In [20]: # Import the numpy and pandas packages import numpy as np import pandas as pd import matplotlib.pyplot as plt Task 1: Reading and Inspection • ### Subtask 1.1: Import and read Import and read the movie database. Store it in a variable called movies. In [64]: #write your code here movies = pd.read csv('IMDB Movies.csv') OrgData = movies movies color director name num critic for reviews duration director facebook likes actor 3 facebook likes actor 2 name actor 1 facebook likes Out[64]: gross genres ... num_user_for_reviews language country content_rating budget title_year actor_2_facebook_likes Action|Adventure|Fantasy|Sci-James Joel David O Color 723.0 178.0 0.0 855.0 1000.0 760505847.0 USA PG-13 237000000.0 2009.0 936.0 3054 English Cameron Moore Orlando PG-13 300000000.0 302.0 169.0 563.0 1000.0 40000.0 309404152.0 Action|Adventure|Fantasy ... 1238 USA 2007.0 5000.0 1 Color Gore Verbinski English Bloom 602.0 148.0 0.0 11000.0 200074175.0 2015.0 393.0 **2** Color Sam Mendes 161.0 Rory Kinnear Action|Adventure|Thriller ... 994 English UK PG-13 245000000.0 Christopher PG-13 250000000.0 813.0 22000.0 23000.0 27000.0 448130642.0 USA 2012.0 23000.0 3 Color 164.0 Christian Bale Action|Thriller ... 2701 English Nolan NaN 131.0 131.0 NaN NaN 12.0 Doug Walker NaN Rob Walker NaN NaN NaN NaN 4 NaN NaN Documentary ... Daphne **5038** Color 87.0 2.0 318.0 637.0 2013.0 470.0 Scott Smith 1.0 NaN Comedy|Drama ... NaN English Canada NaN Zuniga **5039** Color 43.0 43.0 NaN 319.0 841.0 NaN Crime|Drama|Mystery|Thriller USA NaN 593.0 NaN Valorie Curry 359 TV-14 NaN English Benjamin Maxwell **5040** Color 13.0 76.0 0.0 0.0 NaN Drama|Horror|Thriller ... USA 1400.0 2013.0 0.0 0.0 3 English NaN Roberds Moody 489.0 Daniel Henney **5041** Color Daniel Hsia 14.0 100.0 0.0 946.0 10443.0 Comedy|Drama|Romance English USA PG-13 2012.0 719.0 NaN Brian **5042** Color 43.0 90.0 16.0 16.0 85222.0 USA PG 1100.0 2004.0 23.0 86.0 84 English Jon Gunn Documentary ... Herzlinger 5043 rows × 28 columns • ### Subtask 1.2: Inspect the dataframe Inspect the dataframe's columns, shapes, variable types etc. In [65]: #write your code here movies.shape (5043, 28) In [66]: movies.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 5043 entries, 0 to 5042 Data columns (total 28 columns): 5024 non-null object color 4939 non-null object director name num critic for reviews 4993 non-null float64 5028 non-null float64 duration 4939 non-null float64 director facebook likes actor_3_facebook_likes 5020 non-null float64 actor 2 name 5030 non-null object actor_1_facebook_likes 5036 non-null float64 gross 4159 non-null float64
genres 5043 non-null object
actor_1_name 5036 non-null object
movie_title 5043 non-null object
num_voted_users 5043 non-null int64 cast total facebook likes 5043 non-null int64 actor 3 name 5020 non-null object facenumber_in_poster 5030 non-null float64
plot_keywords 4890 non-null object
movie_imdb_link 5043 non-null object
num_user_for_reviews 5023 non-null object
language 5031 non-null object 5030 non-null float64 country 5038 non-null object content_rating 4740 non-null object budget 4551 non-null float64 title_year 4935 non-null float64 actor_2_facebook_likes 5030 non-null float64 5043 non-null float64 imdb score 4714 non-null float64 aspect ratio movie_facebook_likes 5043 non-null int64 dtypes: float64(12), int64(3), object(13) memory usage: 1.1+ MB Task 2: Cleaning the Data • ### Subtask 2.1: Inspect Null values Find out the number of Null values in all the columns and rows. Also, find the percentage of Null values in each column. Round off the percentages upto two decimal places. # Write your code for column-wise null count here movies.isnull().sum(axis=0).sort values(ascending=False) 884 Out[67]: budget 492 329 aspect ratio content_rating 303 plot_keywords 153 title_year 108 director name 104 director facebook likes 104 num_critic_for_reviews 50 actor_3_name 23 actor_3_facebook_likes 23 num_user_for_reviews 20 19 color 15 duration facenumber_in_poster 13 actor 2 name 13 actor_2_facebook likes 13 12 language actor_1_name actor_1_facebook_likes country movie_facebook_likes genres movie title num voted users movie imdb link imdb score cast total facebook likes dtype: int64 • ### Subtask 2.2: Drop unecessary columns For this assignment, you will mostly be analyzing the movies with respect to the ratings, gross collection, popularity of movies, etc. So many of the columns in this dataframe are not required. So it is advised to drop the following columns. color · director facebook likes • actor 1 facebook likes actor_2_facebook_likes actor_3_facebook_likes • actor 2 name cast_total_facebook_likes actor_3_name duration • facenumber in poster content_rating country movie_imdb_link aspect_ratio plot_keywords # Write your code for dropping the columns here. It is advised to keep inspecting the dataframe after each set of opera movies = movies.drop(['color', 'director facebook likes', 'actor 1 facebook likes', 'actor 2 facebook likes', 'actor 3 facebook likes', 'actor 2 name', 'cast total facebook likes', 'actor 3 name', 'duration', 'facenumber in poster', 'content rating', 'country', 'movie imdb link', 'aspect ratio', 'plot keywords'],axis=1) In [71]: movies Out[71]: director_name num_critic_for_reviews movie_title num_voted_users num_user_for_reviews language title_year imdb_score movie_facebook_likes gross actor_1_name budget genres James Cameron 723.0 760505847.0 Action|Adventure|Fantasy|Sci-Fi CCH Pounder Avatar 886204 3054 English 237000000.0 2009.0 7.9 33000 471220 English 300000000.0 2007.0 7.1 Gore Verbinski 302.0 309404152.0 Pirates of the Caribbean: At World's End 1238 0 Action|Adventure|Fantasy Johnny Depp Sam Mendes 602.0 200074175.0 Action|Adventure|Thriller Christoph Waltz 275868 English 245000000.0 2015.0 6.8 85000 Spectre 813.0 448130642.0 1144337 164000 3 Christopher Nolan The Dark Knight Rises 2701 English 250000000.0 2012.0 8.5 Action|Thriller Doug Walker Star Wars: Episode VII - The Force Awakens ... 7.1 Doug Walker NaN NaN 8 NaN NaN NaN 4 Documentary 0 629 Scott Smith Comedy|Drama 2013.0 7.7 84 5038 1.0 NaN Eric Mabius Signed Sealed Delivered 6 English NaN 5039 NaN 43.0 Crime|Drama|Mystery|Thriller The Following 73839 359 NaN 7.5 32000 NaN Natalie Zea English NaN 38 6.3 16 **5040** Benjamin Roberds 13.0 NaN Drama|Horror|Thriller A Plague So Pleasant 3 1400.0 2013.0 Eva Boehnke English 5041 1255 660 Daniel Hsia 14.0 10443.0 Comedy|Drama|Romance Shanghai Calling English 2012.0 6.3 Alan Ruck NaN 5042 4285 456 Jon Gunn 43.0 85222.0 Documentary John August My Date with Drew 84 English 1100.0 2004.0 6.6 5043 rows × 13 columns • ### Subtask 2.3: Drop unecessary rows using columns with high Null percentages Now, on inspection you might notice that some columns have large percentage (greater than 5%) of Null values. Drop all the rows which have Null values for such columns. In [72]: # Write your code for dropping the rows here ## round with 2 for 2 decimal points round(movies.isnull().sum().sort_values(ascending=False)/len(movies)*100,2) 17.53 budget 9.76 2.14 title_year director name 2.06 num critic for reviews 0.99 num_user_for_reviews 0.40 0.24 language actor 1 name 0.14 movie facebook likes 0.00 imdb score 0.00 num voted users 0.00 0.00 movie title 0.00 genres dtype: float64 movies = movies[movies['gross'].notnull()] movies = movies[movies['budget'].notnull()] round(movies.isnull().sum().sort values(ascending=False)/len(movies)*100,2) 0.08 language Out[74]: 0.08 actor 1 name 0.03 num critic for reviews movie facebook likes 0.00 0.00 imdb score title year 0.00 budget 0.00 0.00 num_user_for_reviews 0.00 num voted users 0.00 movie title 0.00 genres 0.00 gross 0.00 director name dtype: float64 • ### Subtask 2.4: Drop unecessary rows Some of the rows might have greater than five NaN values. Such rows aren't of much use for the analysis and hence, should be removed. In [75]: # Write your code for dropping the rows here (movies.isnull().sum(axis=1).sort values(ascending=False) > 5).sum() Out[75]: In [76]: movies = movies[movies.isnull().sum(axis=1).sort values(ascending=False) <= 5]movies c:\users\karan\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:1: UserWarning: Boolean Series key will be reindexed to match DataFrame index. """Entry point for launching an IPython kernel. Out[76]: director_name num_critic_for_reviews budget title_year imdb_score movie_facebook_likes actor_1_name movie_title num_voted_users num_user_for_reviews language gross **0** James Cameron 723.0 760505847.0 2009.0 Action|Adventure|Fantasy|Sci-Fi **CCH Pounder** Avatar 886204 English 237000000.0 7.9 33000 Gore Verbinski 302.0 309404152.0 471220 English 300000000.0 2007.0 0 Action|Adventure|Fantasy Johnny Depp Pirates of the Caribbean: At World's End 7.1 Sam Mendes 602.0 200074175.0 275868 English 245000000.0 2015.0 85000 Action|Adventure|Thriller Christoph Waltz 6.8 Spectre 3 Christopher Nolan 813.0 448130642.0 Action|Thriller The Dark Knight Rises 1144337 2701 English 250000000.0 2012.0 8.5 164000 Tom Hardy 462.0 73058679.0 Action|Adventure|Sci-Fi 212204 English 263700000.0 2012.0 24000 **5** Andrew Stanton Daryl Sabara John Carter 6.6 Primer 5033 Shane Carruth 143.0 424760.0 Drama|Sci-Fi|Thriller Shane Carruth 72639 7000.0 2004.0 7.0 19000 English 5034 70071.0 589 7000.0 2005.0 74 Neill Dela Llana 35.0 lan Gamazon Cavite 35 English 6.3 52055 **5035** Robert Rodriguez 56.0 2040920.0 Action|Crime|Drama|Romance|Thriller Carlos Gallardo El Mariachi 130 Spanish 7000.0 1992.0 6.9 0 5037 **Edward Burns** 4584.0 1338 2011.0 413 14.0 Comedy|Drama Kerry Bishé Newlyweds 9000.0 5042 43.0 85222.0 My Date with Drew 4285 2004.0 6.6 456 Jon Gunn 1100.0 Documentary John August $3891 \text{ rows} \times 13 \text{ columns}$ • ### Subtask 2.5: Fill NaN values You might notice that the language column has some NaN values. Here, on inspection, you will see that it is safe to replace all the missing values with 'English'. In [77]: # Write your code here round(movies.isnull().sum().sort values(ascending=False)/len(movies)*100,2) 0.08 language Out[77]: actor 1 name 0.08 num critic for reviews 0.03 movie facebook likes 0.00 imdb score 0.00 title year 0.00 budget 0.00 num user for reviews 0.00 num voted users 0.00 movie title 0.00 0.00 genres gross 0.00 director name 0.00 dtype: float64 movies.groupby('language').language.count().sort values(ascending=False) language 3707 English 37 French Spanish 26 15 Mandarin 13 German 12 Japanese 10 Hindi Cantonese 8 Italian Korean Portuguese Norwegian Hebrew Persian Dutch Danish Thai Dari Indonesian Aboriginal Icelandic Hungarian Arabic Aramaic Bosnian Telugu Czech Swedish Russian Romanian Dzongkha None Filipino Mongolian Maya Kazakh Vietnamese Zulu Name: language, dtype: int64 movies.language.describe() 3888 count Out[79]: 38 unique English top 3707 freq Name: language, dtype: object movies.language = movies.language.fillna('English') In [81]: round (movies.isnull().sum().sort values (ascending=False) /len(movies) *100,2) actor 1 name Out[81]: num critic for reviews 0.03 movie facebook likes 0.00 imdb score 0.00 title year 0.00 budget 0.00 language 0.00 num user for reviews 0.00 num voted users 0.00 movie title 0.00 genres 0.00 0.00 0.00 director name dtype: float64 • ### Subtask 2.6: Check the number of retained rows You might notice that two of the columns viz. num_critic_for_reviews and actor_1_name have small percentages of NaN values left. You can let these columns as it is for now. Check the number and percentage of the rows retained after completing all the tasks above. # Write your code for checking number of retained rows here len(movies)/len(OrgData) * 100 77.15645449137418 Out[84]: **Checkpoint 1:** You might have noticed that we still have around 77% of the rows! Task 3: Data Analysis • ### Subtask 3.1: Change the unit of columns Convert the unit of the budget and gross columns from \$ to million \$. In [87]: # Write your code for unit conversion here movies['budget'] = movies['budget']/1000000 movies['gross'] = movies['gross']/1000000 In [88]: movies Out[88]: director_name num_critic_for_reviews actor_1_name movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes James Cameron 723.0 760.505847 Action|Adventure|Fantasy|Sci-Fi CCH Pounder 886204 English 237.0000 2009.0 7.9 33000 Gore Verbinski 302.0 309.404152 Johnny Depp Pirates of the Caribbean: At World's End 471220 English 300.0000 2007.0 0 Action|Adventure|Fantasy 6.8 85000 Sam Mendes 602.0 