IMDb Movies Dataset

- 1. How much money should be invested into production
- 2. How long the movie should be
- 3. Which genre should the movie be categorized as
- 4. Best time of year to release the mov

```
In [1]: # import relevant modules
   import numpy as np
   import pandas as pd
   import matplotlib.pyplot as plt
   import seaborn as sns
In [2]: # read in the csv file into a df
   movies_df = pd.read_csv('zippedData/movies.csv')
   movies_df.head()
```

Out[2]:

star	writer	director	votes	score	released	year	genre	rating	name	
Jack Nicholson	Stephen King	Stanley Kubrick	927000.0	8.4	June 13, 1980 (United States)	1980	Drama	R	The Shining	0
Brooke Shields	Henry De Vere Stacpoole	Randal Kleiser	65000.0	5.8	July 2, 1980 (United States)	1980	Adventure	R	The Blue Lagoon	1
Mark Hamill	Leigh Brackett	Irvin Kershner	1200000.0	8.7	June 20, 1980 (United States)	1980	Action	PG	Star Wars: Episode V - The Empire Strikes Back	2
Robert Hays	Jim Abrahams	Jim Abrahams	221000.0	7.7	July 2, 1980 (United States)	1980	Comedy	PG	Airplane!	3
Chevy Chase	Brian Doyle- Murray	Harold Ramis	108000.0	7.3	July 25, 1980 (United States)	1980	Comedy	R	Caddyshack	4

In [3]: # general info on the data - check datatypes and columns with missing rows
 movies_df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 7668 entries, 0 to 7667 Data columns (total 15 columns): Non-Null Count Dtype Column 7668 non-null 0 name object 1 rating 7591 non-null object 2 genre 7668 non-null object 3 year 7668 non-null int64 4 released 7666 non-null object 5 7665 non-null float64 score float64 6 votes 7665 non-null 7 director 7668 non-null object writer 7665 non-null object 7667 non-null 9 star object 10 country 7665 non-null object 11 budget 5497 non-null float64 12 gross 7479 non-null float64 13 company 7651 non-null object 14 runtime 7664 non-null float64 dtypes: float64(5), int64(1), object(9) memory usage: 898.7+ KB

In [4]: # describe method for columns with numerical data
movies_df.describe()

Out[4]:

	year	score	votes	budget	gross	runtime
count	7668.000000	7665.000000	7.665000e+03	5.497000e+03	7.479000e+03	7664.000000
mean	2000.405451	6.390411	8.810850e+04	3.558988e+07	7.850054e+07	107.261613
std	11.153508	0.968842	1.633238e+05	4.145730e+07	1.657251e+08	18.581247
min	1980.000000	1.900000	7.000000e+00	3.000000e+03	3.090000e+02	55.000000
25%	1991.000000	5.800000	9.100000e+03	1.000000e+07	4.532056e+06	95.000000
50%	2000.000000	6.500000	3.300000e+04	2.050000e+07	2.020576e+07	104.000000
75%	2010.000000	7.100000	9.300000e+04	4.500000e+07	7.601669e+07	116.000000
max	2020.000000	9.300000	2.400000e+06	3.560000e+08	2.847246e+09	366.000000

Data-cleaning

```
In [5]: # drop the missing values without a budget
movies_df.dropna(inplace=True)
```

In [6]: # checking dataframe again to make sure there are no longer any missing val
we lost about 2200 movies because they did not specify a production budge
movies_df.info()

<class 'pandas.core.frame.DataFrame'> Int64Index: 5421 entries, 0 to 7652 Data columns (total 15 columns): Non-Null Count Dtype # Column _____ 5421 non-null 0 name object 1 rating 5421 non-null object 2 5421 non-null object genre 3 year 5421 non-null int64 4 released 5421 non-null object 5 float64 score 5421 non-null 6 votes 5421 non-null float64 7 director 5421 non-null object 5421 non-null 8 writer object 9 star 5421 non-null object 5421 non-null object 10 country 11 budget 5421 non-null float64 12 gross 5421 non-null float64 13 company 5421 non-null object 14 runtime 5421 non-null float64 dtypes: float64(5), int64(1), object(9) memory usage: 677.6+ KB

In [7]: movies_df.describe()

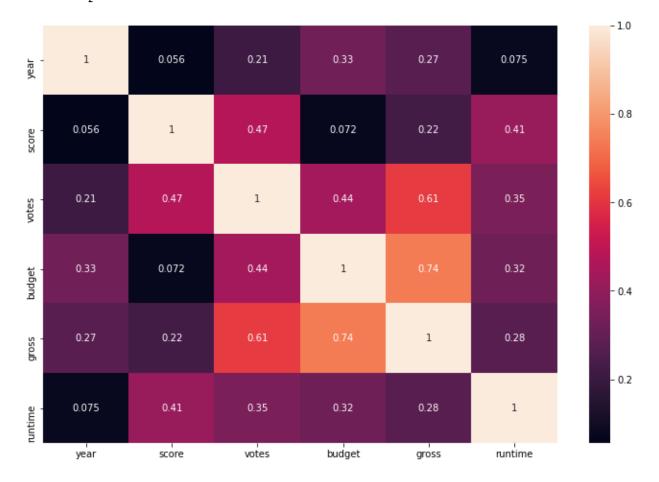
Out[7]:

	year	score	votes	budget	gross	runtime
count	5421.000000	5421.000000	5.421000e+03	5.421000e+03	5.421000e+03	5421.000000
mean	2001.661317	6.392455	1.150661e+05	3.600917e+07	1.032297e+08	108.152739
std	10.540908	0.962730	1.844073e+05	4.157337e+07	1.873027e+08	18.115698
min	1980.000000	1.900000	1.950000e+02	6.000000e+03	3.090000e+02	63.000000
25%	1993.000000	5.800000	1.900000e+04	1.000000e+07	1.073640e+07	95.000000
50%	2002.000000	6.500000	5.200000e+04	2.180000e+07	3.686941e+07	105.000000
75%	2011.000000	7.100000	1.290000e+05	4.500000e+07	1.124625e+08	118.000000
max	2020.000000	9.300000	2.400000e+06	3.560000e+08	2.847246e+09	271.000000

