##1. LOADING DATASET
#importing pandas
import pandas as pd

#Reading the clustering data file

clustering_data = pd.read_csv('C:/Users/shyam/Desktop/UCI/Courses/Quarter 1/Python/Project/Clustering Data (1).csv')

#Checking the shape of the imported dataframe
print(f'The dataset has {clustering_data.shape[0]} rows and {clustering_data.shape[1]} columns')
clustering_data.head()

The dataset has 15144 rows and 90 columns

	uid	PNRLocatorID	avg_amt	round_trip	group_size	group	days_pre_booked	BookingCha
0	504554455244696420493F7C2067657420746869732072	AADMLF	0.019524	0	0.000	0	0.029703	
1	46495853454E44696420493F7C20676574207468697320	AAFBOM	0.081774	1	0.000	0	0.039604	
2	534355545444696420493F7C2067657420746869732072	AAFILI	0.026650	0	0.125	1	0.069307	
3	534355545444696420493F7C2067657420746869732072	AAFILI	0.026650	0	0.125	1	0.069307	
4	44554D4D414E4E44696420493F7C206765742074686973	AAFRQI	0.000000	1	0.000	0	0.035361	

5 rows × 90 columns

##2. DATA PREPROCESSING

#Preparing the data for clustering by copy the previously created dataset
prep_clustering_data = clustering_data.copy()

Saving the uid column before dropping it
save_uid_column = prep_clustering_data['uid']

#Dropping the uid and PNRLocaterID fields since they have string values
prep_clustering_data = prep_clustering_data.drop(['PNRLocatorID', 'uid'], axis=1)

#Checking the shape of the prepared dataframe
print(f'The dataset has {prep_clustering_data.shape[0]} rows and {prep_clustering_data.shape[1]} columns')
prep_clustering_data.head(10)

The dataset has 15144 rows and 88 columns

	avg_amt	round_trip	group_size	group	days_pre_booked	BookingChannel_Other	BookingChannel_Outside_Booking	BookingChannel_Reserva
0	0.019524	0	0.000	0	0.029703	0	1	
1	0.081774	1	0.000	0	0.039604	0	0	
2	0.026650	0	0.125	1	0.069307	0	0	
3	0.026650	0	0.125	1	0.069307	0	0	
4	0.000000	1	0.000	0	0.035361	0	1	
5	0.000000	1	0.125	1	0.050919	0	1	
6	0.000000	1	0.125	1	0.050919	0	1	
7	0.074727	1	0.000	0	0.045262	0	0	
8	0.035414	1	0.000	0	0.082037	0	1	
9	0.035414	1	0.125	1	0.018388	0	1	

10 rows × 88 columns

##3. CLUSTERING

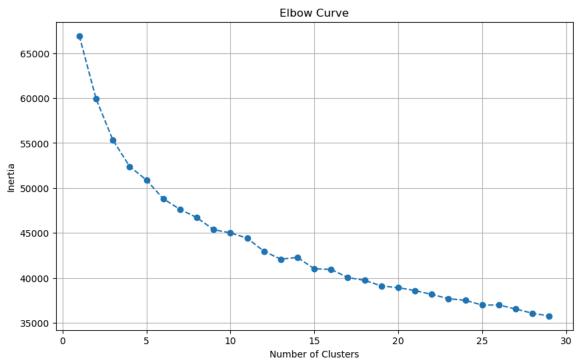
#3.1 Looking at the optimum number of clusters

from sklearn.cluster import KMeans

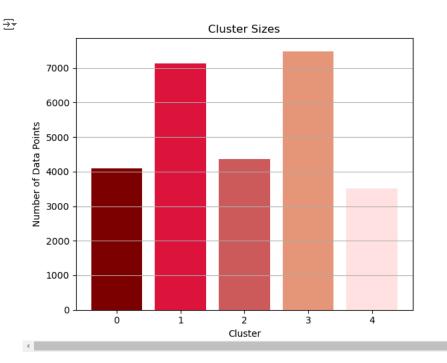
import matplotlib.pyplot as plt

for cluster_num in cluster_range:

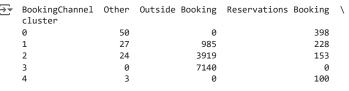
```
print(f'Iteration Number: {cluster_num}')
    kmeans = KMeans(n_clusters=cluster_num, n_init=10)
    kmeans.fit(prep_clustering_data)
    inertia.append(kmeans.inertia_)
plt.figure(figsize=(10,6))
plt.plot(cluster_range, inertia, marker='o', linestyle='--')
plt.title('Elbow Curve')
plt.xlabel('Number of Clusters')
plt.ylabel('Inertia')
plt.grid(True)
plt.show()
→ Iteration Number: 1
     Iteration Number: 2
     Iteration Number: 3
     Iteration Number: 4
     Iteration Number: 5
     Iteration Number: 6
     Iteration Number: 7
     Iteration Number: 8
     Iteration Number: 9
     Iteration Number: 10
     Iteration Number: 11
     Iteration Number: 12
     Iteration Number: 13
     Iteration Number: 14
     Iteration Number: 15
     Iteration Number: 16
     Iteration Number: 17
     Iteration Number: 18
     Iteration Number: 19
     Iteration Number: 20
     Iteration Number: 21
     Iteration Number: 22
     Iteration Number: 23
     Iteration Number: 24
     Iteration Number: 25
     Iteration Number: 26
     Iteration Number: 27
     Iteration Number: 28
     Iteration Number: 29
```

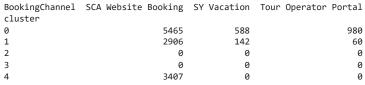


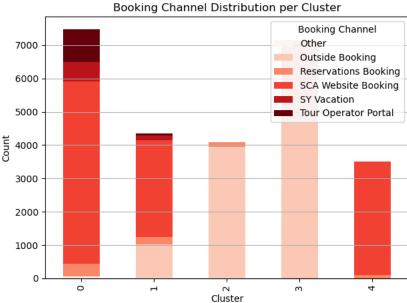
```
#3.2 Applying K-Means, and creating 5 centroids as per the Project requirement
kmeans = KMeans(n_clusters=5, n_init=10)
kmeans.fit(prep_clustering_data)
# attach the dropped UID column and cluster
prep_clustering_data = pd.DataFrame({
         'uid': save_uid_column,
         'cluster': kmeans.labels_
})
print(f'The dataset has {prep_clustering_data.shape[0]} rows and {prep_clustering_data.shape[1]} columns')
prep_clustering_data.head()
 → The dataset has 15144 rows and 2 columns
                                                                                                        uid cluster
           0 504554455244696420493F7C2067657420746869732072...
                                                                                                                            1
           1 46495853454E44696420493F7C20676574207468697320...
                                                                                                                            1
                  534355545444696420493F7C2067657420746869732072...
                  534355545444696420493F7C2067657420746869732072...
           4 44554D4D414E4E44696420493F7C206765742074686973...
                                                                                                                            2
##4. MERGED DATASET FOR VISUALIZATION
# Merging the Clustered dataset with the provided sample_data_transformed.csv
sample_transformed = pd.read_csv('C:/Users/shyam/Desktop/UCI/Courses/Quarter 1/Python/Project/sample_data_transformed.csv')
final_dataframe = sample_transformed.merge(prep_clustering_data[['uid', 'cluster']], on='uid', how='left')
#Checking the shape of the final dataframe and looking at the number of records in each cluster
final_dataframe.head(100)
cluster_sizes = final_dataframe['cluster'].value_counts().sort_index()
print(cluster_sizes)
 cluster
         0
                   7481
                   4348
         1
         2
                   4096
                  7140
         3
         4
                  3510
         Name: count, dtype: int64
          \hbox{C:\shyam\AppData\Local\Temp\ipykernel\_47360\2175281245.py:3: DtypeWarning: Columns (13) have mixed types. Specify dtype option on the supplementation of th
             sample_transformed = pd.read_csv('C:/Users/shyam/Desktop/UCI/Courses/Quarter 1/Python/Project/sample_data_transformed.csv')
         4
##5. ANALYSIS AND VISUALIZATION
from pandas import *
import matplotlib.pyplot as plt
cluster_colors = {
       0: 'maroon',
       1: 'crimson',
       2: 'indianred'
       3: 'darksalmon',
       4: 'mistyrose'
}
cluster_sizes = final_dataframe['cluster'].value_counts().sort_index()
plt.figure(figsize=(6, 5))
for cluster in range(len(cluster_sizes)):
       plt.bar(cluster, cluster_sizes[cluster], color=cluster_colors[cluster])
# Adding plot details
plt.title('Cluster Sizes')
plt.xlabel('Cluster')
plt.ylabel('Number of Data Points')
plt.xticks(ticks=range(len(cluster_sizes)))
plt.grid(axis='y')
plt.tight_layout()
plt.show()
```



```
#Preparing the sample dataframe to be plotted
booking\_channel\_counts = final\_dataframe.groupby(['cluster', 'BookingChannel']).size().unstack(fill\_value=0)
#Printing the sample dataframe
print(booking_channel_counts)
#Creating the Bar plot
booking_channel_counts.plot(kind='bar', stacked=True, colormap='Reds')
# Adding plot details
plt.title('Booking Channel Distribution per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(ticks=range(len(cluster_sizes)))
plt.tight_layout()
plt.grid(axis='y')
plt.legend(title='Booking Channel')
#Displaying the plot
plt.show()
```







```
#Preparing the sample dataframe to be plotted
age_group_counts = final_dataframe.groupby(['cluster', 'age_group']).size().unstack()

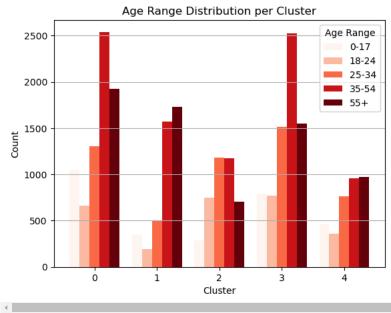
#Printing the sample dataframe
print(age_group_counts)

#Creating the Bar plot
age_group_counts.plot(kind='bar', width=0.8, colormap='Reds')

# Adding plot details
plt.title('Age Range Distribution per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Age Range')

# Displaying the plot
plt.show()
```

```
age_group
          0-17 18-24 25-34 35-54
                                       55+
cluster
0
           1053
                   659
                         1301
                                2540
                                      1928
                                      1729
1
            353
                   191
                          503
                                1572
2
            288
                   749
                         1184
                                1171
                                       704
                                      1548
3
            791
                   767
                         1510
                                2524
                          763
            466
                   356
                                 956
                                       969
```



```
import seaborn as sns
import matplotlib.pyplot as plt

# Create the boxplot without showing outliers
sns.boxplot(x='cluster', y='Age', data=final_dataframe, showfliers=False, color='indianred')

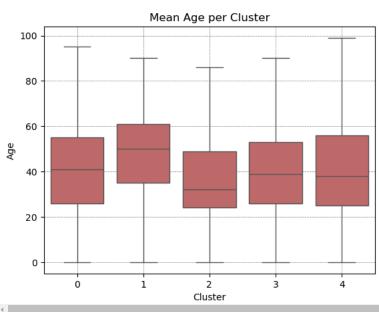
# Show grid lines
plt.grid(True)

# Customize grid appearance (optional)
plt.grid(color='gray', linestyle='--', linewidth=0.5)

# Adding plot details
plt.title('Mean Age per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Age')

# Displaying the plot
plt.show()

Mean Age per Cluster
```

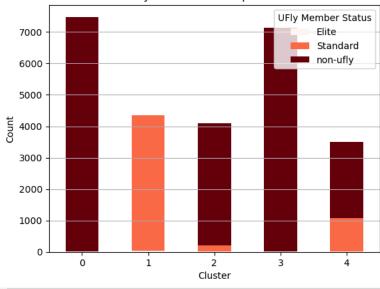


```
#Preparing the sample dataframe to be plotted
member_status_counts = final_dataframe.groupby(['cluster', 'UflyMemberStatus']).size().unstack(fill_value=0)
#Printing the sample dataframe
print(member_status_counts)
#Creating the Bar plot
member_status_counts.plot(kind='bar', stacked=True, colormap='Reds')
# Adding plot details
plt.title('UFly Member Status per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='UFly Member Status')
# Display the plot
plt.show()
→ UflyMemberStatus Elite Standard non-ufly
     cluster
                                            7475
     a
                           6
                                     a
     1
                          45
                                  4302
     2
                           8
                                   192
                                            3896
     3
                          18
                                     0
                                            7122
```

2435 UFly Member Status per Cluster

1054

21

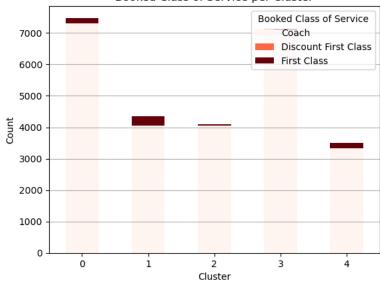


```
#Preparing the sample dataframe to be plotted
booked_class_counts = final_dataframe.groupby(['cluster', 'BkdClassOfService']).size().unstack(fill_value=0)
#Printing the sample dataframe
print(booked_class_counts)
#Creating the Bar plot
booked_class_counts.plot(kind='bar', stacked=True, colormap='Reds')
# Adding plot details
plt.title('Booked Class of Service per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Booked Class of Service')
# Display the plot
plt.show()
```

```
→ BkdClassOfService Coach Discount First Class First Class

    cluster
    0
                        7298
                                                  0
    1
                        4059
                                                  4
                                                             285
    2
                        4055
                                                  0
                                                              41
    3
                        7101
                                                  0
                                                              39
                        3327
                                                             182
    4
                                                  1
```

Booked Class of Service per Cluster



```
#Preparing the sample dataframe to be plotted
gender_code_counts = final_dataframe.groupby(['cluster', 'GenderCode']).size().unstack(fill_value=0)

#Printing the sample dataframe
print(gender_code_counts)

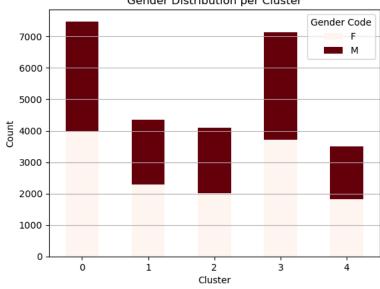
#Creating the Bar plot
gender_code_counts.plot(kind='bar', stacked=True, colormap='Reds')

# Adding plot details
plt.title('Gender Distribution per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Gender Code')

# Displaying the plot
plt.show()
```

```
→ GenderCode
                         Μ
    cluster
                3993
    0
                      3488
    1
                2279
                      2069
    2
                2004
                      2092
    3
                3711
                      3429
                1823
                      1687
```





```
#Preparing the sample dataframe to be plotted
round_trip_counts = final_dataframe.groupby(['cluster', 'round_trip']).size().unstack()

#Printing the sample dataframe
print(round_trip_counts)

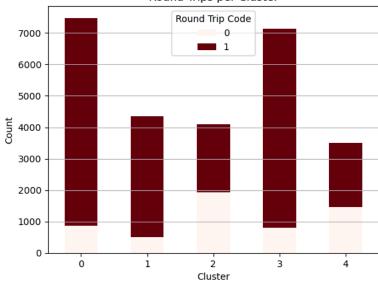
#Creating the Bar plot
round_trip_counts.plot(kind='bar', stacked=True, colormap='Reds')

# Adding plot details
plt.title('Round Trips per Cluster')
plt.xlabel('Cluster')
plt.xlabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Round Trip Code')

# Displaying the plot
plt.show()
```

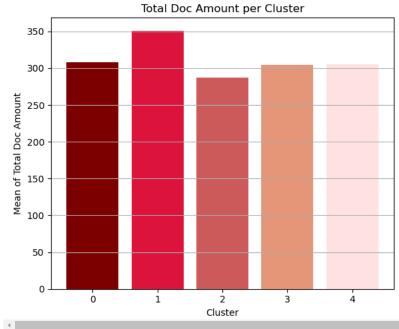
```
→ round_trip
                         1
    cluster
    0
                 865
                      6616
    1
                 509
                      3839
    2
                1935
                      2161
                 794
    3
                      6346
                1464
                      2046
```

Round Trips per Cluster



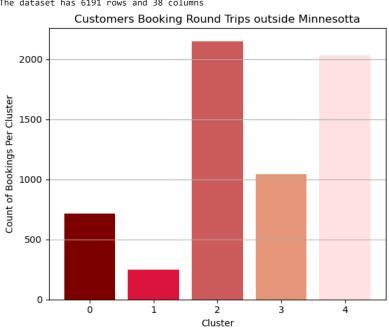
```
cluster_colors = {
    0: 'maroon',
   1: 'crimson',
    2: 'indianred',
    3: 'darksalmon',
    4: 'mistyrose'
}
#Preparing the sample dataframe to be plotted
cluster_revenue = final_dataframe.groupby('cluster')['TotalDocAmt'].mean()
#Creating the Bar plot
plt.figure(figsize=(6, 5))
for cluster in range(len(cluster_revenue)):
    plt.bar(cluster, cluster_revenue[cluster], width = 0.8, color=cluster_colors[cluster])
# Adding plot details
plt.title('Total Doc Amount per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Mean of Total Doc Amount')
plt.xticks(ticks=range(len(cluster_revenue)))
plt.grid(axis='y')
plt.tight_layout()
# Displaying the plot
plt.show()
```





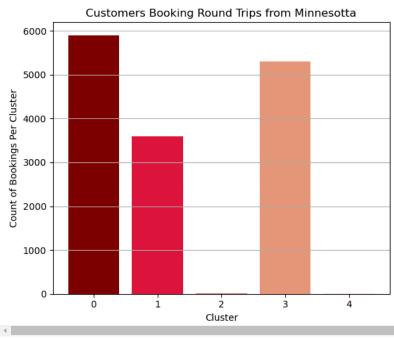
```
#Preparing the sample dataframe to be plotted
msp_origin_dest = final_dataframe.loc[(final_dataframe['round_trip']== 1) & (final_dataframe['true_origins']!= "MSP")]
print(f'The \ dataset \ has \ \{msp\_origin\_dest.shape[0]\} \ rows \ and \ \{msp\_origin\_dest.shape[1]\} \ columns')
cluster_sizes = msp_origin_dest['cluster'].value_counts().sort_index()
#Creating the Bar plot
plt.figure(figsize=(6, 5))
for cluster in range(len(cluster_sizes)):
    plt.bar(cluster, cluster_sizes[cluster], color=cluster_colors[cluster])
# Adding plot details
plt.title('Customers Booking Round Trips outside Minnesotta')
plt.xlabel('Cluster')
plt.ylabel('Count of Bookings Per Cluster')
plt.xticks(ticks=range(len(cluster_revenue)))
plt.grid(axis='y')
plt.tight_layout()
# Displaying the plot
plt.show()
```

The dataset has 6191 rows and 38 columns



```
#Preparing the sample dataframe to be plotted
msp_origin_dest = final_dataframe.loc[(final_dataframe['round_trip']== 1) & (final_dataframe['true_origins']== "MSP")]
print(f'The \ dataset \ has \ \{msp\_origin\_dest.shape[0]\} \ rows \ and \ \{msp\_origin\_dest.shape[1]\} \ columns')
cluster_sizes = msp_origin_dest['cluster'].value_counts().sort_index()
#Creating the Bar plot
plt.figure(figsize=(6, 5))
for cluster in range(len(cluster_sizes)):
    plt.bar(cluster, cluster_sizes[cluster], color=cluster_colors[cluster])
# Adding plot details
plt.title('Customers Booking Round Trips from Minnesotta')
plt.xlabel('Cluster')
plt.ylabel('Count of Bookings Per Cluster')
plt.xticks(ticks=range(len(cluster_revenue)))
plt.grid(axis='y')
plt.tight_layout()
# Displaying the plot
plt.show()
```

→ The dataset has 14817 rows and 38 columns



```
import seaborn as sns
import matplotlib.pyplot as plt

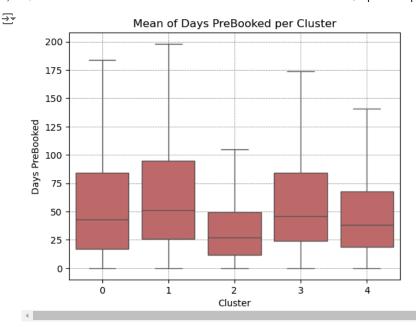
# Create the boxplot without showing outliers
sns.boxplot(x='cluster', y='days_pre_booked', data=final_dataframe, showfliers=False, color='indianred')

# Show grid lines
plt.grid(True)

# Customize grid appearance (optional)
plt.grid(color='gray', linestyle='--', linewidth=0.5)

# Adding plot details
plt.title('Mean of Days PreBooked per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Days PreBooked')

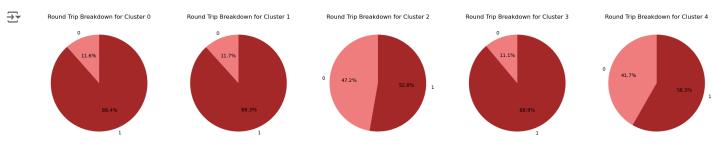
# Displaying the plot
plt.show()
```



```
#Preparing the sample dataframe to be plotted
round_trip_counts = final_dataframe.groupby(['cluster', 'round_trip']).size().unstack()

# Create the pie plot by iterating over each cluster
fig, axes = plt.subplots(1, 5, figsize=(20, 4))
for i, cluster in enumerate(round_trip_counts.index):
    values = round_trip_counts.loc[cluster]
    axes[i].pie(values, labels=values.index, autopct='%1.1f%%', colors=['lightcoral', 'brown'], startangle=90)
    axes[i].axis('equal')
    axes[i].set_title(f'Round Trip Breakdown for Cluster {cluster}')
plt.tight_layout()

# Displaying the plot
plt.show()
```



```
#Preparing the sample dataframe to be plotted
seasonality_counts = final_dataframe.groupby(['cluster', 'seasonality']).size().unstack()

#Printing the sample dataframe
print(seasonality_counts)

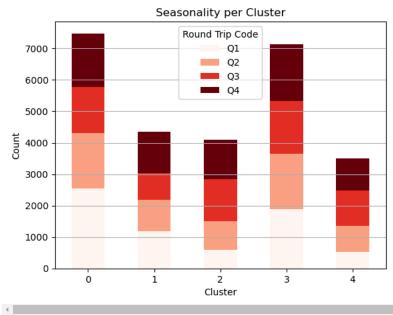
#Creating the Bar plot
seasonality_counts.plot(kind='bar',stacked=True, colormap='Reds')

# Adding plot details
plt.title('Seasonality per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Round Trip Code')

# Displaying the plot
plt.show()
```

```
→ seasonality

                   Q1
                         Q2
                               Q3
                                     Q4
    cluster
    0
                 2543 1758
                             1468
                                   1712
    1
                 1191
                        987
                              848
                                   1322
    2
                  586
                        909
                             1341
                                   1260
                 1879
                       1760
    3
                             1681
                                   1820
    4
                  530
                        815
                             1127
                                   1038
```



```
#Preparing the sample dataframe to be plotted
groupsize_counts = final_dataframe.groupby(['cluster', 'group']).size().unstack()

#Printing the sample dataframe
print(round_trip_counts)

#Creating the Bar plot
groupsize_counts.plot(kind='bar', stacked=True, colormap='Reds')

# Adding plot details
plt.title('Group size per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.legend(title='Group size Code')

# Displaying the plot
plt.show()
```

```
→ group
     cluster
     0
              2171
                    5310
     1
              1797
                    2551
     2
              2724 1372
              2330 4810
     3
              1730 1780
import seaborn as sns
import matplotlib.pyplot as plt
# Create the boxplot without showing outliers
\verb|sns.boxplot(x='cluster', y='group\_size', data=final\_dataframe, showfliers=False, color='indianred')| \\
# Show grid lines
plt.grid(True)
# Customize grid appearance (optional)
plt.grid(color='gray', linestyle='--', linewidth=0.5)
# Adding plot details
plt.title('Group size per Cluster')
plt.xlabel('Cluster')
plt.ylabel('Group size')
# Displaying the plot
plt.show()
₹
```

