In [1]:

```
# importing data to see highest domestic gross in 5 years
import pandas as pd
df= pd.read_csv('bom.movie_gross.csv.gz')
df.head()
df.drop(['foreign_gross','studio'], axis = 1, inplace=True)
df
```

Out[1]:

	title	domestic_gross	year
0	Toy Story 3	415000000.0	2010
1	Alice in Wonderland (2010)	334200000.0	2010
2	Harry Potter and the Deathly Hallows Part 1	296000000.0	2010
3	Inception	292600000.0	2010
4	Shrek Forever After	238700000.0	2010
3382	The Quake	6200.0	2018
3383	Edward II (2018 re-release)	4800.0	2018
3384	El Pacto	2500.0	2018
3385	The Swan	2400.0	2018
3386	An Actor Prepares	1700.0	2018

3387 rows × 3 columns

In [2]: ▶

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3387 entries, 0 to 3386
Data columns (total 3 columns):
#
    Column
                    Non-Null Count Dtype
    ----
                    -----
                                    object
    title
                    3387 non-null
0
 1
    domestic_gross 3359 non-null
                                    float64
 2
                    3387 non-null
                                    int64
    year
dtypes: float64(1), int64(1), object(1)
memory usage: 79.5+ KB
```

In [3]: ▶

```
# sorting data in scending year to see the latest year

dff=df.sort_values(by=['year'], ascending = False)
dff
```

Out[3]:

	title	domestic_gross	year
3386	An Actor Prepares	1700.0	2018
3183	On the Basis of Sex	24600000.0	2018
3176	Tyler Perry's Acrimony	43500000.0	2018
3177	Mary Queen of Scots	16500000.0	2018
3178	The Possession of Hannah Grace	14800000.0	2018
220	After.Life	109000.0	2010
221	Cairo Time	1600000.0	2010
222	Flipped	1800000.0	2010
223	Guzaarish	1000000.0	2010
0	Toy Story 3	415000000.0	2010

3387 rows × 3 columns

In [4]:

Out[4]:

	title	domestic_gross	year
3386	An Actor Prepares	1700.0	2018
3183	On the Basis of Sex	24600000.0	2018
3176	Tyler Perry's Acrimony	43500000.0	2018
3177	Mary Queen of Scots	16500000.0	2018
3178	The Possession of Hannah Grace	14800000.0	2018
1619	Yves Saint Laurent	724000.0	2014
1618	Bears	17800000.0	2014
1621	Like Father, Like Son (2014)	335000.0	2014
1622	Left Behind (2014)	14000000.0	2014
1623	Legends of Oz: Dorothy's Return	8500000.0	2014

1910 rows × 3 columns

In [5]: ▶

```
four_yrs.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1910 entries, 3386 to 1623
Data columns (total 3 columns):
#
     Column
                     Non-Null Count Dtype
                                     object
0
     title
                     1910 non-null
 1
     domestic_gross 1901 non-null
                                     float64
 2
     year
                     1910 non-null
                                     int64
dtypes: float64(1), int64(1), object(1)
memory usage: 59.7+ KB
```

In [6]: ▶

```
# top performing movies in 4 years
top = four_yrs.sort_values(by=['domestic_gross'], ascending = False)
top
```

Out[6]:

	title	domestic_gross	year
1872	Star Wars: The Force Awakens	936700000.0	2015
3080	Black Panther	700100000.0	2018
3079	Avengers: Infinity War	678800000.0	2018
1873	Jurassic World	652300000.0	2015
2758	Star Wars: The Last Jedi	620200000.0	2017
1975	Surprise - Journey To The West	NaN	2015
1739	Lila Lila	NaN	2014
1659	Jessabelle	NaN	2014
1685	Jack and the Cuckoo-Clock Heart	NaN	2014
1681	14 Blades	NaN	2014

1910 rows × 3 columns

In [7]: ▶

```
# slicing top 20 performing movies thorugh domestic gross collection
top_20 =top.iloc[1:21,:]
top_20
```

Out[7]:

	title	domestic_gross	year
3080	Black Panther	700100000.0	2018
3079	Avengers: Infinity War	678800000.0	2018
1873	Jurassic World	652300000.0	2015
2758	Star Wars: The Last Jedi	620200000.0	2017
3082	Incredibles 2	608600000.0	2018
2323	Rogue One: A Star Wars Story	532200000.0	2016
2759	Beauty and the Beast (2017)	504000000.0	2017
2324	Finding Dory	486300000.0	2016
1875	Avengers: Age of Ultron	459000000.0	2015
3081	Jurassic World: Fallen Kingdom	417700000.0	2018
2767	Wonder Woman	412600000.0	2017
2322	Captain America: Civil War	408100000.0	2016
2762	Jumanji: Welcome to the Jungle	404500000.0	2017
2765	Guardians of the Galaxy Vol. 2	389800000.0	2017
2327	The Secret Life of Pets	368400000.0	2016
2326	The Jungle Book (2016)	364000000.0	2016
2330	Deadpool	363100000.0	2016
1878	Inside Out	356500000.0	2015
1874	Furious 7	353000000.0	2015
1489	American Sniper	350100000.0	2014

In [8]:

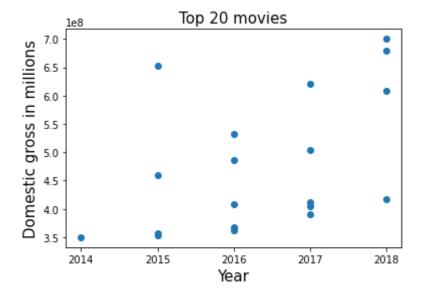
```
# poltting graph to see movies performance

import matplotlib.pyplot as plt
%matplotlib inline
import numpy as np
fig, ax = plt.subplots()
plt.scatter(top_20['year'], top_20['domestic_gross'])
plt.rcParams['figure.figsize'] = [15,15]
plt.xticks(np.arange(2014, 2019, 1))

ax.set_title('Top 20 movies',fontsize=15)
ax.set_ylabel('Year',fontsize=15)
ax.set_ylabel('Domestic gross in millions',fontsize=15)
```

Out[8]:

Text(0, 0.5, 'Domestic gross in millions')



In [9]: ▶

```
# importing titles table to get genres for matching top 20 movies

dfim_basic = pd.read_csv('imdb.title.basics.csv.gz')
dfim_basic.head()
```

Out[9]:

	tconst	primary_title	original_title	start_year	runtime_minutes	genres
0	tt0063540	Sunghursh	Sunghursh	2013	175.0	Action,Crime,Drama
1	tt0066787	One Day Before the Rainy Season	Ashad Ka Ek Din	2019	114.0	Biography,Drama
2	tt0069049	The Other Side of the Wind	The Other Side of the Wind	2018	122.0	Drama
3	tt0069204	Sabse Bada Sukh	Sabse Bada Sukh	2018	NaN	Comedy,Drama
4	tt0100275	The Wandering Soap Opera	La Telenovela Errante	2017	80.0	Comedy,Drama,Fantasy

In [10]:

M

Out[10]:

	tconst	genres
title		
Sunghursh	tt0063540	Action,Crime,Drama
One Day Before the Rainy Season	tt0066787	Biography,Drama
The Other Side of the Wind	tt0069049	Drama
Sabse Bada Sukh	tt0069204	Comedy,Drama
The Wandering Soap Opera	tt0100275	Comedy,Drama,Fantasy
Kuambil Lagi Hatiku	tt9916538	Drama
Rodolpho Teóphilo - O Legado de um Pioneiro	tt9916622	Documentary
Dankyavar Danka	tt9916706	Comedy
6 Gunn	tt9916730	NaN
Chico Albuquerque - Revelações	tt9916754	Documentary

146144 rows × 2 columns

```
In [11]:
```

```
# setting index to prepare for merge
top_20.set_index('title')
```

Out[11]:

	domestic_gross	year
title		
Black Panther	700100000.0	2018
Avengers: Infinity War	678800000.0	2018
Jurassic World	652300000.0	2015
Star Wars: The Last Jedi	620200000.0	2017
Incredibles 2	608600000.0	2018
Rogue One: A Star Wars Story	532200000.0	2016
Beauty and the Beast (2017)	504000000.0	2017
Finding Dory	486300000.0	2016
Avengers: Age of Ultron	459000000.0	2015
Jurassic World: Fallen Kingdom	417700000.0	2018
Wonder Woman	412600000.0	2017
Captain America: Civil War	408100000.0	2016
Jumanji: Welcome to the Jungle	404500000.0	2017
Guardians of the Galaxy Vol. 2	389800000.0	2017
The Secret Life of Pets	368400000.0	2016
The Jungle Book (2016)	364000000.0	2016
Deadpool	363100000.0	2016
Inside Out	356500000.0	2015
Furious 7	353000000.0	2015
American Sniper	350100000.0	2014

In [12]: ▶

```
# cross checking if the Black Panther tconst has come across in merged df
dfim_basic[dfim_basic['title']== 'Black Panther']
```

Out[12]:

	tconst	title	genres
19050	#1825683	Black Panther	Action Adventure Sci-Fi

In [13]: ▶

```
# merging top_20 with dfim_basic to get tconst
genre_join = top_20.merge(dfim_basic, how='left')
genre_join
```

Out[13]:

	title	domestic_gross	year	tconst	genres
0	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi
1	Avengers: Infinity War	678800000.0	2018	tt4154756	Action,Adventure,Sci-Fi
2	Jurassic World	652300000.0	2015	tt0369610	Action,Adventure,Sci-Fi
3	Star Wars: The Last Jedi	620200000.0	2017	tt2527336	Action,Adventure,Fantasy
4	Incredibles 2	608600000.0	2018	tt3606756	Action,Adventure,Animation
5	Rogue One: A Star Wars Story	532200000.0	2016	tt3748528	Action,Adventure,Sci-Fi
6	Beauty and the Beast (2017)	504000000.0	2017	NaN	NaN
7	Finding Dory	486300000.0	2016	tt2277860	Adventure, Animation, Comedy
8	Avengers: Age of Ultron	459000000.0	2015	tt2395427	Action,Adventure,Sci-Fi
9	Jurassic World: Fallen Kingdom	417700000.0	2018	tt4881806	Action,Adventure,Sci-Fi
10	Wonder Woman	412600000.0	2017	tt0451279	Action,Adventure,Fantasy
11	Wonder Woman	412600000.0	2017	tt4028068	Sci-Fi
12	Wonder Woman	412600000.0	2017	tt4283448	Documentary,Drama,Sport
13	Captain America: Civil War	408100000.0	2016	tt3498820	Action,Adventure,Sci-Fi
14	Jumanji: Welcome to the Jungle	404500000.0	2017	tt2283362	Action,Adventure,Comedy
15	Guardians of the Galaxy Vol. 2	389800000.0	2017	tt3896198	Action,Adventure,Comedy
16	The Secret Life of Pets	368400000.0	2016	tt2709768	Adventure, Animation, Comedy
17	The Jungle Book (2016)	364000000.0	2016	NaN	NaN
18	Deadpool	363100000.0	2016	tt1431045	Action,Adventure,Comedy
19	Inside Out	356500000.0	2015	tt1640486	Crime,Drama
20	Inside Out	356500000.0	2015	tt1865538	Documentary
21	Inside Out	356500000.0	2015	tt2064820	Drama
22	Inside Out	356500000.0	2015	tt2071483	Family
23	Inside Out	356500000.0	2015	tt2096673	Adventure, Animation, Comedy
24	Inside Out	356500000.0	2015	tt2608638	Biography,Documentary,History
25	Inside Out	356500000.0	2015	tt6419446	NaN
26	Inside Out	356500000.0	2015	tt8269544	NaN
27	Furious 7	353000000.0	2015	tt2820852	Action, Crime, Thriller
28	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama

```
In [14]:
```

```
# dropping duplicates from above merge.
# getting final data for merging with another df to get these genres directors and writers
gen_final =genre_join.drop_duplicates(['title'], keep ='first')
gen_final
```

Out[14]:

	title	domestic_gross	year	tconst	genres
0	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi
1	Avengers: Infinity War	678800000.0	2018	tt4154756	Action,Adventure,Sci-Fi
2	Jurassic World	652300000.0	2015	tt0369610	Action,Adventure,Sci-Fi
3	Star Wars: The Last Jedi	620200000.0	2017	tt2527336	Action,Adventure,Fantasy
4	Incredibles 2	608600000.0	2018	tt3606756	Action,Adventure,Animation
5	Rogue One: A Star Wars Story	532200000.0	2016	tt3748528	Action,Adventure,Sci-Fi
6	Beauty and the Beast (2017)	504000000.0	2017	NaN	NaN
7	Finding Dory	486300000.0	2016	tt2277860	Adventure, Animation, Comedy
8	Avengers: Age of Ultron	459000000.0	2015	tt2395427	Action,Adventure,Sci-Fi
9	Jurassic World: Fallen Kingdom	417700000.0	2018	tt4881806	Action,Adventure,Sci-Fi
10	Wonder Woman	412600000.0	2017	tt0451279	Action,Adventure,Fantasy
13	Captain America: Civil War	408100000.0	2016	tt3498820	Action,Adventure,Sci-Fi
14	Jumanji: Welcome to the Jungle	404500000.0	2017	tt2283362	Action,Adventure,Comedy
15	Guardians of the Galaxy Vol. 2	389800000.0	2017	tt3896198	Action,Adventure,Comedy
16	The Secret Life of Pets	368400000.0	2016	tt2709768	Adventure, Animation, Comedy
17	The Jungle Book (2016)	364000000.0	2016	NaN	NaN
18	Deadpool	363100000.0	2016	tt1431045	Action,Adventure,Comedy
19	Inside Out	356500000.0	2015	tt1640486	Crime,Drama
27	Furious 7	353000000.0	2015	tt2820852	Action,Crime,Thriller
28	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama

In [15]: ▶

```
# mean of domestic gross by grouping genres
genre_graph = gen_final.groupby('genres').mean()
genre_graph
```

Out[15]:

	domestic_gross	year
genres		
Action,Adventure,Animation	6.086000e+08	2018.000000
Action,Adventure,Comedy	3.858000e+08	2016.666667
Action,Adventure,Fantasy	5.164000e+08	2017.000000
Action,Adventure,Sci-Fi	5.497429e+08	2016.571429
Action,Biography,Drama	3.501000e+08	2014.000000
Action,Crime,Thriller	3.530000e+08	2015.000000
Adventure, Animation, Comedy	4.273500e+08	2016.000000
Crime,Drama	3.565000e+08	2015.000000

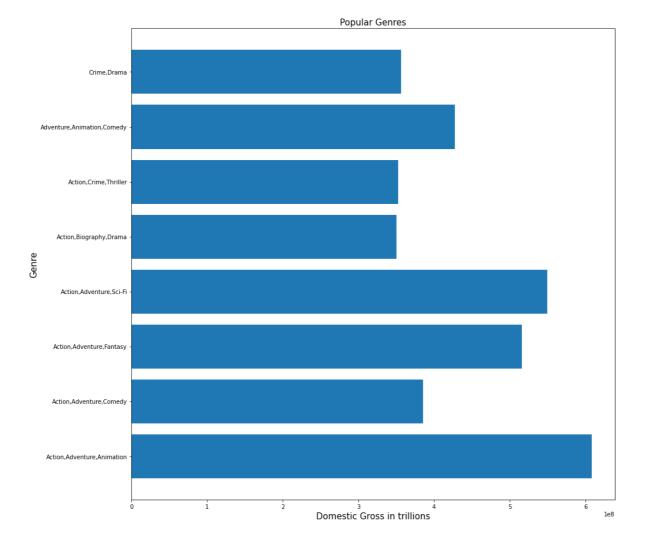
In [16]:

genre_graph = genre_graph.reset_index()

In [17]: ▶

```
# plotting bar graph to see which genre is most popular

import matplotlib.pyplot as pltt
x= genre_graph['genres']
y = genre_graph['domestic_gross']
pltt.barh(x,y)
pltt.title('Popular Genres', fontsize=15)
pltt.ylabel('Genre', fontsize=15)
pltt.xlabel('Domestic Gross in trillions', fontsize=15)
pltt.show()
```



Type *Markdown* and LaTeX: α^2

```
In [18]:
                                                                                         M
 importing datasets to find out popular movies Directors and Writers
crew = pd.read_csv('imdb.title.crew.csv.gz')
principals = pd.read_csv('imdb.title.principals.csv.gz')
```

movies = pd.read_csv('imdb.title.basics.csv.gz')

names = pd.read_csv('imdb.name.basics.csv.gz')

In [19]:

movies.head()

Out[19]:

	tconst	primary_title	original_title	start_year	runtime_minutes	genres
0	tt0063540	Sunghursh	Sunghursh	2013	175.0	Action,Crime,Drama
1	tt0066787	One Day Before the Rainy Season	Ashad Ka Ek Din	2019	114.0	Biography,Drama
2	tt0069049	The Other Side of the Wind	The Other Side of the Wind	2018	122.0	Drama
3	tt0069204	Sabse Bada Sukh	Sabse Bada Sukh	2018	NaN	Comedy,Drama
4	tt0100275	The Wandering Soap Opera	La Telenovela Errante	2017	80.0	Comedy,Drama,Fantasy

In [20]:

crew.head()

Out[20]:

writers	directors	tconst	
nm0899854	nm0899854	tt0285252	0
nm0175726,nm1802864	NaN	tt0438973	1
nm1940585	nm1940585	tt0462036	2
nm0310087,nm0841532	nm0151540	tt0835418	3
nm0284943	nm0089502 nm2291498 nm2292011	tt0878654	4

In [21]:

names.head()

Out[21]:

primary_profe	death_year	birth_year	primary_name	nconst	
miscellaneous,production_manager,pro	NaN	NaN	Mary Ellen Bauder	nm0061671	0
composer,music_department,sound_depar	NaN	NaN	Joseph Bauer	nm0061865	1
miscellaneous,actor,	NaN	NaN	Bruce Baum	nm0062070	2
camera_department,cinematographer,art_depar	NaN	NaN	Axel Baumann	nm0062195	3
production_designer,art_department,set_dec	NaN	NaN	Pete Baxter	nm0062798	4
•					4

In [22]:

movies[movies['primary_title']=='Inside Out']

Out[22]:

	runtime_minutes	start_year	original_title	primary_title	tconst	
Crime	93.0	2011	Inside Out	Inside Out	tt1640486	11735
Docu	74.0	2011	Ti mesa, ti exo	Inside Out	tt1865538	20881
	NaN	2011	Inside Out	Inside Out	tt2064820	26786
	59.0	2011	Inside Out	Inside Out	tt2071483	27024
Adventure, Animation,	95.0	2015	Inside Out	Inside Out	tt2096673	28269
Biography,Documentary	75.0	2013	Inside Out	Inside Out	tt2608638	43781
	NaN	2011	Inside Out	Inside Out	tt6419446	109260
	NaN	2018	Inside Out	Inside Out	tt8269544	132362
•						4

In [23]: ▶

```
# extracting all data related to top 20 movies
example_movies = movies[(movies['primary_title']).isin(['Avengers: Infinity War','Furious 7
example_movies
```

Out[23]:

	runtime_minutes	start_year	original_title	primary_title	tconst	
Action,Adventui	124.0	2015	Jurassic World	Jurassic World	tt0369610	60
Action,Adventure,	141.0	2017	Wonder Woman	Wonder Woman	tt0451279	131
Action, Adventure,	108.0	2016	Deadpool	Deadpool	tt1431045	7543
Crime	93.0	2011	Inside Out	Inside Out	tt1640486	11735
Action,Adventu	134.0	2018	Black Panther	Black Panther	tt1825683	19050
Docu	74.0	2011	Ti mesa, ti exo	Inside Out	tt1865538	20881
	NaN	2011	Inside Out	Inside Out	tt2064820	26786
	59.0	2011	Inside Out	Inside Out	tt2071483	27024
Adventure, Animation,	95.0	2015	Inside Out	Inside Out	tt2096673	28269
Action,Biograph	133.0	2014	American Sniper	American Sniper	tt2179136	31013
Adventure, Animation,	97.0	2016	Finding Dory	Finding Dory	tt2277860	34496
Action,Adventure,	119.0	2017	Jumanji: Welcome to the Jungle	Jumanji: Welcome to the Jungle	tt2283362	34723
Action,Adventu	141.0	2015	Avengers: Age of Ultron	Avengers: Age of Ultron	tt2395427	39010
Action,Adventure,	152.0	2017	Star Wars: Episode VIII - The Last Jedi	Star Wars: The Last Jedi	tt2527336	42223
Biography,Documentary	75.0	2013	Inside Out	Inside Out	tt2608638	43781
Adventure,Animation,	87.0	2016	The Secret Life of Pets	The Secret Life of Pets	tt2709768	45535
Action,Crime	137.0	2015	Furious Seven	Furious 7	tt2820852	47535
Action,Adventu	147.0	2016	Captain America: Civil War	Captain America: Civil War	tt3498820	60379
Action,Adventure,Ar	118.0	2018	Incredibles 2	Incredibles 2	tt3606756	62741
Action,Adventu	133.0	2016	Rogue One	Rogue One: A Star Wars Story	tt3748528	65705
Action,Adventure,	136.0	2017	Guardians of the Galaxy Vol. 2	Guardians of the Galaxy Vol. 2	tt3896198	68538

	runtime_minutes	start_year	original_title	primary_title	tconst	
	60.0	2014	Wonder Woman	Wonder Woman	tt4028068	70470
Action,Adventu	149.0	2018	Avengers: Infinity War	Avengers: Infinity War	tt4154756	72820
Documentary,Dran	75.0	2016	Wonder Woman	Wonder Woman	tt4283448	75016
Action,Adventu	128.0	2018	Jurassic World: Fallen Kingdom	Jurassic World: Fallen Kingdom	tt4881806	84414
	NaN	2011	Inside Out	Inside Out	tt6419446	109260
	NaN	2018	Inside Out	Inside Out	tt8269544	132362

In [29]: ▶

removing duplicate titles from the dataframe

eg_movies = example_movies[-((example_movies['primary_title']=='Inside Out') & (example_mov
eg_movie = eg_movies[-((eg_movies['primary_title']=='Wonder Woman') & (eg_movies['start_yea
eg_movie

Out[29]:

	tconst	primary_title	original_title	start_year	runtime_minutes	ge
60	tt0369610	Jurassic World	Jurassic World	2015	124.0	Action, Adventure,
131	tt0451279	Wonder Woman	Wonder Woman	2017	141.0	Action,Adventure,Fa
7543	tt1431045	Deadpool	Deadpool	2016	108.0	Action,Adventure,Co
19050	tt1825683	Black Panther	Black Panther	2018	134.0	Action, Adventure,
28269	tt2096673	Inside Out	Inside Out	2015	95.0	Adventure,Animation,Co
31013	tt2179136	American Sniper	American Sniper	2014	133.0	Action,Biography,D
34496	tt2277860	Finding Dory	Finding Dory	2016	97.0	Adventure, Animation, Co
34723	tt2283362	Jumanji: Welcome to the Jungle	Jumanji: Welcome to the Jungle	2017	119.0	Action,Adventure,Co
39010	tt2395427	Avengers: Age of Ultron	Avengers: Age of Ultron	2015	141.0	Action, Adventure,
42223	tt2527336	Star Wars: The Last Jedi	Star Wars: Episode VIII - The Last Jedi	2017	152.0	Action,Adventure,Fa
45535	tt2709768	The Secret Life of Pets	The Secret Life of Pets	2016	87.0	Adventure, Animation, Co
47535	tt2820852	Furious 7	Furious Seven	2015	137.0	Action,Crime,T
60379	tt3498820	Captain America: Civil War	Captain America: Civil War	2016	147.0	Action,Adventure,
62741	tt3606756	Incredibles 2	Incredibles 2	2018	118.0	Action,Adventure,Anim
65705	tt3748528	Rogue One: A Star Wars Story	Rogue One	2016	133.0	Action,Adventure,
68538	tt3896198	Guardians of the Galaxy Vol. 2	Guardians of the Galaxy Vol. 2	2017	136.0	Action,Adventure,Co
72820	tt4154756	Avengers: Infinity War	Avengers: Infinity War	2018	149.0	Action,Adventure,
84414	tt4881806	Jurassic World: Fallen Kingdom	Jurassic World: Fallen Kingdom	2018	128.0	Action,Adventure,\$
4						→

In [32]: ▶

```
# Merging to get directors and writers list
eg_movie = eg_movie.merge(crew, on='tconst')
eg_movie
```

Out[32]:

	tconst	primary_title	original_title	start_year	runtime_minutes	genre
0	tt0369610	Jurassic World	Jurassic World	2015	124.0	Action,Adventure,Sci-l
1	tt0451279	Wonder Woman	Wonder Woman	2017	141.0	Action,Adventure,Fantas
2	tt1431045	Deadpool	Deadpool	2016	108.0	Action,Adventure,Comec
3	tt1825683	Black Panther	Black Panther	2018	134.0	Action,Adventure,Sci-I
4	tt2096673	Inside Out	Inside Out	2015	95.0	Adventure, Animation, Comed
5	tt2179136	American Sniper	American Sniper	2014	133.0	Action,Biography,Dram
6	tt2277860	Finding Dory	Finding Dory	2016	97.0	Adventure, Animation, Comed
7	tt2283362	Jumanji: Welcome to the Jungle	Jumanji: Welcome to the Jungle	2017	119.0	Action,Adventure,Comec
8	tt2395427	Avengers: Age of Ultron	Avengers: Age of Ultron	2015	141.0	Action,Adventure,Sci-I
9	tt2527336	Star Wars: The Last Jedi	Star Wars: Episode VIII - The Last Jedi	2017	152.0	Action,Adventure,Fantas
10	tt2709768	The Secret Life of Pets	The Secret Life of Pets	2016	87.0	Adventure, Animation, Comec
11	tt2820852	Furious 7	Furious Seven	2015	137.0	Action,Crime,Thrille
12	tt3498820	Captain America: Civil War	Captain America: Civil War	2016	147.0	Action,Adventure,Sci-I
13	tt3606756	Incredibles 2	Incredibles 2	2018	118.0	Action,Adventure,Animatic
14	tt3748528	Rogue One: A Star Wars Story	Rogue One	2016	133.0	Action,Adventure,Sci-I
15	tt3896198	Guardians of the Galaxy Vol. 2	Guardians of the Galaxy Vol. 2	2017	136.0	Action,Adventure,Comec
16	tt4028068	Wonder Woman	Wonder Woman	2014	60.0	Sci-l
17	tt4154756	Avengers: Infinity War	Avengers: Infinity War	2018	149.0	Action,Adventure,Sci-I
18	tt4283448	Wonder Woman	Wonder Woman	2016	75.0	Documentary,Drama,Spo
19	tt4881806	Jurassic World: Fallen Kingdom	Jurassic World: Fallen Kingdom	2018	128.0	Action,Adventure,Sci-I

In [33]:

```
# Merging
```

director_names = eg_movie.merge(names, left_on='directors', right_on='nconst', how='left')
display(director_names)
director_names.info()

	tconst	primary_title	original_title	start_year	runtime_minutes	ge
0	tt0369610	Jurassic World	Jurassic World	2015	124.0	Action,Adventure,S
1	tt0451279	Wonder Woman	Wonder Woman	2017	141.0	Action,Adventure,Far
2	tt1431045	Deadpool	Deadpool	2016	108.0	Action,Adventure,Cor
3	tt1825683	Black Panther	Black Panther	2018	134.0	Action,Adventure,S
4	tt2096673	Inside Out	Inside Out	2015	95.0	Adventure, Animation, Cor
5	tt2179136	American Sniper	American Sniper	2014	133.0	Action,Biography,Di
6	tt2277860	Finding Dory	Finding Dory	2016	97.0	Adventure, Animation, Cor
7	tt2283362	Jumanji: Welcome to the Jungle	Jumanji: Welcome to the Jungle	2017	119.0	Action,Adventure,Cor
8	tt2395427	Avengers: Age of Ultron	Avengers: Age of Ultron	2015	141.0	Action,Adventure,S
9	tt2527336	Star Wars: The Last Jedi	Star Wars: Episode VIII - The Last Jedi	2017	152.0	Action,Adventure,Far
10	tt2709768	The Secret Life of Pets	The Secret Life of Pets	2016	87.0	Adventure, Animation, Cor
11	tt2820852	Furious 7	Furious Seven	2015	137.0	Action,Crime,Tr
12	tt3498820	Captain America: Civil War	Captain America: Civil War	2016	147.0	Action,Adventure,S
13	tt3606756	Incredibles 2	Incredibles 2	2018	118.0	Action,Adventure,Anim
14	tt3748528	Rogue One: A Star Wars Story	Rogue One	2016	133.0	Action,Adventure,S
15	tt3896198	Guardians of the Galaxy Vol. 2	Guardians of the Galaxy Vol. 2	2017	136.0	Action,Adventure,Cor
16	tt4028068	Wonder Woman	Wonder Woman	2014	60.0	8
17	tt4154756	Avengers: Infinity War	Avengers: Infinity War	2018	149.0	Action,Adventure,S
18	tt4283448	Wonder Woman	Wonder Woman	2016	75.0	Documentary,Drama,

		tconst	primary_title	original_title	start_year	runtime_minutes	ge
,	19	tt4881806	Jurassic World: Fallen Kingdom	Jurassic World: Fallen Kingdom	2018	128.0	Action,Adventure,S
4							>

<class 'pandas.core.frame.DataFrame'>
Int64Index: 20 entries, 0 to 19
Data columns (total 18 columns):

#	Column	Non-Null Count	Dtype
0	tconst	20 non-null	object
1	<pre>primary_title</pre>	20 non-null	object
2	original_title	20 non-null	object
3	start_year	20 non-null	int64
4	runtime_minutes	20 non-null	float64
5	genres	20 non-null	object
6	directors_x	20 non-null	object
7	writers_x	19 non-null	object
8	directors_y	20 non-null	object
9	writers_y	19 non-null	object
10	directors	20 non-null	object
11	writers	19 non-null	object
12	nconst	13 non-null	object
13	primary_name	13 non-null	object
14	birth_year	12 non-null	float64
15	death_year	0 non-null	float64
16	<pre>primary_profession</pre>	13 non-null	object
17	known_for_titles	13 non-null	object
44	Cl+C4/2\ :-+C	1/1\ - +/1/1	

dtypes: float64(3), int64(1), object(14)

memory usage: 3.0+ KB

In [37]:

Dropping duplicate directors

director_list=director_names.dropna(subset=['primary_name'])
director_list

Out[37]:

	tconst	primary_title	original_title	start_year	runtime_minutes	genres
0	tt0369610	Jurassic World	Jurassic World	2015	124.0	Action,Adventure,Sci-Fi
1	tt0451279	Wonder Woman	Wonder Woman	2017	141.0	Action,Adventure,Fantasy
2	tt1431045	Deadpool	Deadpool	2016	108.0	Action,Adventure,Comedy
3	tt1825683	Black Panther	Black Panther	2018	134.0	Action,Adventure,Sci-Fi
5	tt2179136	American Sniper	American Sniper	2014	133.0	Action,Biography,Drama
7	tt2283362	Jumanji: Welcome to the Jungle	Jumanji: Welcome to the Jungle	2017	119.0	Action,Adventure,Comedy
8	tt2395427	Avengers: Age of Ultron	Avengers: Age of Ultron	2015	141.0	Action,Adventure,Sci-Fi
9	tt2527336	Star Wars: The Last Jedi	Star Wars: Episode VIII - The Last Jedi	2017	152.0	Action,Adventure,Fantasy
11	tt2820852	Furious 7	Furious Seven	2015	137.0	Action,Crime,Thriller
13	tt3606756	Incredibles 2	Incredibles 2	2018	118.0	Action,Adventure,Animation
14	tt3748528	Rogue One: A Star Wars Story	Rogue One	2016	133.0	Action,Adventure,Sci-Fi
15	tt3896198	Guardians of the Galaxy Vol. 2	Guardians of the Galaxy Vol. 2	2017	136.0	Action,Adventure,Comedy
19	tt4881806	Jurassic World: Fallen Kingdom	Jurassic World: Fallen Kingdom	2018	128.0	Action,Adventure,Sci-Fi

```
H
In [38]:
   list of Directors
print(director_list.loc[:,'primary_name'])
      Colin Trevorrow
0
1
        Patty Jenkins
           Tim Miller
2
3
         Ryan Coogler
5
       Clint Eastwood
7
          Jake Kasdan
8
          Joss Whedon
         Rian Johnson
9
11
            James Wan
            Brad Bird
13
14
       Gareth Edwards
15
           James Gunn
19
          J.A. Bayona
Name: primary_name, dtype: object
```

```
In [28]: ▶
```

```
# Inspecting data
principals.head()
```

Out[28]:

	tconst	ordering	nconst	category	job	characters
0	tt0111414	1	nm0246005	actor	NaN	["The Man"]
1	tt0111414	2	nm0398271	director	NaN	NaN
2	tt0111414	3	nm3739909	producer	producer	NaN
3	tt0323808	10	nm0059247	editor	NaN	NaN
4	tt0323808	1	nm3579312	actress	NaN	["Beth Boothby"]

M

In [29]:

gen_final

Out[29]:

	title	domestic_gross	year	tconst	genres
0	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi
1	Avengers: Infinity War	678800000.0	2018	tt4154756	Action,Adventure,Sci-Fi
2	Jurassic World	652300000.0	2015	tt0369610	Action,Adventure,Sci-Fi
3	Star Wars: The Last Jedi	620200000.0	2017	tt2527336	Action,Adventure,Fantasy
4	Incredibles 2	608600000.0	2018	tt3606756	Action, Adventure, Animation
5	Rogue One: A Star Wars Story	532200000.0	2016	tt3748528	Action,Adventure,Sci-Fi
6	Beauty and the Beast (2017)	504000000.0	2017	NaN	NaN
7	Finding Dory	486300000.0	2016	tt2277860	Adventure, Animation, Comedy
8	Avengers: Age of Ultron	459000000.0	2015	tt2395427	Action,Adventure,Sci-Fi
9	Jurassic World: Fallen Kingdom	417700000.0	2018	tt4881806	Action,Adventure,Sci-Fi
10	Wonder Woman	412600000.0	2017	tt0451279	Action,Adventure,Fantasy
13	Captain America: Civil War	408100000.0	2016	tt3498820	Action,Adventure,Sci-Fi
14	Jumanji: Welcome to the Jungle	404500000.0	2017	tt2283362	Action,Adventure,Comedy
15	Guardians of the Galaxy Vol. 2	389800000.0	2017	tt3896198	Action,Adventure,Comedy
16	The Secret Life of Pets	368400000.0	2016	tt2709768	Adventure, Animation, Comedy
17	The Jungle Book (2016)	364000000.0	2016	NaN	NaN
18	Deadpool	363100000.0	2016	tt1431045	Action,Adventure,Comedy
19	Inside Out	356500000.0	2015	tt1640486	Crime,Drama
27	Furious 7	353000000.0	2015	tt2820852	Action,Crime,Thriller
28	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama

In [39]: ▶

Merging to get writers codes

writers_codes = gen_final.merge(principals, left_on='tconst', right_on='tconst', how='left'
display(writers_codes)

	title	domestic_gross	year	tconst	genres	ordering	nconst	C
0	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	10.0	nm3234869	С
1	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	1.0	nm1569276	
2	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	2.0	nm0430107	
3	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	3.0	nm2143282	
4	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	4.0	nm1775091	
177	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	5.0	nm0000142	
178	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	6.0	nm0355699	
179	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	7.0	nm4881574	
180	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	8.0	nm6984200	
181	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	9.0	nm7198520	

182 rows × 10 columns

In [40]: ▶

merging to get names of writers from codes

writers=writers_codes.merge(names, left_on='nconst', right_on='nconst', how='left')
display(writers)

	title	domestic_gross	year	tconst	genres	ordering	nconst	C
0	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	10.0	nm3234869	С
1	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	1.0	nm1569276	
2	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	2.0	nm0430107	
3	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	3.0	nm2143282	
4	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	4.0	nm1775091	
177	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	5.0	nm0000142	
178	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	6.0	nm0355699	
179	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	7.0	nm4881574	
180	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	8.0	nm6984200	
181	American Sniper	350100000.0	2014	tt2179136	Action,Biography,Drama	9.0	nm7198520	

182 rows × 15 columns

In [41]:

```
# sorting writers from the dataframe
writers_list = writers['category']== 'writer']
writers_list
```

Out[41]:

	title	domestic_gross	year	tconst	genres	ordering	nconst	category	
6	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	6.0	nm1963288	writer	
7	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	7.0	nm0498278	writer	
8	Black Panther	700100000.0	2018	tt1825683	Action,Adventure,Sci-Fi	8.0	nm0456158	writer	
10	Avengers: Infinity War	678800000.0	2018	tt4154756	Action,Adventure,Sci-Fi	10.0	nm0456158	writer	
17	Avengers: Infinity	678800000.0	2018	tt4154756	Action,Adventure,Sci-Fi	7.0	nm1321655	writer	•

In [42]: ▶

```
# Writers list
w=(writers_list.loc[:,'primary_name'])
w
```

Out[42]:

```
6
                Joe Robert Cole
7
                       Stan Lee
                     Jack Kirby
8
10
                     Jack Kirby
            Christopher Markus
17
                Stephen McFeely
18
19
                       Stan Lee
                     Rick Jaffa
26
27
                  Amanda Silver
28
                 Derek Connolly
29
              Michael Crichton
36
                   George Lucas
                   George Lucas
50
56
                    Chris Weitz
57
                    Tony Gilroy
58
                     John Knoll
59
                    Gary Whitta
               Victoria Strouse
68
69
                   Bob Peterson
77
                       Stan Lee
78
                     Jack Kirby
79
                      Joe Simon
80
                    Jim Starlin
                 Derek Connolly
87
88
                Colin Trevorrow
              Michael Crichton
89
91
                 Harry G. Peter
97
                 Allan Heinberg
98
                    Zack Snyder
99
                    Jason Fuchs
100
       William Moulton Marston
101
                     Jack Kirby
108
            Christopher Markus
109
                Stephen McFeely
110
                      Joe Simon
            Chris Van Allsburg
111
                  Chris McKenna
117
                   Erik Sommers
118
                Scott Rosenberg
119
120
                   Jeff Pinkner
121
                    Jim Starlin
127
                     Dan Abnett
128
                   Andy Lanning
                Steve Englehart
129
130
                      Steve Gan
                     Cinco Paul
138
                     Ken Daurio
139
140
                    Brian Lynch
148
                    Rhett Reese
                   Paul Wernick
149
158
                 Dylan Schaffer
168
                   Chris Morgan
```

169	Gary Scott Thompson
178	Jason Hall
179	Chris Kyle
180	Scott McEwen
181	Jim DeFelice
Name:	<pre>primary_name, dtype: object</pre>

In [43]:	M
w.info()	
<pre><class 'pandas.core.series.series'=""> Int64Index: 57 entries, 6 to 181 Series name: primary_name Non-Null Count Dtype</class></pre>	
57 non-null object dtypes: object(1) memory usage: 912.0+ bytes	
In []:	M
In []:	M
In []:	M
In []:	M
In []:	M
In []:	M
In []:	H
In []:	M
In []:	M

In []:	H
In []:	H
In []:	Н
III [].	VI
In []:	H
In []:	K
In []:	K
In []:	H
In []:	H
In []:	H
In []:	Н
	71
T. []	NI .
In []:	K
In []:	K
In []:	Н