```
In [84]:
             import pandas as pd
            import numpy as np
In [85]:
            df = pd.read excel('hollywood.xlsx')
            df['movie board rating display name'].value counts()
                            448
Out[85]:
                            441
           PG-13
           PG
                            182
           Not Rated
                             83
                             39
           G
           NC-17
                              3
           Name: movie board rating display name, dtype: int64
In [86]:
            df
                                            display_name production_year movie_sequel creative_type
                          id
                                     name
Out[86]:
                                                                                                   Science
               0
                     7950115
                                                                      2009
                                                                                         0
                                    Avatar
                                                    Avatar
                                                                                                    Fiction
                               Harry Potter
                                              Harry Potter
                                   and the
                                                  and the
                   50950115
                                   Deathly
                                                  Deathly
                                                                       2011
                                                                                          1
               1
                                                                                                   Fantasy
                              Hallows: Part
                                             Hallows: Part
                                         Ш
                                             Transformers:
                              Transformers
                                                                                                  Science
                   119870115
                                               Dark of the
                                                                       2011
                                                                                          1
                                         3
                                                                                                   Fiction
                                                    Moon
                  119590115
                                Toy Story 3
                                               Toy Story 3
                                                                       2010
                                                                                          1
                                                                                               Kids Fiction
                                             Pirates of the
                              Pirates of the
                                            Caribbean: On
                    91700115
                                                                       2011
                                                                                                   Fantasy 1
                               Caribbean 4
                                                 Stranger
                                                    Tides
                                                                                             Contemporary
                                 Red State
                  144410115
                                                                       2011
            1191
                                                Red State
                                                                                                    Fiction
                                                                                             Contemporary
            1192 133000115
                                Mission, La
                                                La Mission
                                                                      2009
                                                                                                    Fiction
                                                                                                  Multiple
                                2010 Oscar
                                               2010 Oscar
            1193
                      630115
                                                                      2009
                                                                                                  Creative C
                                    Shorts
                                                   Shorts
                                                                                                    Types
                              sept jours du
                                             Les sept jours
                                                                                             Contemporary
                  133360115
                                                                       2010
            1194
                                 talion, Les
                                                 du talion
                                                                                                    Fiction
                                               Damsels in
                                Damsels in
                                                                                             Contemporary
            1195 146890115
                                                                       2010
                                                                                                    Fiction
                                   Distress
                                                  Distress
```

1196 rows × 15 columns

```
## To check if there is any null value in the dataframe
          df.isnull().any()
         id
                                                 False
Out[87]:
                                                 False
         name
                                                 False
         display name
         production year
                                                 False
         movie sequel
                                                 False
         creative_type
                                                 False
                                                 False
         source
         production method
                                                 False
         genre
                                                 False
                                                 False
         language
                                                 False
         board rating reason
         movie board rating display name
                                                 False
         movie release pattern display name
                                                 False
                                                 False
         total
         Category
                                                 False
         dtype: bool
```

1) What are the top three genre of movies with the highest average earnings?

```
In [88]: movies_grouped_by_genre = df[['total','genre']].groupby('genre',as_index=False]
In [89]: ## This line of code finds total earning by each genre
    total_earnings_by_genre = movies_grouped_by_genre.sum()

In [90]: ## This line of code finds total number of movies by each genre
    total_number_of_movies_by_genre = movies_grouped_by_genre.count()
    total_earnings_by_genre['avg_earnings'] = total_earnings_by_genre['total']/total_earnings_by_genre
Out[90]: genre total avg_earnings
```

aenre total avg_earnings 0 25180 203.064516 Action 1 Adventure 32369 302.514019 2 **Black Comedy** 648 54.000000 3 Comedy 23460 90.230769 Concert/Performance 270 45.000000 5 Documentary 914 16.925926 6 Drama 16493 51.380062 7 Horror 5822 78.675676 8 Multiple Genres 31 3.444444 9 Musical 1718 143.166667 10 Romantic Comedy 5900 73.750000 11 Thriller/Suspense 11935 91.106870 12 Western 485 80.833333

```
In [91]:
    total_earnings_by_genre = total_earnings_by_genre.sort_values(['avg_earnings'])
```

Final Answer

2) Do movies with sequels tend to have higher gross earnings than the movies without sequels? You can pick the last movie to come out for a prequel.

3) Find the proportion (percentage) of movies by ratings for English language and others (all other languages). Are proportions significantly different from each other?

```
In [95]:
          movies by language = df[['name', 'language']].groupby('language')
          movies by language.count()
          english movies = df[df['language']=='English']
          other language movies = df[df['language']!='English']
In [100...
          proportions_by_rating = pd.DataFrame(english_movies['movie_board_rating_displeaders)
          rating labels = proportions by rating.index
          proportions_by_rating['rating'] = rating_labels
          proportions_by_rating.index = [i for i in range(1,len(proportions_by_rating)+
          total = sum(english_movies['movie_board_rating_display_name'].value_counts())
          proportions by rating['total movies'] = proportions by rating['movie board ra
          proportions by rating['proportion(in %)'] = proportions by rating['movie board
          proportions by rating.drop('movie board rating display name',axis=1,inplace =
In [102...
          proportions by rating['proportion(in %)'].round(decimals=2)
          proportions by rating
Out [102...
               rating total_movies proportion(in %)
          1
               PG-13
                             435
                                       38.024476
```

	rating	total_movies	proportion(in %)
2	R	431	37.674825
3	PG	178	15.559441
4	Not Rated	59	5.157343
5	G	38	3.321678
6	NC-17	3	0.262238

```
In [103...
```

proportions_by_rating = pd.DataFrame(other_language_movies['movie_board_ratine rating_labels = proportions_by_rating.index
proportions_by_rating['rating'] = rating_labels
proportions_by_rating.index = [i for i in range(1,len(proportions_by_rating)+total = sum(other_language_movies['movie_board_rating_display_name'].value_coproportions_by_rating['total_movies'] = proportions_by_rating['movie_board_ratine_ratine_by_ratine_ratine_board_ratine_board_ratine_by_ratine_ratine_board_r

In [104...

proportions_by_rating['proportion(in %)'].round(decimals=2)
proportions by rating

Out[104...

	rating	total_movies	proportion(in %)
1	Not Rated	24	46.153846
2	R	17	32.692308
3	PG-13	6	11.538462
4	PG	4	7.692308
5	G	1	1.923077

In []: