

Insights & Data Analysis Report of Movies Metadata.csv Dataset

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The Dataset Contains a total of 5043 rows and 28 columns.

A significant amount of Data is either NULL or NA. To counter this problem, I took help of Python's inbuilt Library "PANDAS".

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Jupyter QtConsole 4.3.1
Python 3.6.3 [Anaconda custom (64-bit)] (default, Oct 15 2017, 03:27:45) [MSC v.1900 64 bit (AMD64)]
Type 'copyright', 'credits' or 'license' for more information
IPython 6.1.0 -- An enhanced Interactive Python. Type '?' for help.

In [1]: import pandas as pd

In [2]: pwd
Out[2]: 'C:\\Users\\HARSHIT'

In [3]: cd Downloads
C:\\Users\\HARSHIT\\Downloads

In [4]: movies = pd.read_csv('movie_metadata.csv')

In [5]: movies.shape
Out[5]: (5043, 28)

In [6]: movies.isnull().sum()
Out[6]:
color                19
director_name        104
num_critic_for_reviews  50
duration             15
director_facebook_likes 104
actor_3_facebook_likes  23
actor_2_name         13
actor_1_facebook_likes  7
gross               884
genres                0
actor_1_name          7
movie_title           0
num_voted_users       0
cast_total_facebook_likes 0
actor_3_name          23
facenumber_in_poster  13
plot_keywords         153
movie_imdb_link        0
num_user_for_reviews  21
language              12
country                5
content_rating        303
budget               492
title_year            108
actor_2_facebook_likes  13
imdb_score            0
aspect_ratio          329
movie_facebook_likes   0
dtype: int64

In [7]: moviesnew = movies.fillna(" ")
```

```

In [8]: moviesnew.isnull().sum()
Out[8]:
color                                0
director_name                        0
num_critic_for_reviews               0
duration                             0
director_facebook_likes              0
actor_3_facebook_likes               0
actor_2_name                         0
actor_1_facebook_likes               0
gross                                0
genres                               0
actor_1_name                         0
movie_title                          0
num_voted_users                     0
cast_total_facebook_likes            0
actor_3_name                         0
facenumber_in_poster                0
plot_keywords                        0
movie_imdb_link                      0
num_user_for_reviews                 0
language                             0
country                              0
content_rating                       0
budget                               0
title_year                           0
actor_2_facebook_likes               0
imdb_score                           0
aspect_ratio                         0
movie_facebook_likes                 0
dtype: int64

In [9]: moviesnew.to_csv('moviesnew.csv' , index=False)

```

Now, the Dataset is free from any NULL or NA values & can be used for Data Visualization and Analysis.

To visualize the Dataset, I used a very useful software called Tableau Public.

Here are some of the insights I gained on visualizing some of the relations in the Dataset.

1. Director and Gross Amount Relation

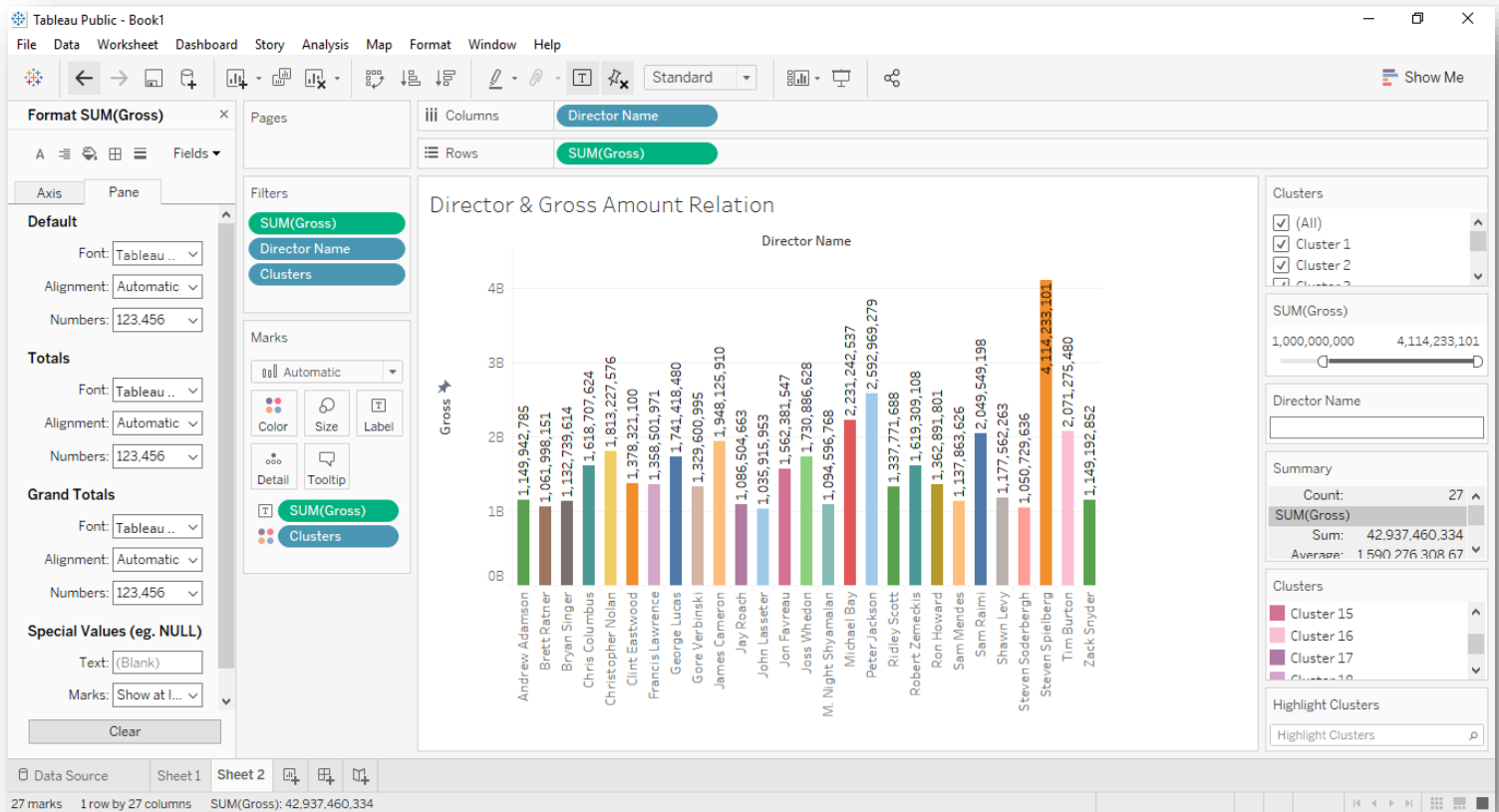
Relation : Director Name and Total Gross Amount.

Visualization By : Column Chart.

Max Gross Amount by : Director Steven Spielberg (\$ 4,114,233,101)

Lowest Gross Amount by : Director Ekachai Uekrongtham (\$ 162)

Insights : Only 27 Directors out of a total of 1879 directors have a total gross amount above \$ 1 billion with highest Gross Amount (\$ 4,114,233,101) being of Director Steven Spielberg and Lowest gross Amount (\$ 1,035,915,953) being of Director John Lasseter.



2. Year and Gross Amount Relation

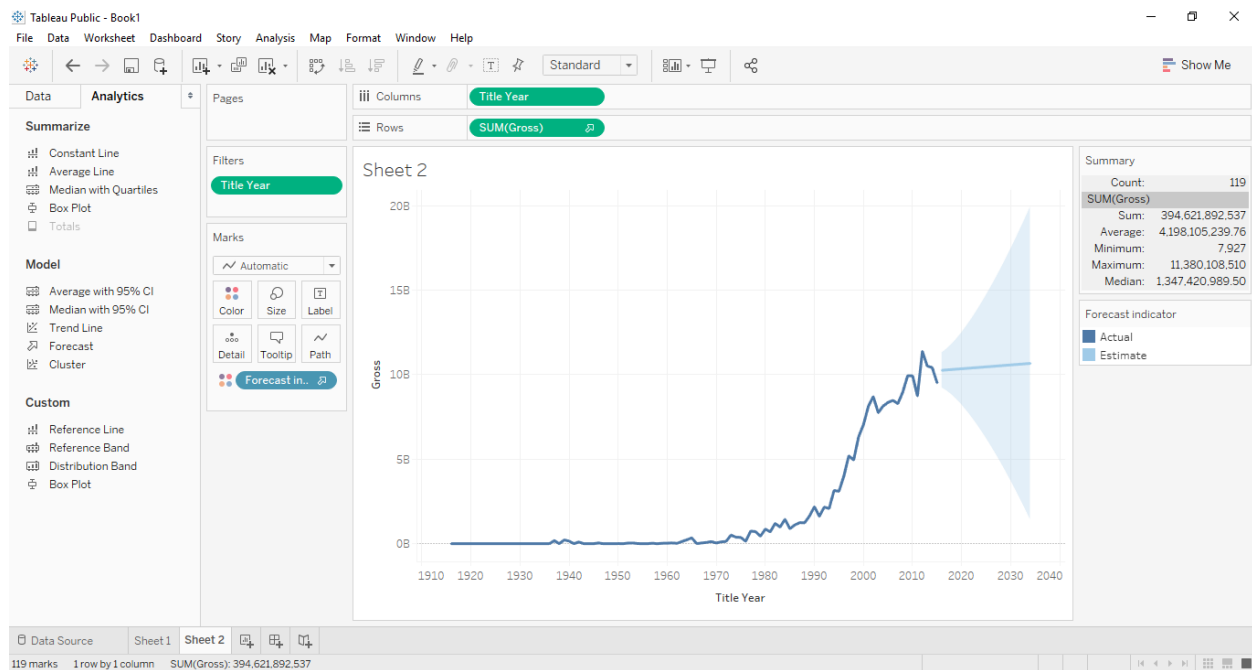
Relation : Year of Release and Total Gross Amount.

Visualization By : Line Chart.

Max Gross Amount in the year : 2012 (\$ 11,380,108,510)

Lowest Gross Amount in the year : 1947 (\$ 7927)

Insights : The Relation Depicts a rise in Total Gross Amount of Movies Released after 1990's with the year 2012 being the Highest Grossing Total for the movies at \$ 11,380,108,510 total gross amount. There is a slight declination after the year 2012 but it does not affect the trend forecast much. The Trend Forecast predicts the gross amount to lie between \$10 bn to \$11 bn between years 2016 and 2040 with maximum High and Lowest Downfall of \$20 bn and \$2 bn respectively.



3. Content Rating and IMDb Score Relation

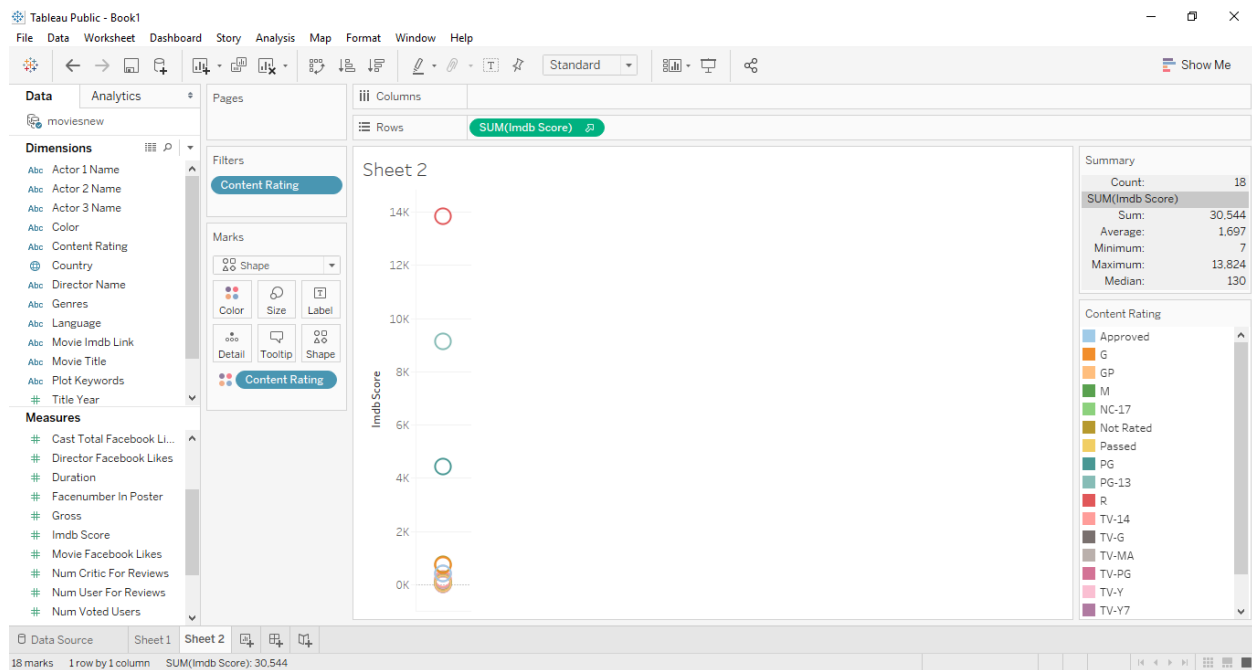
Relation : Content Rating and IMDb Score

Visualization By : Circle Views.

Max IMDb Score is of the Rating : R (13824)

Lowest IMDb Score is of the Rating : TV-Y7 (7)

Insights : The Relation Depicts that most of the content ratings have a low IMDb score between 7 to 731. The 3 Topmost IMDb score is of the rating R with a score of 13824, followed by PG-13 with a score of 9142 and lastly the rating PG with a score of 4412.



4. Facenumber in Poster and Gross Amount Relation

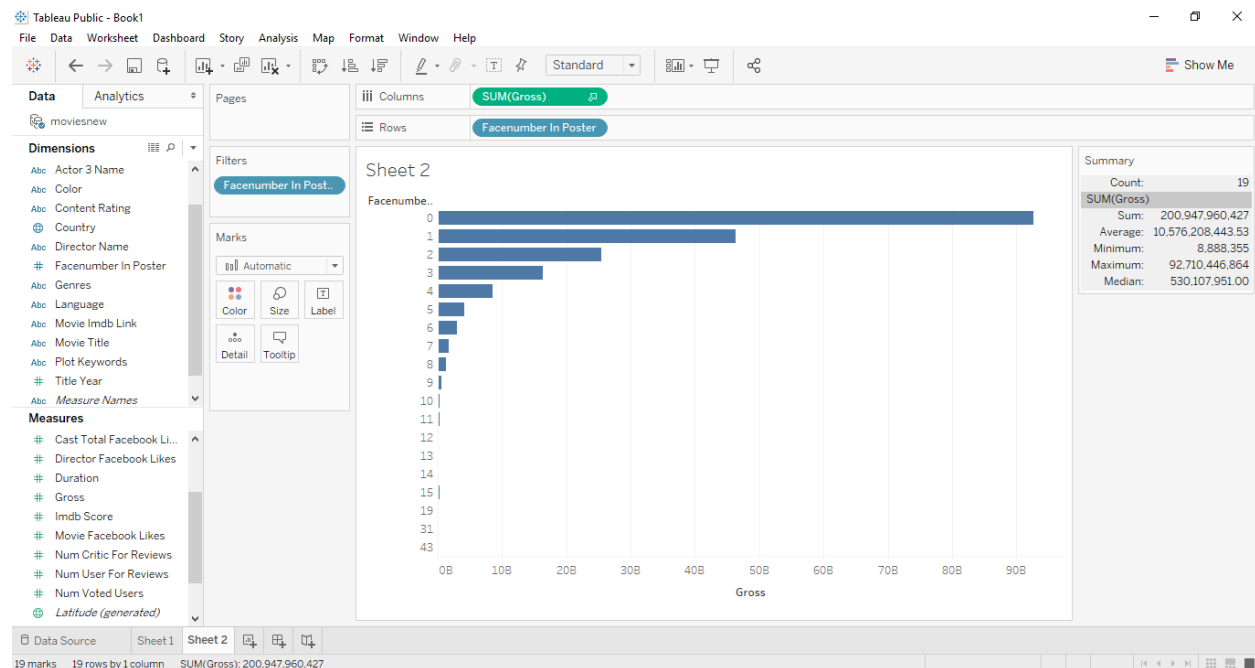
Relation : Facenumber in Poster and Total Gross Amount

Visualization By : Bar Graph.

Max Gross Amount is of Posters having Facenumber : 0 (\$ 92,710,446,864)

Lowest Gross Amount is of Posters having Facenumber: 19 (\$ 8,888,355)

Insights : The Relation Depicts that Movies whose posters have no faces are having the highest total gross amount. The general trend is the total gross amount keeps on decreasing as the number of faces in the poster increases. Valuable insight which can be gained from this relation is that the movie with less number of faces on its poster is supposed to have a high total gross amount.



5. Movie Title and Number of records Relation

Relation : Movie Title and Number of Records

Visualization By : Column Chart.

Max Records won by Movies : 3

Min Record won by Movies : 1

Insights : The Relation Depicts that Movies whose posters have no faces are having the highest total gross amount. The general trend is the total gross amount keeps on decreasing as the

number of faces in the poster increases. Valuable insight which can be gained from this relation is that the movie with less number of faces on its poster is supposed to have a high total gross amount.

