ChandonnetModule08Lab01

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1 Assignment 8

- 1.0.1 Ray Chandonnet
- $1.0.2 \quad 12/15/2022$
- 1.0.3 The libraries you will use are already loaded for you below

```
[3]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from itertools import chain
```

1.1 Question 1

Read in the two Netflix CSV files from /Data/Netflix as pandas dataframes. Print the number of unique genres. This is not as simple as it sounds. You cannot simply find the length of titles['genres'].unique(). You must convert the output of that code to a list, iterate over that list and replace the following characters: []',. Once you have them replace you can split the individual strings to list items and flatten the list. I have already imported the chain() function for you to flatten the list. Look up the documentation to see its usage. There are 19 unique genres, but I want you to write the code to find them.

```
[8]: path = "/users/raychandonnet/Dropbox (Personal)/Merrimack College - MS in Data_U

Science/DSE5002/Week_8/Data/Netflix/"

credits=pd.read_csv(path + "credits.csv") # Read in credits file

titles=pd.read_csv(path + "titles.csv") # Read in titles file

genres=titles['genres'].tolist() # Turn genres column into a list

# now iterate through the list, eliminating brackets, apostrophes and spaces in_U

each text string so you are left with

# a string with each genre separated by commas. Then split that string into a_U

ilist using comma as the separator

# When done, you have a list of lists of genres, parsed properly without the_U

special characters

for i in range(0,len(genres)):

genres[i]=genres[i].replace("[","")

genres[i]=genres[i].replace("","")
```

There are 19 unique non-blank genres found in the title table:

```
Genre
0
    documentation
1
             crime
2
             drama
3
            comedy
4
           fantasy
5
            horror
6
          european
7
          thriller
8
            action
9
             music
10
           romance
11
            family
12
           western
13
               war
14
         animation
15
           history
16
             scifi
17
           reality
18
             sport
```

1.2 Question 2

Print the release year and the imdb score of the highest average score of all movies by year. This is trickier than it sounds. To do this you will need to aggregate the means by year. If you use the simple method you will get a pandas series. The series will need to be converted to a dataframe and the index will need to be set as a column (release year). Once you have done that you can find the numerical index with the highest average imdb score.

```
[9]: # First calculate the mean score by year - note that using "as_index=False"

→ returns a dataframe with normal 0:N index

# rather than a series so no need to convert to a dataframe

avg_by_year=titles.groupby('release_year',as_index=False)['imdb_score'].mean()

avg_by_year=avg_by_year.set_index('release_year') # change the index to the

→ release year
```

```
# Now I use the idxmax method to identify the index with the highest imdb_score;

→ Since the output comes as a series,

# I assume the first value [0] to a high score year variable as an integer
highest_score_year = avg_by_year.idxmax(axis=0)[0]

# Now I can directly access the score for that year using the year as my index
highest_score=avg_by_year['imdb_score'][highest_score_year]
print("The year with the highest average IMDB score was",highest_score_year)
print("The average IMDB score that year was",highest_score)
```

The year with the highest average IMDB score was 1985. The average IMDB score that year was 8.0

1.3 Question 3

There were 208 actors in the movie with the most credited actors. What is the title of that movie? Nulls and NaN values do not count.

```
[14]: # First Create a new table that joins the movie titles with the people credited,
      →by movie id
      titles_and_people=pd.merge(left=titles,right=credits,how='left',on='id')
      # Now slice that table so it is only is only showing type=MOVIE and role = ACTOR
      movies_and_actors = titles_and_people[(titles_and_people["type"] == 'MOVIE') &__
      movies_and_actors.shape
      actors_per_movie=actors_per_movie.set_index('id') # change the index to the id
      actors_per_movie=actors_per_movie.rename({'name':'actors'}, axis=1) # change_
      →column name
      # Now I use the idxmax method to identify the movie with the most actors; Since \Box
      → the output comes as a series,
      # I assign the first value [0] to a movie_id variable
      movie_id = actors_per_movie.idxmax(axis=0)[0]
      actors=actors_per_movie['actors'] [movie_id] # get the number of actors for thatu
      →movie
      movie_match=titles[titles['id']==movie_id] # Find the movie in the titles table_
      movie_name=movie_match.iloc[0]['title'] # pull the movie title from the movie_
      \rightarrow data
      print("The movie with the most actors in it was", movie_name)
      print("That movie had",actors,"actors in it")
      print("I disagree with the statement that the movie has 208 actors credited;")
      print("It has 208 people credited, but one of them is a DIRECTOR and not an,
      →ACTOR")
      nonactors = titles_and_people[(titles_and_people["id"] ==movie_id) &__
      →(titles_and_people["role"] != "ACTOR")]
      nonactors=nonactors[['id','title','name','role']]
```

```
print("The non-actors are:")
print(nonactors)
```

1.4 Question 4

Which movie has the highest IMDB score for the actor Robert De Niro? What year was it made? Create a kdeplot (kernel density estimation to show the distribution of his IMDB movie scores.

```
[17]: # Further slice the titles and people table for only De Niro flicks
     deniro_films = titles_and_people[(titles_and_people["type"] =='MOVIE') &__
      deniro_films.shape
     max_imdb=deniro_films.max(numeric_only=True)['imdb_score'] # identify the max_i
      \hookrightarrowscore
     max_match=deniro_films[(deniro_films['imdb_score']==max_imdb)] #slice the data__
      → frame for only films with that max score
     max_match=max_match[["title","type","release_year","imdb_score","name","role"]]_
      →# reduce columns to only relevant ones
     print("The highest imdb score found for movies featuring Robert De Niro⊔
      →is",max_imdb)
     print("Number of his movies with that score: ",len(max_match))
     print("Movie names and release dates are:")
     print(max_match[["title", "release_year"]])
      # Now build and display simple kdeplot
     data_to_plot = deniro_films['imdb_score']
     res = sns.kdeplot(data_to_plot,
                       color='purple',
                       fill=True)
     plt.title('De Niro Film IMDB Score Distribution')
     plt.ylabel('Density')
     plt.xlabel('IMDB Score')
     plt.show()
```

```
The highest imdb score found for movies featuring Robert De Niro is 8.3

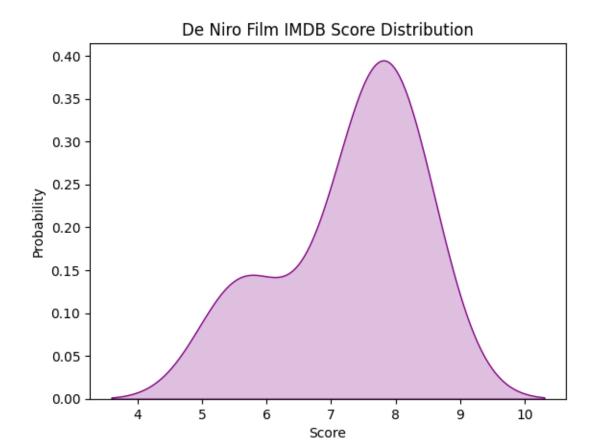
Number of his movies with that score: 2

Movie names and release dates are:

title release_year

Taxi Driver 1976

799 Once Upon a Time in America 1984
```



1.5 Question 5

Create two new boolean columns in the titles dataframe that are true when the description contains war or gangster. Call these columns war_movies and gangster_movies. How many movies are there in both categories? Which category has a higher average IMDB score? Show the IMDB score kernel density estimations of both categories.

```
print("The first way found", first_way_count1, "war movie(s)_
→and",first_way_count2, "gangster movies")
# However, when examining the results, 2 of those three matches include the
⇒string 'war' in a bigger word (specifically
# 'ward' and 'warden'.) Not really the intent. SO I can also do it using a_{\sqcup}
→regex that only searches for the whole word
# 'war', with different results
titles['war_movies'] = titles['description'].str.
titles['gangster_movies'] = titles['description'].str.
second_way_count1 = titles[titles.war_movies == True].shape[0]
second_way_count2 = titles[titles.gangster_movies == True].shape[0]
print("The second way found", second_way_count1, "war movie(s)_
→and",second_way_count2,"gangster movies")
# Now we calculate the mean for the slices of the table that identifies war_{\sqcup}
→ movies and gangster movies
war_IMDB=titles.loc[titles['war_movies']==True,'imdb_score'].mean()
gangster_IMDB=titles.loc[titles['gangster_movies']==True, 'imdb_score'].mean()
print("Average IMDB rating for war movies (using the whole-word search) ∪
→is",war_IMDB)
print("Average IMDB rating for gangster movies (using the whole-word search) ⊔
→is",gangster_IMDB)
if war_IMDB > gangster_IMDB:
   print("Therefore, average rating for war movies is greater")
elif gangster_IMDB > war_IMDB:
   print("Therefore, average rating for gangster movies is greater")
else:
   print("Therefore, their average ratings are the same")
# Finally the graph. I'm probably too much of a perfectionist but I really \Box
→wanted the two plots on the same graph!
# I discovered after a ton of digging that in order to do that, your two series.
→ have to have different names!
# So what I'm doing here is creating two series that slice the IMDB scores for
→war movies and gangster movies,
# delete any NaN values (since some don't have an IMDB score), and the rename_
\hookrightarrow the series. That lets me then plot
# the kernel densitiy on the same graph
war_ratings=titles[titles.war_movies == True]['imdb_score'].dropna() # slice war_
→movie IMDB scores
war_ratings=war_ratings.rename("War Movies") # rename
gangster_ratings=titles[titles.gangster_movies == True]['imdb_score'].dropna() #_U
→slice gangster movie IMDB scores
gangster_ratings=gangster_ratings.rename("Gangster Movies") # rename
fig = sns.kdeplot(war_ratings, fill=True, color="r") # plot war movie scores
```

```
fig = sns.kdeplot(gangster_ratings, fill=True, color="b") # plot gangster movie_

scores

plt.title('IMDB Score Distribution by Category') # make it pretty

plt.ylabel('Density')

plt.xlabel('IMDB Score')

plt.legend(labels=["War Movies", "Gangster Movies"],loc='upper right')

plt.show()
```

The first way found 437 war movie(s) and 35 gangster movies
The second way found 155 war movie(s) and 27 gangster movies
Average IMDB rating for war movies (using the whole-word search) is
6.9315068493150696

Average IMDB rating for gangster movies (using the whole-word search) is 6.344444444444444

Therefore, average rating for war movies is greater

IMDB Score Distribution by Category