200.074175 Action|Adventure|Thriller Christoph Waltz 275868 English 245.0000 2015.0 **3** Christopher Nolan 813.0 448.130642 Action|Thriller Tom Hardy The Dark Knight Rises 1144337 English 250.0000 2012.0 164000 **5** Andrew Stanton 462.0 73.058679 Action|Adventure|Sci-Fi Daryl Sabara John Carter 212204 English 263.7000 2012.0 6.6 24000 72639 7.0 5033 Shane Carruth 0.424760 Drama|Sci-Fi|Thriller Shane Carruth English 0.0070 2004.0 19000 5034 Neill Dela Llana 35.0 0.070071 Ian Gamazon Cavite English 0.0070 2005.0 52055 1992.0 6.9 **5035** Robert Rodriguez 56.0 2.040920 Action|Crime|Drama|Romance|Thriller Carlos Gallardo El Mariachi Spanish 0.0070 456 5042 Jon Gunn 43.0 0.085222 My Date with Drew 4285 0.0011 2004.0 Documentary John August 3891 rows × 13 columns • ### Subtask 3.2: Find the movies with highest profit 1. Create a new column called profit which contains the difference of the two columns: gross and budget. 2. Sort the dataframe using the profit column as reference. 3. Extract the top ten profiting movies in descending order and store them in a new dataframe - top10 # Write your code for creating the profit column here movies['profit'] = movies['gross'] - movies['budget'] Out[89]: profit director_name num_critic_for_reviews movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes gross genres actor_1_name Action|Adventure|Fantasy|Sci-Fi 33000 523.505847 James Cameron 723.0 760.505847 CCH Pounder 886204 3054 English 237.0000 2009.0 7.9 Avatar Gore Verbinski 302.0 309.404152 Johnny Depp Pirates of the Caribbean: At World's End 471220 1238 English 300.0000 2007.0 7.1 9.404152 Action|Adventure|Fantasy 0 English 245.0000 6.8 2 Sam Mendes 602.0 200.074175 275868 994 2015.0 85000 -44.925825 Action|Adventure|Thriller Christoph Waltz Spectre 1144337 2012.0 8.5 **3** Christopher Nolan 813.0 448.130642 Action|Thriller The Dark Knight Rises 2701 English 250.0000 164000 198.130642 Tom Hardy **5** Andrew Stanton 462.0 73.058679 Action|Adventure|Sci-Fi Daryl Sabara John Carter 212204 738 English 263.7000 2012.0 6.6 24000 -190.641321 5033 Shane Carruth 143.0 0.424760 Drama|Sci-Fi|Thriller Shane Carruth Primer 72639 371 English 0.0070 2004.0 7.0 19000 0.417760 6.3 5034 Neill Dela Llana 35.0 0.070071 Cavite 589 35 0.0070 2005.0 74 0.063071 Thriller Ian Gamazon English 5035 Robert Rodriguez 2.040920 Action|Crime|Drama|Romance|Thriller Carlos Gallardo El Mariachi 52055 130 1992.0 6.9 0 2.033920 56.0 Spanish 0.0070 5037 **Edward Burns** 14.0 0.004584 Comedy|Drama Kerry Bishé Newlyweds 1338 English 0.0090 2011.0 6.4 413 -0.004416 John August 6.6 5042 Jon Gunn 43.0 0.085222 My Date with Drew 4285 84 2004.0 456 0.084122 Documentary English 0.0011 $3891 \text{ rows} \times 14 \text{ columns}$ In [90]: # Write your code for sorting the dataframe here movies.sort values(by='profit',ascending=False) Out[90]: budget title_year imdb_score movie_facebook_likes profit director_name num_critic_for_reviews movie_title num_voted_users num_user_for_reviews gross genres actor_1_name language **0** James Cameron 523.505847 723.0 760.505847 Action|Adventure|Fantasy|Sci-Fi **CCH Pounder** 886204 3054 237.000000 2009.0 7.9 33000 Avatar English 644.0 652.177271 Action|Adventure|Sci-Fi|Thriller Bryce Dallas Howard Jurassic World 418214 1290 150.000000 7.0 150000 502.177271 29 Colin Trevorrow English 2015.0 315.0 658.672302 7.7 458.672302 James Cameron Drama|Romance Leonardo DiCaprio Titanic 793059 2528 200.000000 1997.0 26000 26 English Harrison Ford Star Wars: Episode IV - A New Hope 282.0 460.935665 Action|Adventure|Fantasy|Sci-Fi 911097 1470 8.7 33000 449.935665 3024 George Lucas English 11.000000 1977.0 7.9 **3080** Steven Spielberg 215.0 434.949459 Family|Sci-Fi **Henry Thomas** E.T. the Extra-Terrestrial 281842 515 English 10.500000 1982.0 34000 424.449459 2334 Katsuhiro Ôtomo 105.0 0.410388 Action|Adventure|Animation|Family|Sci-Fi|Thriller William Hootkins Steamboy 13727 79 2127.519898 2004.0 6.9 973 -2127.109510 Japanese Adventure|Animation|Fantasy Hayao Miyazaki 2.298191 Minnie Driver Princess Mononoke 221552 570 2400.000000 8.4 11000 -2397.701809 2323 174.0 Japanese 1997.0 5603 7.1 3005 0.195888 Drama|Romance|War Marcell Nagy **Fateless** 45 Hungarian 2500.000000 2005.0 607 -2499.804112 Lajos Koltai 73.0 3859 Chan-wook Park 202.0 0.211667 Crime|Drama Min-sik Choi Lady Vengeance 53508 131 Korean 4200.000000 2005.0 7.7 4000 -4199.788333 68883 7.0 7000 -12213.298588 2988 Joon-ho Bong 363.0 2.201412 Comedy|Drama|Horror|Sci-Fi The Host 279 12215.500000 2006.0 Doona Bae Korean $3891 \text{ rows} \times 14 \text{ columns}$ # Write your code for top10 movies top10 = movies.sort values(by='profit', ascending=False).head(10) top10 Out[91]: movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes director_name num_critic_for_reviews actor_1_name profit gross genres Action|Adventure|Fantasy|Sci-Fi James Cameron 33000 523.505847 723.0 760.505847 **CCH Pounder** Avatar 886204 3054 English 237.0 2009.0 7.9 418214 7.0 644.0 652.177271 Action|Adventure|Sci-Fi|Thriller Bryce Dallas Howard Jurassic World 1290 150.0 2015.0 150000 502.177271 29 Colin Trevorrow English 315.0 658.672302 26 James Cameron Drama|Romance Leonardo DiCaprio Titanic 793059 2528 English 200.0 1997.0 7.7 26000 458.672302 282.0 460.935665 Action|Adventure|Fantasy|Sci-Fi Star Wars: Episode IV - A New Hope 911097 1470 1977.0 8.7 33000 449.935665 3024 George Lucas Harrison Ford English 11.0 281842 10.5 1982.0 7.9 34000 424.449459 3080 Steven Spielberg 215.0 434.949459 Family|Sci-Fi Henry Thomas E.T. the Extra-Terrestrial 515 English 703.0 623.279547 Chris Hemsworth 995415 1722 220.0 2012.0 8.1 123000 403.279547 794 Joss Whedon Action|Adventure|Sci-Fi The Avengers English 8.1 123000 403.279547 703.0 623.279547 The Avengers 995415 1722 2012.0 17 Joss Whedon Action|Adventure|Sci-Fi Chris Hemsworth English 220.0 186.0 422.783777 Adventure|Animation|Drama|Family|Musical 509 Roger Allers Matthew Broderick The Lion King 644348 656 English 45.0 1994.0 8.5 17000 377.783777 320.0 474.544677 Action|Adventure|Fantasy|Sci-Fi Natalie Portman Star Wars: Episode I - The Phantom Menace 534658 3597 1999.0 6.5 13000 359.544677 240 George Lucas English 115.0 **66** Christopher Nolan 645.0 533.316061 Action|Crime|Drama|Thriller The Dark Knight 1676169 4667 185.0 2008.0 9.0 37000 348.316061 Christian Bale English • ### Subtask 3.3: Drop duplicate values After you found out the top 10 profiting movies, you might have notice a duplicate value. So, it seems like the dataframe has duplicate values as well. Drop the duplicate values from the dataframe and repeat Subtask 3.2. # Write your code for dropping duplicate values here movies.drop_duplicates(keep='first',inplace=True) In [93]: # Write code for repeating subtask 2 here movies Out[93]: director_name num_critic_for_reviews actor_1_name movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes profit James Cameron 723.0 760.505847 Action|Adventure|Fantasy|Sci-Fi CCH Pounder 886204 English 237.0000 2009.0 7.9 33000 523.505847 302.0 309.404152 Johnny Depp Pirates of the Caribbean: At World's End English 300.0000 Gore Verbinski 471220 1238 2007.0 7.1 9.404152 Action|Adventure|Fantasy 602.0 200.074175 Sam Mendes Action|Adventure|Thriller Christoph Waltz Spectre 275868 English 245.0000 2015.0 6.8 85000 -44.925825 **3** Christopher Nolan 813.0 448.130642 Action|Thriller Tom Hardy The Dark Knight Rises 1144337 2701 English 250.0000 2012.0 8.5 164000 198.130642 Andrew Stanton 462.0 73.058679 Action|Adventure|Sci-Fi Daryl Sabara John Carter 212204 738 English 263.7000 2012.0 6.6 24000 -190.641321 7.0 5033 Shane Carruth 143.0 0.424760 Drama|Sci-Fi|Thriller Shane Carruth Primer 72639 371 English 0.0070 2004.0 19000 0.417760 5034 Neill Dela Llana 35.0 0.070071 Thriller Ian Gamazon Cavite 589 35 English 0.0070 2005.0 6.3 0.063071 6.9 5035 Robert Rodriguez 2.040920 Action|Crime|Drama|Romance|Thriller Carlos Gallardo El Mariachi 52055 130 0.0070 1992.0 2.033920 Spanish 5037 **Edward Burns** 14.0 0.004584 Comedy|Drama Kerry Bishé Newlyweds 1338 0.0090 2011.0 6.4 413 -0.004416 English 5042 Jon Gunn 43.0 0.085222 John August My Date with Drew 4285 English 0.0011 2004.0 6.6 456 0.084122 Documentary 3856 rows × 14 columns In [94]: top10 = movies.sort values(by='profit', ascending=False).head(10) top10 Out[94]: director_name num_critic_for_reviews gross genres actor_1_name movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes profit James Cameron 723.0 760.505847 Action|Adventure|Fantasy|Sci-Fi **CCH Pounder** Avatar 886204 3054 237.0 2009.0 7.9 33000 523.505847 English 150000 502.177271 29 Colin Trevorrow 644.0 652.177271 Action|Adventure|Sci-Fi|Thriller Bryce Dallas Howard Jurassic World 418214 1290 English 150.0 2015.0 7.0 James Cameron 315.0 658.672302 Drama|Romance Leonardo DiCaprio Titanic 793059 2528 English 200.0 1997.0 7.7 26000 458.672302 Action|Adventure|Fantasy|Sci-Fi 282.0 460.935665 Harrison Ford Star Wars: Episode IV - A New Hope 911097 1470 English 11.0 1977.0 8.7 33000 449.935665 George Lucas Steven Spielberg Family|Sci-Fi 34000 281842 703.0 623.279547 Joss Whedon Action|Adventure|Sci-Fi Chris Hemsworth 995415 1722 English 220.0 2012.0 8.1 123000 403.279547 17 The Avengers 186.0 422.783777 Adventure|Animation|Drama|Family|Musical The Lion King 509 Roger Allers Matthew Broderick 644348 656 45.0 1994.0 8.5 17000 377.783777 English 240 George Lucas 320.0 474.544677 Action|Adventure|Fantasy|Sci-Fi Natalie Portman Star Wars: Episode I - The Phantom Menace 534658 3597 English 115.0 1999.0 6.5 13000 359.544677 **66** Christopher Nolan 645.0 533.316061 Action|Crime|Drama|Thriller Christian Bale The Dark Knight 1676169 4667 185.0 2008.0 9.0 37000 348.316061 English 439 **Gary Ross** 673.0 407.999255 Adventure|Drama|Sci-Fi|Thriller Jennifer Lawrence The Hunger Games 701607 1959 78.0 2012.0 7.3 140000 329.999255 English In []: Checkpoint 2: You might spot two movies directed by James Cameron in the list. • ### Subtask 3.4: Find IMDb Top 250 1. Create a new dataframe IMDb_Top_250 and store the top 250 movies with the highest IMDb Rating (corresponding to the column: imdb_score). Also make sure that for all of these movies, the num_voted_users is greater than 25,000. Also add a Rank column containing the values 1 to 250 indicating the ranks of the corresponding films. 2. Extract all the movies in the IMDb_Top_250 dataframe which are not in the English language and store them in a new dataframe named Top_Foreign_Lang_Film. # Write your code for extracting the top 250 movies as per the IMDb score here. Make sure that you store it in a new dataframe # and name that dataframe as 'IMDb Top 250' IMDb_Top_250 = movies[movies['num_voted_users'] > 25000].sort_values(by='imdb_score',ascending=False).head(250) IMDb_Top_250 Out[101.. director name num critic for reviews actor_1_name movie_title num_voted_users num_user_for_reviews budget title_year imdb_score movie_facebook_likes profit gross genres language 3.341469 1937 199.0 28.341469 Frank Darabont Crime|Drama Morgan Freeman The Shawshank Redemption 1689764 1994.0 9.3 108000 4144 English 25.00 3466 Francis Ford Coppola 208.0 134.821952 The Godfather 1155770 2238 6.00 1972.0 9.2 43000 128.821952 Crime|Drama Al Pacino English Robert De Niro 149.0 57.300000 790926 1974.0 9.0 44.300000 2837 Francis Ford Coppola Crime|Drama The Godfather: Part II 650 13.00 14000 English Action|Crime|Drama|Thriller 4667 2008.0 9.0 Christopher Nolan 645.0 533.316061 The Dark Knight 1676169 37000 348.316061 Christian Bale English 185.00 4498 6.100000 503509 780 1966.0 8.9 Sergio Leone 181.0 Western Clint Eastwood The Good, the Bad and the Ugly Italian 1.20 20000 4.900000 4931 John Carney 232.0 9.437933 Drama|Music|Romance Glen Hansard Once 90827 329 English 0.18 2007.0 7.9 26000 9.257933 287.0 128.067808 Action|Drama|Romance 217740 2000.0 7.9 2605 Chen Chang Crouching Tiger, Hidden Dragon 1641 15.00 0 113.067808 Ang Lee Mandarin 3029 David O. Russell 275869 2010.0 7.9 36000 68.571803 410.0 93.571803 Biography|Drama|Sport The Fighter 389 Christian Bale English 25.00 16000 36.362352 2177 111.0 56.362352 **Edward Scissorhands** 357581 1990.0 7.9 Tim Burton Fantasy|Romance Johnny Depp 588 English 20.00 2487 72.000000 Drama|Family|Musical|Romance 66959 1964.0 7.9 55.000000 George Cukor 82.0 My Fair Lady 258 17.00 0 Jeremy Brett English 250 rows × 14 columns In [102... IMDb_Top_250['Rank'] = IMDb_Top_250['imdb_score'].rank(method='first',ascending=False) IMDb_Top_250 Out[102... movie_title num_voted_users num_user_for_reviews language budget title_year imdb_score movie_facebook_likes director_name num_critic_for_reviews profit Rank gross genres actor_1_name 199.0 28.341469 1937 1994.0 Frank Darabont The Shawshank Redemption 1689764 9.3 108000 3.341469 1.0 Crime|Drama Morgan Freeman 4144 English 25.00 **3466** Francis Ford Coppola 208.0 134.821952 The Godfather 1155770 2238 6.00 1972.0 9.2 43000 128.821952 2.0 Crime|Drama Al Pacino English The Godfather: Part II Robert De Niro 790926 650 1974.0 9.0 2837 Francis Ford Coppola 149.0 57.300000 Crime|Drama 14000 44.300000 3.0 English 13.00 645.0 533.316061 Action|Crime|Drama|Thriller Christian Bale 1676169 4667 2008.0 9.0 Christopher Nolan The Dark Knight 37000 348.316061 4.0 English 185.00 6.100000 780 1966.0 8.9 20000 4498 Sergio Leone 181.0 Clint Eastwood The Good, the Bad and the Ugly 503509 Italian 1.20 4.900000 5.0 Western 4931 John Carney 232.0 9.437933 Drama|Music|Romance Glen Hansard 90827 329 0.18 2007.0 7.9 26000 9.257933 246.0 English 2605 287.0 128.067808 217740 1641 Mandarin 2000.0 7.9 0 113.067808 247.0 Action|Drama|Romance Chen Chang Crouching Tiger, Hidden Dragon 15.00 Ang Lee 410.0 93.571803 275869 2010.0 7.9 36000 68.571803 248.0 3029 Christian Bale The Fighter 389 25.00 David O. Russell Biography|Drama|Sport English 111.0 56.362352 2177 357581 588 1990.0 7.9 16000 36.362352 249.0 Tim Burton Fantasy|Romance Johnny Depp **Edward Scissorhands** English 20.00 55.000000 250.0 2487 George Cukor 72.000000 Drama|Family|Musical|Romance 66959 258 1964.0 7.9 82.0 My Fair Lady 17.00 0 Jeremy Brett English 250 rows × 15 columns IMDb_Top_250[IMDb_Top_250['language']!='English'] Out[106.. director name num critic for reviews gross budget title_year imdb_score movie_facebook_likes movie_title num_voted_users num_user_for_reviews profit Rank actor_1_name language genres The Good, the Bad and the Clint Eastwood 4498 6.100000 503509 1.200000 1966.0 8.9 20000 4.900000 5.0 Sergio Leone 181.0 Western 780 Italian Ugly Takashi Shimura 4747 153.0 0.269061 Action|Adventure|Drama Seven Samurai 229012 596 2.000000 1954.0 8.7 11000 -1.730939 17.0 Akira Kurosawa Japanese 4029 7.563397 533200 3.300000 2002.0 8.7 28000 4.263397 20.0 Fernando Meirelles 214.0 Crime|Drama Alice Braga City of God 749 Portuguese 2373 10.049886 417971 19.000000 2001.0 28000 -8.950114 23.0 246.0 Adventure | Animation | Family | Fantasy 902 8.6 Hayao Miyazaki Bunta Sugawara Spirited Away Japanese Florian Henckel von The Lives of Others 4259 215.0 11.284657 Sebastian Koch 259379 2.000000 2006.0 8.5 39000 9.284657 35.0 Drama|Thriller 407 German Donnersmarck 4921 0.925402 Children of Heaven 27882 130 0.180000 1997.0 8.5 0.745402 39.0 Majid Majidi 46.0 Drama|Family Bahare Seddiqi Persian 2323 2.298191 221552 1997.0 8.4 11000 47.0 174.0 Adventure|Animation|Fantasy Minnie Driver 570 2400.000000 -2397.701809 Hayao Miyazaki Princess Mononoke Japanese Das Boot 2970 11.433134 168203 14.000000 1981.0 11000 Wolfgang Petersen 96.0 Adventure|Drama|Thriller|War Jürgen Prochnow 426 8.4 -2.566866 49.0 German 4105 305.0 2.181290 356181 3.000000 2003.0 43000 -0.818710 57.0 Chan-wook Park Min-sik Choi Oldboy 809 8.4 Drama|Mystery|Thriller Korean 4659 354.0 7.098492 Shahab Hosseini 151812 264 0.500000 2011.0 48000 6.598492 58.0 Asghar Farhadi Drama|Mystery A Separation Persian 8.4 Baahubali: The Beginning 1329 6.498000 62756 18.026148 2015.0 21000 -11.528148 59.0 44.0 Action|Adventure|Drama|Fantasy|War Tamannaah Bhatia 410 8.4 S.S. Rajamouli Telugu 1298 33.201661 534262 77.000000 2001.0 39000 -43.798339 242.0 1314 61.0 Jean-Pierre Jeunet Comedy|Romance Mathieu Kassovitz Amélie French 8.4 2734 260.0 0.026435 111841 6.000000 1927.0 8.3 12000 -5.973565 65.0 Fritz Lang Drama|Sci-Fi Brigitte Helm Metropolis 413 German -3.189032 73.0 4033 349.0 0.610968 170155 3.800000 2012.0 60000 The Hunt 249 8.3 Thomas Vinterberg Drama Thomas Bo Larsen Danish Thomas Oliver Hirschbiegel 2829 192.0 5.501940 248354 13.500000 2004.0 8.3 14000 -7.998060 81.0 Biography|Drama|History|War Downfall 564 German Kretschmann 2551 406.0 37.623143 467234 1083 13.500000 2006.0 8.2 27000 24.123143 86.0 Guillermo