

In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
df = pd.read_csv('Black_Friday_sale.csv')
df
```

Out[2]:

	User_ID	Product_ID	Gender	Age	Occupation	City_Category	Stay_In_Current_City_
0	1000001	P00069042	F	0-17	10	A	
1	1000001	P00248942	F	0-17	10	A	
2	1000001	P00087842	F	0-17	10	A	
3	1000001	P00085442	F	0-17	10	A	
4	1000002	P00285442	M	55+	16	C	
...
550063	1006033	P00372445	M	51-55	13	B	
550064	1006035	P00375436	F	26-35	1	C	
550065	1006036	P00375436	F	26-35	15	B	
550066	1006038	P00375436	F	55+	1	C	
550067	1006039	P00371644	F	46-50	0	B	

550068 rows × 12 columns



In [3]:

```
df.describe
```

Out[3]:

<bound method NDFrame.describe of					User_ID	Product_ID	Gender	Age
Occupation	City_Category	\						
0	1000001	P00069042	F	0-17	10		A	
1	1000001	P00248942	F	0-17	10		A	
2	1000001	P00087842	F	0-17	10		A	
3	1000001	P00085442	F	0-17	10		A	
4	1000002	P00285442	M	55+	16		C	
...	
550063	1006033	P00372445	M	51-55	13		B	
550064	1006035	P00375436	F	26-35	1		C	
550065	1006036	P00375436	F	26-35	15		B	
550066	1006038	P00375436	F	55+	1		C	
550067	1006039	P00371644	F	46-50	0		B	
Stay_In_Current_City_Years				Marital_Status	Product_Category_1 \			
0			2		0		3	
1			2		0		1	
2			2		0		12	
3			2		0		12	
4			4+		0		8	
...			
550063			1		1		20	
550064			3		0		20	
550065			4+		1		20	
550066			2		0		20	
550067			4+		1		20	
Product_Category_2			Product_Category_3	Purchase				
0		NaN		NaN	8370			
1		6.0		14.0	15200			
2		NaN		NaN	1422			
3		14.0		NaN	1057			
4		NaN		NaN	7969			
...				
550063		NaN		NaN	368			
550064		NaN		NaN	371			
550065		NaN		NaN	137			
550066		NaN		NaN	365			
550067		NaN		NaN	490			

[550068 rows x 12 columns]>

In [4]:

```
df.head()
```

Out[4]:

	User_ID	Product_ID	Gender	Age	Occupation	City_Category	Stay_In_Current_City_Years
0	1000001	P00069042	F	0-17	10	A	2
1	1000001	P00248942	F	0-17	10	A	2
2	1000001	P00087842	F	0-17	10	A	2
3	1000001	P00085442	F	0-17	10	A	2
4	1000002	P00285442	M	55+	16	C	4+

In [5]:

```
df.columns
```

Out[5]:

```
Index(['User_ID', 'Product_ID', 'Gender', 'Age', 'Occupation', 'City_Category',
      'Stay_In_Current_City_Years', 'Marital_Status', 'Product_Category_1',
      'Product_Category_2', 'Product_Category_3', 'Purchase'],
      dtype='object')
```

In [6]:

```
df.nunique()
```

Out[6]:

```
User_ID          5891
Product_ID       3631
Gender            2
Age              7
Occupation       21
City_Category     3
Stay_In_Current_City_Years  5
Marital_Status    2
Product_Category_1  20
Product_Category_2  17
Product_Category_3  15
Purchase        18105
dtype: int64
```

In [7]:

```
df.isna().sum()
```

Out[7]:

```
User_ID          0
Product_ID       0
Gender           0
Age              0
Occupation       0
City_Category    0
Stay_In_Current_City_Years  0
Marital_Status   0
Product_Category_1  0
Product_Category_2 173638
Product_Category_3 383247
Purchase         0
dtype: int64
```

In [8]:

```
df.drop('Product_Category_2',axis=1,inplace=True)
df.drop('Product_Category_3',axis=1,inplace=True)
```

In [9]:

```
from sklearn.preprocessing import LabelEncoder

labenc=LabelEncoder()
```

In [10]:

```
df['Gender']=labenc.fit_transform(df['Gender'])
df['Gender']
```

Out[10]:

```
0      0
1      0
2      0
3      0
4      1
..
550063  1
550064  0
550065  0
550066  0
550067  0
Name: Gender, Length: 550068, dtype: int32
```

In [11]:

```
df['Age']=labenc.fit_transform(df['Age'])  
df['Age']
```

Out[11]:

```
0      0  
1      0  
2      0  
3      0  
4      6  
..  
550063  5  
550064  2  
550065  2  
550066  6  
550067  4  
Name: Age, Length: 550068, dtype: int32
```

In [12]:

```
selected_col=['Age','Purchase']
```

In [13]:

```
from sklearn.preprocessing import StandardScaler  
  
scaler=StandardScaler()  
scaleddf=scaler.fit_transform(df[selected_col])  
scaleddf
```

Out[13]:

```
array([[ -1.84424754,  -0.1779729 ],  
       [ -1.84424754,   1.1817558 ],  
       [ -1.84424754,  -1.56119326],  
       ...,  
       [ -0.36673935,  -1.81701338],  
       [  2.58827703,  -1.77162273],  
       [  1.11076884,  -1.7467375 ]])
```

In [14]:

```

from sklearn.cluster import KMeans
inertia_value=[]
k_values=range(1,11)

for i in k_values:
    kmeans=KMeans(n_clusters=i,random_state=0)
    kmeans.fit(scaledddf)
    inertia_value.append(kmeans.inertia_)

    # Plot the Elbow method to choose the optimal number of clusters
plt.plot(k_values,inertia_value,marker='o')
plt.title('Elbow Method for Optimal Number of Clusters')
plt.xlabel('No: of clusters(k)')
plt.ylabel('Inertia')
plt.show()

```

modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here.

Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations> (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)

dtype = np.float

C:\Users\hp\Anaconda3\lib\site-packages\sklearn\metrics\pairwise.py:56: DeprecationWarning: `np.float` is a deprecated alias for the builtin `float`. To silence this warning, use `float` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here.

Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations> (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)

dtype = np.float

C:\Users\hp\Anaconda3\lib\site-packages\sklearn\metrics\pairwise.py:56: DeprecationWarning: `np.float` is a deprecated alias for the builtin `float`. To silence this warning, use `float` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here.

Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations> (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)

In [16]:

```

k=5
kmeans=KMeans(n_clusters=5,random_state=0)
df['cluster']=kmeans.fit_predict(df[selected_col])
df.head()

```

[rg/devdocs/release/1.20.0-notes.html#deprecations](https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations) (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)
 dtype = np.float
 C:\Users\hp\Anaconda3\lib\site-packages\sklearn\metrics\pairwise.py:56:
 DeprecationWarning: `np.float` is a deprecated alias for the builtin `float`. To silence this warning, use `float` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here.
 Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations> (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)
 dtype = np.float
 C:\Users\hp\Anaconda3\lib\site-packages\sklearn\metrics\pairwise.py:56:
 DeprecationWarning: `np.float` is a deprecated alias for the builtin `float`. To silence this warning, use `float` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here.
 Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations> (<https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>)

In [17]:

```

k=5
cluster_characteristics = []
for cluster_id in range(k):
    cluster_data = df[df['cluster']==cluster_id]

    dominant_age_group=cluster_data['Age'].mode()[0]
    product_of_interest=cluster_data['Product_ID'].mode()[0]

    cluster_characteristics.append({
        'Cluster': cluster_id,
        'Dominant_Age_Group': dominant_age_group,
        'Product_Of_Interest': product_of_interest
    })

cluster_characteristics_df=pd.DataFrame(cluster_characteristics)
cluster_characteristics_df

```

Out[17]:

	Cluster	Dominant_Age_Group	Product_Of_Interest
0	0	2	P00145042
1	1	2	P00265242
2	2	2	P00051442
3	3	2	P00025442
4	4	2	P00102642

In [18]:

```
df.groupby('Age')['Product_ID'].agg(pd.Series.mode).to_frame()
```

Out[18]:

	Product_ID
Age	
0	P00255842
1	P00265242
2	P00265242
3	P00025442
4	P00265242
5	P00265242
6	P00265242

In []:

```
df.head()
```

In [21]:

```
max_sales_category=df.groupby('Product_Category_1')['Purchase'].sum().idxmax()
max_sales_value=df.groupby('Product_Category_1')['Purchase'].sum().max()

print(" Category of Product with Highest Sales: Product Category", max_sales_category)
print("Highest Sales Value:", max_sales_value)
```

Category of Product with Highest Sales: Product Category 1
Highest Sales Value: 1910013754

In [20]:

```
max_sales_product=df.groupby('Product_ID')['Purchase'].sum().idxmax()
max_sales_value=df.groupby('Product_ID')['Purchase'].sum().max()

print(" ID of Highest Sales Product:", max_sales_product)
print(" Sales Value:", max_sales_value)
```

Product ID with Highest Sales: P00025442
Highest Sales Value: 27995166

In []: