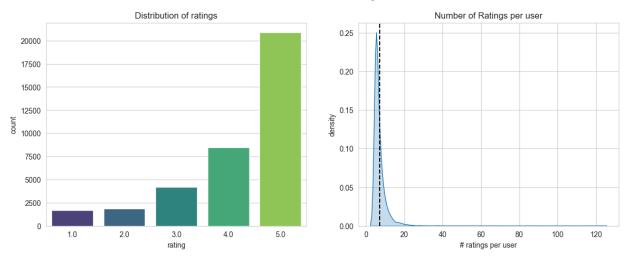
```
In [2]: import numpy as np
        import pandas as pd
        import sklearn
        import matplotlib.pyplot as plt
        import seaborn as sns
In [3]: ratings = pd.read csv("data/amazon/instant video ratings.csv")
In [4]: #How many items are in the data set? How many users? How many ratings?
        #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: 37126
        Number of unique users: 5130
        Number of unique items: 1685
In [5]: #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[5]:
                            user n_ratings
        0 A0705654XT5UCAYOY7TH
                                        8
        1 A099898949AFPOGMFDCB
        2
                   A1004HZ4AR10UI
                                        6
        3
                                        5
                   A10072IQFY9167
        4
                  A100UD67AHFODS
                                        5
In [6]: 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()
```



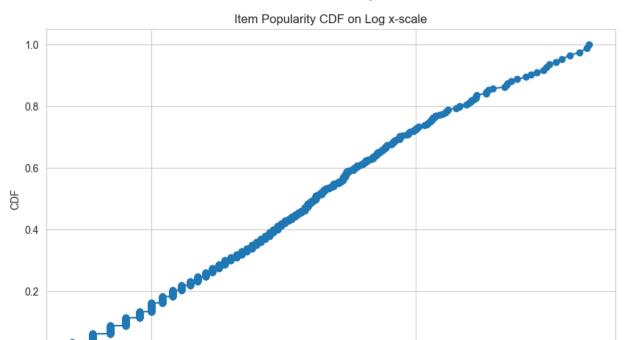
```
In [7]: #Find most and least active users for all 4 dataset. How many ratings did they
    user_active = ratings.groupby('user')['rating'].count()
    most_active = user_active.idxmax()
    least_active = user_active.idxmin()

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

The most active user is AV6QDP8Q0ONK4 with 123 ratings. The least active user is A10072IQFY9167 with 5 ratings.

10¹



Number of Ratings per Item (Log Scale)

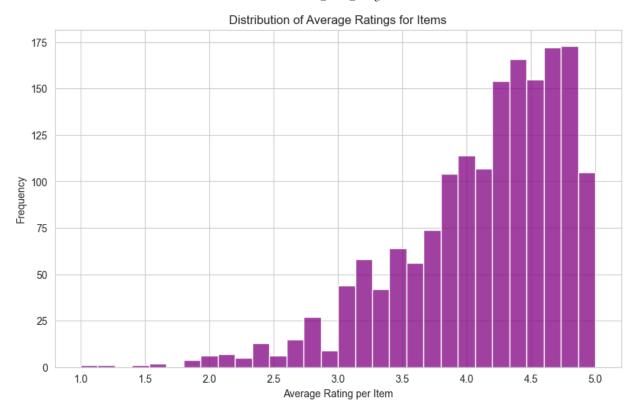
10²

In [9]: #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()

0.0



```
In [10]: #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
1557 B00I3MPDP4
                        455
1257 B00DAHSVYC
                        444
1549 B00I3MMN4I
                        417
1001 B00APE00H4
                        384
1184 B00CDBTQCW
                        358
1555 B00I3MNGCG
                        341
1556 B00I3MNVBW
                        322
1057 B00B8P809K
                        313
1551 B00I3MMTS8
                        306
1183 B00CDBR1P6
                        288
```

```
In [11]: #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
959
      B00A1ZV2KG
                          5.0
1247 B00D6MQ6ZM
                          5.0
1268 B00DI2NTK0
                          5.0
258
      B00332YHLU
                          5.0
462
      B004SKQHYQ
                          5.0
878
      B009DNVLH6
                          5.0
      B003336P1Y
                          5.0
259
1031 B00B09HOGI
                          5.0
914
      B009KZYJ62
                          5.0
415
      B004E2BY8W
                          5.0
```

In [12]: #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
264
      B00366JI90
                     4.852734
                                    4.918367
265
      B0036D8LPO
                     4.847376
                                    4.941176
1291 B00DTOYIIE
                                    4.836735
                     4.821133
402
      B004BHISOI
                     4.820348
                                    5.000000
      B002A493NY
                     4.801191
190
                                    4.885714
268
      B0036DCJKW
                     4.796891
                                    4.863636
1388 B00F3KFFPM
                     4.795483
                                    4.835616
1529 B00HQQOLSS
                     4.784680
                                    4.944444
401
      B004BHGDYK
                     4.780425
                                    5.000000
1190 B00CE10BD4
                     4.773810
                                     4.846154
```

```
In []:
```