel("Quantity")
```

Out[25]:

Text(0, 0.5, 'Quantity')



In [26]:

```
km.cluster_centers_
```

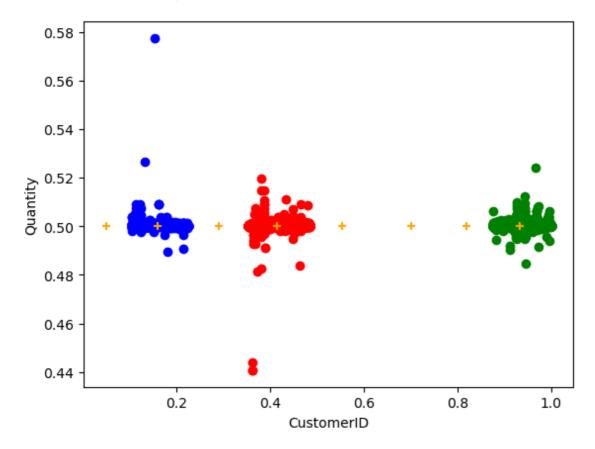
Out[26]:

In [27]:

```
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="+")
plt.xlabel("CustomerID")
plt.ylabel("Quantity")
```

Out[27]:

Text(0, 0.5, 'Quantity')



In [28]:

```
k_rng=range(1,10)
sse=[]
```

```
In [29]:
```

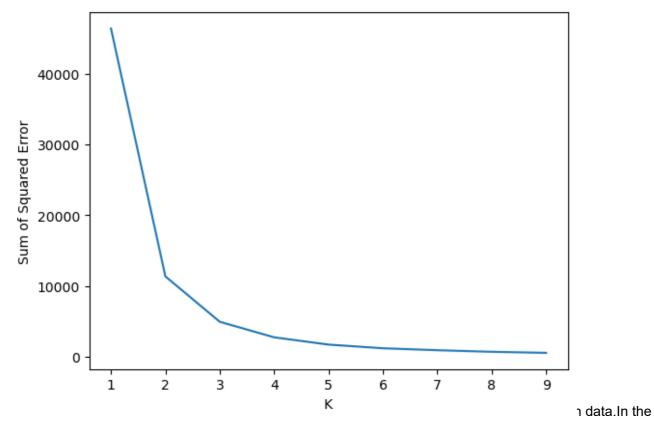
```
for k in k_rng:
    km=KMeans(n_clusters=k)
    km.fit(df[["CustomerID","Quantity"]])
    sse.append(km.inertia_)
#km.inertia_ will give you the value of sum of square error
print(sse)
plt.plot(k_rng,sse)
plt.xlabel("K")
plt.ylabel("Sum of Squared Error")
```

- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
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- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
- C:\Users\sruth\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
 0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(

[46374.84553398371, 11336.06582016775, 4921.040005132341, 2723.5169350354 81, 1695.0703798884356, 1178.5471813165186, 903.0078177312912, 677.262375 9205824, 528.3702588048828]

Out[29]:

Text(0, 0.5, 'Sum of Squared Error')



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 []: