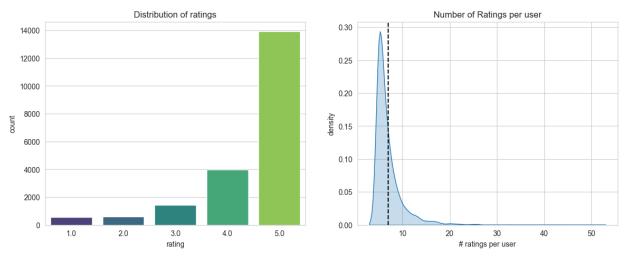
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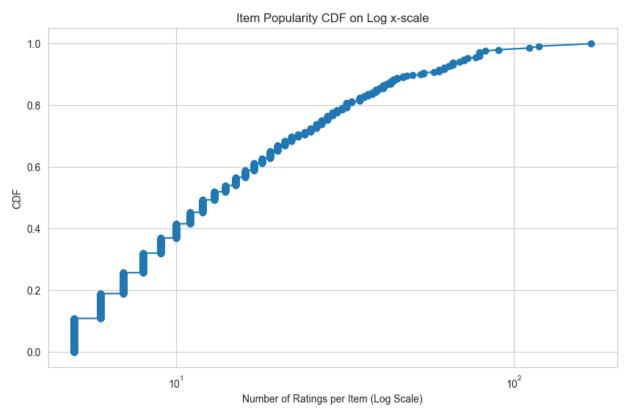
```
In [1]: import numpy as np
        import pandas as pd
        import sklearn
        import matplotlib.pyplot as plt
        import seaborn as sns
In [2]: ratings = pd.read csv('data/amazon/automotive ratings.csv')
In [3]: #How many items are in the data set? How many users? How many ratings?
        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: 20473
        Number of unique users: 2928
        Number of unique items: 1835
In [4]: #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[4]:
                            user n_ratings
        0 A00473363TJ8YSZ3YAGG9
                                        5
        1 A0473259F6GQNBD88IYN
                                       11
        2 A09567722SXKYZTV7OFCJ
                                        5
                                        5
        3
                 A10063PJ5C9WQQ
                 A100UD67AHFODS
In [5]: 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()
```



The most active user is A2V1J3JT500ZFO with 51 ratings. The least active user is A00473363TJ8YSZ3YAGG9 with 5 ratings.

```
In [7]: #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()
```

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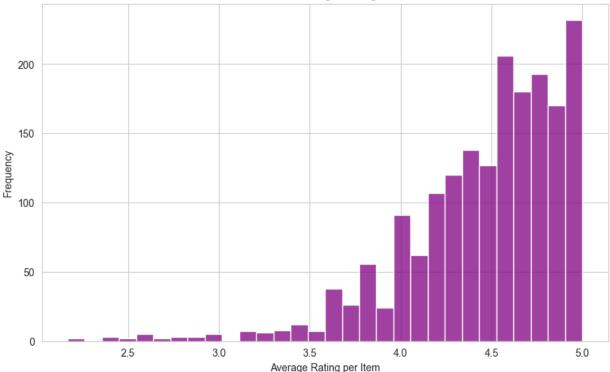


```
In [8]: #What is the distribution of average ratings for items?

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 [9]:
        #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
477
      B000CITK8S
                         169
1656 B007TG7HFO
                         118
1089 B001V8U12M
                         111
1141 B002BC4N5I
                          90
1198 B002OUMVWY
                          82
866
      B0014Y82UQ
                          79
705
      B000NCOKZQ
                          79
                          79
273
      B0009IQZFM
177
      B00068XCOU
                          79
1012 B001LHVOVK
                          77
```

```
In [10]: #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)
```

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The 10 items with the highest average ratings are:

```
item avg rating
461
      B000CB6FNS
                          5.0
1390 B003YJJS2U
                          5.0
1381 B003WDDLTO
                          5.0
829
      B0010DZZQA
                          5.0
599
      B000FR5XQE
                          5.0
828
      B0010DZZPG
                          5.0
600
      B000FRLO9Y
                          5.0
1635 B007B8JMZI
                          5.0
343
      B000B8WCBG
                          5.0
171
      B00067BVDC
                          5.0
```

```
In [11]: #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^^-|}{|R_i|+\gamma*r^^-|}}{|R_i|+\gamma*r^^-|}{|R_i|+\gamma*r^^-|}}{|R_i|+\gamma*r^^-|}{|R_i|+\gamma*r^^-|}}{|R_i|+\gamma*r^^-|}{|R_i|+\gamma*r^^-|}}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_i|+\gamma*r^-|}{|R_
```

The 10 items with the highest damped average ratings are:

```
item
                  damped mean
                               undamped mean
      B000601ICE
                     4.898431
                                     5.000000
196
684
      B000M61JFC
                     4.874248
                                     5.000000
658
      B000J19XSA
                     4.867960
                                     5.000000
600
      B000FRL09Y
                     4.861011
                                     5.000000
      B00062YZZS
161
                     4.861011
                                     5.000000
1519 B004YJQE0S
                     4.859969
                                     4.952381
425
      B000C3XD90
                     4.853289
                                     5.000000
1683 B008FSQ130
                     4.848300
                                     4.947368
129
      B0002UEOLO
                     4.845307
                                     4.920000
706
      B000NCS7GE
                     4.844659
                                     5.000000
```

```
In [ ]:
```