

#### Introduction to Data Management PROJECT REPORT

(Project Semester August-December 2020)

#### PROJECT REPORT ON

### Analysing IMDB data from 2006 to 2016

Submitted by

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Programme: Bachelor of Technology

Section: KM100

Course Code: INT217

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**DECLARATION** 

I, Rounak Agarwal, student of Computer Science & Engineering under CSE/IT

Discipline at, Lovely Professional University, Punjab, hereby declare that all the

information furnished in this project report is based on my own intensive work and

is genuine.

Date: 05-11-2022

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## **ACKNOWLEDGEMENT**

Primarily I'd thank god for being able to complete my project with success. Then I'd like to thank my mentor Ms. Veerpal Kaur, whose valuable guidance has been the ones that helped me patch this project and make it full proof success in contribution towards the completion of this project.

Last but not least I'd rather thanks to Lovely Professional University, and my parent's inspiration, who gave me this golden opportunity to learn many new things, to learn another aspects of life.

Rounak Agarwal

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# **INTRODUCTION**

Data management is important because the data your organization creates is a very valuable resource.
very variable resource.
The last thing you want to do is spend time and resources collecting data and
business intelligence, only to lose or misplace that information.
In that case, you would then have to spend time and resources again to get that
same business intelligence you already had.
And on that data analysis is carried out which show visualization of our
problems in efficient way.
Data Analysis is a process of inspecting, cleansing, transforming, and
modeling data with the goal of discovering useful information, informing
conclusions, and supporting decision- making.
This project is based on such data analysis on IMDb data from 2006 to 2016
IMDb is an online database of information related to films, television
programs, home videos, video games, and streaming content online -
including cast, production crew and personal biographies, plot summaries,
trivia, ratings, and fan and critical reviews.
This IMDb dataset contains 12 data fields.

# OBJECTIVES/SCOPE OF ANALYSIS

After analysis of the dataset, the aim of this project is to give answer of given

#### objectives in easy way:

- Total revenues earned by movies in each year
- Total Movies released in each year
- Top 5 IMDb rating movies
- Top 7 IMDb metascore movies
- Top 5 revenue movies from whole dataset or each year
- Most voted director from whole dataset or in each year
- Top 5 longest movies from whole dataset or in each year

## **SOURCE OF DATASET:**

Source of dataset: https://www.kaggle.com/PromptCloudHQ/imdb-data

The dataset is based on 1000 popular movies on IMDb from 2006 to 2016.
The columns included in the dataset are given below:
□ Rank
☐ Movie rank order
□ Title
☐ The title of the film
☐ Category of the film
☐ Description
☐ Summary of the film
☐ Director
☐ The name of the film's director
□ Actors
☐ Name of the Cast
□ Year
☐ Released year of the film
□ Runtime
☐ Duration of film in minutes
□ Rating
$\square$ IMDB rating given by the user $\square$
Votes
☐ Number of votes
□ Revenue

TD . 1		•	.11.
Total	revenues	1n	millione
1 Otal	1C V CHUCS	111	IIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIIII

#### ☐ Metascores

☐ An aggregated average of critic scores. Values are between 0 and 100.

Higher scores represent positive reviews.

## Sample of dataset with data fields is given below:

1	Α	В	C	D	E	F	G	Н	1	J	K	L
1 F	lank	Title	Genre	Description	Director	Actors	Year	Runtime	( Rating	Votes	Revenue	Metascore
2		1 Guardians of the Galaxy	Action,Adventure,Sci-Fi	A group of intergalactic criminal	James Gunn	Chris Pratt, Vin Di	2014	121	8.1	757074	333.13	76
3		2 Prometheus	Adventure, Mystery, Sci-Fi	Following clues to the origin of	Ridley Scott	Noomi Rapace, Lo	2012	124	7	485820	126.46	65
4		3 Split	Horror,Thriller	Three girls are kidnapped by a n	M. Night Shyamalan	James McAvoy, A	2016	117	7.3	157606	138.12	62
5		4 Sing	Animation,Comedy,Family	In a city of humanoid animals, a	Christophe Lourdelet	Matthew McCona	2016	108	7.2	60545	270.32	59
6		5 Suicide Squad	Action, Adventure, Fantasy	A secret government agency rec	David Ayer	Will Smith, Jared	2016	123	6.2	393727	325.02	40
7		6 The Great Wall	Action, Adventure, Fantasy	European mercenaries searchin	Yimou Zhang	Matt Damon, Tian	2016	103	6.1	56036	45.13	42
8		7 La La Land	Comedy, Drama, Music	A jazz pianist falls for an aspiring	Damien Chazelle	Ryan Gosling, Emi	2016	128	8.3	258682	151.06	93
9		8 Mindhorn	Comedy	A has-been actor best known fo	Sean Foley	Essie Davis, Andre	2016	89	6.4	2490		71
10		9 The Lost City of Z	Action, Adventure, Biography	A true-life drama, centering on	I James Gray	Charlie Hunnam,	2016	141	7.1	7188	8.01	78
11		10 Passengers	Adventure, Drama, Romance	A spacecraft traveling to a distar	Morten Tyldum	Jennifer Lawrence	2016	116	7	192177	100.01	41
12		11 Fantastic Beasts and Where	Adventure, Family, Fantasy	The adventures of writer Newt	David Yates	Eddie Redmayne,	2016	133	7.5	232072	234.02	66
13	1	12 Hidden Figures	Biography, Drama, History	The story of a team of female A	Theodore Melfi	Taraji P. Henson,	2016	127	7.8	93103	169.27	74
14	1	13 Rogue One	Action,Adventure,Sci-Fi	The Rebel Alliance makes a risk	Gareth Edwards	Felicity Jones, Die	2016	133	7.9	323118	532.17	65
15	1	14 Moana	Animation, Adventure, Comedy	In Ancient Polynesia, when a te	Ron Clements	Auli'i Cravalho, De	2016	107	7.7	118151	248.75	81
16	1	15 Colossal	Action,Comedy,Drama	Gloria is an out-of-work party gi	Nacho Vigalondo	Anne Hathaway, J	2016	109	6.4	8612	2.87	70
17		16 The Secret Life of Pets	Animation, Adventure, Comedy	The quiet life of a terrier named	Chris Renaud	Louis C.K., Eric Sto	2016	87	6.6	120259	368.31	61
18	1	17 Hacksaw Ridge	Biography, Drama, History	WWII American Army Medic De	Mel Gibson	Andrew Garfield,	2016	139	8.2	211760	67.12	71
19		18 Jason Bourne	Action,Thriller	The CIA's most dangerous forme	Paul Greengrass	Matt Damon, Tom	2016	123	6.7	150823	162.16	58
20		19 Lion	Biography.Drama	A five-vear-old Indian boy gets	Garth Davis	Dev Patel. Nicole	2016	118	8.1	102061	51.69	69

# ETL PROCESS:

ETL is a process that extracts the data from different source systems, then
transforms the data (like applying calculations, concatenations, etc.) and
finally loads the data into the Data Warehouse system.
Full form of ETL is Extract, Transform and Load.

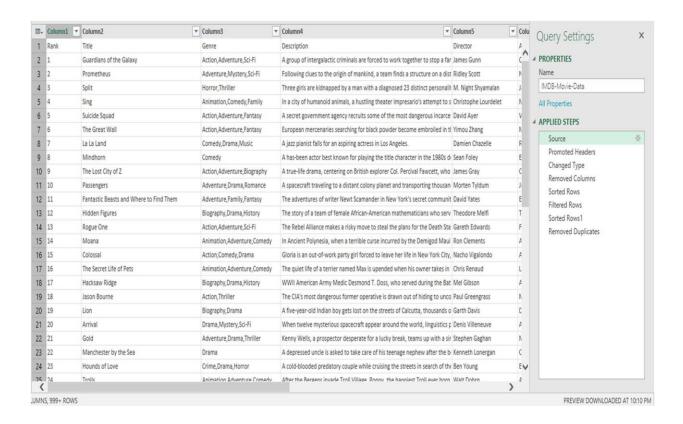
- ☐ The triple combination of ETL provides crucial functions that are many times combined into a single application or suite of tools that help in the following areas:
  - Enhances Business Intelligence solutions for decision making.
  - Allows verification of data transformation, aggregation and calculations rules.
  - Allows sample data comparison between source and target system.
  - Helps to improve productivity as it codifies and reuses without additional technical skills.
- ☐ Initially, the raw dataset was arranged as shown in given picture:

1	Rank	Title	Genre	Description	Director	Actors	Year	Runtime (	Rating	Votes	Revenue	Metascore
2	1	Guardians	Action,Ad	A group o	James Gui	Chris Prat	2014	121	8.1	757074	333.13	76
3	2	Promethe	Adventure	Following	Ridley Sco	Noomi Ra	2012	124	7	485820	126.46	65
4	3	Split	Horror,Th	Three girl:	M. Night S	James Mc	2016	117	7.3	157606	138.12	62
5	4	Sing	Animation	In a city of	Christoph	Matthew	2016	108	7.2	60545	270.32	59
6	5	Suicide So	Action,Ad	A secret g	David Aye	Will Smith	2016	123	6.2	393727	325.02	40
7	6	The Great	Action,Ad	European	Yimou Zha	Matt Dam	2016	103	6.1	56036	45.13	42
8	7	La La Land	Comedy,D	A jazz piar	Damien C	Ryan Gosl	2016	128	8.3	258682	151.06	93
9	8	Mindhorn	Comedy	A has-bee	Sean Fole	Essie Davi	2016	89	6.4	2490		71
10	9	The Lost C	Action,Ad	A true-life	James Gra	Charlie Hu	2016	141	7.1	7188	8.01	78
11	10	Passenger	Adventure	A spacecra	Morten Ty	Jennifer L	2016	116	7	192177	100.01	41
12	11	Fantastic I	Adventure	The adver	David Yate	Eddie Red	2016	133	7.5	232072	234.02	66
13	12	Hidden Fi	Biography	The story	Theodore	Taraji P. H	2016	127	7.8	93103	169.27	74
14	13	Rogue On	Action,Ad	The Rebel	Gareth Ed	Felicity Jo	2016	133	7.9	323118	532.17	65
15	14	Moana	Animation	In Ancient	Ron Clem	Auli'i Crav	2016	107	7.7	118151	248.75	81
16	15	Colossal	Action,Co	Gloria is a	Nacho Vig	Anne Hath	2016	109	6.4	8612	2.87	70
17	16	The Secre	Animation	The quiet	Chris Rena	Louis C.K.,	2016	87	6.6	120259	368.31	61
18	17	Hacksaw F	Biography	WWII Ame	Mel Gibso	Andrew G	2016	139	8.2	211760	67.12	71
19	18	Jason Bou	Action,Th	The CIA's	Paul Gree	Matt Dam	2016	123	6.7	150823	162.16	58
20	19	Lion	Biography	A five-yea	Garth Dav	Dev Patel	2016	118	8.1	102061	51.69	69
21	20	Arrival	Drama, My	When twe	Denis Vill	Amy Adan	2016	116	8	340798	100.5	81
22	21	Gold	Adventure	Kenny We	Stephen 0	Matthew	2016	120	6.7	19053	7.22	49
23	22	Manchest	Drama	A depress	Kenneth L	Casey Affl	2016	137	7.9	134213	47.7	96

Steps taken to clean dataset thorough ETL process

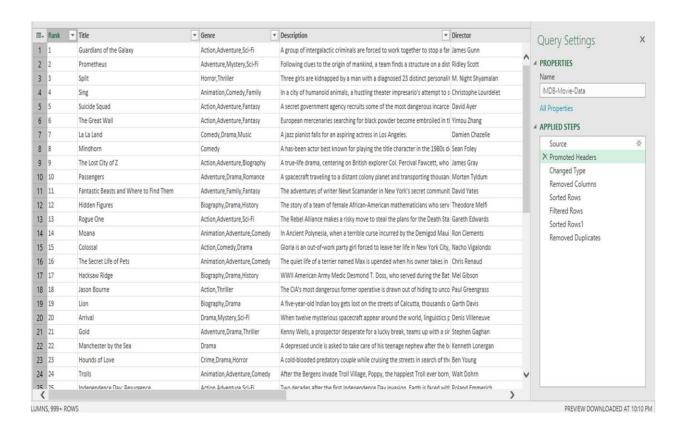
#### Step 1:

☐ Firstly, open blank excel file and go to Data from Tab and select New query and entered data csv file from it.



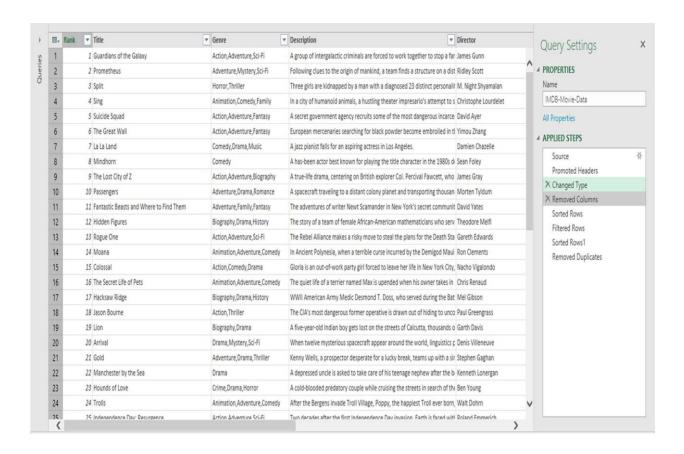
### Step 2:

☐ After entering the query editor, data fields title will promote into header



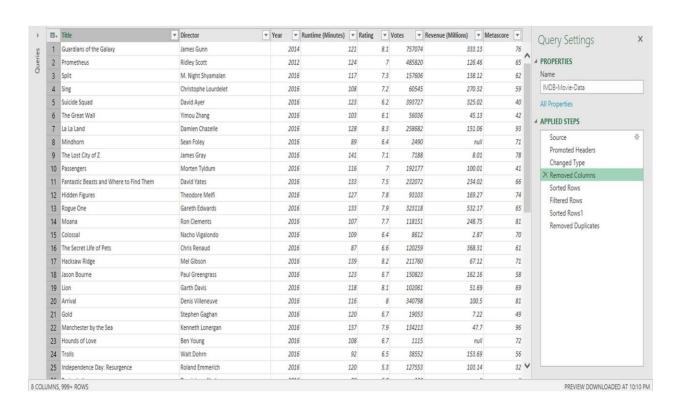
#### Step 3:

☐ Changed data type will change the type of data fields such as metascores converted from text into number, Revenue converted from text into currency



#### Step 4:

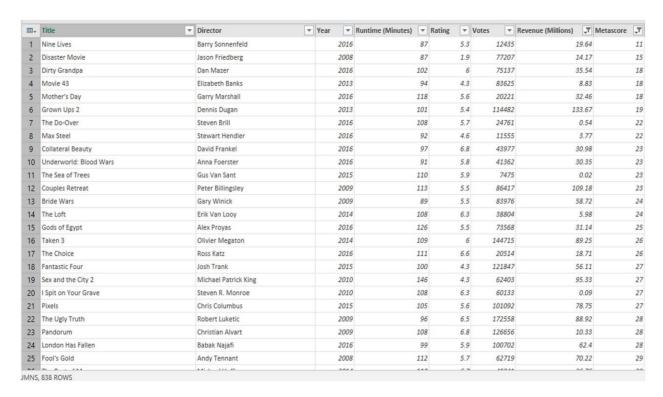
- □ Now, from query editor select Home tab and remove column which are not in required for analysis. Column which are removed from dataset are given below:
  - Rank
  - Genre
  - Description
  - Actor



#### Step 5:

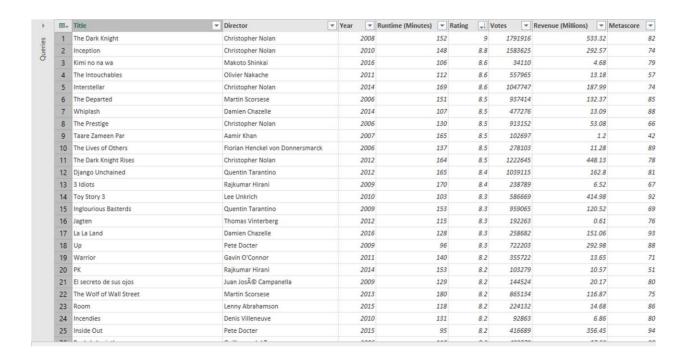
□ Now to remove null values we first sort the dataset show that null values comes on the top and after that we can delete it through Delete rows from Home tab.





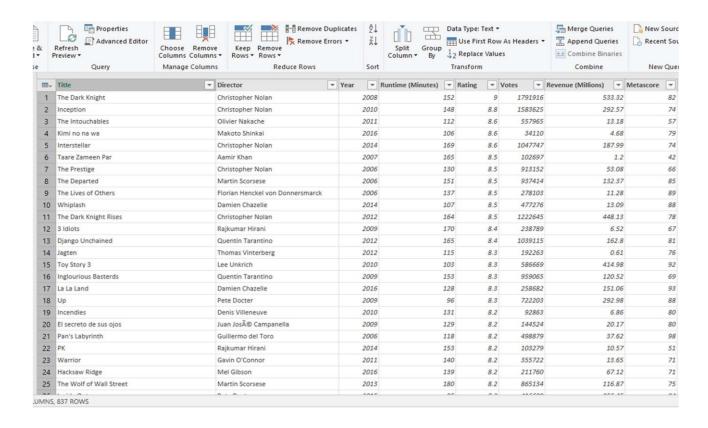
#### Step 6:

☐ Finally, sort the dataset through filter of rating data fields



Step 7:

☐ Remove duplicates from the data using duplicate option rom home tab



Finally, after cleaning the data, the final dataset sample is shown below:

Title	Director	Year 💌	Runtime (Minutes)	Rating 💌	Votes 💌	Revenue (Millions) 🔽 I	Metascore 💌
The Dark Knight	Christopher Nolan	2008	31.00:00:00	9	1791916	533.32	82
Inception	Christopher Nolan	2010	27.00:00:00	8.8	1583625	292.57	74
Interstellar	Christopher Nolan	2014	17.00:00:00	8.6	1047747	187.99	74
Kimi no na wa	Makoto Shinkai	2016	15.00:00:00	8.6	34110	4.68	79
The Intouchables	Olivier Nakache	2011	21.00:00:00	8.6	557965	13.18	57
The Prestige	Christopher Nolan	2006	9.00:00:00	8.5	913152	53.08	66
Taare Zameen Par	Aamir Khan	2007	13.00:00:00	8.5	102697	1.2	42
The Dark Knight Rises	Christopher Nolan	2012	12.00:00:00	8.5	1222645	448.13	78
Whiplash	Damien Chazelle	2014	16.00:00:00	8.5	477276	13.09	88
The Lives of Others	Florian Henckel von Donnersmarck	2006	16.00:00:00	8.5	278103	11.28	89
The Departed	Martin Scorsese	2006	30.00:00:00	8.5	937414	132.37	85
Django Unchained	Quentin Tarantino	2012	13.00:00:00	8.4	1039115	162.8	81
3 Idiots	Rajkumar Hirani	2009	18.00:00:00	8.4	238789	6.52	67
Koe no katachi	Naoko Yamada	2016	8.00:00:00	8.4	2421	6.98	80
Up	Pete Docter	2009	5.00:00:00	8.3	722203	292.98	88
Inglourious Basterds	Quentin Tarantino	2009	1.00:00:00	8.3	959065	120.52	69
Toy Story 3	Lee Unkrich	2010	12.00:00:00	8.3	586669	414.98	92
Jagten	Thomas Vinterberg	2012	24.00:00:00	8.3	192263	0.61	76
La La Land	Damien Chazelle	2016	7.00:00:00	8.3	258682	151.06	93
Paint It Black	Amber Tamblyn	2016	5.00:00:00	8.3	61	131.1	71
PK	Rajkumar Hirani	2014	1.00:00:00	8.2	103279	10.57	51
Room	Lenny Abrahamson	2015	27.00:00:00	8.2	224132	14.68	86
Pan's Labyrinth	Guillermo del Toro	2006	27.00:00:00	8.2	498879	37.62	98
Warrior	Gavin O'Connor	2011	19.00:00:00	8.2	355722	13.65	71
Hacksaw Ridge	Mel Gibson	2016	18.00:00:00	8.2	211760	67.12	71
Incendies	Denis Villeneuve	2010	10.00:00:00	8.2	92863	6.86	80
El secreto de sus ojos	Juan José Campanella	2009	8.00:00:00	8.2	144524	20.17	80
The Wolf of Wall Street	Martin Scorsese	2013	28.00:00:00	8.2	865134	116.87	75

# Analysis on dataset

## 1. Most Rating Movies

#### Introduction

 $\square$  By performing this analysis, we will Sum of Rating earn by movies from 2006 to 2016.

#### Description

☐ The analysis based on the year, revenue of the dataset

Specific requirements, functions and formulas

- $\square$  Pivot table is used for the analysis.
- ☐ Sum function is used in pivot table for the sum of the rating in the pivot table

#### Analysis results

Row Labels 📑 S	um of Rating
Inception	8.8
Interstellar	8.6
Kimi no na wa	8.6
The Dark Knight	9
The Intouchables	8.6
Grand Total	43.6

#### Visualization



#### 1. Most Metascore Movies

#### Introduction

 $\square$  By performing this analysis, we will Sum of Rating earn by movies from 2006 to 2016.

#### Description

 $\hfill\Box$  The analysis based on the year, revenue of the dataset

Specific requirements, functions and formulas

- $\Box$  Pivot table is used for the analysis.
- ☐ Sum function is used in pivot table for the sum of the rating in the pivot table

### Analysis results

Row Labels	<b>I</b> Sum of Metascore
Boyhood	100
Moonlight	99
Pan's Labyrinth	98
12 Years a Slave	96
Ratatouille	96
Gravity	96
Manchester by the S	ea 96
Grand Total	681

#### Visualization



## 3. Total revenues(millions) in each year

#### Introduction

 $\square$  By performing this analysis, we will Total revenues earn by movies from 2006 to 2016.

#### Description

 $\Box$  The analysis based on the year, revenue of the dataset

#### Specific requirements, functions and formulas

- $\square$  Pivot table is used for the analysis.
- ☐ Sum function is used in pivot table for the sum of the revenues in the pivot table

### Analysis results

Row Labels	▼ Sum of Revenue (Millions)
2006	3575.28
2007	4095.26
2008	4681.22
2009	5208.39
2010	5989.65
2011	5341.09
2012	6826.39
2013	7399.11
2014	7945.33
2015	8799.09
2016	11002.09
<b>Grand Total</b>	70862.9

#### Visualization



## 4. Total Movies released in each year

#### Introduction

☐ By performing this analysis, we will get Number of movies released in each year

### Description

 $\Box$  The analysis based on the year, title of the dataset

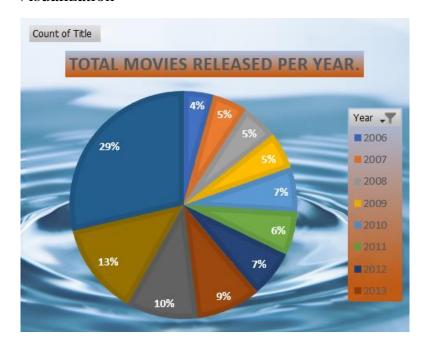
Specific requirements, functions and formulas

- $\square$  Pivot table is used for the analysis.
- ☐ Count function is used in pivot table for the counting of the title in the pivot table in their respective year

#### Analysis results

Row Labels 🔻	Count of Title
2006	40
2007	46
2008	49
2009	49
2010	59
2011	58
2012	62
2013	86
2014	95
2015	123
2016	268
Grand Total	935

#### Visualization



### 5. Top 5 Revenue Movies

#### Introduction

□ By performing this analysis, we will get Top 5 revenue movies of all time or selected year in the slicer

#### Description

- $\Box$  The analysis is based on title, revenue of the dataset
- $\Box$  Filter is used to show top 5 movies
- ☐ Slicer is connected with the graph to get each year top 5 revenue movies

### Specific requirements, functions and formulas

- $\square$  Pivot table is used for the analysis.
- □ Sum function is used in pivot table for the sum of revenue of the title

#### Analysis results

	-
Row Labels	Sum of Revenue (Millions)2
The Dark Knight	533.32
The Avengers	623.28
Star Wars: Episode VII - The Force Awakens	936.63
Jurassic World	652.18
Avatar	760.51
Grand Total	3505.92

#### Visualization



#### 6. Most voted Director

#### Introduction

☐ By performing this analysis, we will get Top 5 Most voted director from the dataset

#### Description

- $\ \square$  The analysis is based on Director, Votes of the dataset
- $\Box$  Filter is used to show top 5 most voted director
- ☐ Slicer is connected with the graph to get each year top 5 most voted director

## Specific requirements, functions and formulas

- ☐ Pivot table is used for the analysis.
- Sum function is used in pivot table for the sum of votes for the correspondence director.

### Analysis results

Row Labels 🔻 S	um of Votes
Christopher Nolan	6559085
David Fincher	2309652
David Yates	2033854
J.J. Abrams	2174937
Martin Scorsese	2966524
Michael Bay	1706531
Peter Jackson	1698695
Quentin Tarantino	2339350
Ridley Scott	2080074
Zack Snyder	2301544
Grand Total	26170246

#### Visualization



### 7. Top 5 Longest Movies

#### Introduction

☐ By performing this analysis, we will get Top 5 longest movies of all time or selected year in the slicer

#### Description

- $\square$  The analysis is based on title, runtimes of the dataset
- $\Box$  Filter is used to show top 5 movies
- ☐ Slicer is connected with the graph to get each year top 5 revenue movies

### Specific requirements, functions and formulas

- $\Box$  Pivot table is used for the analysis.
- □ Sum function is used in pivot table for the sum of runtimes of the title

#### Analysis results

Row Labels	Sum of Runtime (Minutes)
The Hateful Eight	187
The Wolf of Wall Stree	t 180
La vie d'Adèle	180
Cloud Atlas	172
3 Idiots	170
Grand Total	222

#### Visualization



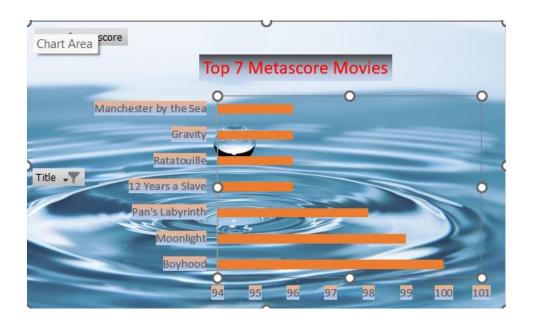
# List of Analysis with results

- 1. Top 5 rating movies
  - $\Box$  Top 5 rating movies:
    - 1) The Dark knight
      - 2) Inception
    - 3) The Intouchables
      - 4) Interstellar
      - 5) Kimi no na wa



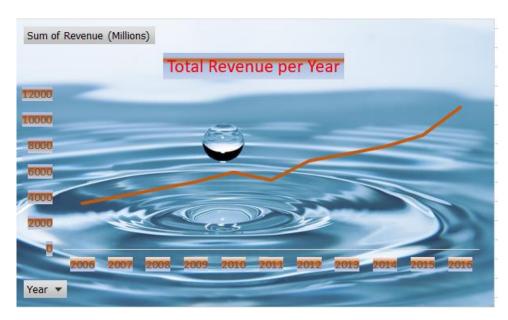
### $2.\Box$ Top 7Metascores movies:

- 1) Boyhood
- 2) Moonlight
- 3) Pan's Labyrinth
- 4) 12 Years a slave
  - 5) Ratatouille
- 6) Manchester by the Sea
  - 7) Gravity

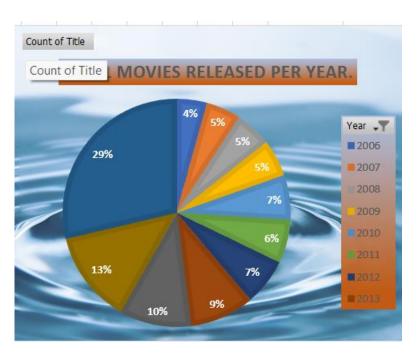


## 3. Total revenues(millions) in each year

☐ The Highest Revenue was recorded in 2016 with total \$11002.09 million. Whereas the lowest Revenue was recorded in 2006 with total \$3575.28 million.

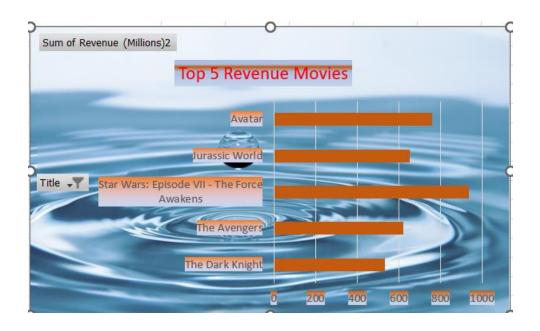


- 4. Total Movies released in each year
- ☐ The highest number of movies was released in 2016 with total 198 movies.
- ☐ While, the lowest number of movies was released in 2006 with total 40 movies.



### 5. Top 5 Revenue Movies

- $\Box$  The top 5 revenues are given below:
  - 1) Star Wars: Episode VII The Force Awakens
  - 2) Avatar
  - 3) Jurassic World
  - 4) The Avengers
  - 5) The Dark Knight



#### 6. Most voted Director

☐ The Most Voted Director in the dataset is Christopher Nolan. ☐ After that, the second most voted Director is Martin Scorsese

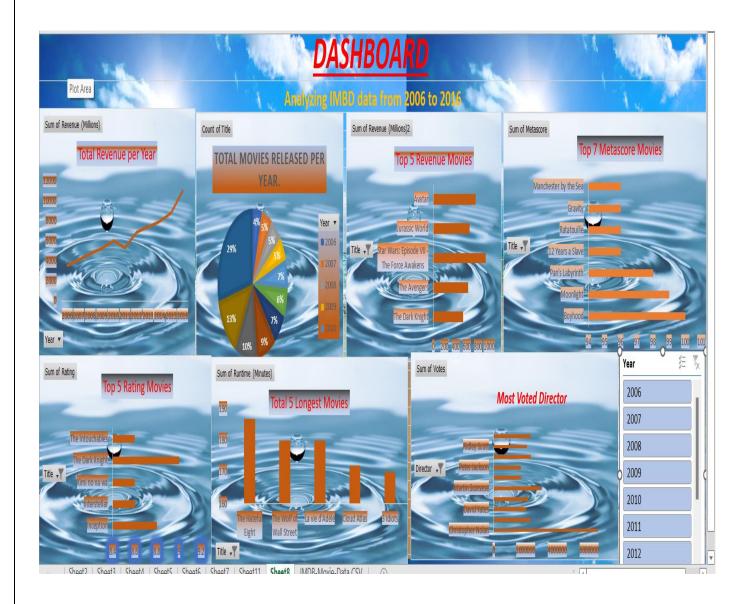


## 7. Top 5 Longest Movies

- $\Box$  The top 5 longest movies are given below:
  - 1) The Hateful Eight
  - 2) The Wolf of Wall Street
  - 3) La Vie d'Adèle
  - 4) Cloud Atlas
  - 5) 3 Idiots



## **FINAL DASHBOARD:**



## **BIBLIOGRAPHY:**

☐ Dataset source:

https://www.kaggle.com/PromptCloudHQ/imdb-data

## ☐ Dashboard Background Image:

 $https://hubblecontent.osi.office.net/contentsvc/ui/index.html?picker=1\&fb=1\&p=0\&t=0\&a=1\&idp=live\&lid=2\&taskpane=0\&ht=0\&license=Perpetual2019\&lcid=1033\&syslcid=16393\&uilcid=1033\&app=1\&ver=16\&tl=2\&build=16.0.15726\&searchtype=Image\&cfd=1\&moss=0\&ins=1\&albm=0\&eurl=0\&msel=0\&sl=all\&dontforcehomeurl=1\&setlang=en-US\&pi=1\&style=0\&correlationid=b7e15125-ee12-40c8-8291-c036c6356f48\&uid=ee4cd33305f92e0\_LiveId\&cw=683\&ch=402\&cp=bing\&comcall=1\&fparty=1\&mkt=en-IN$