Python Lab-10 Name- Jatin Karthik Matriculation pp. 212201
Matriculation no313301 In [1]: import pandas as pd import numpy as np import matplotlib.pyplot as plt from collections import Counter import seaborn as sns from sklearn.decomposition import NMF import math as Math
In [2]: All_data=pd.read_csv(r"D:\Hildesheim Universitaet\Semester 1\ML LAB\Exercise_10\ml-100k\ml-100
2 22 377 1 878887116 3 244 51 2 880606923 4 166 346 1 886397596 99995 880 476 3 880175444 99996 716 204 5 879795543
9997 276 1090 1 874795795 9998 13 225 2 882399156 9999 12 203 3 879959583 100000 rows × 4 columns In [3]: user_data=pd.read_csv(r"D:\Hildesheim Universitaet\Semester 1\ML LAB\Exercise_10\ml-100k\ullend-100k\u
names=["userId", "age", "gender", "occupation", "zipcode"], sep=" ") #userdata=userdata.replace(['F', 'M'], [1,0]) user_data
3 4 24 M technician 43537 4 5 33 F other 15213 938 939 26 F student 33319 939 940 32 M administrator 02215 940 941 20 M student 97229
941 942 48 F librarian 78209 942 943 22 M student 77841 943 rows × 5 columns In [4]: item_data=pd.read_csv(r"D:\Hildesheim Universitaet\Semester 1\ML LAB\Exercise_10\ml-100k\ml-100k\u.item", names=["movieid", "movietitle", "releasedate", "videoreleasedate", "IMDbURL", "unknown", "Action", "Adventure", "Animation", "Action", "Adventure", "Animation", "Action", "Action", "Action", "Adventure", "Animation", "Action", "Actio
"Children's", "Comedy", "Crime", "Documentary", "Drama", "Fantasy", "Film-Noir", "Horror", "Musical", "Mystery", "Romance", "Sci-Fi", "Thriller", "War", "Western"], sep=" ", encoding="latin-1") Out[4]: movieid movieitle releasedate videoreleasedate videoreleasedate NaN http://us.imdb.com/M/title-exact?Toy%20Story%2 NaN http://us.imdb.com/M/title-exact?GoldenEye%20(no 1 out[4]: novieid novieitle releasedate videoreleasedate videoreleasedate videoreleasedate NaN http://us.imdb.com/M/title-exact?Toy%20Story%2 no 1 out[4]: novieid novieitle releasedate videoreleasedate videoreleasedate videoreleasedate NaN http://us.imdb.com/M/title-exact?Coy%20Story%2 no 1 out[4]: novieid novieitle releasedate videoreleasedate videoreleasedate videoreleasedate NaN http://us.imdb.com/M/title-exact?Coy%20Story%2 no 0 out[4]: novieid novieitle releasedate videoreleasedate videoreleasedate NaN http://us.imdb.com/M/title-exact?Coy%20Story%2 no 0 out[4]: novieid novieitle releasedate Nusical Mystery Romance Sci-Fi Thriller War Western Var Western Out[4]: novieid novieitle novieitle novieitle releasedate videoreleasedate Nan http://us.imdb.com/M/title-exact?Coy%20Story%2 no 0 out[4]: novieid novieitle releasedate Nusical Mystery Romance Sci-Fi Thriller War Western Out[4]: novieid novieitle noviei
2 3 Four Rooms (1995) 01-Jan-1995 NaN http://us.imdb.com/M/title-exact?Four%20Rooms% 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
print(item_data.info()) <class 'pandas.core.frame.dataframe'=""> RangeIndex: 100000 entries, 0 to 99999 Data columns (total 4 columns): # Column Non-Null Count Dtype </class>
dtypes: int64(4) memory usage: 3.1 MB None <class 'pandas.core.frame.dataframe'=""> RangeIndex: 943 entries, 0 to 942 Data columns (total 5 columns): # Column Non-Null Count Dtype</class>
2 gender 943 non-null object 3 occupation 943 non-null object 4 zipcode 943 non-null object dtypes: int64(2), object(3) memory usage: 37.0+ KB None <class 'pandas.core.frame.dataframe'=""> RangeIndex: 1682 entries, 0 to 1681 Data columns (total 24 columns): # Column Non-Null Count Dtype</class>
0 movieid 1682 non-null int64 1 movietitle 1682 non-null object 2 releasedate 1681 non-null object 3 videoreleasedate 0 non-null float64 4 IMDbURL 1679 non-null object 5 unknown 1682 non-null int64 6 Action 1682 non-null int64 7 Adventure 1682 non-null int64 8 Animation 1682 non-null int64 9 Children's 1682 non-null int64 10 Comedy 1682 non-null int64
11 Crime
22 War 1682 non-null int64 23 Western 1682 non-null int64 dtypes: float64(1), int64(20), object(3) memory usage: 315.5+ KB None 1) Showcase how the ratings vary across users, as an example consider whether the plot is able to tell if most ratings are only from a handful of users. The below graph shows where the median or mean of the most ratings were by users. I also printed the maximum movies rated by one user and the minimum one.
<pre>In [6]: merge1=pd.merge(All_data, user_data, on='userId', how='left') In [7]: movies_rated = All_data.groupby("userId").size().sort_values(ascending=False) print(f"Max movies rated by one user: {max(movies_rated)}\nMin movies rated by one user: {min(movies_rated)}") All_data.userId.value_counts().plot.box(figsize=(12, 5)) plt.title("Number of Movies rated by a Single user", fontsize=16) plt.show() Max movies rated by one user: 737</pre>
Number of Movies rated by a Single user Number of Movies rated by a Single user 700 -
300 - 200 - 100 - 0 -
In [8]: #showing only the top 10 most users who rated ID_plot=movies_rated[0:25] ID_plot.plot(kind='bar') plt.xlabel('different user IDs') plt.ylabel('Ratings') Out[8]: Text(0, 0.5, 'Ratings')
700 - 600 - 500 - 500 - 80 400 - 200 -
2) Showcase how the ratings vary across items. In [22]: item_list=All_data["itemId"].value_counts() items=All_data.groupby('itemId').size().sort_values(ascending=False)
<pre>item_list10=items[0:10] # I show only the top 10 most highly rated items here b_plot=item_list10 b_plot.plot(kind='bar') plt.title("ratings variation across items") plt.xlabel('Item IDs') plt.ylabel('number of ratings that item ID got') plt.show()</pre>
ratings variation across items From 100 -
item_list_last10=items[-10:] #I show only the least 10 most rated items here which is all 1 b_plot=item_list_last10 b_plot.plot(kind='bar') plt.title("ratings variation across items")
plt.xlabel('Item IDs') plt.ylabel('number of ratings that item ID got') plt.show() ratings variation across items
1451 1451 1451 1451 1451 1451 1451 1451
3) Are there genres that are more highly rated than others? Yes, drama is the most highly rated genre In [11]: freq_array=pd.Series() for col in item_data.columns[5:]: freq_array[col]=sum(item_data[col].values) bar_plot=freq_array
bar_plot.plot(kind='bar') plt.title("mosted watched genre type") plt.show() <ipython-input-11-2940e0721dbe>:1: DeprecationWarning: The default dtype for empty Series will be 'object' instead of 'float64' in a future version. Specify a dtype explicitly to silence this warning. freq_array=pd.Series() mosted watched genre type 700 600</ipython-input-11-2940e0721dbe>
500 - 400 - 300 - 200 - 100 -
4) What age groups prefer what genres based on ratings? You can bin respective ages to your preference In []:
In [1]: merge1.sort_values(by='age') out[12]: merge1.tort_values(by='age') out[12]: merge1.tort_values(by='age') out[12]: merge1.sort_values(by='age') out[12]
50845 30 313 5 885941156 7 M student 55436 16651 30 231 2 875061066 7 M student 55436 3845 30 181 4 875060217 7 M student 55436
79034 481 199 5 885828543 73 M retired 37771 61198 481 70 5 885828389 73 M retired 37771 45783 481 163 4 885828389 73 M retired 37771 100000 rows × 8 columns In [13]: print(sorted(merge1['age'].unique())) [7, 10, 11, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65,
8, 69, 70, 73] My age group preferenc eis as follows:- "children"0-12 yrs, "teenagers"13-19 yrs, "young adults"20-25 yrs, "adults" 26-50 yrs, "old-aged"50-74 yrs In [14]: bins = pd.IntervalIndex.from_tuples([(0, 12), (13, 19), (20, 25), (26,50), (50,74)]) temp1=merge1.groupby(pd.cut(merge1["age"], bins, labels=["children", "teenagers", "young adults", "adults", "old-aged"])).sum() temp1 but[14]: userld itemld rating timestamp age
age (0,12] 23694 37989 346 89104808397 908 (13,19] 3710022 3534800 26611 6737172558577 136968 (20,25] 8824618 7863350 62768 16109434303043 422336 (26,50] 26255045 23402810 200431 49890840449957 2024878 (50,74] 3953840 4061819 34839 8379476974700 532576
categories={'children':346, 'teenagers':26611, 'young adults':62768, 'adults':200431, 'old-aged':34839} names = list(categories.keys()) values = list(categories.values()) plt.bar(range(len(categories)), values, tick_label=names) plt.title('bar plot of age categories vs ratings by them respectively') plt.show() bar plot of age categories vs ratings by them respectively
20000 - 175000 - 15000 - 125000 - 100000 - 150
In []: I have displayed just a few more graphs to more explore the dataset. This graph shows which numbered rating were used the mos. The bar plot shows that 4.0 was the most used rating.
All_data["rating"].hist() plt.suptitle("Rating Histogram") plt.xlabel('ratings out of 5') plt.ylabel('number of such ratings') plt.show() Rating Histogram 35000 30000 Rating Histogram
30000
The below plot shows variation of movies watched based on a person's occupation. We see that students watch the most movies En [17]: ba_plot=user_data["occupation"].value_counts() ba_plot=ba_plot.sort_index() ba_plot.plot(kind='bar') plt.title("frequency of watching movies depending on occupation")
frequency of watching movies depending on occupation 175 150 125
ministrator artist doctor equication of control of cont
The below graph shows how the gender distribution stands wrt movie watching user_data=user_data.replace(['F', 'M'], [1,0]) user_data.head() user_data["gender"].hist() plt.suptitle("Gender Histogram")
plt.xlabel('0 is Male and 1 is Female') plt.show() Gender Histogram 700 600 500
400 200 100 0.0 0.2 0.4 0.6 0.8 10 0 is Male and 1 is Female
In []: