

```
In [121... import pandas as pd
from sklearn.cluster import KMeans
from sklearn.model_selection import train_test_split
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from sklearn_extra.cluster import KMedoids
```

```
In [122... df=pd.read_csv("C:/Users/NITISH BOKKA/Downloads/archive (16)/sales_data_sample.c
```

```
In [123... df.isnull().sum()
```

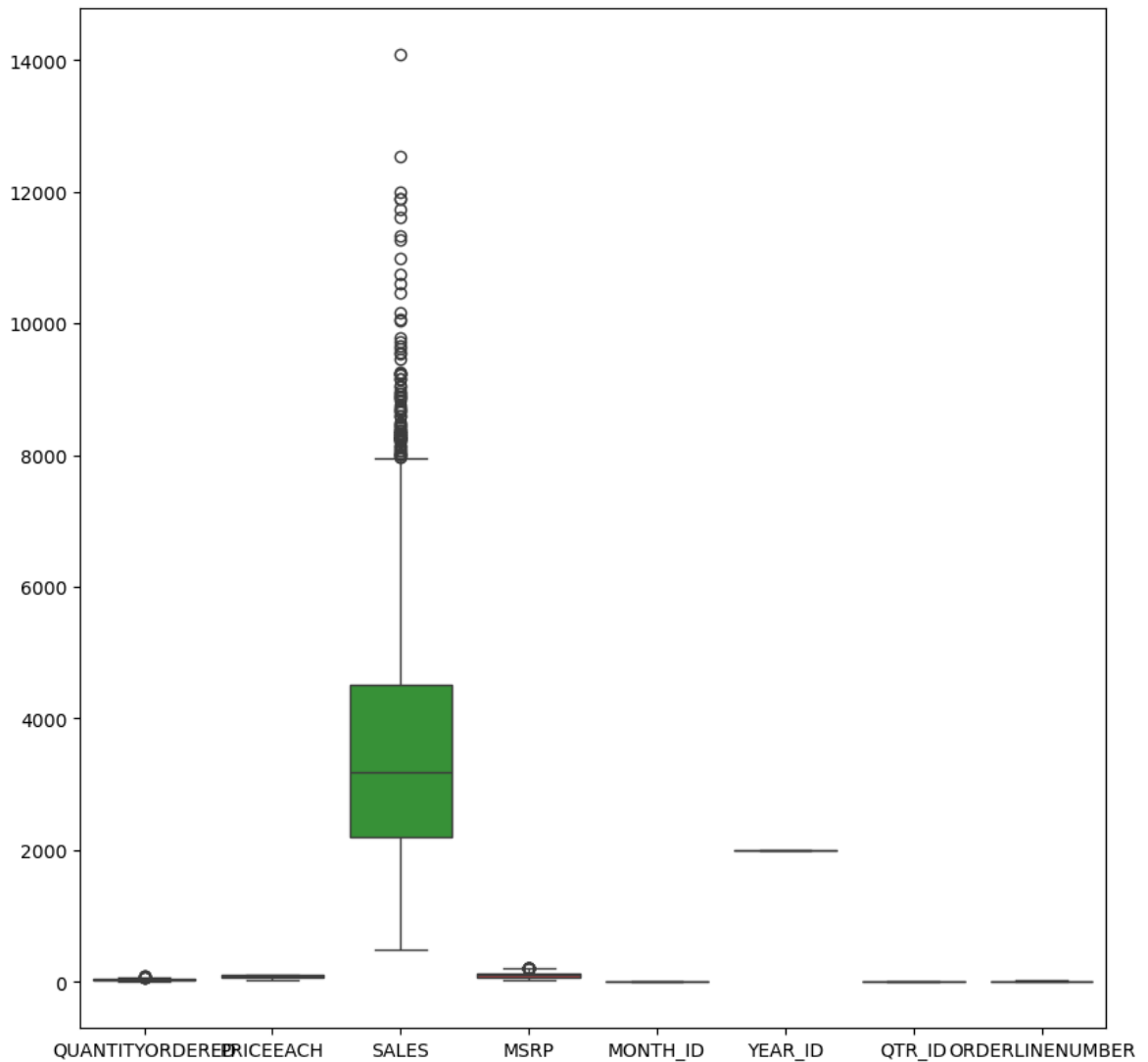
```
Out[123... ORDERNUMBER          0
QUANTITYORDERED       0
PRICEEACH             0
ORDERLINENUMBER       0
SALES                 0
ORDERDATE             0
STATUS               0
QTR_ID               0
MONTH_ID             0
YEAR_ID              0
PRODUCTLINE          0
MSRP                 0
PRODUCTCODE          0
CUSTOMERNAME         0
PHONE                0
ADDRESSLINE1         0
ADDRESSLINE2         0
CITY                 0
STATE                1486
POSTALCODE           76
COUNTRY              0
TERRITORY            1074
CONTACTLASTNAME      0
CONTACTFIRSTNAME     0
DEALSIZE             0
dtype: int64
```

```
In [124... df['ADDRESSLINE2']=df['ADDRESSLINE2'].bfill()
df['POSTALCODE']=df['POSTALCODE'].ffill()
df['STATE']=df['STATE'].ffill()
df['TERRITORY']=df['TERRITORY'].bfill()
```

```
In [125... df.isnull().sum()
```

```
Out[125... 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     5
            CITY             0
            STATE            0
            POSTALCODE       0
            COUNTRY          0
            TERRITORY        1
            CONTACTLASTNAME  0
            CONTACTFIRSTNAME 0
            DEALSIZE         0
            dtype: int64
```

```
In [126... num_cols=df[['QUANTITYORDERED', 'PRICEEACH', 'SALES', 'MSRP', 'MONTH_ID', 'YEAR_ID', '
plt.figure(figsize=(10,10))
sns.boxplot(num_cols)
plt.show()
```



```
In [127... Q1=df['QUANTITYORDERED'].quantile(0.25)
Q3=df['QUANTITYORDERED'].quantile(0.75)
IQR=Q3-Q1
lower_quartile=Q1-1.5*IQR
upper_quartile=Q3+1.5*IQR
df=df[(df['QUANTITYORDERED']>=lower_quartile) & (df['QUANTITYORDERED']<=upper_qu
```

