## RECOMMENDING MICROSOFT MOVIES TO MAKE

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#### Overview

A one-paragraph overview of the project, including the business problem, data, methods, results and recommendations.

Microsoft is planning to create a movie production company and would like to get recommendations on the type of films that the type of films to produce according to the gross trending movies in the box office. Using Python and the provided box office data I will separate the works per studio see what they have dona and what has worked for them the methods i have used are data exploration, dealt with missing values, explored categorical data, identified outliers, performed correlation, and visualized the data

the results I found are that films that people tend to love according to gross sales

Boyhood- Coming of Age story

Heartbreaker- comedy drama

Jurassic World Universe -Sci-fi Adventure Supported by a sequel to the first jurassic park boosting sales in the same year

Dark Knight Rises- Action

Harry Potter and the Deathly Hallows Part 2 - Fantasy Adventure also boosted by a prequel the part one of it

#### **Business Problem**

Microsoft is planning to create a movie production studio and would like to get recommendations on the type of films that the type of films to produce. They have no clue of where to begin and would like to utilize data from the gross trending movies in the box office to influence the kind of movies they should make

i chose to ask the following question; according to the top studio houses: which film has the highest international grossing

domestic grossing	which film has the highest
belong to	which genre do these films
released	what year were these films

this will help them come up wih genres that they can start producing on. as they are new to the business and want a large ammount of people to view their films for commercial success, following the mass produced films and genres will put them in the billboards

## **Data Understanding**

Describe the data being used for this project.

the data being used is the box\_office data as it will help us to see what the domestic and international viewers like to see, then from there we can emulate

the data consists of the title, studio name, domestic gross ammount, foreign gross ammount and the year the film premiered

i intend to categorise the data accoring to the studios and find out their properties

```
In [1]: # Import standard packages
   import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   import os
   import sqlite3
   %matplotlib inline
```

```
In [2]: #loading all csv files to use for analysis
box_office = pd.read_csv(r'C:\Users\DANIEL\Desktop\PROJECT 1 ASSETS\DATA\bom.movie_gross.csv')
tmdb = pd.read_csv(r'C:\Users\DANIEL\Desktop\PROJECT 1 ASSETS\DATA\tmdb.movies.csv')
the_numbers = pd.read_csv(r'C:\Users\DANIEL\Desktop\PROJECT 1 ASSETS\DATA\tn.movie_budgets.csv'
```

# Begin with the Box Office data to understand it clean it and get it ready for manipulation and visualization

```
box_office
In [3]:
Out[3]:
                                                   title
                                                            studio domestic_gross foreign_gross year
                                                                                      652000000.0 2010
                                             Toy Story 3
                                                                       415000000.0
             1
                               Alice in Wonderland (2010)
                                                               BV
                                                                       334200000.0
                                                                                      691300000.0 2010
                Harry Potter and the Deathly Hallows Part 1
                                                                       296000000.0
                                                                                      664300000.0 2010
             3
                                              Inception
                                                                       292600000.0
                                                                                      535700000.0 2010
                                                               WB
             4
                                      Shrek Forever After
                                                             P/DW
                                                                       238700000.0
                                                                                      513900000.0 2010
          3382
                                             The Quake
                                                            Magn.
                                                                             6200.0
                                                                                             NaN 2018
          3383
                               Edward II (2018 re-release)
                                                                                             NaN 2018
                                                               FM
                                                                             4800.0
          3384
                                                El Pacto
                                                              Sony
                                                                             2500.0
                                                                                             NaN 2018
          3385
                                              The Swan Synergetic
                                                                             2400.0
                                                                                             NaN 2018
          3386
                                       An Actor Prepares
                                                              Grav.
                                                                             1700.0
                                                                                             NaN 2018
```

3387 rows × 5 columns

```
In [4]: #checking out the different data types in Box Office csv file
box_office.info()
```

3 foreign\_gross 2037 non-null float64 4 year 3387 non-null int64 dtypes: float64(2), int64(1), object(2)

memory usage: 132.4+ KB

In [5]: #a brief statistical description of the data
box\_office.describe()

Out[5]: domestic\_gross foreign\_gross year count 3.359000e+03 2.037000e+03 3387.000000 2.874585e+07 7.503550e+07 2013.958075 mean std 6.698250e+07 1.373874e+08 2.478141 min 1.000000e+02 6.000000e+02 2010.000000 25% 1.200000e+05 3.700000e+06 2012.000000 50% 1.400000e+06 1.900000e+07 2014.000000 75% 2.790000e+07 7.550000e+07 2016.000000 max 9.367000e+08 9.605000e+08 2018.000000

In [6]: #getting info on any missing data. this way i can identify how to make it uniform and usable #then i can use it to identify outliers and do the necessary changes according to the findings box\_office.isnull().sum()

Out[6]: title 0
studio 5
domestic\_gross 28
foreign\_gross 1350
year 0
dtype: int64

DROPPING AND FILLING IN MISSING VALUES

In [7]: box\_office = box\_office.dropna(subset=['studio'])
 box\_office

#i have chosen to drop the rows in the studio column that were missing
#this is because the data in there cannot be extroplated using mean or mode as it is not an int
#if i physically added new data(random) it wouldn't affect the analysis greatly
#so i chose to frop them

Out[7]:		title	studio	domestic_gross	foreign_gross	year
	0	Toy Story 3	BV	415000000.0	652000000.0	2010
	1	Alice in Wonderland (2010)	BV	334200000.0	691300000.0	2010
	2	Harry Potter and the Deathly Hallows Part 1	WB	296000000.0	664300000.0	2010
	3	Inception	WB	292600000.0	535700000.0	2010
	4	Shrek Forever After	P/DW	238700000.0	513900000.0	2010
	•••					
	3382	The Quake	Magn.	6200.0	NaN	2018
	3383	Edward II (2018 re-release)	FM	4800.0	NaN	2018
	3384	El Pacto	Sony	2500.0	NaN	2018
	3385	The Swan	Synergetic	2400.0	NaN	2018
	3386	An Actor Prepares	Grav.	1700.0	NaN	2018

