```
In [1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

In [2]: data = pd.read_csv("IMDB-Movie-Data.csv")
```

1. Display top 5 rows of dataset

In [60]:	data.head(5)										
Out[60]:	Rank	Title	Genre	Description	Director	Actors	Yea				
	0 1	Guardians of the Galaxy	Action,Adventure,Sci-Fi	A group of intergalactic criminals are forced	James Gunn	Chris Pratt, Vin Diesel, Bradley Cooper, Zoe S	201				
	1 2	Prometheus	Adventure, Mystery, Sci-Fi	Following clues to the origin of mankind, a te	Ridley Scott	Noomi Rapace, Logan Marshall- Green, Michael Fa	201				
	2 3	Split	Horror,Thriller	Three girls are kidnapped by a man with a diag	M. Night Shyamalan	James McAvoy, Anya Taylor-Joy, Haley Lu Richar	201				
	3 4	Sing	Animation, Comedy, Family	In a city of humanoid animals, a hustling thea	Christophe Lourdelet	Matthew McConaughey,Reese Witherspoon, Seth Ma	201				
	4 5	Suicide Squad	Action, Adventure, Fantasy	A secret government agency recruits some of th	David Ayer	Will Smith, Jared Leto, Margot Robbie, Viola D	201				
4							•				

2. Display last 5 rows of dataset

In [61]: data.tail(5)

ut[61]:		Rank	Title	Genre	Description	Director	Actors	Year	Ru (Mir
	995	996	Secret in Their Eyes	Crime,Drama,Mystery	A tight-knit team of rising investigators, alo	Billy Ray	Chiwetel Ejiofor, Nicole Kidman, Julia Roberts	2015	
	996	997	Hostel: Part II	Horror	Three American college students studying abroa	Eli Roth	Lauren German, Heather Matarazzo, Bijou Philli	2007	
	997	998	Step Up 2: The Streets	Drama, Music, Romance	Romantic sparks occur between two dance studen	Jon M. Chu	Robert Hoffman, Briana Evigan, Cassie Ventura,	2008	
	998	999	Search Party	Adventure,Comedy	A pair of friends embark on a mission to reuni	Scot Armstrong	Adam Pally, T.J. Miller, Thomas Middleditch,Sh	2014	
	999	1000	Nine Lives	Comedy, Family, Fantasy	A stuffy businessman finds himself trapped ins	Barry Sonnenfeld	Kevin Spacey, Jennifer Garner, Robbie Amell,Ch	2016	
									•

3. Find Shape of Our Dataset (Number of Rows And Number of Columns)

```
In [5]: data.shape
Out[5]: (1000, 12)
In [6]: print("No of Rows :",data.shape[0])
    print("No of Columns :",data.shape[1])
    No of Rows : 1000
    No of Columns : 12
```

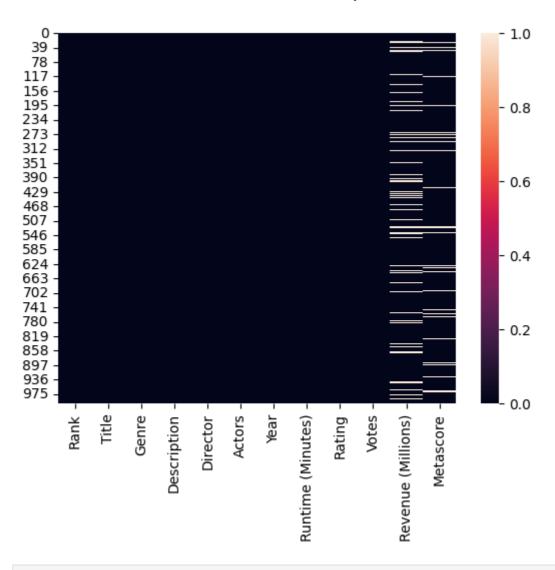
4. Getting Information About Our Dataset Like Total Number Rows, Total Number of Columns, Datatypes of Each Column And Memory Requirement

```
In [7]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 12 columns):
    Column
                      Non-Null Count Dtype
0
    Rank
                      1000 non-null
                                     int64
    Title
1
                     1000 non-null object
 2
    Genre
                     1000 non-null object
    Description
                     1000 non-null object
 3
 4
    Director
                     1000 non-null object
 5
                     1000 non-null object
  Actors
                     1000 non-null int64
 6
  Year
 7
    Runtime (Minutes) 1000 non-null int64
                     1000 non-null
 8
    Rating
                                     float64
9
    Votes
                     1000 non-null
                                     int64
10 Revenue (Millions) 872 non-null
                                     float64
11 Metascore
                      936 non-null
                                     float64
dtypes: float64(3), int64(4), object(5)
memory usage: 93.9+ KB
```

5. Check Missing Values In The Dataset

```
data.isnull().values.any()
 Out[8]: True
         data.isnull().value_counts('Metascore')
 In [9]:
 Out[9]: Metascore
         False
                   936
                    64
          True
         dtype: int64
In [10]: data.isnull().sum()
Out[10]: Rank
                                  0
         Title
                                  0
         Genre
         Description
         Director
                                  0
         Actors
         Year
                                  0
         Runtime (Minutes)
                                  0
                                  0
         Rating
         Votes
         Revenue (Millions)
                                128
         Metascore
                                 64
         dtype: int64
In [11]: sns.heatmap(data.isnull())
Out[11]: <AxesSubplot: >
```



```
In [12]:
         data.isnull().sum() * 100 / len(data)
Out[12]:
         Rank
                                  0.0
          Title
                                  0.0
          Genre
                                  0.0
          Description
                                  0.0
         Director
                                  0.0
          Actors
                                  0.0
                                  0.0
          Year
          Runtime (Minutes)
                                  0.0
          Rating
                                  0.0
          Votes
                                  0.0
          Revenue (Millions)
                                 12.8
          Metascore
                                  6.4
          dtype: float64
```

6. Drop All The Missing Values

```
In [13]: df = data.dropna(axis = 0)
In [14]: df.isnull().sum()
```

```
Out[14]: Rank
                                  0
          Title
                                  0
          Genre
                                  0
          Description
                                 0
          Director
                                 0
                                 0
          Actors
          Year
          Runtime (Minutes)
          Rating
          Votes
          Revenue (Millions)
                                  0
          Metascore
          dtype: int64
```

In [15]: df.isnull().values.any()

Out[15]: False

7. Check For Duplicate Data

```
In [16]: data.duplicated().sum()
Out[16]: 0
In [17]: data.duplicated().values.any()
Out[17]: False
In [18]: df1 = data.drop_duplicates()
```

8. Get Overall Statistics About The DataFrame

```
data.describe()
In [19]:
Out[19]:
                                                 Runtime
                                                                                         Revenue
                         Rank
                                                                                Votes
                                       Year
                                                                Rating
                                                                                                    Metasco
                                                (Minutes)
                                                                                        (Millions)
           count 1000.000000
                                              1000.000000
                                                                                                   936.0000
                                1000.000000
                                                           1000.000000
                                                                        1.000000e+03
                                                                                       872.000000
           mean
                    500.500000
                                2012.783000
                                               113.172000
                                                              6.723200
                                                                        1.698083e+05
                                                                                        82.956376
                                                                                                    58.9850
                    288.819436
                                   3.205962
                                                18.810908
                                                              0.945429
                                                                        1.887626e+05
                                                                                       103.253540
                                                                                                    17.1947
              std
             min
                      1.000000
                                2006.000000
                                                66.000000
                                                              1.900000
                                                                        6.100000e+01
                                                                                         0.000000
                                                                                                    11.0000
             25%
                    250.750000
                                2010.000000
                                               100.000000
                                                              6.200000
                                                                        3.630900e+04
                                                                                        13.270000
                                                                                                    47.0000
             50%
                    500.500000
                                2014.000000
                                               111.000000
                                                              6.800000
                                                                       1.107990e+05
                                                                                        47.985000
                                                                                                    59.5000
             75%
                    750.250000
                                2016.000000
                                               123.000000
                                                              7.400000
                                                                        2.399098e+05
                                                                                      113.715000
                                                                                                    72.0000
                  1000.000000
                                2016.000000
                                               191.000000
                                                              9.000000
                                                                       1.791916e+06
                                                                                       936.630000
                                                                                                   100.0000
```

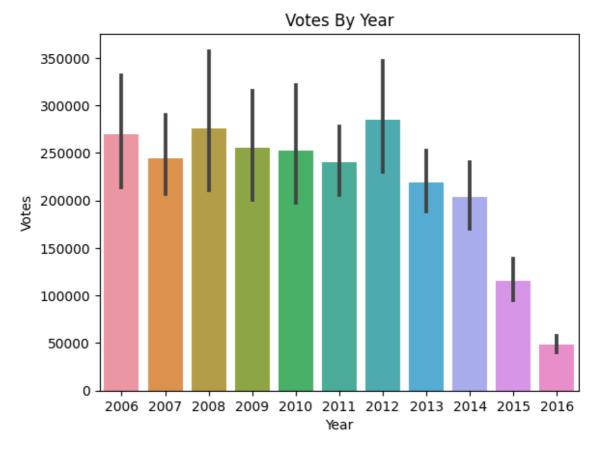
Count 1000.000000 10000 10000 10000 10000 10000 10000 10000 10000 10000 10000 10000 10000 1	In [20]:	<pre>data.describe(include = 'all')</pre>							
top NaN 999 207 1000 644 996 NaN top NaN The Host Action,Adventure,Sci-Fi Intergalactic criminals are forced Ridley Scott lutherson, Liam Hemsw NaN freq NaN 2 50 1 8 2 NaN mean 500.500000 NaN NaN NaN NaN NaN NaN NaN 2012.783000 std 288.819436 NaN NaN NaN NaN NaN NaN NaN NaN NaN 2016.000000 25% 250.750000 NaN 2016.000000 50% 500.500000 NaN NaN </th <th>Out[20]:</th> <th></th> <th>Rank</th> <th>Title</th> <th>Genre</th> <th>Description</th> <th>Director</th> <th>Actors</th> <th>Year</th>	Out[20]:		Rank	Title	Genre	Description	Director	Actors	Year
top NaN The Host Action,Adventure,Sci-Fi A group of intergalactic criminals are forced. Ridley Scott Jennifer Lawrence, Josh Putcherson, Liam Hemsw NaN freq NaN 2 50 1 8 2 NaN mean 500.500000 NaN NaN NaN NaN NaN NaN NaN 2012.783000 std 288.819436 NaN NaN NaN NaN NaN NaN NaN 3.205962 min 1.000000 NaN NaN NaN NaN NaN NaN NaN 2010.000000 206.000000 25% 250.750000 NaN 2014.000000 2014.000000 2014.000000 2014.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000 2016.000000		count	1000.000000	1000	1000	1000	1000	1000	1000.000000
top NaN The Host Action,Adventure,Sci-Flow A group of intergalactic criminals are forced Ridley Scott Lawrence, Hutcherson, Liam Hemsw NaN freq NaN 2 50 1 8 2 NaN mean 500.500000 NaN NaN NaN NaN NaN NaN 2012.78300C std 288.819436 NaN NaN NaN NaN NaN NaN NaN 3.205962 min 1.000000 NaN NaN NaN NaN NaN NaN 2010.00000 25% 250.750000 NaN NaN NaN NaN NaN NaN NaN NaN NaN 2014.00000 50% 500.50000 NaN 2016.00000 75% 750.250000 NaN 2016.00000 <th rowspan="3"></th> <th>unique</th> <th>NaN</th> <th>999</th> <th>207</th> <th>1000</th> <th>644</th> <th>996</th> <th>NaN</th>		unique	NaN	999	207	1000	644	996	NaN
mean 500.500000 NaN NaN NaN NaN 2012.78300C std 288.819436 NaN NaN NaN NaN NaN NaN 3.205962 min 1.000000 NaN NaN NaN NaN NaN 2006.00000C 25% 250.750000 NaN NaN NaN NaN NaN NaN 2014.00000C 50% 500.500000 NaN NaN NaN NaN NaN NaN 2016.00000C 75% 750.250000 NaN NaN NaN NaN NaN NaN NaN NaN 2016.00000C		top	NaN			intergalactic criminals are forced	-	Lawrence, Josh Hutcherson, Liam	NaN
std 288.819436 NaN NaN NaN NaN NaN 3.205962 min 1.000000 NaN NaN NaN NaN NaN 2006.000000 25% 250.750000 NaN NaN NaN NaN NaN 2010.000000 50% 500.500000 NaN NaN NaN NaN NaN NaN 2014.000000 75% 750.250000 NaN NaN NaN NaN NaN NaN NaN NaN NaN 2016.000000		freq	NaN	2	50	1	8	2	NaN
min 1.000000 NaN NaN NaN NaN 2006.000000 25% 250.750000 NaN NaN NaN NaN NaN 2010.000000 50% 500.500000 NaN NaN NaN NaN NaN 2014.000000 75% 750.250000 NaN NaN NaN NaN NaN NaN NaN 2016.000000 max 1000.000000 NaN NaN NaN NaN NaN NaN 2016.000000		mean	500.500000	NaN	NaN	NaN	NaN	NaN	2012.783000
25% 250.750000 NaN NaN NaN NaN 2010.000000 50% 500.500000 NaN NaN NaN NaN NaN 2014.000000 75% 750.250000 NaN NaN NaN NaN NaN NaN 2016.000000 max 1000.000000 NaN NaN NaN NaN NaN 2016.000000		std	288.819436	NaN	NaN	NaN	NaN	NaN	3.205962
50% 500.500000 NaN NaN NaN NaN 2014.000000 75% 750.250000 NaN NaN NaN NaN NaN 2016.000000 max 1000.000000 NaN NaN NaN NaN NaN 2016.000000		min	1.000000	NaN	NaN	NaN	NaN	NaN	2006.000000
75% 750.250000 NaN NaN NaN NaN NaN NaN 2016.0000000 max 1000.000000 NaN NaN NaN NaN NaN NaN 2016.0000000		25%	250.750000	NaN	NaN	NaN	NaN	NaN	2010.000000
max 1000.000000 NaN NaN NaN NaN NaN 2016.000000		50%	500.500000	NaN	NaN	NaN	NaN	NaN	2014.000000
		75%	750.250000	NaN	NaN	NaN	NaN	NaN	2016.000000
	4	max	1000.000000	NaN	NaN	NaN	NaN	NaN	

9. Display Title of The Movie Having Runtime Greater Than or equal to 180 Minutes

10. In Which Year There Was The Highest Average Voting?

```
In [22]: data.groupby('Year')['Votes'].mean().sort_values(ascending = False)
```

```
Out[22]: Year
          2012
                  285226.093750
          2008
                  275505.384615
          2006
                  269289.954545
          2009
                  255780.647059
          2010
                  252782.316667
          2007
                  244331.037736
          2011
                  240790.301587
          2013
                  219049.648352
          2014
                  203930.224490
          2015
                  115726.220472
          2016
                   48591.754209
          Name: Votes, dtype: float64
In [23]:
         sns.barplot(data=data, x='Year',y='Votes')
          plt.title('Votes By Year')
          plt.show()
```

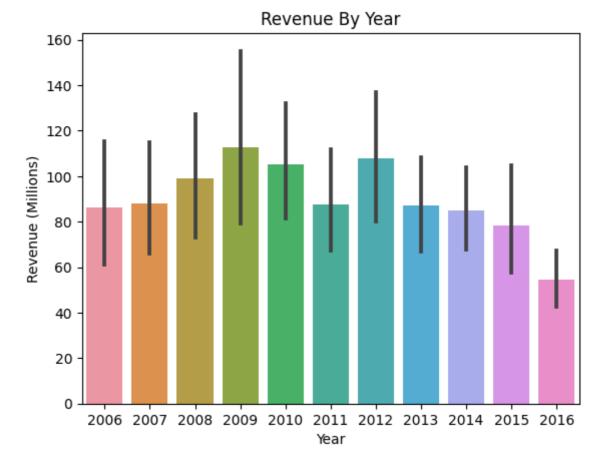


11. In Which Year There Was The Highest Average Revenue?

```
In [24]: data.groupby('Year')['Revenue (Millions)'].mean().sort_values(ascending = False)
```

```
Out[24]: Year
          2009
                  112.601277
                  107.973281
          2012
          2010
                  105.081579
          2008
                   99.082745
          2007
                   87.882245
          2011
                   87.612258
          2013
                   87.121818
          2006
                   86.296667
          2014
                   85.078723
          2015
                   78.355044
          2016
                   54.690976
          Name: Revenue (Millions), dtype: float64
          sns.barplot(data=data, x='Year',y='Revenue (Millions)')
In [25]:
          plt.title('Revenue By Year')
          plt.show()
```





12. Find The Average Rating For Each **Director**

```
data.groupby('Director')['Rating'].mean().sort_values(ascending = False)
```

```
Out[26]: Director
         Nitesh Tiwari
                              8.80
         Christopher Nolan
                              8.68
         Olivier Nakache
                              8.60
         Makoto Shinkai
                              8.60
         Aamir Khan
                              8.50
                              . . .
         Micheal Bafaro
                              3.50
         Jonathan Holbrook
                              3.20
         Shawn Burkett
                             2.70
         James Wong
                              2.70
         Jason Friedberg
                              1.90
         Name: Rating, Length: 644, dtype: float64
```

13. Display Top 10 Lengthy Movies Title and Runtime

```
df.columns
In [27]:
Out[27]: Index(['Rank', 'Title', 'Genre', 'Description', 'Director', 'Actors', 'Year',
                  'Runtime (Minutes)', 'Rating', 'Votes', 'Revenue (Millions)',
                  'Metascore'],
                 dtype='object')
In [28]: dataTop10 = df.sort_values('Runtime (Minutes)', ascending = False)
          dataTop10.head(10)[['Title','Runtime (Minutes)']].set_index('Title')
Out[28]:
                                               Runtime (Minutes)
                                         Title
                              The Hateful Eight
                                                            187
                         The Wolf of Wall Street
                                                            180
                                 La vie d'Adèle
                                                            180
                                   Cloud Atlas
                                                            172
                                       3 Idiots
                                                            170
          Pirates of the Caribbean: At World's End
                                                            169
                                    Interstellar
                                                            169
              The Hobbit: An Unexpected Journey
                                                            169
             The Curious Case of Benjamin Button
                                                            166
                  Transformers: Age of Extinction
                                                            165
In [29]:
          top10_len = data.nlargest(10,'Runtime (Minutes)')[['Title','Runtime (Minutes)']]
          .set index('Title')
```

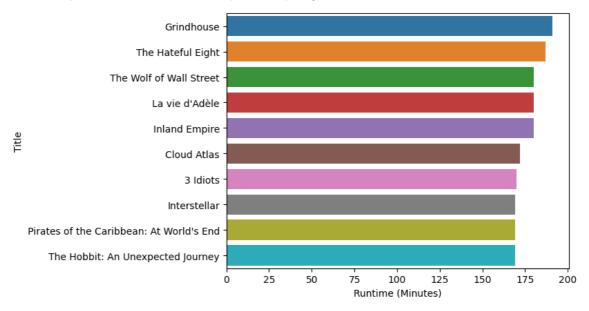
```
top10_len
```

Out[29]: Runtime (Minutes)

Title	
Grindhouse	191
The Hateful Eight	187
The Wolf of Wall Street	180
La vie d'Adèle	180
Inland Empire	180
Cloud Atlas	172
3 Idiots	170
Interstellar	169
Pirates of the Caribbean: At World's End	169
The Hobbit: An Unexpected Journey	169

In [30]: sns.barplot(data=top10_len, x='Runtime (Minutes)', y=top10_len.index)

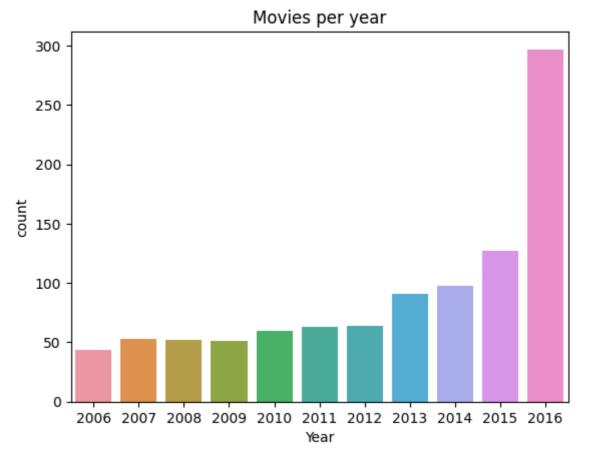
Out[30]: <AxesSubplot: xlabel='Runtime (Minutes)', ylabel='Title'>



14. Display Number of Movies Per Year

```
In [31]: df_MoviesPerYear = data.groupby('Year')['Rank'].count()
    df_MoviesPerYear
```

```
Out[31]: Year
          2006
                    44
                    53
          2007
                    52
          2008
          2009
                    51
          2010
                    60
          2011
                    63
          2012
                    64
          2013
                    91
          2014
                   98
          2015
                  127
          2016
                  297
          Name: Rank, dtype: int64
          data['Year'].value_counts()
In [32]:
Out[32]:
          2016
                  297
          2015
                  127
          2014
                   98
          2013
                    91
          2012
                    64
          2011
                    63
                    60
          2010
          2007
                    53
          2008
                    52
          2009
                    51
                    44
          2006
          Name: Year, dtype: int64
          sns.countplot(data=data, x='Year')
In [33]:
          plt.title('Movies per year')
          plt.show()
```



15. Find Most Popular Movie Title (Highest Revenue)

			ad(2)						
ut[34]:	R	Rank	Title	Genre	Description	Director	Actors	Year	Runtime (Minutes)
	0	1	Guardians of the Galaxy	Action,Adventure,Sci- Fi	A group of intergalactic criminals are forced	James Gunn	Chris Pratt, Vin Diesel, Bradley Cooper, Zoe S	2014	121
	1	2	Prometheus	Adventure,Mystery,Sci- Fi	Following clues to the origin of mankind, a te	Ridley Scott	Noomi Rapace, Logan Marshall- Green, Michael Fa	2012	124
n [35]:	data	a.nla	argest(5,'R	evenue (Millions)')	[['Title',	Revenue	(Millions	s)']]	
ut[35]:									
				Title	Revenue (Mil	lions)			
	50	Star V	Wars: Episode \		Revenue (Mil	lions) 936.63			
	50 87	Star V	Wars: Episode \	VII - The Force Awakens Avatar	g				
		Star V	Wars: Episode \	VII - The Force Awakens	9	936.63			
	87	Star V	Wars: Episode \	VII - The Force Awakens Avatar	5 7	936.63 760.51			
	87 85	Star V	Wars: Episode \	VII - The Force Awakens Avatar Jurassic World	9 7 6	936.63 760.51 552.18			
n [36]:	87 85 76 54			VII - The Force Awakens Avatar Jurassic World The Avengers	9 7 6 9	936.63 760.51 552.18 523.28 533.32	(Millions)']][['	['Title',
n [36]: ut[36]:	87 85 76 54			VII - The Force Awakens Avatar Jurassic World The Avengers The Dark Knight (Millions)'].max()	9 7 6 9	936.63 760.51 552.18 523.28 533.32	(Millions))][['(['Title',
	87 85 76 54 data	a[dat	ca['Revenue	VII - The Force Awakens Avatar Jurassic World The Avengers The Dark Knight (Millions)'].max()	5 == data['F Revenue (Mil	936.63 760.51 552.18 523.28 533.32	(Millions))']][['(['Title',

Out[37]: Revenue (Millions)

Title

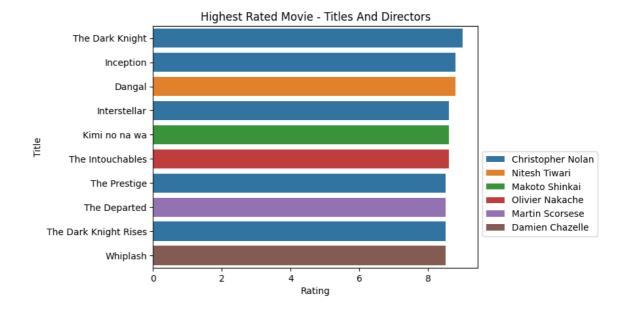
Title	
A Kind of Murder	0.00
Dead Awake	0.01
Wakefield	0.01
Lovesong	0.01
Love, Rosie	0.01

16. Display Top 10 Highest Rated Movie Titles And its Directors

```
In [38]: top10_highrated_mvi = data.nlargest(10,'Rating')[['Title','Rating','Director']].
top10_highrated_mvi
```

Out[38]: Rating **Director Title The Dark Knight** 9.0 Christopher Nolan Inception Christopher Nolan Nitesh Tiwari Dangal 8.8 8.6 Christopher Nolan Interstellar Kimi no na wa 8.6 Makoto Shinkai The Intouchables Olivier Nakache 8.6 The Prestige Christopher Nolan 8.5 The Departed Martin Scorsese 8.5 The Dark Knight Rises 8.5 Christopher Nolan Whiplash 8.5 Damien Chazelle

```
In [39]: sns.barplot(data = top10_highrated_mvi, x='Rating', y=top10_highrated_mvi.index,
    plt.title('Highest Rated Movie - Titles And Directors')
    plt.legend(bbox_to_anchor=[1, 0.5], loc=2)
    plt.show()
```



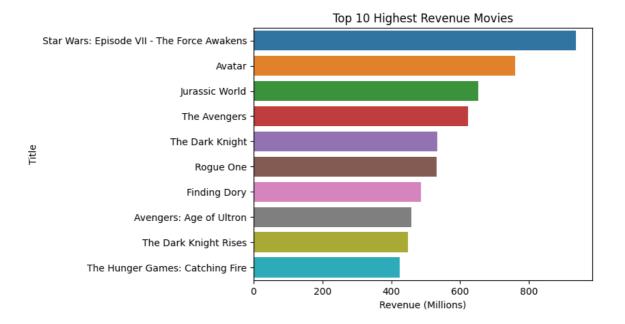
17. Display Top 10 Highest Revenue Movie Titles

In [40]: top10HighRevenue_mvi = data.nlargest(10,'Revenue (Millions)')[['Title','Revenue top10HighRevenue_mvi

Out[40]: Revenue (Millions)

Title	
Star Wars: Episode VII - The Force Awakens	936.63
Avatar	760.51
Jurassic World	652.18
The Avengers	623.28
The Dark Knight	533.32
Rogue One	532.17
Finding Dory	486.29
Avengers: Age of Ultron	458.99
The Dark Knight Rises	448.13
The Hunger Games: Catching Fire	424.65

```
In [41]: sns.barplot(data = top10HighRevenue_mvi, x='Revenue (Millions)' , y = top10HighR
    plt.title('Top 10 Highest Revenue Movies')
    plt.show()
```



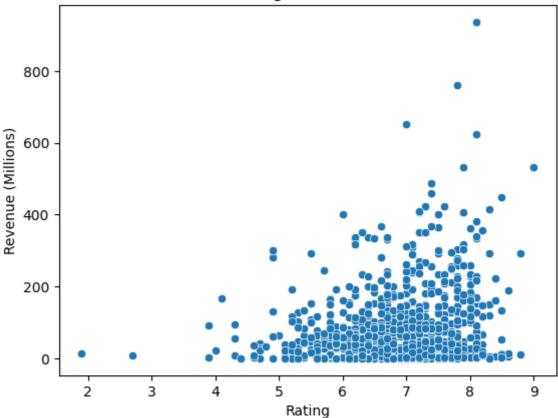
18. Find Average Rating of Movies Year Wise

```
data.groupby('Year')['Rating'].mean().sort_values(ascending = False)
Out[42]: Year
          2007
                  7.133962
                  7.125000
          2006
          2009
                  6.960784
                  6.925000
          2012
          2011
                  6.838095
          2014
                  6.837755
          2010
                  6.826667
          2013
                  6.812088
          2008
                  6.784615
          2015
                  6.602362
          2016
                  6.436700
          Name: Rating, dtype: float64
```

19. Does Rating Affect The Revenue?

```
In [43]: sns.scatterplot(data = data, x='Rating', y= 'Revenue (Millions)' )
   plt.title('Rating vs Revenue')
   plt.show()
```





Yes, Rating directly affects the Revenue.

20. Classify Movies Based on Ratings [Excellent, Good, and Average]

```
In [44]: def rating(rating):
    if rating >= 7.0:
        return 'Excellent'
    elif rating >= 6.0:
        return 'Good'
    else:
        return 'Average'
In [45]: data['Rating_Cat'] = data['Rating'].apply(rating)
data.head(10)
```

Out[45]:

	Rank	Title	Genre	Description	Director	Actors	γ
0	1	Guardians of the Galaxy	Action, Adventure, Sci-Fi	A group of intergalactic criminals are forced	James Gunn	Chris Pratt, Vin Diesel, Bradley Cooper, Zoe S	2
1	2	Prometheus	Adventure,Mystery,Sci-Fi	Following clues to the origin of mankind, a te	Ridley Scott	Noomi Rapace, Logan Marshall- Green, Michael Fa	2
2	3	Split	Horror,Thriller	Three girls are kidnapped by a man with a diag	M. Night Shyamalan	James McAvoy, Anya Taylor-Joy, Haley Lu Richar	2
3	4	Sing	Animation, Comedy, Family	In a city of humanoid animals, a hustling thea	Christophe Lourdelet	Matthew McConaughey,Reese Witherspoon, Seth Ma	2
4	5	Suicide Squad	Action, Adventure, Fantasy	A secret government agency recruits some of th	David Ayer	Will Smith, Jared Leto, Margot Robbie, Viola D	2
5	6	The Great Wall	Action, Adventure, Fantasy	European mercenaries searching for black powde	Yimou Zhang	Matt Damon, Tian Jing, Willem Dafoe, Andy Lau	2
6	7	La La Land	Comedy, Drama, Music	A jazz pianist falls for an aspiring actress i	Damien Chazelle	Ryan Gosling, Emma Stone, Rosemarie DeWitt, J	2
7	8	Mindhorn	Comedy	A has-been actor best known for playing the ti	Sean Foley	Essie Davis, Andrea Riseborough, Julian Barrat	2
8	9	The Lost City of Z	Action, Adventure, Biography	A true-life drama, centering on British explor	James Gray	Charlie Hunnam, Robert Pattinson, Sienna Mille	2
9	10	Passengers	Adventure, Drama, Romance	A spacecraft traveling to a distant colony pla	Morten Tyldum	Jennifer Lawrence, Chris Pratt, Michael Sheen,	2

21. Count Number of Action Movies

```
In [46]: len(data[data['Genre'].str.contains('Action', case = False)])
Out[46]: 303
```

22. Find Unique Values From Genre

```
In [47]: data['Genre']
Out[47]: 0
                  Action, Adventure, Sci-Fi
                 Adventure, Mystery, Sci-Fi
          2
                          Horror, Thriller
          3
                  Animation, Comedy, Family
                 Action, Adventure, Fantasy
          995
                      Crime, Drama, Mystery
          996
                                    Horror
          997
                      Drama, Music, Romance
          998
                         Adventure, Comedy
          999
                    Comedy, Family, Fantasy
          Name: Genre, Length: 1000, dtype: object
In [56]: list1 = []
          for value in data['Genre']:
              list1.append(value.split(','))
In [57]: list_d =[]
          for item in list1:
              for item1 in item:
                  list_d.append(item1)
In [58]: uni_list = []
          for item in list d:
              if item not in uni_list:
                  uni_list.append(item)
In [59]: uni_list
```

```
Out[59]: ['Action',
            'Adventure',
            'Sci-Fi',
            'Mystery',
           'Horror',
            'Thriller',
           'Animation',
           'Comedy',
           'Family',
           'Fantasy',
           'Drama',
           'Music',
           'Biography',
           'Romance',
           'History',
           'Crime',
            'Western',
            'War',
           'Musical',
           'Sport']
```

23. How Many Films of Each Genre Were Made?

```
from collections import Counter
In [55]: Counter(list_d)
Out[55]: Counter({'Action': 303,
                    'Adventure': 259,
                   'Sci-Fi': 120,
                   'Mystery': 106,
                   'Horror': 119,
                    'Thriller': 195,
                   'Animation': 49,
                   'Comedy': 279,
                   'Family': 51,
                   'Fantasy': 101,
                   'Drama': 513,
                   'Music': 16,
                   'Biography': 81,
                   'Romance': 141,
                   'History': 29,
                   'Crime': 150,
                    'Western': 7,
                   'War': 13,
                   'Musical': 5,
                    'Sport': 18})
 In [ ]:
```