```
import pandas as pd
 In [1]:
           import numpy as np
           import matplotlib.pyplot as plt
           import seaborn as sns
           from sklearn.preprocessing import MinMaxScaler, StandardScaler
           df = pd_read_csv("TechElectro_Customer_Data.csv")
In [25]:
In [26]:
           df.head()
Out[26]:
                                                            AnnualIncome
              CustomaleerID Age Gender MaritalStatus
                                                                            TotalPurchases PreferredCategory
                                                                    (USD)
           0
                        1001
                                58
                                       male
                                                  Divorced
                                                                    73598
                                                                                        53
                                                                                                    Appliances
           1
                        1002
                                                   Married
                                                                    31717
                                                                                        87
                                                                                                    Appliances
                                32
                                       male
           2
                        1003
                                55
                                     Female
                                                  Divorced
                                                                    26952
                                                                                        29
                                                                                                    Appliances
           3
                        1004
                                                   Married
                                                                    38031
                                                                                       87
                                                                                                    Electronics
                                       male
                                32
           4
                        1005
                                                   Married
                                                                    43231
                                                                                        18
                                                                                                    Appliances
                                32
                                     Female
In [27]:
           df.isnull().sum()
                               0
           CustomaleerID
Out[27]:
           Age
                            0
           Gender
                             0
           MaritalStatus
           AnnualIncome (USD)
           TotalPurchases
           PreferredCategory
           dtype: int64
In [28]:
           numerical_cols = ['Age', 'AnnualIncome (USD)', 'TotalPurchases']
           minmax_scaler = MinMaxScaler()
           standard_scaler = StandardScaler()
           df[numerical_cols] = minmax_scaler.fit_transform(df[numerical_cols])
           df.head()
In [29]:
Out[29]:
                                                                AnnualIncome
              CustomaleerID
                                  Age Gender MaritalStatus
                                                                                TotalPurchases PreferredCategory
                                                                        (USD)
           0
                              0.851064
                        1001
                                           male
                                                      Divorced
                                                                     0.745837
                                                                                      0.525641
                                                                                                       Appliances
                        1002
                              0.297872
                                           male
                                                       Married
                                                                     0.096459
                                                                                      0.961538
                                                                                                       Appliances
           2
                        1003
                              0.787234
                                                      Divorced
                                                                     0.022576
                                                                                      0.217949
                                                                                                       Appliances
                                         Female
           3
                        1004
                              0.297872
                                                                                      0.961538
                                                                                                        Electronics
                                           male
                                                       Married
                                                                     0.194359
           4
                        1005 0.297872
                                         Female
                                                       Married
                                                                     0.274987
                                                                                      0.076923
                                                                                                       Appliances
```

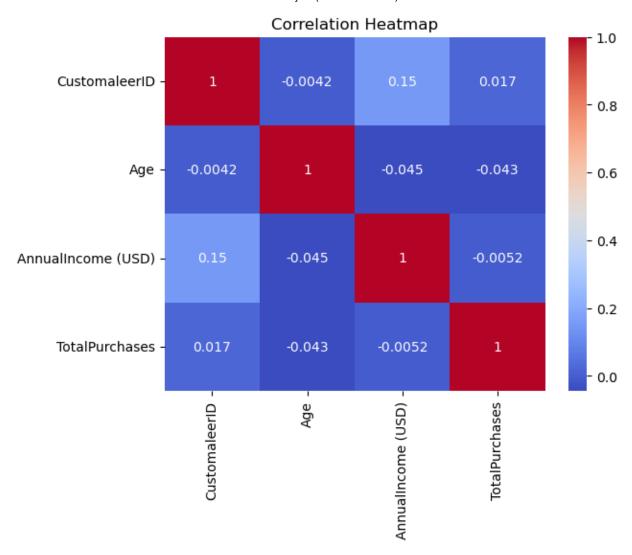
### **Encoding Categorical Variables:**

```
categorical_col = 'PreferredCategory'
In [32]:
           df_encoded = pd_get_dummies(df, columns=[categorical_col], drop_first=True)
           df_encoded.head()
In [33]:
Out[33]:
                                                                AnnualIncome
                                   Age Gender MaritalStatus
              CustomaleerID
                                                                                TotalPurchases PreferredCategory_
                                                                        (USD)
           0
                        1001
                              0.851064
                                           male
                                                      Divorced
                                                                      0.745837
                                                                                      0.525641
                        1002
                              0.297872
                                           male
                                                       Married
                                                                      0.096459
                                                                                      0.961538
           2
                                                      Divorced
                                                                     0.022576
                                                                                      0.217949
                        1003
                              0.787234
                                         Female
           3
                        1004
                              0.297872
                                           male
                                                       Married
                                                                      0.194359
                                                                                      0.961538
           4
                        1005 0.297872
                                                       Married
                                                                     0.274987
                                                                                      0.076923
                                         Female
```

### **Correlation heatmap**

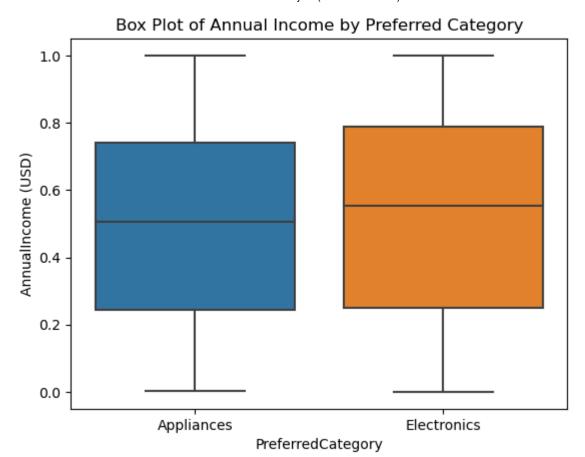
```
In [34]: correlation_matrix = df.corr()
    sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
    plt.title('Correlation Heatmap')
    plt.show()
```

C:\Users\Qurrat\AppData\Local\Temp\ipykernel\_14748\2298098936.py:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning. correlation\_matrix = df.corr()



# Box plot for annual income by preferred category

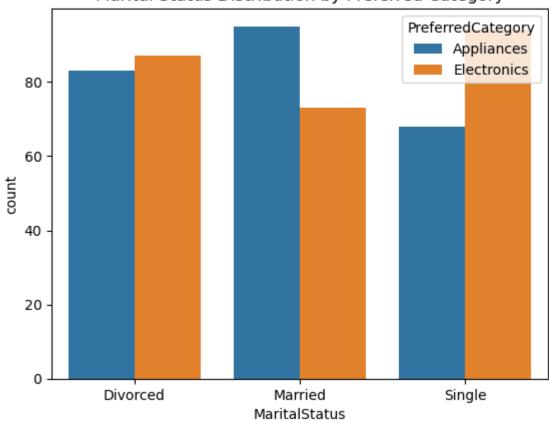
In [35]: sns.boxplot(data=df, x='PreferredCategory', y='AnnualIncome (USD)')
plt.title('Box Plot of Annual Income by Preferred Category')
plt.show()



## Bar plot for marital status distribution

In [36]: sns.countplot(data=df, x='MaritalStatus', hue='PreferredCategory')
plt.title('Marital Status Distribution by Preferred Category')
plt.show()

#### Marital Status Distribution by Preferred Category



```
In [ ]:
In [48]:
           from sklearn.cluster import KMeans
           from sklearn.preprocessing import StandardScaler
           columns_for_clustering = ['Age', 'AnnualIncome (USD)', 'TotalPurchases', 'PreferredCategory_Electronics']
           scaler = StandardScaler()
In [49]:
           df scaled = scaler.fit transform(df encoded[columns for clustering])
           kmeans = KMeans(n clusters=k, random state=42)
           df_encoded['Cluster'] = kmeans.fit_predict(df_scaled)
           cluster_sizes = df_encoded['Cluster'].value_counts()
           print("Cluster Sizes:", cluster_sizes)
           C:\anaconda\Lib\site-packages\sklearn\cluster\_kmeans.py:870: FutureWarning: The default value of `n_i
           nit` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
            warnings.warn(
           C:\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=2.
            warnings.warn(
           Cluster Sizes: 2 246
           0
              131
               123
           1
           Name: Cluster, dtype: int64
           plt_figure(figsize=(10, 6))
In [50]:
           sns.scatterplot(data=df_encoded, x='Age', y='AnnualIncome (USD)', hue='Cluster', palette='viridis')
```

plt.show()

plt.title('Cluster Distribution by Age and Annual Income')