TREATING MISSING VALUES USING MEDIAN AS THE DATA COULD CONTAIN OUTLIERS MAKING IT SKEWED I ALSO USED MEDIAN AS THE DATA IS NOT DISTRIBUTED

```
In [8]:
           box_office['domestic_gross'] = box_office['domestic_gross'].fillna(box_office['domestic_gross']
           box office
           box_office['foreign_gross'] = box_office['foreign_gross'].fillna(box_office['foreign_gross'].me
           box office
          <ipython-input-8-e1a4cc905a53>:1: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
          Try using .loc[row_indexer,col_indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/i
          ndexing.html#returning-a-view-versus-a-copy
            box office['domestic gross'] = box office['domestic gross'].fillna(box office['domestic gros
          s'].median())
          <ipython-input-8-e1a4cc905a53>:3: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
          Try using .loc[row_indexer,col_indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/i
          ndexing.html#returning-a-view-versus-a-copy
            box_office['foreign_gross'] = box_office['foreign_gross'].fillna(box_office['foreign_gros
          s'].median())
 Out[8]:
                                               title
                                                       studio
                                                              domestic_gross foreign_gross
             0
                                          Toy Story 3
                                                          BV
                                                                 415000000.0
                                                                               652000000.0
                                                                                          2010
             1
                             Alice in Wonderland (2010)
                                                                  334200000.0
                                                                               691300000.0 2010
                                                                               664300000.0 2010
                Harry Potter and the Deathly Hallows Part 1
                                                          WB
                                                                 296000000.0
             3
                                           Inception
                                                                  292600000.0
                                                                               535700000.0 2010
                                                          WB
             4
                                   Shrek Forever After
                                                        P/DW
                                                                 238700000.0
                                                                               513900000.0 2010
                                                                                          2018
          3382
                                                                      6200.0
                                                                                19000000.0
                                          The Quake
                                                        Magn.
          3383
                             Edward II (2018 re-release)
                                                          FM
                                                                      4800.0
                                                                                19000000.0
                                                                                          2018
          3384
                                            El Pacto
                                                                      2500.0
                                                                                19000000.0 2018
                                                         Sony
          3385
                                           The Swan Synergetic
                                                                      2400.0
                                                                                19000000.0
                                                                                          2018
          3386
                                                                      1700.0
                                                                                19000000.0 2018
                                    An Actor Prepares
                                                         Grav
         3382 rows × 5 columns
 In [9]:
           box office.isnull().sum()
 Out[9]: title
                             0
          studio
                             0
          domestic_gross
                             0
          foreign_gross
                             0
          year
                             0
          dtype: int64
In [10]:
           box_office.duplicated().sum()
Out[10]: 0
In [11]:
           box_office['studio'].value_counts().head(15)
         IFC
                     166
Out[11]:
          Uni.
                     147
```

WB

Fox

Magn.

140

136

136

```
SPC
          123
           110
Sony
BV
           106
           103
LGF
           101
Par.
Eros
           89
Wein.
           77
            74
CL
            68
Strand
FoxS
            67
Name: studio, dtype: int64
```

```
In [12]: box_office['studio'].value_counts().tail(10)
```

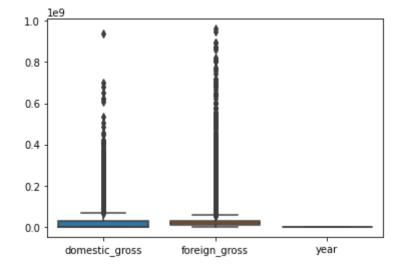
```
Out[12]: CP
                        1
          Gaatri
                        1
          SHO
                        1
          Libre
                        1
          Proud
                        1
          ELS
                        1
          Abk.
                        1
          CineGalaxy
                        1
          Trafalgar
          Truly
```

Name: studio, dtype: int64

USING SCATTER PLOT TO VISUALIZE OUTLIERS

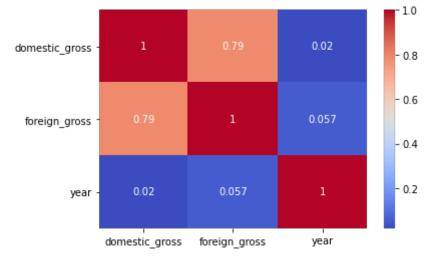
```
In [13]: sns.boxplot(data=box_office)
   #i decided not to drop the outliers as nothing is out proportional range
```

#### Out[13]: <AxesSubplot:>



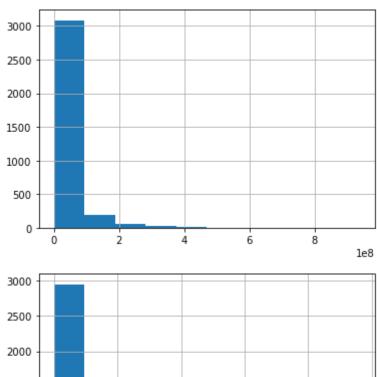
```
In [14]: correlation_matrix = box_office.corr()

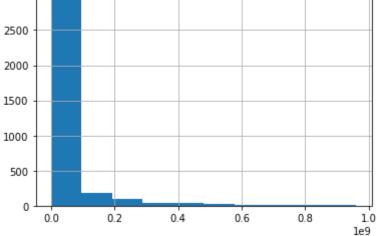
# Visualize correlation matrix
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
plt.show()
```

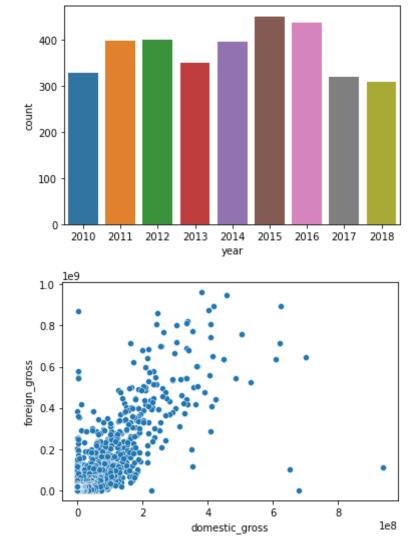


THERE IS A POSITIVE CORRELATION BETWEEN DOMESTIC AND FOREIGN GROSS AMOUNTS OF 0.79 THERE IS LITTLE TO NO CORRELATION BETWEEN YEAR AND

```
In [15]: box_office['domestic_gross'].hist()
    plt.show()
    box_office['foreign_gross'].hist()
    plt.show()
    sns.countplot(x='year', data=box_office)
    plt.show()
    sns.scatterplot(x='domestic_gross', y= 'foreign_gross', data=box_office)
    plt.show()
```







# FOR THIS DATA SET: CATEGORIZATION BY STUDIO 1. IFC STUDIO

TITLES BY STUDIO IFC

THERE ARE 166 TITLES BY IFC STUDIO

THE HIGHEST DOMESTIC GROSSING FILM BY THEM IS "BOYHOOD"

GROSSING AT: 25400000.0

PREMIERED IN 2014

ACCORDING TO THE BAR GRAPH 2014 WAS THE YEAR WITHT THE MOST DOMESTIC GROSS INCOME

THE HIGHEST FOREIGN GROSSING FILM IS "HEARTBREAKER"

GROSSING AT: 46900000.0

PREMIERED IN 2010

In [16]:

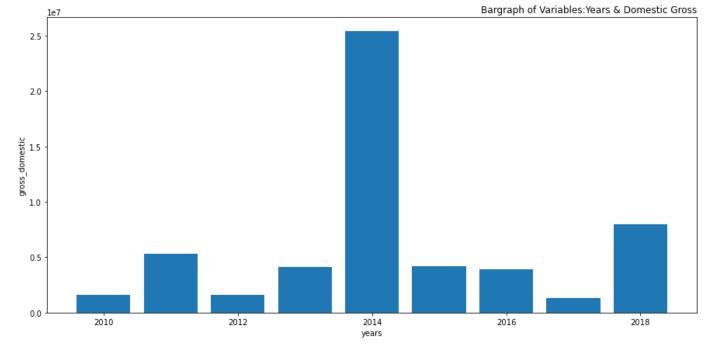
ACCORDING TO THE BAR GRAPH 2010 WAS THE YEAR WITH THE MOST FOREIGN GROSS INCOME

studio1 = box\_office.loc[box\_office['studio']=='IFC']
studio1

```
110
                             Heartbreaker
                                            IFC
                                                       504000.0
                                                                  46900000.0 2010
           112 The Good, the Bad, the Weird
                                            IFC
                                                       128000.0
                                                                  44100000.0 2010
           151
                              Soul Kitchen
                                            IFC
                                                       277000.0
                                                                  17600000.0 2010
                           Looking for Eric
                                            IFC
                                                        55800.0
                                                                  11500000.0 2010
           166
           190
                                            IFC
                                                                   5100000.0 2010
                                  Vincere
                                                       619000.0
                                            IFC
                                                                  19000000.0 2018
          3324
                             Ghost Stories
                                                       149000.0
          3335
                             Mary Shelley
                                            IFC
                                                       109000.0
                                                                  19000000.0 2018
                                                                  19000000.0 2018
          3344
                   The House That Jack Built
                                            IFC
                                                        0.00088
          3361
                               A Ciambra
                                            IFC
                                                        41900.0
                                                                  19000000.0 2018
          3374
                               The Escape
                                            IFC
                                                        14000.0
                                                                  19000000.0 2018
         166 rows × 5 columns
In [17]:
           highest_gross = studio1['domestic_gross'].max()
           print('the highest domestic grossing film by IFC studio is:', highest_gross)
          the highest domestic grossing film by IFC studio is: 25400000.0
           target = 25400000.0
In [18]:
           row = studio1[studio1['domestic_gross']== target]
           if not row.empty:
               film = row['title'].values[0]
               print(target, film)
               print('no',target)
          25400000.0 Boyhood
In [19]:
           target = 'Boyhood'
           row = studio1.loc[studio1['title']==target]
           if not row.empty:
               year = row['year'].values[0]
               print(target, year)
           else:
               print('no')
          Boyhood 2014
In [20]:
           type(studio1)
Out[20]: pandas.core.frame.DataFrame
In [21]:
           studio1 = pd.DataFrame(studio1)
           years = studio1['year']
           grossd = studio1['domestic_gross']
           grossf = studio1['foreign_gross']
           fig = plt.figure(figsize= (15,7))
           plt.bar(years, grossd)
           plt.xlabel ('years')
           plt.ylabel('gross_domestic')
           plt.title('Bargraph of Variables:Years & Domestic Gross', loc = "right")
           plt.show()
```

title studio domestic\_gross foreign\_gross

Out[16]:



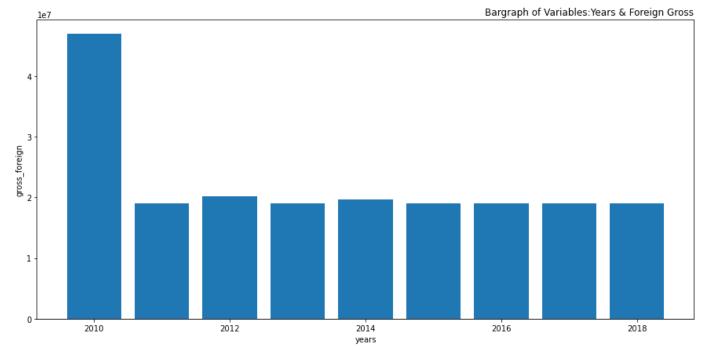
```
In [22]: studio1 = pd.DataFrame(studio1)

years = studio1['year']
grossd = studio1['domestic_gross']
grossf = studio1['foreign_gross']

fig = plt.figure(figsize= (15,7))

plt.bar(years, grossf)
plt.xlabel ('years')
plt.ylabel('gross_foreign')
plt.title('Bargraph of Variables:Years & Foreign Gross', loc = "right")

plt.show()
```



```
In [23]: highest_gross = studio1['foreign_gross'].max()
    print('the highest domestic grossing film by IFC studio is:', highest_gross)
```

the highest domestic grossing film by IFC studio is: 46900000.0

```
In [24]: target = 46900000.0
    row = studio1[studio1['foreign_gross']== target]
    if not row.empty:
        film = row['title'].values[0]
        print(target, film)
    else:
        print('no',target)
```

46900000.0 Heartbreaker

```
In [25]: target = 'Heartbreaker'
    row = studio1.loc[studio1['title']==target]
    if not row.empty:
        year = row['year'].values[0]
        print(target, year)
    else:
        print('no')
```

Heartbreaker 2010

### 2. UNIVERSAL STUDIOS

STUDIO UNIVERSAL

THERE ARE TITLES BY UNIVERSAL STUDIOS

THE HIGHEST DOMESTIC GROSSING FILM BY THEM IS "JURASSIC WORLD"

GROSSING AT: 652300000.0

PREMIERED IN 2015

ACCORDING TO THE BAR GRAPH 2015 WAS THE YEAR WITH THE MOST DOMESTIC GROSS INCOME

THE HIGHEST FOREIGN GROSSING FILM IS "JURASSIC WORLD: FALLEN KINGDOM"

GROSSING AT: 891800000.0

PREMIERED IN 2018

ACCORDING TO THE BAR GRAPH 2018 WAS THE YEAR WITH THE MOST FOREIGN GROSS INCOME

```
In [26]: studio2 = box_office.loc[box_office['studio']=='Uni.']
    studio2
```

Out[26]:	title	studio	domestic_gross	foreign_gross	year
8	Despicable Me	Uni.	251500000.0	291600000.0	2010
18	Robin Hood	Uni.	105300000.0	216400000.0	2010
20	Little Fockers	Uni.	148400000.0	162200000.0	2010
49	The Wolfman	Uni.	62000000.0	77800000.0	2010
66	Green Zone	Uni.	35100000.0	59800000.0	2010
		•••			
3148	Mortal Engines	Uni.	16000000.0	67700000.0	2018
3172	Breaking In (2018)	Uni.	46800000.0	4600000.0	2018
3219	Welcome to Marwen	Uni.	10800000.0	2100000.0	2018

```
title studio domestic_gross foreign_gross year
          3289 Schindler's List (2018 re-release)
                                             Uni.
                                                       833000.0
                                                                   19000000.0 2018
          3369
                               Loving Pablo
                                             Uni.
                                                        22000.0
                                                                  19000000.0 2018
         147 rows × 5 columns
          highest_gross = studio2['domestic_gross'].max()
In [27]:
          print('the highest domestic grossing film by Universal studio is:', highest_gross)
          the highest domestic grossing film by Universal studio is: 652300000.0
          target = 652300000.0
In [28]:
          row = studio2[studio2['domestic_gross']== target]
          if not row.empty:
               film = row['title'].values[0]
               print(target, film)
          else:
               print('no',target)
          652300000.0 Jurassic World
```

target = 'Jurassic World' In [29]:

```
row = studio2.loc[studio2['title']==target]
if not row.empty:
   year = row['year'].values[0]
   print(target, year)
   print('no')
```

Jurassic World 2015

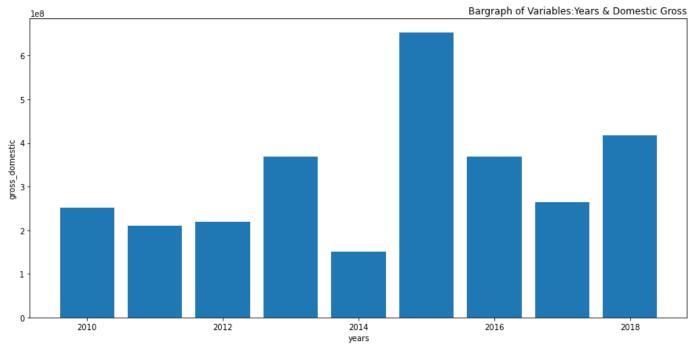
```
In [30]: studio2 = pd.DataFrame(studio2)

years = studio2['year']
grossd = studio2['domestic_gross']
grossf = studio2['foreign_gross']

fig = plt.figure(figsize= (15,7))

plt.bar(years, grossd)
plt.xlabel ('years')
plt.ylabel('gross_domestic')
plt.title('Bargraph of Variables:Years & Domestic Gross', loc = "right")

plt.show()
```



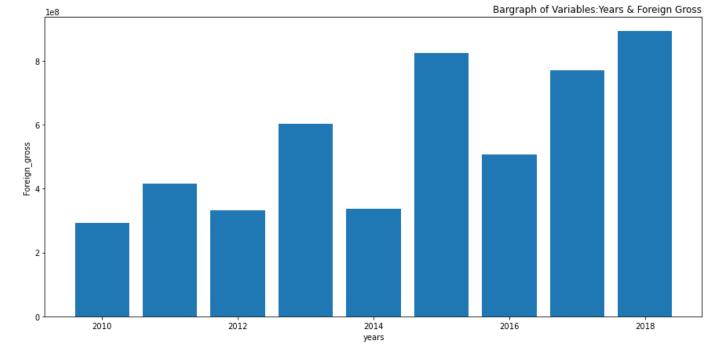
```
In [31]: studio2 = pd.DataFrame(studio2)

years = studio2['year']
grossd = studio2['domestic_gross']
grossf = studio2['foreign_gross']

fig = plt.figure(figsize= (15,7))

plt.bar(years, grossf)
plt.xlabel ('years')
plt.ylabel('Foreign_gross')
plt.ylabel('Foreign_gross')
plt.title('Bargraph of Variables:Years & Foreign Gross', loc = "right")

plt.show()
```



```
In [32]: highest_gross = studio2['foreign_gross'].max()
    print('the highest domestic grossing film by Universal studio is:', highest_gross)
```

the highest domestic grossing film by Universal studio is: 891800000.0

```
In [33]: target = 891800000
  row = studio2[studio2['foreign_gross']== target]
  if not row.empty:
     film = row['title'].values[0]
     print(target, film)
  else:
     print('no',target)
```

891800000 Jurassic World: Fallen Kingdom

```
In [34]: target = 'Jurassic World: Fallen Kingdom'
    row = studio2.loc[studio2['title']==target]
    if not row.empty:
        year = row['year'].values[0]
        print(target, year)
    else:
        print('no')
```

Jurassic World: Fallen Kingdom 2018

### 3. WARNER BROS

WARNER BROTHERS

THE HIGHEST DOMESTIC GROSSING FILM BY THEM IS "DARK NIGHT RISES"

GROSSING AT: 448100000.0

PREMIERED IN 2012

ACCORDING TO THE BAR GRAPH 2012 WAS THE YEAR WITH THE MOST DOMESTIC GROSS INCOME

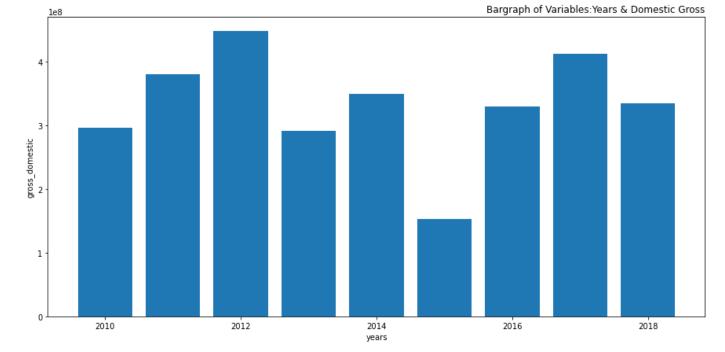
THE HIGHEST FOREIGN GROSSING FILM IS " Harry Potter and the Deathly Hallows Part 2"

GROSSING AT: 960500000.0

PREMIERED IN 2011

```
studio3 = box_office.loc[box_office['studio']=='WB']
In [35]:
           studio3
Out[35]:
                                                title studio domestic_gross foreign_gross
                                                                                        year
             2 Harry Potter and the Deathly Hallows Part 1
                                                               296000000.0
                                                                            664300000.0 2010
                                                       WB
             3
                                           Inception
                                                       WB
                                                               292600000.0
                                                                            535700000.0 2010
            10
                              Clash of the Titans (2010)
                                                               163200000.0
                                                                            330000000.0 2010
                                                       WB
            35
                                           Due Date
                                                       WB
                                                               100500000.0
                                                                            111200000.0 2010
            37
                                           Yogi Bear
                                                       WB
                                                               100200000.0
                                                                            101300000.0 2010
          3161
                                           12 Strong
                                                       WB
                                                                45800000.0
                                                                             21600000.0 2018
                                     The 15:17 to Paris
                                                                36300000.0
                                                                             20800000.0 2018
          3167
                                                       WB
          3170
                           Teen Titans Go! To The Movies
                                                       WB
                                                                29800000.0
                                                                             22300000.0 2018
          3209
                               They Shall Not Grow Old
                                                                18000000.0
                                                                             19000000.0 2018
                                                       WB
                  2001: A Space Odyssey (2018 re-release)
                                                                 3200000.0
                                                                             19000000.0 2018
          3264
                                                       WB
         140 rows × 5 columns
In [36]:
           highest_gross = studio3['domestic_gross'].max()
           print('the highest domestic grossing film by WarnerBros is:', highest_gross)
          the highest domestic grossing film by WarnerBros is: 448100000.0
In [37]:
           target = 448100000.0
           row = studio3[studio3['domestic_gross']== target]
           if not row.empty:
               film = row['title'].values[0]
               print(target, film)
           else:
               print('no',target)
          448100000.0 The Dark Knight Rises
In [38]:
           target = 'The Dark Knight Rises'
           row = studio3.loc[studio3['title']==target]
           if not row.empty:
               year = row['year'].values[0]
               print(target, year)
           else:
               print('no')
          The Dark Knight Rises 2012
In [39]:
          studio3 = pd.DataFrame(studio3)
           years = studio3['year']
           grossd = studio3['domestic_gross']
           grossf = studio3['foreign_gross']
           fig = plt.figure(figsize= (15,7))
           plt.bar(years, grossd)
           plt.xlabel ('years')
           plt.ylabel('gross_domestic')
           plt.title('Bargraph of Variables:Years & Domestic Gross', loc = "right")
```

plt.show()



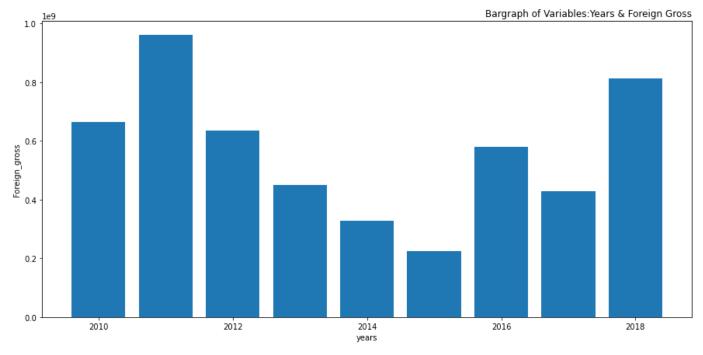
```
In [40]: studio3 = pd.DataFrame(studio3)

years = studio3['year']
grossd = studio3['domestic_gross']
grossf = studio3['foreign_gross']

fig = plt.figure(figsize= (15,7))

plt.bar(years, grossf)
plt.xlabel ('years')
plt.ylabel('Foreign_gross')
plt.ylabel('Foreign_gross')
plt.title('Bargraph of Variables:Years & Foreign Gross', loc = "right")

plt.show()
```



```
In [41]: highest_gross = studio3['foreign_gross'].max()
    print('the highest domestic grossing film by WarnerBros is:', highest_gross)
```

the highest domestic grossing film by WarnerBros is: 960500000.0

```
In [42]: target = 960500000.0
   row = studio3[studio3['foreign_gross']== target]
   if not row.empty:
       film = row['title'].values[0]
       print(target, film)
   else:
       print('no',target)
```

960500000.0 Harry Potter and the Deathly Hallows Part 2

```
In [43]: target = 'Harry Potter and the Deathly Hallows Part 2'
    row = studio3.loc[studio3['title']==target]
    if not row.empty:
        year = row['year'].values[0]
        print(target, year)
    else:
        print('no')
```

Harry Potter and the Deathly Hallows Part 2 2011

# OVERVIEW PLOT OF BOX OFFICE DATA FRAME

```
In [44]: #plotting the domestic gross over years using plt.bar

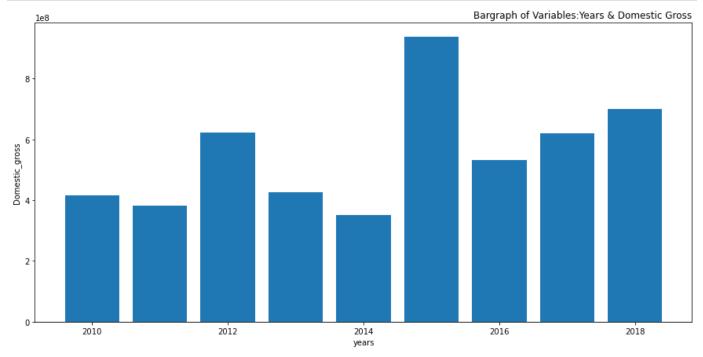
b = pd.DataFrame(box_office)

year= box_office['year']
   dgross = box_office['domestic_gross']
   fgross = box_office['foreign_gross']

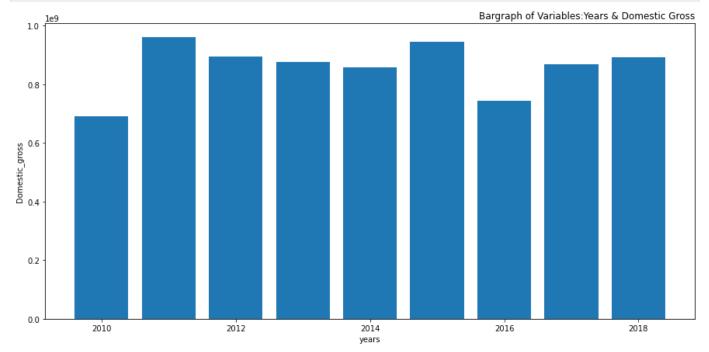
fig = plt.figure(figsize=(15,7))

plt.bar(year, dgross)
   plt.xlabel ('years')
   plt.ylabel('Domestic_gross')
   plt.title('Bargraph of Variables:Years & Domestic Gross', loc = "right")

plt.show()
```



overall in 2015 experienced the highest domestic grossing



overall in 2011 and 2015 experienced the highest domestic grossing

#### CONCLUSION

domestic gross and foreign gross are postively correlated at 0.79

i analysed the top 3 studios: IFC, WARNER BROS AND UNIVERSAL STUDIOS

WARNER BROS: "The Dark Knight Rises" was the highest domestic grossing film in 2012, contributing to the highest domestic gross income that year. On the other hand, "Harry Potter and the Deathly Hallows Part 2" was the highest foreign grossing film in 2011, contributing to the highest foreign gross income in that year.

UNIVERSAL STUDIOS "Jurassic World" and "Jurassic World: Fallen Kingdom" were the highest domestic and foreign grossing films, respectively, for Universal Studios. These films contributed to the years 2015 and 2018 being the years with the highest domestic and foreign gross income, according to the provided bar graph.

IFC STUDIOS "Boyhood" and "Heartbreaker" were the highest domestic and foreign grossing films, respectively, for IFC Studio. These films contributed to 2014 being the year with the highest domestic gross income and 2010 being the year with the highest foreign gross income, according to the provided bar graph

if time permitted i would comment better and use the budgets data to come up with profit margins that these films have and use it to analyse our expected commercial success

```
Out[46]:
                release_date
                                                             production_budget domestic_gross worldwide_gross
          0
             1
                Dec 18, 2009
                                                      Avatar
                                                                   $425,000,000
                                                                                  $760,507,625
                                                                                                $2,776,345,279
                              Pirates of the Caribbean: On Stranger
                    May 20,
          1
             2
                                                                   $410,600,000
                                                                                  $241,063,875
                                                                                                $1,045,663,875
                       2011
          2
             3
                  Jun 7, 2019
                                                 Dark Phoenix
                                                                   $350,000,000
                                                                                   $42,762,350
                                                                                                 $149,762,350
          3
             4
                                        Avengers: Age of Ultron
                 May 1, 2015
                                                                   $330,600,000
                                                                                  $459,005,868
                                                                                                $1,403,013,963
                                   Star Wars Ep. VIII: The Last Jedi
                Dec 15, 2017
                                                                   $317,000,000
                                                                                  $620,181,382
                                                                                                $1,316,721,747
In [47]:
          the_numbers.info()
           #brief overview of data
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5782 entries, 0 to 5781
          Data columns (total 6 columns):
               Column
                                   Non-Null Count
                                                    Dtype
               -----
               id
           0
                                   5782 non-null
                                                    int64
           1
               release_date
                                   5782 non-null
                                                    object
           2
                                   5782 non-null
               movie
                                                    object
           3
               production_budget 5782 non-null
                                                    object
           4
                                   5782 non-null
               domestic_gross
                                                    object
               worldwide gross
                                   5782 non-null
                                                    object
          dtypes: int64(1), object(5)
          memory usage: 271.2+ KB
In [48]:
           #transforming the object type to integer type
           the_numbers['production_budget'] = pd.to_numeric(the_numbers['production_budget'].str.replace('
           the_numbers['domestic_gross'] = pd.to_numeric(the_numbers['domestic_gross'].str.replace('[\$,]'
           the_numbers['worldwide_gross'] = pd.to_numeric(the_numbers['worldwide_gross'].str.replace('[\$,
In [49]:
           the_numbers.info()
           #confirming changes
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5782 entries, 0 to 5781
          Data columns (total 6 columns):
           #
               Column
                                   Non-Null Count
                                                    Dtype
           0
               id
                                   5782 non-null
                                                    int64
                                   5782 non-null
           1
               release_date
                                                    object
           2
                                   5782 non-null
               movie
                                                    object
               production_budget 5782 non-null
           3
                                                    int64
           4
               domestic_gross
                                   5782 non-null
                                                    int64
               worldwide_gross
                                   5782 non-null
                                                    int64
          dtypes: int64(4), object(2)
          memory usage: 271.2+ KB
In [50]:
          the_numbers['production_budget'] = the_numbers['production_budget'].astype(float)
           the numbers['domestic gross'] = the numbers['domestic gross'].astype(float)
           the_numbers['worldwide_gross'] = the_numbers['worldwide_gross'].astype(float)
           #changing numeric to float type
In [51]:
          the_numbers.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5782 entries, 0 to 5781
          Data columns (total 6 columns):
           #
               Column
                                   Non-Null Count Dtype
           0
               id
                                   5782 non-null
                                                    int64
                                   5782 non-null
           1
               release_date
                                                    object
           2
                                   5782 non-null
                                                    object
               movie
           3
               production_budget 5782 non-null
                                                    float64
           4
               domestic_gross
                                   5782 non-null
                                                    float64
               worldwide gross
                                   5782 non-null
                                                    float64
          dtypes: float64(3), int64(1), object(2)
```

memory usage: 271.2+ KB

In [52]: the\_numbers.describe()

#### Out[52]: id production\_budget domestic\_gross worldwide\_gross

		production_budget	uoest.e_g. ess	g. 655
count	5782.000000	5.782000e+03	5.782000e+03	5.782000e+03
mean	50.372363	3.158776e+07	4.187333e+07	9.148746e+07
std	28.821076	4.181208e+07	6.824060e+07	1.747200e+08
min	1.000000	1.100000e+03	0.000000e+00	0.000000e+00
25%	25.000000	5.000000e+06	1.429534e+06	4.125415e+06
50%	50.000000	1.700000e+07	1.722594e+07	2.798445e+07
75%	75.000000	4.000000e+07	5.234866e+07	9.764584e+07
max	100.000000	4.250000e+08	9.366622e+08	2.776345e+09

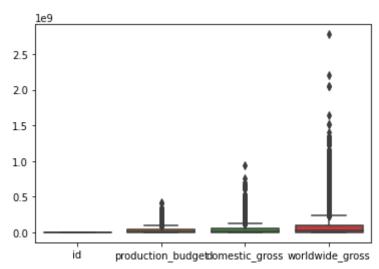
```
In [53]: the_numbers.isnull().sum()
#the_numbers.duplicated().sum()
```

```
Out[53]: id 0 release_date 0 movie 0 production_budget 0 domestic_gross 0 worldwide_gross 0 dtype: int64
```

In [54]: pd.set\_option('display.float\_format', lambda x: '%.3f' % x)
#this removes the scientific notation of putting numbers in powers of ten for readability purpo

In [55]: sns.boxplot(data=the\_numbers)
 #the worldwide gross is definetly higher than domestic gross

#### Out[55]: <AxesSubplot:>



```
In [56]: the_numbers['domestic_gross'].isnull().mean()
#there are no missing values to affect non null mean
```

Out[56]: 0.0

**DEALING WITH OUTLIERS** 

```
In [57]:
           # Calculate the mean of non-null values in the 'domestic_gross' column
           mean_domestic_gross = the_numbers['domestic_gross'][the_numbers['domestic_gross'] != 0].mean()
           # Replace zeros with the mean
           the_numbers['domestic_gross'] = the_numbers['domestic_gross'].replace(0, mean_domestic_gross)
           #why for some reason i was geting negative infinity in my data, so i replaced zeros with their
           # Calculate the mean of non-null values in the 'worldwide gross' column
In [58]:
           mean_worldwide_gross = the_numbers['worldwide_gross'][the_numbers['worldwide_gross'] != 0].mean_
           # Replace zeros with the mean
           the_numbers['worldwide_gross'] = the_numbers['worldwide_gross'].replace(0, mean_worldwide_gross
In [59]:
           the numbers
Out[59]:
                 id
                     release_date
                                                         movie production_budget
                                                                                   domestic_gross
                                                                                                  worldwide_gross
                     Dec 18, 2009
                                                         Avatar
                                                                     425000000.000
                                                                                     760507625.000
                                                                                                    2776345279.000
                         May 20,
                                       Pirates of the Caribbean: On
                  2
              1
                                                                     410600000.000
                                                                                     241063875.000
                                                                                                     1045663875.000
                            2011
                                                   Stranger Tides
                  3
                      Jun 7, 2019
                                                    Dark Phoenix
                                                                     350000000.000
                                                                                     42762350.000
                                                                                                     149762350.000
              2
              3
                      May 1, 2015
                                           Avengers: Age of Ultron
                                                                     330600000.000
                                                                                    459005868.000
                                                                                                     1403013963.000
                                                                     317000000.000
                                                                                    620181382.000
              4
                  5
                     Dec 15, 2017
                                     Star Wars Ep. VIII: The Last Jedi
                                                                                                    1316721747.000
                                                                                     46257465.790
          5777 78
                     Dec 31, 2018
                                                         Red 11
                                                                          7000.000
                                                                                                      97687996.115
          5778
                79
                      Apr 2, 1999
                                                       Following
                                                                          6000.000
                                                                                         48482.000
                                                                                                        240495.000
          5779
                80
                      Jul 13, 2005
                                     Return to the Land of Wonders
                                                                          5000.000
                                                                                          1338.000
                                                                                                          1338,000
          5780
                81
                     Sep 29, 2015
                                             A Plague So Pleasant
                                                                          1400.000
                                                                                      46257465.790
                                                                                                      97687996.115
          5781 82
                                              My Date With Drew
                                                                          1100.000
                                                                                       181041.000
                                                                                                        181041.000
                      Aug 5, 2005
          5782 rows × 6 columns
```

there are no null values so thers's nothing to be removed or filled

#### PROFIT MARGINS

CREATING TABLE DOMESTIC PROFIT MARGIN (D\_PROFIT\_MARGIN\_PERCENTAGE)

CREATING TABLE WORLDWIDE PROFIT MARGIN (W\_PROFIT\_MARGIN\_PERCENTAGE)

THESE COLUMNS WILL SHOW US WHICH FILM HAS THE HIGHEST PROFIT MARGIN PERCENTAGE: DOMESTIC AND WORLDWIDE

OPTIMAL = 40

In [61]:

```
profit_d = the_numbers['domestic_gross'] - the_numbers['production_budget']
In [60]:
          d_profit_margin_percentage = (profit_d / the_numbers['domestic_gross']) * 100
          the_numbers = the_numbers.assign(d_profit_margin_percentage=d_profit_margin_percentage)
          profit_w = the_numbers['worldwide_gross'] - the_numbers['production_budget']
          w_profit_margin_percentage = (profit_w / the_numbers['worldwide_gross']) * 100
          the_numbers = the_numbers.assign(w_profit_margin_percentage=w_profit_margin_percentage)
          the_numbers
```

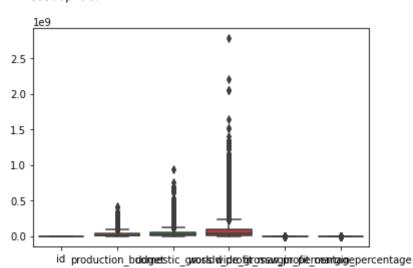
Out[61]: id release date movie production\_budget domestic\_gross worldwide\_gross d\_profit\_margin\_percenta

	id	release_date	movie	production_budget	domestic_gross	worldwide_gross	d_profit_margin_percenta
0	1	Dec 18, 2009	Avatar	425000000.000	760507625.000	2776345279.000	44.1
1	2	May 20, 2011	Pirates of the Caribbean: On Stranger Tides	410600000.000	241063875.000	1045663875.000	-70.3
2	3	Jun 7, 2019	Dark Phoenix	35000000.000	42762350.000	149762350.000	-718.4
3	4	May 1, 2015	Avengers: Age of Ultron	330600000.000	459005868.000	1403013963.000	27.9
4	5	Dec 15, 2017	Star Wars Ep. VIII: The Last Jedi	317000000.000	620181382.000	1316721747.000	48.8
•••							
5777	78	Dec 31, 2018	Red 11	7000.000	46257465.790	97687996.115	99.9
5778	79	Apr 2, 1999	Following	6000.000	48482.000	240495.000	87.6
5779	80	Jul 13, 2005	Return to the Land of Wonders	5000.000	1338.000	1338.000	-273.6
5780	81	Sep 29, 2015	A Plague So Pleasant	1400.000	46257465.790	97687996.115	99.9
5781	82	Aug 5, 2005	My Date With Drew	1100.000	181041.000	181041.000	99.3

5782 rows × 8 columns

#### In [62]: sns.boxplot(data=the\_numbers)

#### Out[62]: <AxesSubplot:>



In [63]:	<pre>the_numbers.describe()</pre>				
----------	-----------------------------------	--	--	--	--

Out[63]:		id	production_budget	domestic_gross	$worldwide\_gross$	d_profit_margin_percentage	w_profit_margi
	count	5782.000	5782.000	5782.000	5782.000	5782.000	
	mean	50.372	31587757.097	46257465.790	97687996.115	-3733.481	

	50%	50.000	17000000.000	27688002.000	38637682.000	16.375
	75%	75.000	4000000.000	52348661.500	97687996.115	65.470
	max	100.000	425000000.000	936662225.000	2776345279.000	99.997
In [64]:	the_n	umbers.isn	a().sum()			
Out[64]:	domest worldw d_prof	tion_budge ic_gross ide_gross it_margin_ it_margin_	0 0 0 t 0 0 percentage 0 percentage 0			
In [65]:	the_n	umbers['d_	profit_margin_p	ercentage'].de	scribe()	
Out[65]:	count mean std min 25% 50% 75% max Name:	-3733 64079 -3267873 -120 16 65	.646	e, dtype: floa	t64	

66881753.042

5609102.250

388.000

28.821

1.000

25.000

std

min

25%

41812076.827

5000000.000

1100.000

production\_budget domestic\_gross worldwide\_gross d\_profit\_margin\_percentage w\_profit\_margi

26.000

64079.646

-120.723

-3267873.856

173088700.211

8210838.250

# THIS SHOWS THAT THERE IS A NEGATIVE MEAN OF THE DOMESTIC PROFIT MARGIN (-3733)

THIS MEANS THAT THE STUDIOS HAVE A HIGH PROBABILITY OF MAKING LOSES DOMESTICALLY LETS SEE HOW THAT FAIRS WORLDWIDE

```
the_numbers['w_profit_margin_percentage'].describe()
In [66]:
                      5782.000
         count
Out[66]:
                     -2317.354
          mean
          std
                     54679.452
          min
                  -3846053.846
          25%
                       -29.252
          50%
                        52.118
          75%
                        78.911
                        99.999
         Name: w_profit_margin_percentage, dtype: float64
```

# THIS SHOWS THAT THERE IS A NEGATIVE MEAN OF THE FOREIGN PROFIT MARGIN (-2317)

THIS MEANS THAT THE STUDIOS HAVE A HIGH PROBABILITY OF MAKING LOSES INTERNATIONALLY MY CODE IS WRONG ON THIS MEAN THING