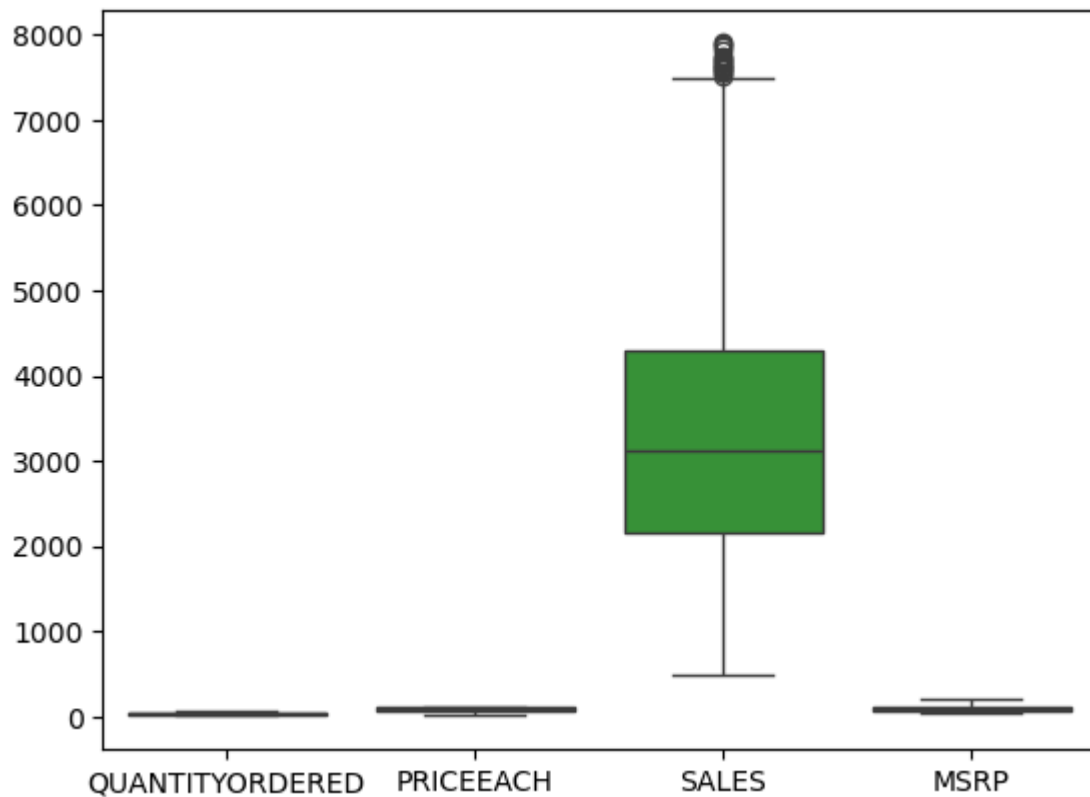
```
In [128... Q1=df['PRICEEACH'].quantile(0.25)
Q3=df['PRICEEACH'].quantile(0.75)
IQR=Q3-Q1
lower_quartile=Q1-1.5*IQR
upper_quartile=Q3+1.5*IQR
df=df[(df['PRICEEACH']>=lower_quartile) & (df['PRICEEACH']<=upper_quartile)]
```

```
In [129... Q1=df['SALES'].quantile(0.25)
Q3=df['SALES'].quantile(0.75)
IQR=Q3-Q1
lower_quartile=Q1-1.5*IQR
upper_quartile=Q3+1.5*IQR
df=df[(df['SALES']>=lower_quartile) & (df['SALES']<=upper_quartile)]
```

```
In [130... Q1=df['MSRP'].quantile(0.25)
Q3=df['MSRP'].quantile(0.75)
IQR=Q3-Q1
lower_quartile=Q1-1.5*IQR
```

```
upper_quartile=Q3+1.5*IQR
df=df[(df['MSRP']>=lower_quartile) & (df['MSRP']<=upper_quartile)]
```

```
In [131... num_cols=df[['QUANTITYORDERED', 'PRICEEACH', 'SALES', 'MSRP']]
sns.boxplot(num_cols)
plt.show()
```



```
In [132... X=df[['QUANTITYORDERED', 'PRICEEACH', 'SALES', 'MSRP', 'MONTH_ID', 'YEAR_ID', 'QTR_ID']]
scaler=StandardScaler()
X_scaled=scaler.fit_transform(X)
```

```
In [133... k_range=range(1,21)
inertia_list=[]
for i in k_range:
    kmn=KMeans(n_clusters=i, random_state=42)
    kmn.fit(X_scaled)
    inertia_list.append(kmn.inertia_)
    print(i, kmn.inertia_)
```

```

1 21608.000000000007
2 16641.968372583986
3 13612.012474606314
4 12084.10539529429
5 10979.658325322398
6 10208.235629943432
7 9695.764847973925
8 9198.201876745454
9 8740.722878030774
10 8449.597922175664
11 8069.016953571581
12 7788.595622825555
13 7542.94850594196
14 7293.653760449979
15 7102.3075339065435
16 6963.875367712442
17 6886.105986466975
18 6667.932645102546
19 6504.34582857871
20 6407.093498463479

```

```

In [134... print("KMeans using Euclidean")
            kmn=KMeans(n_clusters=16)
            kmn.fit(X_scaled)
            print(kmn.inertia_)

```

```

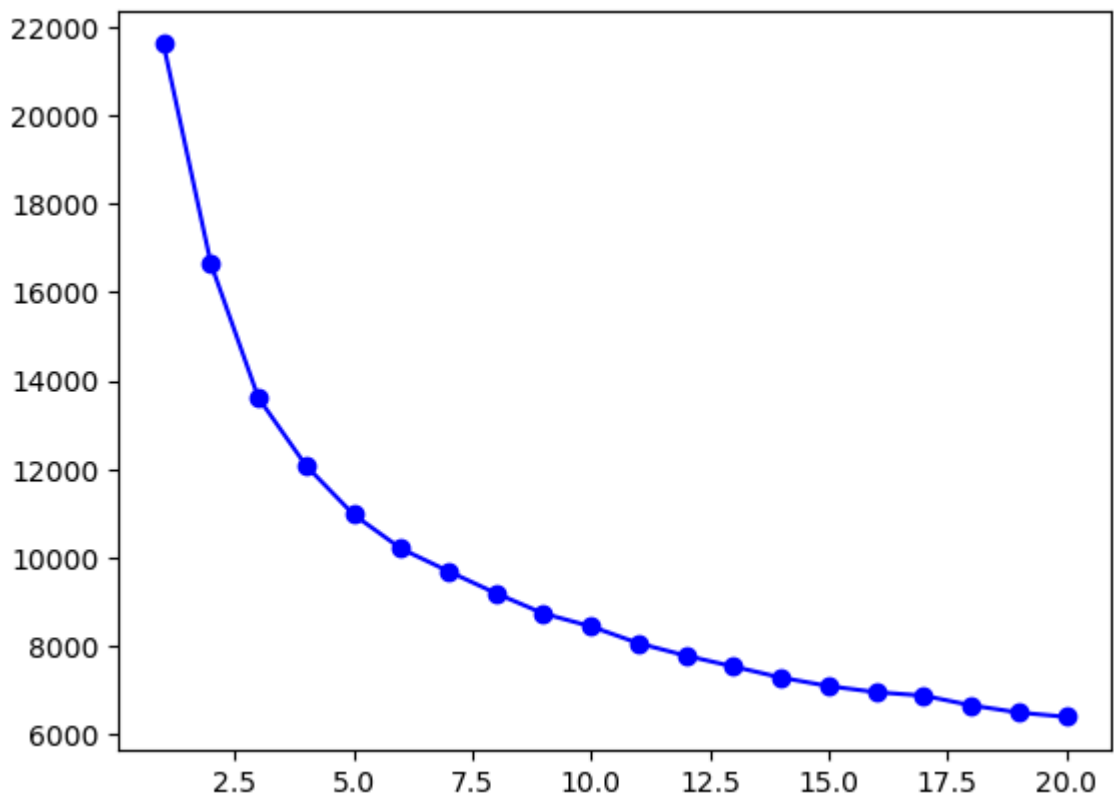
KMeans using Euclidean
7030.311017948243

```

```

In [135... plt.plot(k_range, inertia_list,'bo-')
            plt.show()

```



```

In [136... k_range=range(1,21)
            k_manhattan=[]
            inertia_list=[]

```

```

for i in k_range:
    kmn=KMedoids(n_clusters=i, metric='manhattan', random_state=42)
    kmn.fit(X_scaled)
    k_manhattan.append(kmn.inertia_)
    print(i, kmn.inertia_)

```

```

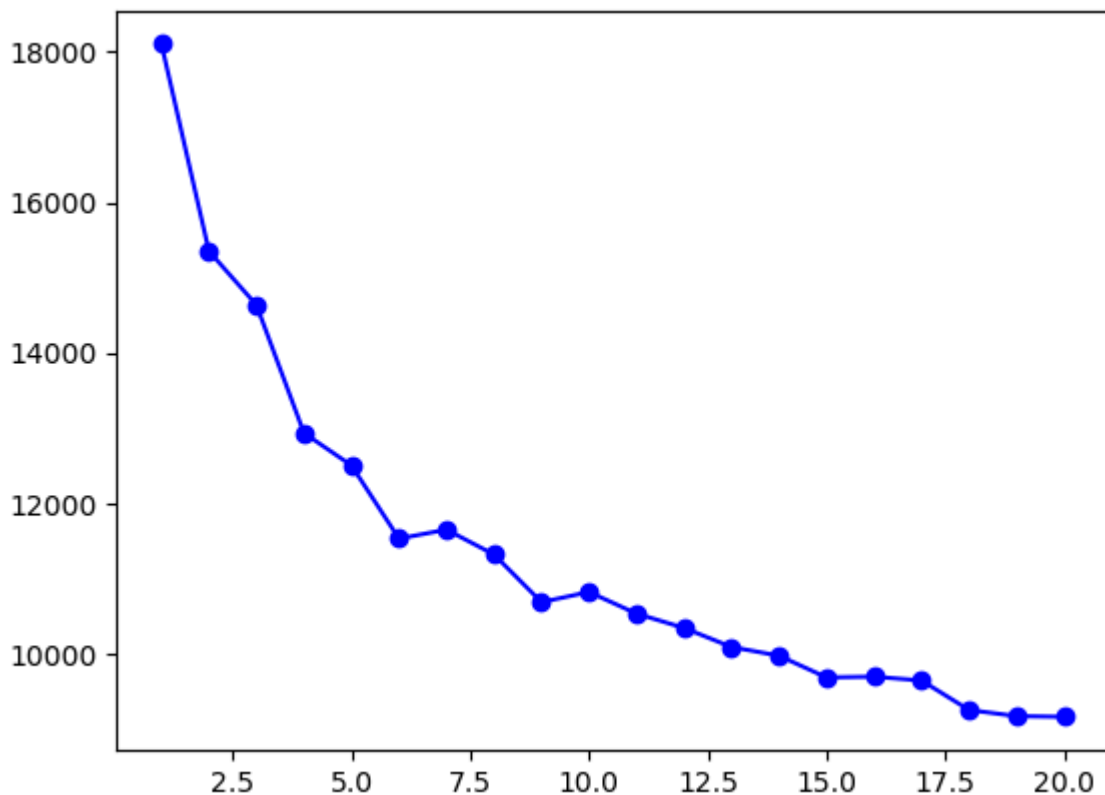
1 18105.313940642394
2 15351.000711074905
3 14633.84880579897
4 12932.104647630194
5 12500.296195025103
6 11527.150473404772
7 11646.616961969872
8 11316.27791752853
9 10682.229800159712
10 10817.487452167577
11 10529.845008100248
12 10339.20748714991
13 10085.709608251495
14 9969.896468579764
15 9680.933295636409
16 9691.009555800505
17 9640.34806520279
18 9248.003626357504
19 9166.397880454051
20 9157.289220324637

```

```

In [137... plt.plot(k_range, k_manhattan, 'bo-')
plt.show()

```



```

In [171... print("KMeans using Manhattan")
kmn=KMedoids(n_clusters=18, metric='manhattan', random_state=42)
kmn.fit(X_scaled)
print(kmn.inertia_)

```

KMeans using Manhattan
9248.003626357504

In []: