### **Project: Investigate a Dataset (TMDB movies)**

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### Introduction

1-In order to complete my Investigate a dataset project i am using a dataset called TMDB movies file downloaded from the udacity.

2-the shape of the dataset is 10866 rows with 21 columns.

### Questions that can be used to analyse the dataset are:

- 1-Movies with profit and loss.
- 2-Movies with lower and higher budget.
- 3-Movies with lower and higher revenue.
- 4-Most of the genres.
- 5-Most of the cast.
- 6-Movie with shortest and longest runtime.

#### In [1]:

```
# Use this cell to set up import statements for all of the packages that you
# plan to use.

# Remember to include a 'magic word' so that your visualizations are plotted
# inline with the notebook. See this page for more:
# http://ipython.readthedocs.io/en/stable/interactive/magics.html
```

#### In [2]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
```

### **Data Wrangling**

--Data Wrangling is used to clean the dataset for better results. As per my observation from the dataset there are some of the cleanliness need to be done.

### **General Properties**

### In [3]:

# Load your data and print out a few lines. Perform operations to inspect data # types and look for instances of missing or possibly errant data. df=pd.read\_csv('tmdb-movies.csv') df.head(10)

### Out[3]:

	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	hi
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	281957	tt1663202	9.110700	135000000	532950503	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	hi
6	87101	tt1340138	8.654359	155000000	440603537	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	
7	286217	tt3659388	7.667400	108000000	595380321	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	
8	211672	tt2293640	7.404165	74000000	1156730962	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	
9	150540	tt2096673	6.326804	175000000	853708609	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	

10 rows × 21 columns

```
In [4]:

df.shape

Out[4]:

(10866, 21)

In [5]:

df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 21 columns):

id 10866 non-null int64 imdb\_id 10856 non-null object popularity 10866 non-null float64 budget 10866 non-null int64 revenue 10866 non-null int64 original\_title 10866 non-null object 10790 non-null object cast 2936 non-null object homepage director 10822 non-null object 8042 non-null object tagline 9373 non-null object keywords overview 10862 non-null object 10866 non-null int64 runtime genres 10843 non-null object production\_companies 9836 non-null object release\_date 10866 non-null object 10866 non-null int64 vote\_count vote\_average 10866 non-null float64 10866 non-null int64 release\_year

dtypes: float64(4), int64(6), object(11)

10866 non-null float64

10866 non-null float64

memory usage: 1.7+ MB

budget\_adj

revenue adj

### In [6]:

df\_clean=df.copy()
df\_clean.head(10)

### Out[6]:

	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	hi
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	281957	tt1663202	9.110700	135000000	532950503	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	hi
6	87101	tt1340138	8.654359	155000000	440603537	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	
7	286217	tt3659388	7.667400	108000000	595380321	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	
8	211672	tt2293640	7.404165	74000000	1156730962	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	
9	150540	tt2096673	6.326804	175000000	853708609	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	

In [7]:

df\_clean.shape

Out[7]:

(10866, 21)

In [8]:

df\_clean.describe()

Out[8]:

	id	popularity	budget	revenue	runtime	vote_count	١
count	10866.000000	10866.000000	1.086600e+04	1.086600e+04	10866.000000	10866.000000	1
mean	66064.177434	0.646441	1.462570e+07	3.982332e+07	102.070863	217.389748	
std	92130.136561	1.000185	3.091321e+07	1.170035e+08	31.381405	575.619058	
min	5.000000	0.000065	0.000000e+00	0.000000e+00	0.000000	10.000000	
25%	10596.250000	0.207583	0.000000e+00	0.000000e+00	90.000000	17.000000	
50%	20669.000000	0.383856	0.000000e+00	0.000000e+00	99.000000	38.000000	
75%	75610.000000	0.713817	1.500000e+07	2.400000e+07	111.000000	145.750000	
max	417859.000000	32.985763	4.250000e+08	2.781506e+09	900.000000	9767.000000	
4						)	•

#### In [9]:

```
df clean.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 21 columns):
id
                        10866 non-null int64
imdb id
                        10856 non-null object
popularity
                        10866 non-null float64
budget
                        10866 non-null int64
                        10866 non-null int64
revenue
                        10866 non-null object
original_title
                        10790 non-null object
cast
homepage
                        2936 non-null object
                        10822 non-null object
director
                        8042 non-null object
tagline
keywords
                        9373 non-null object
overview
                        10862 non-null object
runtime
                        10866 non-null int64
                        10843 non-null object
genres
production_companies
                        9836 non-null object
                        10866 non-null object
release_date
                        10866 non-null int64
vote_count
                        10866 non-null float64
vote_average
release_year
                        10866 non-null int64
                        10866 non-null float64
budget_adj
revenue_adj
                        10866 non-null float64
dtypes: float64(4), int64(6), object(11)
memory usage: 1.7+ MB
```

### Data Cleaning (Replace this with more specific notes!)

As per my observation made from the dataset, There are some cleaning, calculation and change of datatype of some columns are need to be done.

- 1-There are 4 columns that are duplicated values in the original title column from the dataset.
- 2-There are some unwanted columns that need to be droped.
- 3-caluclation for profit and profit adj(from columns budget,revenue,budget adj,revenue adj)
- 4-changinf the datatypes of some columns.

Note:-for the currency related columns i am going to assume that the currency will be in Dollors, as dollors is used as the international currency.

```
In [10]:
```

```
# After discussing the structure of the data and any problems that need to be cleaned, perform those cleaning steps in the second part of this section.
```

### Finding sum of null values in the Dataset:

```
In [11]:
```

original\_title 0 76 cast homepage 7930 44 director tagline 2824 keywords 1493 overview 4 runtime 0 genres 23 production\_companies 1030 release\_date 0 0 vote\_count vote\_average 0 release\_year 0

revenue\_adj
dtype: int64

budget\_adj

### Finding the number of Duplicated values in the original\_titles column:

0

0

### In [15]:

```
df_clean.original_title.duplicated(False).value_counts()
```

### Out[15]:

False 10294 True 572

Name: original\_title, dtype: int64

### In [20]:

```
df_clean[df_clean.original_title.duplicated(False)]
Out[20]:
```

		cast	original_title	revenue	budget	popularity	imdb_id	id	
₩	http://	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	The Revenant	532950503	135000000	9.110700	tt1663202	281957	5
		Lily James Cate Blanchett Richard Madden Helen	Cinderella	542351353	95000000	5.556818	tt1661199	150689	18
•		Jim Parsons Rihanna Steve Martin Jennifer Lope	Home	368871007	135000000	2.976436	tt2224026	228161	46
	)								∢

### Comparing two columns to find the duplicate values columns are original\_title and cast:

### In [21]:

```
df_clean.duplicated(['original_title','cast'],False).value_counts()
```

### Out[21]:

False 10862 True 4 dtype: int64

```
In [23]:

df_clean[df_clean.duplicated(['original_title','cast'],False)]
```

Out[23]:

	id	imdb_id	popularity	budget	revenue	original_title	cast	homepaς
2089	42194	tt0411951	0.596430	30000000	967000	TEKKEN	Jon Foo Kelly Overton Cary- Hiroyuki Tagawa Ian	Na
2090	42194	tt0411951	0.596430	30000000	967000	TEKKEN	Jon Foo Kelly Overton Cary- Hiroyuki Tagawa Ian	Na
4063	28004	tt0359639	0.371510	6000000	0	Madea's Family Reunion	Tyler Perry Blair Underwood Lynn Whitfield Bor	Na
6701	16781	tt0455612	0.552267	6000000	57231524	Madea's Family Reunion	Tyler Perry Blair Underwood Lynn Whitfield Bor	Na
4 rows	s × 21 c	olumns						
4								•

Droping the duplicate values except the first occurrence and checking the duplicate values:

```
In [31]:

df_clean.drop_duplicates(['original_title','cast'],keep='first',inplace=True)

In [32]:

df_clean[df_clean.duplicated(['original_title','cast'])]

Out[32]:

id imdb_id popularity budget revenue original_title cast homepage director tagline ...

O rows × 21 columns
```

Dropping unwanted columns such as id,imdb\_id,keywords,overview,production\_companies:

### In [33]:

df\_clean.drop(['id','imdb\_id','homepage','keywords','overview','production\_companies','home

### In [34]:

df\_clean.head(10)

### Out[34]:

	popularity	budget	revenue	original_title	cast	director	runtime
0	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	124
1	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	120
2	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	119
3	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	136
4	9.335014	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle	James Wan	137
5	9.110700	135000000	532950503	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	Alejandro González Iñárritu	156
6	8.654359	155000000	440603537	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	Alan Taylor	125
7	7.667400	108000000	595380321	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	Ridley Scott	141
8	7.404165	74000000	1156730962	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	Kyle Balda Pierre Coffin	91
9	6.326804	175000000	853708609	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	Pete Docter	94
4							•

### Change of datat types to some of the columns:

1-budget\_adj column from float64 to int64.

2-revenue\_adj column from float6 to int 64.

3-release\_date from str to datetime.

```
In [35]:

df_clean['budget_adj']=df_clean['budget_adj'].astype('int64')

In [36]:

df_clean['revenue_adj']=df_clean['revenue_adj'].astype('int64')

In [37]:

df_clean['release_date']=pd.to_datetime(df_clean['release_date'],format='%m/%d/%y')
```

### In [38]:

```
df_clean.dtypes
```

#### Out[38]:

popularity float64 budget int64 int64 revenue original\_title object object cast director object runtime int64 object genres release\_date datetime64[ns] int64 vote\_count float64 vote\_average int64 release\_year int64 budget\_adj revenue adj int64 dtype: object

### Caluclation for profit and profit\_adj:

```
In [39]:
```

```
df_clean['profit']=df_clean['revenue']-df_clean['budget']
```

```
In [40]:
```

```
df_clean['profit_adj']=(df_clean['revenue_adj']-df_clean['budget_adj']).astype('int64')
```

### In [41]:

df\_clean.head(10)

### Out[41]:

	popularity	budget	revenue	original_title	cast	director	runtime
0	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi	Colin Trevorrow	124
1	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays- Byrne Nic	George Miller	120
2	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel	Robert Schwentke	119
3	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D	J.J. Abrams	136
4	9.335014	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle	James Wan	137
5	9.110700	135000000	532950503	The Revenant	Leonardo DiCaprio Tom Hardy Will Poulter Domhn	Alejandro González Iñárritu	156
6	8.654359	155000000	440603537	Terminator Genisys	Arnold Schwarzenegger Jason Clarke Emilia Clar	Alan Taylor	125
7	7.667400	108000000	595380321	The Martian	Matt Damon Jessica Chastain Kristen Wiig Jeff	Ridley Scott	141
8	7.404165	74000000	1156730962	Minions	Sandra Bullock Jon Hamm Michael Keaton Allison	Kyle Balda Pierre Coffin	91
9	6.326804	175000000	853708609	Inside Out	Amy Poehler Phyllis Smith Richard Kind Bill Ha	Pete Docter	94
4							•

### In [42]:

df\_clean.budget.replace((0),np.NaN,inplace=True)

### **Exploratory Data Analysis**

**Tip**: Now that you've trimmed and cleaned your data, you're ready to move on to exploration. Compute statistics and create visualizations with the goal of addressing the research questions that you posed in the Introduction section. It is recommended that you be systematic with your approach. Look at one variable at a time, and then follow it up by looking at relationships between variables.

# Question 1-Movies with profit and loss.

### In [63]:

movie\_profit=pd.DataFrame(df\_clean.loc[df\_clean.profit.idxmax()])
movie\_loss=pd.DataFrame(df\_clean.loc[df\_clean.profit.idxmin()])
pd.concat([movie\_profit,movie\_loss],axis=1)

### Out[63]:

	1386	2244
popularity	9.43277	0.25054
budget	2.37e+08	4.25e+08
revenue	2781505847	11087569
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
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
vote_count	8458	74
vote_average	7.1	6.4
release_year	2009	2010
budget_adj	240886902	425000000
revenue_adj	2827123750	11087569
profit	2544505847	-413912431
profit_adj	2586236848	-413912431

# Question 2-Movies with lower and higher budget.

### In [67]:

```
movie_budget_high=pd.DataFrame(df_clean.loc[df_clean.budget.idxmax()])
movie_budget_low=pd.DataFrame(df_clean.loc[df_clean.budget.idxmin()])
pd.concat([movie_budget_high,movie_budget_low],axis=1)
```

### Out[67]:

	2244	1151
popularity	0.25054	0.177102
budget	4.25e+08	1
revenue	11087569	0
original_title	The Warrior's Way	Fear Clinic
cast	Kate Bosworth Jang Dong-gun Geoffrey Rush Dann	Thomas Dekker Robert Englund Cleopatra Coleman
director	Sngmoo Lee	Robert Hall
runtime	100	95
genres	Adventure Fantasy Action Western Thriller	Horror
release_date	2010-12-02 00:00:00	2014-10-31 00:00:00
vote_count	74	15
vote_average	6.4	4.1
release_year	2010	2014
budget_adj	425000000	0
revenue_adj	11087569	0
profit	-413912431	-1
profit_adj	-413912431	0

# Question 3-Movies with lower and higher revenue.

### In [72]:

```
movie_revenue_high=pd.DataFrame(df_clean.loc[df_clean.revenue.idxmax()])
movie_revenue_low=pd.DataFrame(df_clean.loc[df_clean.revenue.idxmin()])
pd.concat([movie_revenue_high,movie_revenue_low],axis=1)
```

### Out[72]:

	1386	48
popularity	9.43277	2.93234
budget	2.37e+08	3e+07
revenue	2781505847	0
original_title	Avatar	Wild Card
cast	Sam Worthington Zoe Saldana Sigourney Weaver S	Jason Statham Michael Angarano Milo Ventimigli
director	James Cameron	Simon West
runtime	162	92
genres	Action Adventure Fantasy Science Fiction	Thriller Crime Drama
release_date	2009-12-10 00:00:00	2015-01-14 00:00:00
vote_count	8458	481
vote_average	7.1	5.3
release_year	2009	2015
budget_adj	240886902	27599987
revenue_adj	2827123750	0
profit	2544505847	-30000000
profit_adj	2586236848	-27599987

# Question 4-Movie with shortest and longest runtime.

### In [74]:

```
movie_runtime_high=pd.DataFrame(df_clean.loc[df_clean.runtime.idxmax()])
movie_runtime_low=pd.DataFrame(df_clean.loc[df_clean.runtime.idxmin()])
pd.concat([movie_runtime_high,movie_runtime_low],axis=1)
```

### Out[74]:

	3894	92
popularity	0.006925	1.87604
budget	NaN	NaN
revenue	0	0
original_title	The Story of Film: An Odyssey	Mythica: The Necromancer
cast	Mark Cousins Jean-Michel Frodon Cari Beauchamp	Melanie Stone Adam Johnson Kevin Sorbo Nicola
director	Mark Cousins	A. Todd Smith
runtime	900	0
genres	Documentary	Fantasy Action Adventure
release_date	2011-09-03 00:00:00	2015-12-19 00:00:00
vote_count	14	11
vote_average	9.2	5.4
release_year	2011	2015
budget_adj	0	0
revenue_adj	0	0
profit	0	0
profit_adj	0	0

# Count of genres in the dataset:

### In [125]:

```
def genres(column):
    genres=df_clean[column].str.cat(sep='|')
    genres=pd.Series(genres.split("|"))
    count=genres.value_counts()
    return count
genres_count=genres('genres')
genres_count
```

### Out[125]:

Drama	4759
Comedy	3792
Thriller	2907
Action	2384
Romance	1711
Horror	1637
Adventure	1471
Crime	1354
Family	1231
Science Fiction	1229
Fantasy	916
Mystery	810
Animation	699
Documentary	520
Music	408
History	334
War	270
Foreign	188
TV Movie	167
Western	165
dtype: int64	

# Count of casts in the dataset:

### In [174]:

```
def cast(column):
    cast=df_clean[column].str.cat(sep='|')
    cast=pd.Series(cast.split("|"))
    count=cast.value_counts()
    return count
cast_count=cast('cast')
cast_count
```

### Out[174]:

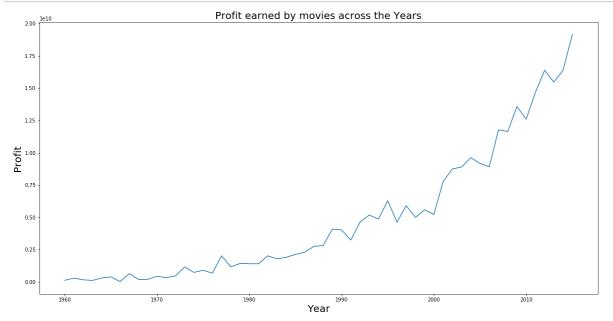
Out[174]:	
Robert De Niro	72
Samuel L. Jackson	71
Bruce Willis	62
Nicolas Cage	61
Michael Caine	53
Robin Williams	51
John Cusack	50
Morgan Freeman	49
John Goodman	49
Susan Sarandon	48
Liam Neeson	48
Alec Baldwin	47
Julianne Moore	47
Tom Hanks	46
Christopher Walken	46
Gene Hackman	46
Johnny Depp	46
Willem Dafoe	45
Sylvester Stallone	45
Dennis Quaid	45
Clint Eastwood	44
Meryl Streep	44
Ed Harris	44
Donald Sutherland	44
Robert Downey Jr.	43
Woody Harrelson	43
Keanu Reeves	43
Antonio Banderas	43
Ewan McGregor	43
Eddie Murphy	42
. ,	• •
Matt Corboy	1
Monica Swinn	1
Bernie McInerney	1
Gisele Fraga	1
Gérard Darmon	1
George Williams	1
Barry Nelson	1
Yuta Hiraoka	1
ShÅ Hayami	1
Yukihide Benny	1
David Bagby	1
Rita Tushingham	1
Lymari Nadal	1
Daniela Ruah	1
Kevin Clash	1
Monique Alexander	1
Andrew Fullerton	1
Joyce Jameson	1

Pete Smith 1 Jesse Garcia Thorsten Kaye 1 Jake Siegel 1 Michel Bouquet 1 Michael Polley 1 Kris Black 1 Will Tranfo 1 Bill Johnson Dean Marshall 1 Alexa Nikolas Ayrton Senna Length: 19026, dtype: int64

## Profit earned by movies across the year:

### In [180]:

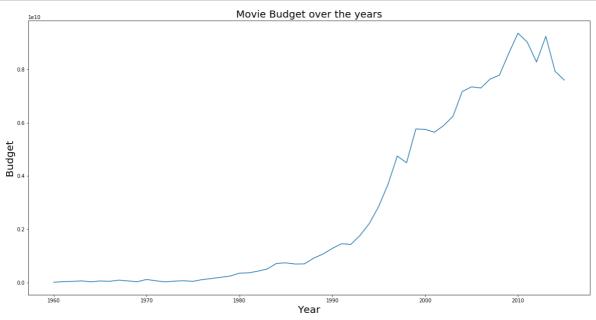
```
profit_year=df_clean.groupby('release_year')['profit'].sum()
plt.figure(figsize=(20,10));
plt.xlabel('Year',fontsize=20)
plt.ylabel('Profit',fontsize=20)
plt.title('Profit earned by movies across the Years',fontsize=20)
plt.plot(profit_year);
```



# Movie budget Over the years:

#### In [181]:

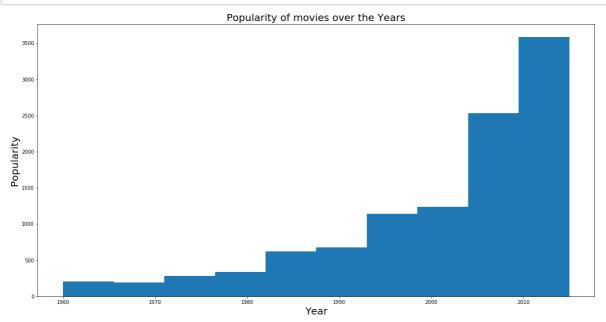
```
budget_year=df_clean.groupby('release_year')['budget'].sum()
plt.figure(figsize=(20,10));
plt.xlabel('Year',fontsize=20);
plt.ylabel('Budget',fontsize=20);
plt.title('Movie Budget over the years',fontsize=20);
plt.plot(budget_year);
```



# Popularity of the movies over the years:

#### In [179]:

```
popularity=df_clean.groupby('popularity')['release_year'].mean()
plt.figure(figsize=(20,10));
plt.xlabel('Year',fontsize=20);
plt.ylabel('Popularity',fontsize=20);
plt.title('Popularity of movies over the Years',fontsize=20)
plt.hist(popularity);
```



### **Conclusions**

**Tip**: Finally, summarize your findings and the results that have been performed. Make sure that you are clear with regards to the limitations of your exploration. If you haven't done any statistical tests, do not imply any statistical conclusions. And make sure you avoid implying causation from correlation!

**Tip**: Once you are satisfied with your work here, check over your report to make sure that it is satisfies all the areas of the rubric (found on the project submission page at the end of the lesson). You should also probably remove all of the "Tips" like this one so that the presentation is as polished as possible.

### **Submitting your Project**

Before you submit your project, you need to create a .html or .pdf version of this notebook in the workspace here. To do that, run the code cell below. If it worked correctly, you should get a return code of 0, and you should see the generated .html file in the workspace directory (click on the orange Jupyter icon in the upper left).

Alternatively, you can download this report as .html via the **File > Download as** submenu, and then manually upload it into the workspace directory by clicking on the orange Jupyter icon in the upper left, then using the Upload button.

Once you've done this, you can submit your project by clicking on the "Submit Project" button in the lower right here. This will create and submit a zip file with this .ipynb doc and the .html or .pdf version you created. Congratulations!

```
In [ ]:
```

```
from subprocess import call
call(['python', '-m', 'nbconvert', 'Investigate_a_Dataset.ipynb'])
```