In [46]: # Importing Pandas
 import pandas as pd
 import csv # read and write csv files
 from datetime import datetime # operations to parse dates
 import matplotlib.pyplot as plt
 % matplotlib inline
 import numpy as np
 import seaborn as sns

In [34]: #reading tmdb csv file and storing that to a variable
 tmdb = pd.read_csv('tmdb-movies.csv')

#calling out first 100 rows (excluding headers) of tmdb database
 tmdb.head(101)

#lets give a list of movies that needs to be deleted
 del_col = ['id', 'imdb_id', 'popularity', 'budget_adj', 'revenue_adj',
 'homepage', 'keywords', 'overview', 'production_companies', 'vote_count'
 , 'vote_average']

#deleting the columns from the database
 movie_data = tmdb.drop(del_col, 1)
 #now take a look at this new dataset
 movie_data.head()

Out[34]:

	budget	revenue	original_title	cast	director	tagline	runtime
C	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open.	124
1	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day.	120
2	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Can Destroy You	119
3	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	Every generation has a story.	136
4	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle 	James Wan	Vengeance Hits Home	137

```
In [36]: # Describing TMDB Data
movie data.describe()
```

Out[36]:

	budget	revenue	runtime	release_year
count	1.086600e+04	1.086600e+04	10866.000000	10866.000000
mean	1.462570e+07	3.982332e+07	102.070863	2001.322658
std	3.091321e+07	1.170035e+08	31.381405	12.812941
min	0.000000e+00	0.000000e+00	0.000000	1960.000000
25%	0.000000e+00	0.000000e+00	90.000000	1995.000000
50%	0.000000e+00	0.000000e+00	99.000000	2006.000000
75%	1.500000e+07	2.400000e+07	111.000000	2011.000000
max	4.250000e+08	2.781506e+09	900.000000	2015.000000

```
In [37]: # Perform operations to inspect data
# types and look for instances of missing or possibly errant data.
movie_data.info()
```

```
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 10 columns):
budget
                 10866 non-null int64
revenue
                 10866 non-null int64
                 10866 non-null object
original title
cast
                  10790 non-null object
director
                 10822 non-null object
tagline
                  8042 non-null object
runtime
                 10866 non-null int64
genres
                 10843 non-null object
release date
                10866 non-null object
release year
                 10866 non-null int64
dtypes: int64(4), object(6)
memory usage: 849.0+ KB
```

<class 'pandas.core.frame.DataFrame'>

```
In [38]: sum(movie_data.duplicated())
```

Out[38]: 1

```
In [39]: movie_data.drop_duplicates(inplace=True)
```

In [40]: movie_data.info()

<class 'pandas.core.frame.DataFrame'> Int64Index: 10865 entries, 0 to 10865 Data columns (total 10 columns): 10865 non-null int64 budget revenue 10865 non-null int64 original_title 10865 non-null object cast 10789 non-null object director 10821 non-null object tagline 8041 non-null object runtime 10865 non-null int64 10842 non-null object genres release_date 10865 non-null object release_year 10865 non-null int64 dtypes: int64(4), object(6) memory usage: 933.7+ KB

In [41]: movie_data.isnull().sum()

0 Out[41]: budget 0 revenue original_title 0 cast 76 director 44 tagline 2824 runtime 0 genres 23 0 release date release_year 0 dtype: int64

In [42]: movie data.describe()

Out[42]:

	budget	revenue	runtime	release_year
count	1.086500e+04	1.086500e+04	10865.000000	10865.000000
mean	1.462429e+07	3.982690e+07	102.071790	2001.321859
std	3.091428e+07	1.170083e+08	31.382701	12.813260
min	0.000000e+00	0.000000e+00	0.000000	1960.000000
25%	0.000000e+00	0.000000e+00	90.000000	1995.000000
50%	0.000000e+00	0.000000e+00	99.000000	2006.000000
75%	1.500000e+07	2.400000e+07	111.000000	2011.000000
max	4.250000e+08	2.781506e+09	900.000000	2015.000000

In [43]: movie_data.head()

Out[43]:

	budget	revenue	original_title	cast	director	tagline	runtime
0	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open.	124
1	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day.	120
2	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Can Destroy You	119
3	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	Every generation has a story.	136
4	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle 	James Wan	Vengeance Hits Home	137

In [44]: movie_data.drop_duplicates(keep = 'first', inplace = True)

In [47]: #giving list of column names that needs to be checked
 check_row = ['budget', 'revenue']

#this will replace the value of '0' to NaN of columns given in the list
movie data[check row] = movie data[check row].replace(0, np.NaN)

#now we will drop any row which has NaN values in any of the column of the list (check_row)

movie_data.dropna(subset = check_row, inplace = True)

In [48]: movie_data

Out[48]:

	budget	revenue	original_title	cast	director	
0	150000000.0	1.513529e+09	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	-
1	150000000.0	3.784364e+08	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	۱ [
2	110000000.0	2.952382e+08	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	[
3	200000000.0	2.068178e+09	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	E
4	190000000.0	1.506249e+09	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle	James Wan	\
5	135000000.0	5.329505e+08	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	Alejandro González Iñárritu	(r f
6	155000000.0	4.406035e+08	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	Alan Taylor	F
7	108000000.0	5.953803e+08	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	Ridley Scott	E
8	74000000.0	1.156731e+09	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	Kyle Balda Pierre Coffin	E ł
9	175000000.0	8.537086e+08	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	Pete Docter	1
10	245000000.0	8.806746e+08	Spectre	Daniel Craig Christoph Waltz Léa Seydoux Ralp	Sam Mendes	Æ
11	176000003.0	1.839877e+08	Jupiter Ascending	Mila Kunis Channing Tatum Sean Bean Eddie Redm	Lana Wachowski Lilly Wachowski	E

	budget	revenue	original_title	cast	director
12	15000000.0	3.686941e+07	Ex Machina	Domhnall Gleeson Alicia Vikander Oscar Isaac S	Alex Garland
13	88000000.0	2.436371e+08	Pixels	Adam Sandler Michelle Monaghan Peter Dinklage	Chris Columbus
14	280000000.0	1.405036e+09	Avengers: Age of Ultron	Robert Downey Jr. Chris Hemsworth Mark Ruffalo	Joss Whedon
15	44000000.0	1.557601e+08	The Hateful Eight	Samuel L. Jackson Kurt Russell Jennifer Jason	Quentin Tarantino
16	48000000.0	3.257714e+08	Taken 3	Liam Neeson Forest Whitaker Maggie Grace Famke	Olivier Megaton
17	130000000.0	5.186022e+08	Ant-Man	Paul Rudd Michael Douglas Evangeline Lilly Cor	Peyton Reed
18	95000000.0	5.423514e+08	Cinderella	Lily James Cate Blanchett Richard Madden Helen	Kenneth Branagh
19	160000000.0	6.505234e+08	The Hunger Games: Mockingjay - Part 2	Jennifer Lawrence Josh Hutcherson Liam Hemswor	Francis Lawrence
20	190000000.0	2.090357e+08	Tomorrowland	Britt Robertson George Clooney Raffey Cassidy	Brad Bird
21	30000000.0	9.170983e+07	Southpaw	Jake Gyllenhaal Rachel McAdams Forest Whitaker	Antoine Fuqua
22	110000000.0	4.704908e+08	San Andreas	Dwayne Johnson Alexandra Daddario Carla Gugino	Brad Peyton
23	40000000.0	5.696515e+08	Fifty Shades of Grey	Dakota Johnson Jamie Dornan Jennifer Ehle Eloi	Sam Taylor- Johnson

	budget	revenue	original_title	cast	director
24	28000000.0	1.333465e+08	The Big Short	Christian Bale Steve Carell Ryan Gosling Brad	Adam McKay
25	150000000.0	6.823301e+08	Mission: Impossible - Rogue Nation	Tom Cruise Jeremy Renner Simon Pegg Rebecca Fe	Christopher McQuarrie
26	68000000.0	2.158636e+08	Ted 2	Mark Wahlberg Seth MacFarlane Amanda Seyfried	Seth MacFarlane
27	81000000.0	4.038021e+08	Kingsman: The Secret Service	Taron Egerton Colin Firth Samuel L. Jackson Mi	Matthew Vaughn
28	20000000.0	8.834647e+07	Spotlight	Mark Ruffalo Michael Keaton Rachel McAdams Lie	Tom McCarthy
29	61000000.0	3.112569e+08	Maze Runner: The Scorch Trials	Dylan O'Brien Kaya Scodelario Thomas Brodie-Sa	Wes Ball
10690	8200000.0	1.632143e+08	The Sound of Music	Julie Andrews Christopher Plummer Eleanor Park	Robert Wise
10691	14000000.0	1.117219e+08	Doctor Zhivago	Omar Sharif Julie Christie Geraldine Chaplin R	David Lean
10692	5600000.0	2.995000e+07	Those Magnificent Men in Their Flying Machines	Stuart Whitman Sarah Miles James Fox Alberto S	Ken Annakin
10716	20000000.0	1.200000e+07	The Greatest Story Ever Told	Max von Sydow Michael Anderson Jr. Carroll Bak	George Stevens
10724	7000000.0	8.197449e+07	On Her Majesty's Secret Service	George Lazenby Diana Rigg Telly Savalas Gabrie	Peter R. Hunt

	budget	revenue	original_title	cast	director
10725	6000000.0	1.023089e+08	Butch Cassidy and the Sundance Kid	Paul Newman Robert Redford Katharine Ross Stro	George Roy Hill
10727	3600000.0	4.478505e+07	Midnight Cowboy	Dustin Hoffman Jon Voight Sylvia Miles John Mc	John Schlesinger
10728	6244087.0	6.386410e+05	The Wild Bunch	Ernest Borgnine William Holden Robert Ryan Edm	Sam Peckinpah
10755	6000000.0	1.818138e+08	Grease	John Travolta Olivia Newton-John Stockard Chan	Randal Kleiser
10756	20000000.0	1.878840e+08	Jaws 2	Roy Scheider Lorraine Gary Murray Hamilton Jos	Jeannot Szwarc
10757	650000.0	5.500000e+07	Dawn of the Dead	David Emge Ken Foree Scott H. Reiniger Gaylen	George A. Romero
10758	55000000.0	3.002180e+08	Superman	Marlon Brando Gene Hackman Christopher Reeve N	Richard Donner
10759	300000.0	7.000000e+07	Halloween	Donald Pleasence Jamie Lee Curtis P.J. Soles N	John Carpenter
10760	2700000.0	1.410000e+08	Animal House	John Belushi Tim Matheson John Vernon Verna Bl	John Landis
10762	15000000.0	5.000000e+07	The Deer Hunter	Robert De Niro John Cazale John Savage Christo	Michael Cimino
10770	2300000.0	3.500000e+07	Midnight Express	Brad Davis Irene Miracle Bo Hopkins Randy Quai	Alan Parker

	budget	revenue	original_title	cast	director
10771	4000000.0	3.047142e+07	The Lord of the Rings	Christopher Guard William Squire Michael Schol	Ralph Bakshi
10775	7920000.0	1.456008e+07	Death on the Nile	Peter Ustinov Mia Farrow Simon MacCorkindale L	John Guillermin
10777	11.0	1.100000e+01	F.I.S.T.	Sylvester Stallone Rod Steiger Peter Boyle Mel	Norman Jewison
10778	5000000.0	7.230000e+06	Force 10 from Navarone	Harrison Ford Robert Shaw Barbara Bach Edward	Guy Hamilton
10779	12000000.0	2.276508e+07	Convoy	Kris Kristofferson Ali MacGraw Ernest Borgnine	Sam Peckinpah
10780	3500000.0	2.404653e+07	Invasion of the Body Snatchers	Donald Sutherland Brooke Adams Leonard Nimoy V	Philip Kaufman
10788	24000000.0	2.104905e+07	The Wiz	Diana Ross Michael Jackson Nipsey Russell Ted	Sidney Lumet
10791	6800000.0	2.651836e+07	Damien: Omen II	William Holden Lee Grant Jonathan Scott- Taylor	Don Taylor Mike Hodges
10793	1000000.0	3.713768e+06	Watership Down	John Hurt Richard Briers Michael Graham Cox Jo	Martin Rosen
10822	7500000.0	3.373669e+07	Who's Afraid of Virginia Woolf?	Elizabeth Taylor Richard Burton George Segal S	Mike Nichols
10828	3000000.0	1.300000e+07	Torn Curtain	Paul Newman Julie Andrews Lila Kedrova Hansjö	Alfred Hitchcock
10829	4653000.0	6.000000e+06	El Dorado	John Wayne Robert Mitchum James Caan Charlene	Howard Hawks

	budget	revenue	original_title	cast	director
10835	12000000.0	2.000000e+07	The Sand Pebbles	Steve McQueen Richard Attenborough Richard Cre	Robert Wise
10848	5115000.0	1.200000e+07	Fantastic Voyage	Stephen Boyd Raquel Welch Edmond O'Brien Donal	Richard Fleischer

3854 rows × 10 columns

In [49]: #replacing 0 with NaN of runtime column of the dataframe movie_data['runtime'] = movie_data['runtime'].replace(0, np.NaN)

> #calling the column which need to be formatted in datetime and storing t hose values in them

movie_data.release_date = pd.to_datetime(movie_data['release_date'])

#showing the dataset movie_data.head()

Out[49]:

	budget	revenue	original_title	cast	director	tagline	runti
0	150000000.0	1.513529e+09	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open.	124
1	150000000.0	3.784364e+08	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day.	120
2	110000000.0	2.952382e+08	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Can Destroy You	119
3	200000000.0	2.068178e+09	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	Every generation has a story.	136
4	190000000.0	1.506249e+09	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle 	James Wan	Vengeance Hits Home	137

```
In [50]: movie_data.dtypes
                                   float64
Out[50]: budget
         revenue
                                   float64
         original_title
                                    object
         cast
                                    object
         director
                                    object
         tagline
                                    object
         runtime
                                     int64
         genres
                                    object
         release_date
                            datetime64[ns]
         release year
                                     int64
         dtype: object
In [51]: #applymap function changes the columns data type to the type 'argument'
          we pass
         change coltype = ['budget', 'revenue']
         movie data[change_coltype] = movie_data[change_coltype].applymap(np.int6
          4)
         #shwoing the datatypes of all columns
         movie_data.dtypes
Out[51]: budget
                                     int64
                                     int64
         revenue
         original_title
                                    object
         cast
                                    object
         director
                                    object
         tagline
                                    object
         runtime
                                     int64
         genres
                                    object
         release date
                            datetime64[ns]
         release year
                                     int64
         dtype: object
In [52]: movie data.rename(columns = {'budget' : 'budget (in US$)', 'revenue' :
          'revenue (in US$)'}, inplace = True)
```

In [53]: movie_data

Out[53]:

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
0	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trev
1	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George N
2	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentk
3	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrar
4	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle	James Wa
5	135000000	532950503	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	Alejandro GonzÃįlez IñÃįrritu
6	155000000	440603537	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	Alan Taylo
7	108000000	595380321	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	Ridley Sc
8	74000000	1156730962	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	Kyle Balda Pie Coffin
9	175000000	853708609	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	Pete Doct
10	245000000	880674609	Spectre	Daniel Craig Christoph Waltz Léa Seydoux Ralp	Sam Men
11	176000003	183987723	Jupiter Ascending	Mila Kunis Channing Tatum Sean Bean Eddie Redm	Lana Wachows Wachows

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
12	15000000	36869414	Ex Machina	Domhnall Gleeson Alicia Vikander Oscar Isaac S	Alex Garla
13	88000000	243637091	Pixels	Adam Sandler Michelle Monaghan Peter Dinklage	Chris Columbus
14	280000000	1405035767	Avengers: Age of Ultron	Robert Downey Jr. Chris Hemsworth Mark Ruffalo	Joss Whe
15	44000000	155760117	The Hateful Eight	Samuel L. Jackson Kurt Russell Jennifer Jason	Quentin Tarantino
16	48000000	325771424	Taken 3	Liam Neeson Forest Whitaker Maggie Grace Famke	Olivier Me
17	130000000	518602163	Ant-Man	Paul Rudd Michael Douglas Evangeline Lilly Cor	Peyton Ro
18	95000000	542351353	Cinderella	Lily James Cate Blanchett Richard Madden Helen	Kenneth Branagh
19	160000000	650523427	The Hunger Games: Mockingjay - Part 2	Jennifer Lawrence Josh Hutcherson Liam Hemswor	Francis Lawrence
20	190000000	209035668	Tomorrowland	Britt Robertson George Clooney Raffey Cassidy	Brad Bird
21	30000000	91709827	Southpaw	Jake Gyllenhaal Rachel McAdams Forest Whitaker	Antoine F
22	110000000	470490832	San Andreas	Dwayne Johnson Alexandra Daddario Carla Gugino	Brad Pey
23	4000000	569651467	Fifty Shades of Grey	Dakota Johnson Jamie Dornan Jennifer Ehle Eloi	Sam Taylo Johnson

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
24	28000000	133346506	The Big Short	Christian Bale Steve Carell Ryan Gosling Brad	Adam Mc
25	150000000	682330139	Mission: Impossible - Rogue Nation	mpossible - Renner Simon	
26	68000000	215863606	Ted 2	Mark Wahlberg Seth Ted 2 MacFarlane Amanda Seyfried	
27	81000000	403802136	Kingsman: The Secret Service	Taron Egerton Colin Firth Samuel L. Jackson Mi	Matthew Vaughn
28	20000000	88346473	Spotlight	Mark Ruffalo Michael Keaton Rachel McAdams Lie	Tom McC
29	61000000	311256926	Maze Runner: The Scorch Trials	Dylan O'Brien Kaya Scodelario Thomas Brodie-Sa	Wes Ball
10690	8200000	163214286	The Sound of Music	Julie Andrews Christopher Plummer Eleanor Park	Robert W
10691	14000000	111721910	Doctor Zhivago	Omar Sharif Julie Christie Geraldine Chaplin R	David Lea
10692	5600000	29950000	Those Magnificent Men in Their Flying Machines	Stuart Whitman Sarah Miles James Fox Alberto S	Ken Anna
10716	20000000	12000000	The Greatest Story Ever Told Max von Sydow Michael Anderson Jr. Carroll Bak		George Stevens
10724	7000000	81974493	On Her Majesty's Secret Service	George Lazenby Diana Rigg Telly Savalas Gabrie	Peter R. F

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
10725	6000000	102308889	Butch Cassidy and the Sundance Kid	Paul Newman Robert Redford Katharine Ross Stro	George R
10727	3600000	44785053	Midnight Cowboy	Dustin Hoffman Jon Voight Sylvia Miles John Mc	John Schlesing
10728	6244087	638641	The Wild Bunch	Ernest Borgnine William Holden Robert Ryan Edm	Sam Pecl
10755	6000000	181813770	Grease	John Travolta Olivia Newton-John Stockard Chan	Randal Kl
10756	20000000	187884007	Jaws 2	Roy Scheider Lorraine Gary Murray Hamilton Jos	Jeannot Szwarc
10757	650000	55000000	Dawn of the Dead	David Emge Ken Foree Scott H. Reiniger Gaylen	George A Romero
10758	55000000	300218018	Marlon Brando Gene Superman Hackman Christopher Reeve N		Richard C
10759	300000	70000000	Halloween	Donald Pleasence Jamie Lee Curtis P.J. Soles N	John Carı
10760	2700000	141000000	Animal House	John Belushi Tim Matheson John Vernon Verna Bl	John Lan
10762	15000000	50000000	The Deer Hunter	Robert De Niro John Cazale John Savage Christo	Michael C
10770	2300000	35000000	Midnight Express	Brad Davis Irene Miracle Bo Hopkins Randy Quai	Alan Park

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
10771	4000000	30471420	The Lord of the Rings	Christopher Guard William Squire Michael Schol	Ralph Bal
10775	7920000	14560084	Death on the Nile	Peter Ustinov Mia Farrow Simon MacCorkindale L	John Guil
10777	11	11	F.I.S.T.	Sylvester Stallone Rod Steiger Peter Boyle Mel	Norman Jewison
10778	5000000	7230000	Force 10 from Navarone	Harrison Ford Robert Shaw Barbara Bach Edward	Guy Ham
10779	12000000	22765081	Convoy	Kris Kristofferson Ali MacGraw Ernest Borgnine	
10780	3500000	24046533	Invasion of the Body Snatchers	Donald Sutherland Brooke Adams Leonard Nimoy V	Philip Kau
10788	24000000	21049053	The Wiz	Diana Ross Michael Jackson Nipsey Russell Ted	Sidney Lu
10791	6800000	26518355	Damien: Omen II	William Holden Lee Grant Jonathan Scott- Taylor	Don Taylor Mik Hodges
10793	1000000	3713768	Watership Down	John Hurt Richard Briers Michael Graham Cox Jo	Martin Ro
10822	7500000	33736689	Who's Afraid of Virginia Woolf? Elizabeth Taylor Richard Burton George Segal S		Mike Nich
10828	3000000	13000000	Paul Newman Julie Torn Curtain Andrews Lila Kedrova Hansjö		Alfred Hitchcock
10829	4653000	6000000	El Dorado	John Wayne Robert Mitchum James Caan Charlene	Howard H

	budget_(in_US\$)	revenue_(in_US\$)	original_title	cast	di
10835	12000000	20000000	The Sand Pebbles	Steve McQueen Richard Attenborough Richard Cre	Robert W
10848	5115000	12000000	Fantastic Voyage	Stephen Boyd Raquel Welch Edmond O'Brien Donal	Richard Fleischer

3854 rows × 10 columns

Out[54]:

	budget_(in_US\$)	revenue_(in_US\$)	profit	original_title	cast	direc
0	150000000	1513528810	1363528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrc
1	150000000	378436354	228436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller
2	110000000	295238201	185238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwen
3	200000000	2068178225	1868178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams
4	190000000	1506249360	1316249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle 	James Wan

Q1: Which movie had the greatest and least budget?

```
In [59]: def highest_lowest(column_name):
    #highest
    #taking the index value of the highest number in profit column
    highest_id = movie_data[column_name].idxmax()
    #calling by index number, storing that row info to a variable
    highest_details = pd.DataFrame(movie_data.loc[highest_id])

#lowest
    #same processing as above
    lowest_id = movie_data[column_name].idxmin()
    lowest_details = pd.DataFrame(movie_data.loc[lowest_id])

#concatenating two dataframes
    two_in_one_data = pd.concat([highest_details, lowest_details], axis
= 1)

return two_in_one_data
```

In [60]: highest_lowest('budget_(in_US\$)')

Out[60]:

	2244	2618
budget_(in_US\$)	425000000	1
revenue_(in_US\$)	11087569	100
profit	-413912431	99
original_title	The Warrior's Way	Lost & Found
cast	Kate Bosworth Jang Dong-gun Geoffrey Rush Dann	David Spade Sophie Marceau Ever Carradine Step
director	Sngmoo Lee	Jeff Pollack
tagline	Assassin. Hero. Legend.	A comedy about a guy who would do anything to
runtime	100	95
genres	Adventure Fantasy Action Western Thriller	Comedy Romance
release_date	2010-12-02 00:00:00	1999-04-23 00:00:00
release_year	2010	1999

Ans: The Warrior's Way

Q2: Which movie earns the most and least profit?

In [58]: #calling the function and passing the argument
highest_lowest('profit')

Out[58]:

	1386	2244
budget_(in_US\$)	237000000	425000000
revenue_(in_US\$)	2781505847	11087569
profit	2544505847	-413912431
original_title	Avatar	The Warrior's Way
cast	Sam Worthington Zoe Saldana Sigourney Weaver S	Kate Bosworth Jang Dong-gun Geoffrey Rush Dann
director	James Cameron	Sngmoo Lee
tagline	Enter the World of Pandora.	Assassin. Hero. Legend.
runtime	162	100
genres	Action Adventure Fantasy Science Fiction	Adventure Fantasy Action Western Thriller
release_date	2009-12-10 00:00:00	2010-12-02 00:00:00
release_year	2009	2010

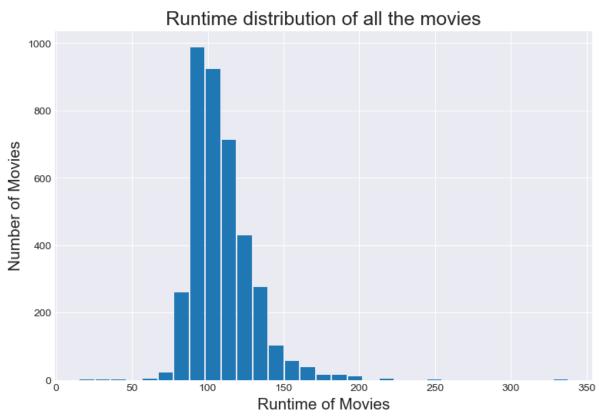
Ans: Avatar with \$2781505847

Q3: What is the average runtime of all movies?

Out[62]: 109.22029060716139

Ans: 109.22029060716139

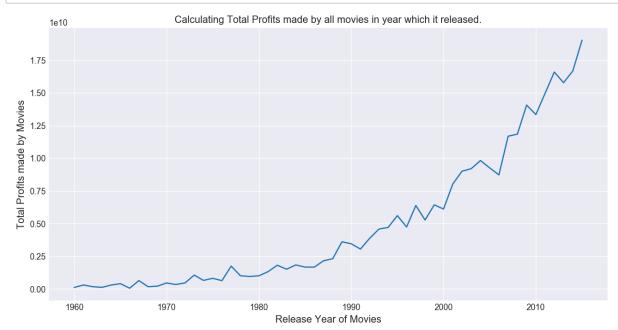
```
#plotting a histogram of runtime of movies
#gives styles to bg plot
sns.set_style('darkgrid')
#chaging the label size, this will change the size of all plots that we
 plot from now!
plt.rc('xtick', labelsize = 10)
plt.rc('ytick', labelsize = 10)
#giving the figure size(width, height)
plt.figure(figsize=(9,6), dpi = 100)
#x-axis label name
plt.xlabel('Runtime of Movies', fontsize = 15)
#y-axis label name
plt.ylabel('Number of Movies', fontsize=15)
#title of the graph
plt.title('Runtime distribution of all the movies', fontsize=18)
#giving a histogram plot
plt.hist(movie_data['runtime'], rwidth = 0.9, bins =31)
#displays the plot
plt.show()
```



Opinion: as you can see the tallest bar here is time interval between 85-100 min(approx) and around 1000 movies out of 3855 movies have the runtime between these time intervals. So we can also say from this graph that mode time of movies is around 85-110 min, has the highest concentration of data points around this time interval. The distribution of this graph is positively skewed or right skewed!

Q4: In which year we had the most movies making profits?

```
profits_each year = movie_data.groupby('release_year')['profit'].sum()
#giving the figure size(width, height)
plt.figure(figsize=(12,6), dpi = 130)
#labeling x-axis
plt.xlabel('Release Year of Movies', fontsize = 12)
#labeling y-axis
plt.ylabel('Total Profits made by Movies', fontsize = 12)
#title of a the plot
plt.title('Calculating Total Profits made by all movies in year which it
 released.')
#plotting what needs to be plotted
plt.plot(profits_each_year)
#showing the plot
plt.show()
#shows which year made the highest profit
profits each year.idxmax()
```



Out[68]: 2015

Opinion: The year 2015, shows us the highest peak, having the highest profit than in any year, of more than 18 billion dollars. Not every year had same amount of movies released, the year 2015 had the most movie releases than in any other year. The more old the movies, the more less releases at that year (atleast this is what the dataset shows us).

```
In [69]: #storing the values in the the form of DataFrame just to get a clean and
   better visual output
   profits_each_year = pd.DataFrame(profits_each_year)
   profits_each_year.tail()
```

Out[69]:

	profit
release_year	
2011	14966694704
2012	16596845507
2013	15782743325
2014	16676201357
2015	19032145273

Q5: Which directer directed most films?

```
In [70]: def extract_data(column_name):
    #will take a column, and separate the string by '|'
    all_data = profit_movie_data[column_name].str.cat(sep = '|')

#giving pandas series and storing the values separately
    all_data = pd.Series(all_data.split('|'))

#this will us value in descending order
    count = all_data.value_counts(ascending = False)

return count
```

```
In [72]: #assinging new dataframe which holds values only of movies having profit
    $100k or more
    profit_movie_data = movie_data.query('profit >= 50000000')

#reindexing new dataframe
    profit_movie_data.index = range(len(profit_movie_data))
    #will initialize dataframe from 1 instead of 0
    profit_movie_data.index = profit_movie_data.index + 1

#showing the dataset
    profit_movie_data.head(2)
```

Out[72]:

	budget_(in_US\$)	revenue_(in_US\$)	profit	original_title	cast	director
1	150000000	1513528810	1363528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow
2	150000000	378436354	228436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller

```
In [73]: #this will variable will store the return value from a function
    director_count = extract_data('director')
    #shwoing top 5 values
    director_count.head()
```

Out[73]: Steven Spielberg 23
Robert Zemeckis 13
Clint Eastwood 12
Tim Burton 11
Ron Howard 10
dtype: int64

Q6: Which genre were more successful?

```
In [74]: #this will variable will store the return value from a function
    genre_count = extract_data('genres')
    #shwoing top 5 values
    genre_count.head()
```

```
Out[74]: Comedy 492
Drama 481
Action 464
Thriller 405
Adventure 379
dtype: int64
```

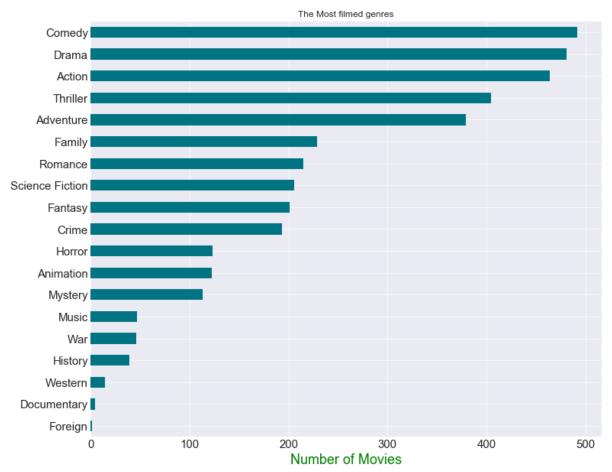
```
In [75]: genre_count.sort_values(ascending = True, inplace = True)
    #initializing plot
    ax = genre_count.plot.barh(color = '#007482', fontsize = 15)

#giving a title
    ax.set(title = 'The Most filmed genres')

#x-label
    ax.set_xlabel('Number of Movies', color = 'g', fontsize = '18')

#giving the figure size(width, height)
    ax.figure.set_size_inches(12, 10)

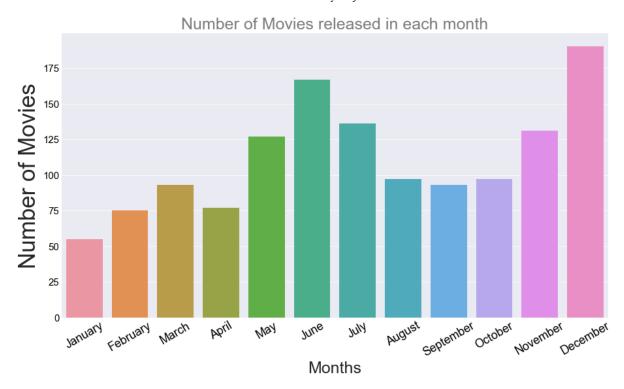
#shwoing the plot
    plt.show()
```



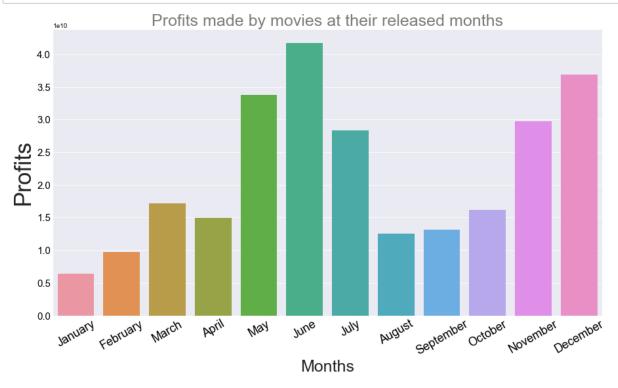
Opinion: Another amazing results. Action, Drama and Comedy genres are the most as visualized but Comedy takes the prize, about 492 movies have genres comedy which make \$50M+ in profit. In comparison, even Adventure and Thriller really play the role.

Q7: Which month released highest number of movies in all of the years? And which month made the most profit?

```
In [84]: ###### Which month released highest number of movies in all of the year
         #giving a new dataframe which gives 'release-date' as index
         index release date = profit movie data.set index('release date')
         #now we need to group all the data by month, since release date is in fo
         rm of index, we extract month from it
         groupby index = index_release_date.groupby([(index_release_date.index.mo
         nth)])
         #this will give us how many movies are released in each month
         monthly_movie_count = groupby_index.profit.count()
         #converting table to a dataframe
         monthly movie count= pd.DataFrame(monthly movie count)
         #qiving a list of months
         month_list = ['January', 'February', 'March', 'April', 'May', 'June', 'J
         uly', 'August', 'September', 'October', 'November', 'December']
         monthly movie count bar = sns.barplot(x = monthly movie count.index, y =
          monthly movie count.profit, data = monthly movie count)
         #setting size of the graph
         monthly movie count bar.figure.set_size_inches(15,8)
         #setting the title and customizing
         monthly movie count bar.axes.set title('Number of Movies released in eac
         h month', fontsize = 25, alpha = 0.6)
         #setting x-label
         monthly movie count bar.set xlabel("Months", fontsize = 25)
         #setting y-label
         monthly movie count bar.set ylabel("Number of Movies", fontsize = 35)
         #customizing axes values
         monthly movie count bar.tick params(labelsize = 15, labelcolor="black")
         #rotating the x-axis values to make it readable
         monthly movie count bar.set xticklabels(month list, rotation = 30, size
         = 18)
         #shows the plot
         plt.show()
```



```
######## which month made the most profit
#now since the data is grouped by month, we add 'profit' values to respe
ctive months, saving all this to a new var
monthly_profit = groupby_index.profit.sum()
#converting table to a dataframe
monthly profit = pd.DataFrame(monthly profit)
#giving seaborn bar plot to visualize the data
#giving values to our graph
monthly profit bar = sns.barplot(x = monthly profit.index, y = monthly p
rofit.profit, data = monthly profit)
#setting size of the graph
monthly profit_bar.figure.set_size_inches(15,8)
#setting the title and customizing
monthly_profit_bar.axes.set_title('Profits made by movies at their relea
sed months',fontsize = 25, alpha = 0.6)
#setting x-label
monthly_profit_bar.set_xlabel("Months", fontsize = 25)
#setting y-label
monthly profit bar.set ylabel("Profits", fontsize = 35)
#customizing axes values
monthly profit bar.tick params(labelsize = 15, labelcolor="black")
#rotating the x-axis values to make it readable
monthly profit bar.set xticklabels(month list, rotation = 30, size = 18)
#shows the plot
plt.show()
```



Opinion: Seeing the both visualizations of both graphs we see similar trend. Where there are more movie released there is more profit and vice versa but just not for one month i.e December. December is the month where most movie release but when compared to profits it ranks second. This means that december month has high release rate but less profit margin. The month of June where we have around 165 movie releases, which is second highest, is the highest in terms of making profits.

Reference: https://matplotlib.org/users/pyplot-tutorial.html)
https://matplotlib/bar-charts/)
https://plot.ly/matplotlib/bar-charts/)
https://cs231n.github.io/python-numpy-tutorial/)
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https://stackoverflow.com/questions/11346283/renaming-columns-in-pandas)