del Toro Drama|Fantasy|War Ivana Baquero Pan's Labyrinth Spanish 20.167424 4000 131831 2.000000 2009.0 8.2 33000 18.167424 97.0 262.0 The Secret in Their Eyes 231 Juan José Campanella Drama|Mystery|Thriller Ricardo Darín Spanish 3550 Denis Villeneuve 226.0 6.857096 Drama|Mystery|War 80429 156 6.800000 2010.0 8.2 37000 0.057096 98.0 Lubna Azabal Incendies French 2047 4.710455 Christian Bale 214091 24.000000 2004.0 8.2 13000 -19.289545 100.0 212.0 Adventure | Animation | Family | Fantasy 330 Hayao Miyazaki Howl's Moving Castle Japanese 157.0 2.086345 64556 10.000000 2004.0 -7.913655 110.0 2830 140 Spanish 8.1 Alejandro Amenábar Biography|Drama|Romance Belén Rueda The Sea Inside Tae Guk Gi: The 2914 86.0 1.110186 31943 12.800000 2004.0 8.1 -11.689814 119.0 Action|Drama|War Min-sik Choi Je-kyu Kang 224 Korean Brotherhood of War 4461 1.647780 Ulrich Thomsen 65951 258 1.300000 1998.0 5000 0.347780 127.0 98.0 Drama The Celebration Danish 8.1 Thomas Vinterberg 3553 142.0 0.008060 Action|Crime|Drama|Thriller 81644 107 Portuguese 4.000000 2007.0 8.1 11000 José Padilha Elite Squad -3.991940 143.0 Wagner Moura 3423 150.0 0.439162 106160 1100.000000 1988.0 0 -1099.560838 147.0 Katsuhiro Ôtomo Action|Animation|Sci-Fi Mitsuo Iwata Akira 430 8.1 Japanese 4267 157.0 173551 361 2000.0 11000 3.383834 152.0 Alejandro G. Iñárritu 5.383834 Drama|Thriller Adriana Barraza **Amores Perros** 2.000000 8.1 Spanish Catherine 4.443403 70194 7.300000 2007.0 14000 -2.856597 166.0 3456 242.0 Animation|Biography|Drama|War Persepolis 158 8.0 Vincent Paronnaud French Deneuve 210.0 4.018695 Shah Rukh Khan 69759 235 2010.0 8.0 27000 -7.981305 175.0 3344 Karan Johar Adventure|Drama|Thriller My Name Is Khan Hindi 12.000000 Fernanda 4144 Walter Salles 71.0 5.595428 **Central Station** 28951 257 Portuguese 2.900000 1998.0 8.0 2.695428 177.0 Drama Montenegro 4284 2.283276 Animation|Biography|Documentary|Drama|History|War 46107 156 1.500000 2008.0 8.0 0.783276 181.0 Ari Folman 231.0 Ari Folman Waltz with Bashir Hebrew 3.300000 183.0 4897 122.0 3.500000 147566 0.200000 1964.0 Sergio Leone Action|Drama|Western Clint Eastwood A Fistful of Dollars 235 Italiar 8.0 1171 Yimou Zhang 283.0 0.084961 149414 31.000000 2002.0 7.9 -30.915039 212.0 841 Action|Adventure|History Jet Li Hero Mandarin -5.246069 226.0 2863 Clint Eastwood 251.0 13.753931 Drama|History|War Yuki Matsuzaki Letters from Iwo Jima 132149 316 19.000000 2006.0 7.9 5000 Japanese 3264 447.0 0.225377 70382 190 8.900000 2012.0 7.9 33000 -8.674623 229.0 Michael Haneke Drama|Romance Isabelle Huppert French Amour 3510 Yash Chopra 2.921738 Drama|Musical|Romance Shah Rukh Khan 34449 119 7.000000 2004.0 7.9 2000 -4.078262 233.0 29.0 Veer-Zaara Hindi Jean-Baptiste 3677 Christophe Barratier 112.0 3.629758 Drama|Music The Chorus 44151 5.500000 2004.0 7.9 -1.870242 237.0 110 French Maunier -0.278739 243.0 4415 94.0 1.221261 Crime|Drama|Thriller 38215 1.500000 2000.0 7.9 Fabián Bielinsky Ricardo Darín Nine Queens 125 Spanish 4 Months, 3 Weeks and 2 44763 4640 Cristian Mungiu 233.0 1.185783 Drama Anamaria Marinca 0.590000 2007.0 7.9 14000 0.595783 244.0 172 Romanian Days Crouching Tiger, Hidden Action|Drama|Romance 0 113.067808 247.0 2605 287.0 128.067808 Chen Chang 217740 15.000000 2000.0 7.9 Ang Lee 1641 Mandarin Dragon **Checkpoint 3:** Can you spot Veer-Zaara in the dataframe? • ### Subtask 3.5: Find the best directors 1. Group the dataframe using the director_name column. 2. Find out the top 10 directors for whom the mean of imdb_score is the highest and store them in a new dataframe top10director . In [109... # Write your code for extracting the top 10 directors here top10director = movies.groupby('director name').imdb score.mean().sort values(ascending=False).head(10) top10director director name Out[109... Charles Chaplin 8.600000 8.600000 Tony Kaye 8.500000 Ron Fricke 8.500000 Damien Chazelle Majid Majidi 8.500000 Alfred Hitchcock 8.500000 Sergio Leone 8.433333 8.425000 Christopher Nolan Asghar Farhadi 8.400000 Richard Marquand 8.400000 Name: imdb score, dtype: float64 **Checkpoint 4:** No surprises that Damien Chazelle (director of Whiplash and La La Land) is in this list.