Project: TMDb Movies project UDACITY

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Introduction

This data set contains information about 10,000 movies collected from The Movie Database (TMDb), including user ratings and revenue. Certain columns, like 'cast' and 'genres', contain multiple values separated by pipe (|) characters. There are some odd characters in the 'cast' column. Don't worry about cleaning them. You can leave them as is. The final two columns ending with "_adj" show the budget and revenue of the associated movie in terms of 2010 dollars, accounting for inflation over time.

The final two columns ending with "_adj" show the budget and revenue of the associated movie in terms of 2010 dollars, accounting for inflation over time.

SOURCE TO DATASET

https://d17h27t6h515a5.cloudfront.net/topher/2017/October/59dd1c4c_tmdb-movies/tmdb-movies.csv (https://d17h27t6h515a5.cloudfront.net/topher/2017/October/59dd1c4c_tmdb-movies/tmdb-movies.csv).

Questions:

- 1. Top 5 most expensive movies
- 2. What is the average runtime, budget, revenue of the movies
- 3. What is the movie with the highest andlowest profit
- 4. Which year had most profitable movies
- 5. The most popular movie before the movie with the highest profit was released
- 6. Is there a correlation between how popular a movie is and the profit?

Data Wrangling

The data wrangling process will be done using pandas. Some methods will be used to explore the dataset and draw intuition about the dataset to understand the shape, datatype and other parameters** First o all I will load my data and use several codes to draw insight as to the shape, datatypes, NAN values etc present in the data

```
In [2]:
            df = pd.read_csv('tmdb-movies.csv')
            # looking at null values and datatypes so I can know what cleanin techniques
            df.info()
            <class 'pandas.core.frame.DataFrame'>
            RangeIndex: 10866 entries, 0 to 10865
            Data columns (total 21 columns):
                 Column
                                       Non-Null Count Dtype
                 _ _ _ _ _
             0
                 id
                                       10866 non-null int64
             1
                 imdb id
                                       10856 non-null object
             2
                 popularity
                                       10866 non-null float64
             3
                 budget
                                       10866 non-null int64
             4
                 revenue
                                       10866 non-null int64
             5
                                       10866 non-null object
                 original_title
             6
                                       10790 non-null object
                 cast
             7
                 homepage
                                                       object
                                       2936 non-null
             8
                 director
                                       10822 non-null object
             9
                 tagline
                                       8042 non-null
                                                       object
                 keywords
                                       9373 non-null
                                                       object
                 overview
                                       10862 non-null object
             11
             12
                 runtime
                                       10866 non-null int64
             13
                 genres
                                       10843 non-null object
                 production_companies 9836 non-null
                                                       object
             15
                 release_date
                                       10866 non-null object
                 vote count
                                       10866 non-null int64
             17
                 vote average
                                       10866 non-null float64
             18
                release year
                                       10866 non-null int64
             19
                 budget adj
                                       10866 non-null float64
                 revenue adj
                                       10866 non-null float64
             20
            dtypes: float64(4), int64(6), object(11)
            memory usage: 1.7+ MB
            #Exploring the shape of the dataset
            df.shape
```

```
In [3]:
```

Out[3]: (10866, 21)

observations

- The data set has "10866" rows and "21" columns
- · There are some null values in the data set
- data type looks consistent but release date is stored as a string
- budget and revenue are floats
- there are some 0 values in our data sets

```
# Confirming that release date is stored as a string
In [4]:
            type(df['release_date'][0])
```

Out[4]: str

In [5]:

exploring the frist 5 rows to have a sense of my data
df.head()

Out[5]:

	id	imdb_id	popularity	budget	revenue	original_title	cast	
0	135397	tt0369610	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	
1	76341	tt1392190	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	
2	262500	tt2908446	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	http://v
3	140607	tt2488496	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	
4	168259	tt2820852	9.335014	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle 	

5 rows × 21 columns

Out[6]: 1

In [7]: M df[df.duplicated()]

Out[7]:

							Jon FoolKelly	
2090	42194	tt0411951	0.59643	30000000	967000	TEKKEN	Overton Cary- Hiroyuki Tagawallan	NaN

budget revenue original_title

1 rows × 21 columns

id

imdb_id popularity

cast homepage

Data Cleaning -

Thes list below showsthe different cleaning techniques that will be employed to make the dataset appropriate for Exploratory data Analysis

- Changing release date column from string to datetime data type.
- · Removing the duplicated values.
- Changing format of budget and revenue columns.
- Remove unnecessary columns such as 'imdb_id', 'budget_adj', 'revenue_adj', 'homepage', production_company, 'keywords', 'overview', 'production_companies', 'vote_count' and 'vote_average'.
- replace all the values from '0' to NAN in 'budget' and 'revenue' columns, then removing them.

Changing release date column from string to date type data.

To change the release_date to datetime I will have to use the datetime function of pandas

Now I need to confirm that my release date is now a date datatype

```
In [9]:

    df.info()

            <class 'pandas.core.frame.DataFrame'>
            RangeIndex: 10866 entries, 0 to 10865
            Data columns (total 21 columns):
             #
                 Column
                                      Non-Null Count Dtype
            - - -
                 _____
                                       -----
                                                      ----
             0
                 id
                                       10866 non-null
                                                      int64
                                       10856 non-null object
             1
                 imdb id
             2
                 popularity
                                       10866 non-null float64
             3
                 budget
                                       10866 non-null int64
             4
                 revenue
                                      10866 non-null int64
             5
                 original_title
                                      10866 non-null object
             6
                 cast
                                      10790 non-null object
             7
                 homepage
                                       2936 non-null
                                                      object
             8
                                      10822 non-null object
                 director
             9
                 tagline
                                      8042 non-null
                                                      object
             10
                keywords
                                      9373 non-null
                                                      object
             11
                overview
                                      10862 non-null object
                runtime
             12
                                       10866 non-null int64
             13
                genres
                                      10843 non-null object
             14
                production companies 9836 non-null
                                                      object
             15
                release date
                                      10866 non-null datetime64[ns]
             16
                vote count
                                      10866 non-null int64
             17
                vote average
                                      10866 non-null float64
                                      10866 non-null int64
             18
                release_year
             19
                budget adj
                                      10866 non-null float64
                                      10866 non-null float64
                 revenue adj
            dtypes: datetime64[ns](1), float64(4), int64(6), object(10)
            memory usage: 1.7+ MB
```

Removing duplicate data

Before proceeding withthe analysis some duplicate values need to be dropped so they don't affect the outcome of the analysis

```
In [10]: # removing duplicate data
    df.drop_duplicates(keep ='first', inplace=True)
    # checking for duplicated value
    df.duplicated().sum()
Out[10]: 0
```

```
    df.nunique()

In [11]:
    Out[11]: id
                                        10865
              imdb_id
                                        10855
              popularity
                                        10814
              budget
                                          557
              revenue
                                        4702
              original_title
                                        10571
              cast
                                        10719
              homepage
                                         2896
                                        5067
              director
              tagline
                                        7997
              keywords
                                        8804
              overview
                                        10847
              runtime
                                          247
              genres
                                         2039
              production_companies
                                         7445
              release_date
                                        5909
              vote_count
                                        1289
              vote_average
                                           72
              release_year
                                           56
              budget_adj
                                         2614
              revenue_adj
                                         4840
              dtype: int64
```

Changing format of budget and revenue columns.

```
▶ | change type=['budget', 'revenue']
In [12]:
             #changing data type
             df[change type]=df[change type].applymap(np.int64)
             #printing data types of the dataset to see the changed information
             df.dtypes
    Out[12]: id
                                                int64
              imdb id
                                               object
              popularity
                                              float64
             budget
                                                int64
             revenue
                                                int64
                                               object
             original_title
              cast
                                               object
                                               object
             homepage
              director
                                               object
                                               object
             tagline
              keywords
                                               object
              overview
                                               object
             runtime
                                                int64
              genres
                                               object
                                               object
              production_companies
                                      datetime64[ns]
             release date
             vote count
                                                int64
             vote_average
                                              float64
              release year
                                                int64
              budget adj
                                              float64
              revenue adj
                                              float64
              dtype: object
```

Removing unnecessary columns

Since I won't be using all columns in my analysis, I will be dropping the columns that I won't be needing at all throught the course of the EDA because they are not relevant to answering the posed questions

C:\Users\Simeon\AppData\Local\Temp\ipykernel_2248\2768113811.py:4: FutureWa rning: In a future version of pandas all arguments of DataFrame.drop except for the argument 'labels' will be keyword-only.

df = df.drop(col_to_del, 1)

Out[13]:

	id	popularity	budget	revenue	original_title	cast	director	taglin€
0	135397	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	The park is open
1	76341	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	What a Lovely Day
2	262500	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	One Choice Car Destroy You
4								•

Replacing the '0' in our column to NAN so we can drop it

After exploring the first and last few rows in budget and revenue columns, there are several entries have '0' value, so firstly I will convert those values to NAN then I will remove them from the dataset.

After removing them we ned to check to see the that the amount of columns have reduced to be sure the affected columns have beendropped

```
Int64Index: 3854 entries, 0 to 10848
Data columns (total 12 columns):
     Column
                      Non-Null Count Dtype
     ----
                      -----
                                       ----
0
     id
                      3854 non-null
                                       int64
                   3854 non-null
3854 non-null
     popularity
 1
                                       float64
                      3854 non-null float64
 2
     budget
     revenue 3854 non-null float64
 3
 4
     original_title 3854 non-null object
             3850 non-null
r 3853 non-null
3574 non-null
3854 non-null
3854 non-null
 5
     cast
                                       object
 6
     director
                                       object
    tagline
 7
                                      object
 8
     runtime
                                       int64
 9
     genres
                                       object
10 release_date 3854 non-null 11 release_year 3854 non-null
                                       datetime64[ns]
                                       int64
dtypes: datetime64[ns](1), float64(3), int64(3), object(5)
memory usage: 391.4+ KB
```

Exploratory Data Analysis

Research Question 1 (Top 5 most expensive movies)

To get the top 5 most expensive movies we need to sort the dataset based on budget in descending order. This will give us the list from most expensive to the least expensive

In [16]: •

expensive_movies = df.sort_values(by='budget',ascending=False).head()
expensive_movies

Out[16]:

	id	popularity	budget	revenue	original_title	cast	director
2244	46528	0.250540	425000000.0	1.108757e+07	The Warrior's Way	Kate Bosworth Jang Dong- gun Geoffrey Rush Dann	Sngmoo Lee
3375	1865	4.955130	380000000.0	1.021683e+09	Pirates of the Caribbean: On Stranger Tides	Johnny Depp Penélope Cruz Geoffrey Rush Ian M	Rob Marshall
7387	285	4.965391	300000000.0	9.610000e+08	Pirates of the Caribbean: At World's End	Johnny Depp Orlando Bloom Keira Knightley Geof	Gore Verbinski
14	99861	5.944927	280000000.0	1.405036e+09	Avengers: Age of Ultron	Robert Downey Jr. Chris Hemsworth Mark Ruffalo	Joss Whedon
6570	1452	1.957331	270000000.0	3.910812e+08	Superman Returns	Brandon Routh Kevin Spacey Kate Bosworth James	Bryan Singer

Out[17]:

	id	popularity	budget	revenue	original_title	cast	director	tagline
3581	59296	0.520430	1.0	1378.0	Love, Wedding, Marriage	Mandy Moore Kellan Lutz Jessica Szohr Autumn F	Dermot Mulroney	Here comes the ride.
2618	39964	0.090186	1.0	100.0	Lost & Found	David Spade Sophie Marceau Ever Carradine Step	Jeff Pollack	A comedy about a guy who would do anything to
8944	14373	0.464188	2.0	16.0	Death Wish 2	Charles Bronson Jill Ireland Vincent Gardenia	Michael Winner	First His Wife. Now His Daughter. It's Time To
10050	20701	0.317091	3.0	16.0	Tales from the Darkside: The Movie	Rae Dawn Chong Christian Slater Deborah Harry	John Harrison	From the depths of four twisted minds.
2398	39356	0.028456	3.0	43.0	Воу	James Rolleston Craig Hall Taika Waitit Te Ah	Taika Waititi	Summer, Girls, Gangs, Drugs its not easy b

Movie ID 2244 The Warrior's Way shows the largest budget i.e 425000000

Whereas Love, Wedding, Marriage with id no. 3581 shows the smallest budget i.e 1 dollar

Research Question 2 (What is the average runtime, budget, revenue of the movies)

To avoid repetition I am going to create a simple function called avg i.e average to calculate the average runtime, budget and revenue. This will give us an idea of where majority of the movies fall in terms of budget, revenue and runtime.

No unit was given for runtime, budget and revenue so we would assume runtime is in minutes, revenue and budget are in dollars

On the average the runtime, budget and revenue are 109.2 mins, 37203697 dollars, 107686616 dollars

Research Question 3 (What is the movie with the highest and lowest profit)

We don't have a column represents the total profits of the movies in our dataset to find the most and least profit movies so I will answer this question in two steps:

A. Calculating the profit of each movie

I'm going to add a new column to the dataset represents the total profit every movie in our dataset made by subtracting the revenue from the budget.

```
In [22]: # Finding profit
df['profit'] = df['revenue']-df['budget']
```

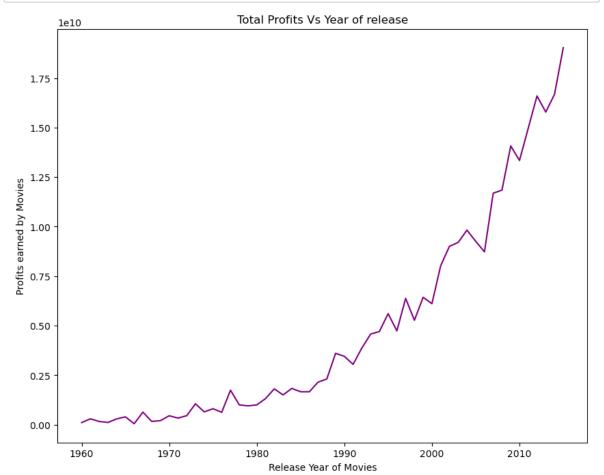
B. From the profit column calculate movies with most and least profit¶

```
In [23]:
              # Which movie has highest profit?
             highest_profit_movie = df.sort_values(by='profit',ascending=False)
             highest profit movie.iloc[0]
    Out[23]: id
                                                                               19995
              popularity
                                                                            9.432768
              budget
                                                                         237000000.0
              revenue
                                                                        2781505847.0
              original_title
                                                                              Avatar
                                 Sam Worthington | Zoe Saldana | Sigourney Weaver | S...
              cast
                                                                       James Cameron
              director
                                                        Enter the World of Pandora.
              tagline
              runtime
                                          Action | Adventure | Fantasy | Science Fiction
              genres
              release date
                                                                2009-12-10 00:00:00
              release_year
                                                                                2009
                                                                        2544505847.0
              profit
              Name: 1386, dtype: object
In [24]:
             lowest_profit_movie = df.sort_values(by='profit')
              lowest profit movie.iloc[0]
    Out[24]: id
                                                                               46528
              popularity
                                                                             0.25054
              budget
                                                                         425000000.0
                                                                          11087569.0
              revenue
              original title
                                                                  The Warrior's Way
                                 Kate Bosworth|Jang Dong-gun|Geoffrey Rush|Dann...
              cast
                                                                          Sngmoo Lee
              director
              tagline
                                                            Assassin. Hero. Legend.
              runtime
                                         Adventure | Fantasy | Action | Western | Thriller
              genres
              release date
                                                                2010-12-02 00:00:00
                                                                                2010
              release_year
              profit
                                                                        -413912431.0
              Name: 2244, dtype: object
```

Movie Avatar with id 19995 shows the highest earned profit i.e 237000000.

WhereasThe Warrior's Way movie with id 46528 shows the lowest earned profit i.e -413912431

Research Question 4 (Which year had most profitable movies)



```
In [26]: ► #To find which year made the highest profit?
profits_year.idxmax()
```

Out[26]: 2015

Graphically we cannot easily spot that 2015 is the year with the highest profit but using the idxmax() helps confirm that . However an important observation from the graph is that we can see an upwards trend of profit Year on Year

Research Question 5 (The most popular movie before the movie with the highest profit was released)

Avatar was the movies with the highst profit in 2009 but before then what movie was the most popular?

```
# in this case the highest profit earning movie was Avatar released in 2009
In [27]:
             popular_movie_2010_idx = df[df["release_year"]<=2009]["popularity"].idxmax()</pre>
             df.loc[popular movie 2010 idx]
    Out[27]: id
                                                                                  11
                                                                           12.037933
              popularity
              budget
                                                                          11000000.0
                                                                         775398007.0
              revenue
              original_title
                                                                           Star Wars
                                Mark Hamill Harrison Ford | Carrie Fisher | Peter ...
              cast
              director
                                                                        George Lucas
                                      A long time ago in a galaxy far, far away...
             tagline
              runtime
                                                  Adventure | Action | Science Fiction
              genres
                                                                1977-03-20 00:00:00
              release date
              release year
                                                                                1977
              profit
                                                                         764398007.0
             Name: 1329, dtype: object
```

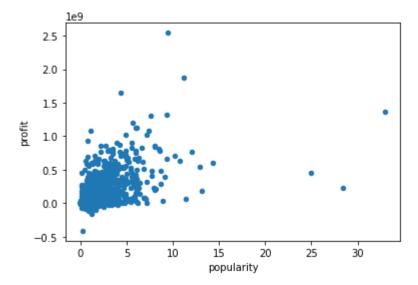
The most popular movie was Star Wars released in 1977 before Avatar

Research Question 5 (Is there a correlation between popularity and profit?)

We have looked at Avatar and Star Wars in previos sections but does the popularity of a movie affect how much profit is earned from it?

Firstly, I will create a scatter plot between popularity and profit to visually see the corelation and then use the corr() to get the exact value

```
In [28]: # creating a scatterplot between popularity and profit
df.plot(x = 'popularity', y = 'profit', kind = 'scatter');
```



```
In [29]: # getiing the quantitative value for the correlation between profit and popul
corr = df['popularity'].corr(df['profit'])
corr
```

Out[29]: 0.5962013674920571

Conclusions

To summarize the whole analysis:

SUMMARY

The top 5 most expensive movies are:

- · The Warrior's Way
- · Pirates of the Caribbean: On Stranger Tides
- · Pirates of the Caribbean: At World's End
- · Avengers: Age of Ultron
- · Superman Returns

The average:

- runtime is 109.220291
- budget is 37203697.0
- revenue is 107686616.0
- Movie with the highest profit is 'Avatar', While Movie with the lowest profit is 'The Warrior's Way'

 The year with the most profit is 2015 While there is a 0.59 correllation between popularity and profit

CONCLUSION

- We can observe an ongoing upwards trend in Total profit of movies every year.
- There is a moderate relationship between popularity of a movie and profit made
- The most expensive movies made belong to the genres of Action|Adventure|Fantasy/Sci-Fi

LIMITATIONS

Since the budget and revenue column do not have currency unit, it might be possible different movies have budget in different currency according to the country they were produced in. So an inconsistency appears here which can state the complete analysis wrong. Dropping the rows with missing values might have also affected the overall analysis.