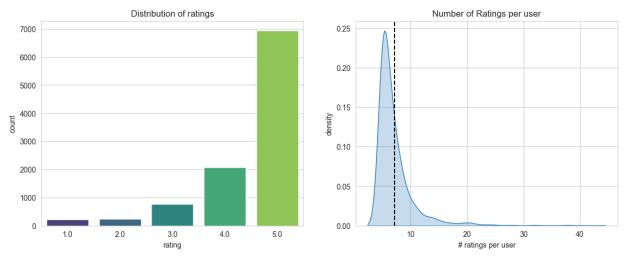
```
In [38]: import numpy as np
         import pandas as pd
         import sklearn
         import matplotlib.pyplot as plt
          import seaborn as sns
In [39]: ratings = pd.read csv('data/amazon/musical instruments ratings.csv')
         ratings.head()
Out[39]:
                       user
                                  item rating
              A2IBPI20UZIR0U 1384719342
                                          5.0
             A14VAT5EAX3D9S 1384719342
                                          5.0
         2 A195EZSQDW3E21 1384719342
                                          5.0
         3 A2C00NNG1ZQQG2 1384719342
                                          5.0
             A94QU4C90B1AX 1384719342
                                          5.0
In [40]: #How many items are in the data set? How many users? How many ratings?
         n ratings = len(ratings)
         n_users = ratings['user'].nunique()
         n items = ratings['item'].nunique()
         print(f'Number of ratings: {n ratings}')
         print(f'Number of unique users: {n users}')
         print(f'Number of unique items: {n items}')
         Number of ratings: 10261
         Number of unique users: 1429
         Number of unique items: 900
In [41]: #User activity:
         #What is the distribution of ratings-per-user?
         user_freq = ratings[['user', 'rating']].groupby('user').count().reset_index()
         user freq.columns = ['user', 'n ratings']
         user freq.head()
Out[41]:
                              user n_ratings
         0 A00625243BI8W1SSZNLMD
                                          8
          1
                   A10044ECXDUVKS
                                          6
         2
                   A102MU6ZC9H1N6
                                          6
         3
                   A109JTUZXO61UY
                                          5
         4
                  A109ME7C09HM2M
                                          5
In [42]: sns.set style("whitegrid")
         plt.figure(figsize=(14,5))
         plt.subplot(1,2,1)
         ax = sns.countplot(x="rating", data=ratings, palette="viridis")
         plt.title("Distribution of ratings")
```

```
plt.subplot(1,2,2)
ax = sns.kdeplot(user_freq['n_ratings'], fill=True, legend=False)
plt.axvline(user_freq['n_ratings'].mean(), color="k", linestyle="--")
plt.xlabel("# ratings per user")
plt.ylabel("density")
plt.title("Number of Ratings per user")
plt.show()
```



```
In [48]: #Find most and least active users for all 4 dataset. How many ratings did they
    user_active = ratings.groupby('user')['rating'].count()
    #user_active.columns = ['n_ratings']

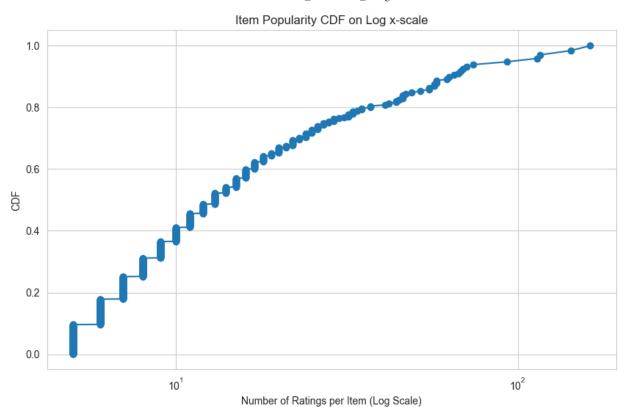
most_active = user_active.idxmax()
    least_active = user_active.idxmin()

most_active_ratings = user_active[most_active]
    least_active_ratings = user_active[least_active]

print(f"The most active user is {most_active} with {most_active_ratings} rating
    print(f"The least active user is {least_active} with {least_active_ratings} rating
```

The most active user is ADH008UVJOT10 with 42 ratings. The least active user is A109JTUZX061UY with 5 ratings.

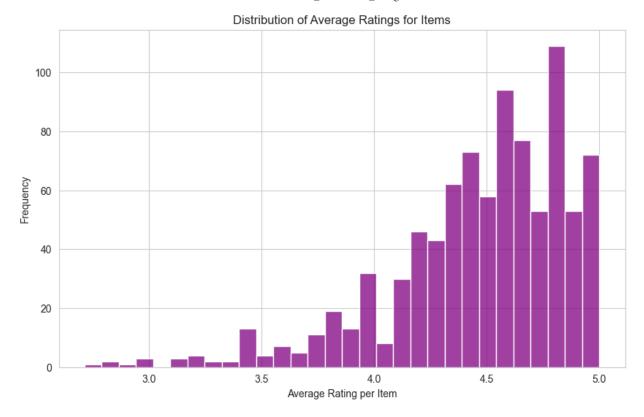
```
In [49]: #Item statistics:
    #What is the item popularity curve (the distribution of ratings-per-item)? A Cl
    item_freq = ratings.groupby('item')['rating'].count().reset_index()
    item_freq.columns = ['item', 'n_ratings']
    item_freq_sort = item_freq.sort_values(by='n_ratings')
    cdf = np.cumsum(item_freq_sort['n_ratings']) / np.sum(item_freq_sort['n_ratings'])
    plt.figure(figsize=(10, 6))
    plt.plot(item_freq_sort['n_ratings'], cdf, marker='o')
    plt.xscale('log')
    plt.xlabel('Number of Ratings per Item (Log Scale)')
    plt.ylabel('CDF')
    plt.title('Item Popularity CDF on Log x-scale')
    plt.show()
```



```
In [50]: #What is the distribution of average ratings for items?

avg_item_rating = ratings.groupby('item')['rating'].mean().reset_index()
avg_item_rating.columns = ['item', 'avg_rating']

sns.set_style("whitegrid")
plt.figure(figsize=(10,6))
sns.histplot(avg_item_rating['avg_rating'], bins=30, kde=False, color='purple'
plt.xlabel('Average Rating per Item')
plt.ylabel('Frequency')
plt.title("Distribution of Average Ratings for Items")
plt.show()
```



```
In [52]: #Non-personalized recommendation
    #What are the 10 most popular items (the items with the most ratings)? Show the
    item_pop = ratings.groupby('item')['rating'].count().reset_index()
    item_pop.columns = ['item', 'n_ratings']

top10_item = item_pop.sort_values(by='n_ratings', ascending=False).head(10)

print('The most popular items based on the number of ratings are: \n')
print(top10_item[['item', 'n_ratings']])
```

The most popular items based on the number of ratings are:

```
item n ratings
706
    B003VWJ2K8
                        163
     B0002E1G5C
94
                        143
156
    B0002F7K7Y
                        116
707
     B003VWKPHC
                        114
     B0002H0A3S
                         93
198
44
     B0002CZVXM
                         74
258
    B0006NDF8A
                         71
    B0009G1E0K
285
                         69
     B0002E2KPC
                         68
116
     B0002GLDQM
167
                         67
```

```
In [53]: #What are the 10 items with the highest average ratings (with their titles and
   item_avg_rating = ratings.groupby('item')['rating'].mean().reset_index()
   item_avg_rating.columns = ['item', 'avg_rating']
   top10_avg_rate = item_avg_rating.sort_values(by='avg_rating', ascending=False)
   print('The 10 items with the highest average ratings are: \n')
   print(top10_avg_rate)
```

The 10 items with the highest average ratings are:

```
item avg rating
0
     1384719342
                        5.0
                        5.0
    B003AJMPW4
669
604
    B0027842S4
                        5.0
641 B002RLLD88
                        5.0
288 B0009IEB0I
                        5.0
283
    B0009EU01G
                        5.0
282 B0009EQOES
                        5.0
649 B0033P106S
                        5.0
275
    B0007Y3XGW
                        5.0
274
    B0007XWUQ2
                        5.0
```

```
In [54]: #What are the 10 movies with the highest damped average ratings, with a Bayesia
    #is computed by: $r^-i=\frac{\Sum{r_{ui}\in R_i}r_{ui}}+\gamma*r^-)}{|R_i|+\gamma*r^-}}{|R_i|+\gamma*r^-}}

overall_mean_rating = ratings['rating'].mean()
gamma = 5

item_stats = ratings.groupby('item')['rating'].agg(['sum', 'count']).reset_index
item_stats['damped_mean'] = (item_stats['sum'] + gamma * overall_mean_rating)
item_stats['undamped_mean'] = item_stats['sum'] / item_stats['count']

top10_damped_items = item_stats.sort_values(by='damped_mean', ascending=False).

print('The 10 items with the highest damped average ratings are: \n')
print(top10_damped_items[['item', 'damped_mean', 'undamped_mean']])
```

The 10 items with the highest damped average ratings are:

```
item damped mean
                              undamped mean
594
    B001W99HE8
                    4.897749
                                   5.000000
                                    4.962963
504
     B0015RIN6U
                    4.888866
                                   5.000000
424
    B000RKL8R2
                    4.849631
649
     B0033P106S
                    4.829581
                                    5.000000
604
     B0027842S4
                    4.829581
                                    5.000000
120
    B0002E2XCW
                                    4.866667
                    4.828874
718 B0042EZH6W
                    4.826366
                                    4.888889
387 B000L6GD04
                    4.820089
                                    4.864865
438 B000RY68PA
                    4.817408
                                    5.000000
669 B003AJMPW4
                    4.817408
                                    5.000000
```

```
In [ ]:
```