Correlation

In [8]: # visualize the correlation between columns with a heatmap using seaborn
 plt.figure(figsize=(12,8))
 sns.heatmap(movies_df.corr(), annot=True)

Out[8]: <AxesSubplot:>



Which month has the highest revenue?

```
In [9]: # get a look at the last 10 rows of the data
# all recent movies that released during 2020
movies_df.tail(5)
```

Out[9]:

	name	rating	genre	year	released	score	votes	director	writer	star	CC
7648	Bad Boys for Life	R	Action	2020	January 17, 2020 (United States)	6.6	140000.0	Adil El Arbi	Peter Craig	Will Smith	l !
7649	Sonic the Hedgehog	PG	Action	2020	February 14, 2020 (United States)	6.5	102000.0	Jeff Fowler	Pat Casey	Ben Schwartz	l !
7650	Dolittle	PG	Adventure	2020	January 17, 2020 (United States)	5.6	53000.0	Stephen Gaghan	Stephen Gaghan	Robert Downey Jr.	l ;
7651	The Call of the Wild	PG	Adventure	2020	February 21, 2020 (United States)	6.8	42000.0	Chris Sanders	Michael Green	Harrison Ford	C
7652	The Eight Hundred	Not Rated	Action	2020	August 28, 2020 (United States)	6.8	3700.0	Hu Guan	Hu Guan	Zhi- zhong Huang	

The 'released' column is a string object containing parentheses with the country the film was released in

- Will perform data-cleaning to isolate the date on its own and convert to a pandas datetime object

```
In [10]: # new column to remove the country of origin from the release date
         movies df['release'] = movies df['released'].str.split('(').str[0]
         movies df['release']
Out[10]: 0
                     June 13, 1980
         1
                      July 2, 1980
                     June 20, 1980
         2
                      July 2, 1980
         3
                     July 25, 1980
                  January 17, 2020
         7648
                 February 14, 2020
         7649
         7650
                  January 17, 2020
         7651
                 February 21, 2020
                   August 28, 2020
         7652
         Name: release, Length: 5421, dtype: object
```

- In [11]: # create a new column called 'release_date' that will be stored as a dateti
 movies_df['release_date'] = pd.to_datetime(movies_df['release'])
- In [12]: # now drop the released and release columns since we have our release_date
 movies_df.drop(['released', 'release'], axis=1, inplace=True)
- In [13]: # get a look at the movies_df now that we created a new release_date column
 movies_df

Out[13]:

	name	rating	genre	year	score	votes	director	writer	star	coun
0	The Shining	R	Drama	1980	8.4	927000.0	Stanley Kubrick	Stephen King	Jack Nicholson	Uni Kingd
1	The Blue Lagoon	R	Adventure	1980	5.8	65000.0	Randal Kleiser	Henry De Vere Stacpoole	Brooke Shields	Uni Sta
2	Star Wars: Episode V - The Empire Strikes Back	PG	Action	1980	8.7	1200000.0	Irvin Kershner	Leigh Brackett	Mark Hamill	Uni Sta
3	Airplane!	PG	Comedy	1980	7.7	221000.0	Jim Abrahams	Jim Abrahams	Robert Hays	Uni Sta
4	Caddyshack	R	Comedy	1980	7.3	108000.0	Harold Ramis	Brian Doyle- Murray	Chevy Chase	Uni Sta
7648	Bad Boys for Life	R	Action	2020	6.6	140000.0	Adil El Arbi	Peter Craig	Will Smith	Uni Sta
7649	Sonic the Hedgehog	PG	Action	2020	6.5	102000.0	Jeff Fowler	Pat Casey	Ben Schwartz	Uni Sta
7650	Dolittle	PG	Adventure	2020	5.6	53000.0	Stephen Gaghan	Stephen Gaghan	Robert Downey Jr.	Uni Sta
7651	The Call of the Wild	PG	Adventure	2020	6.8	42000.0	Chris Sanders	Michael Green	Harrison Ford	Cana
7652	The Eight Hundred	Not Rated	Action	2020	6.8	3700.0	Hu Guan	Hu Guan	Zhi- zhong Huang	Ch

5421 rows × 15 columns

To get a chart of the average movie gross by month, I first have to create a month column. This 'month' column will be extracted from the recently created release_date column

```
In [14]: # create a month column
movies_df['month'] = movies_df['release_date'].dt.month
movies_df
```

Out[14]:

	name	rating	genre	year	score	votes	director	writer	star	cour
0	The Shining	R	Drama	1980	8.4	927000.0	Stanley Kubrick	Stephen King	Jack Nicholson	Uni Kingd
1	The Blue Lagoon	R	Adventure	1980	5.8	65000.0	Randal Kleiser	Henry De Vere Stacpoole	Brooke Shields	Uni Sta
2	Star Wars: Episode V - The Empire Strikes Back	PG	Action	1980	8.7	1200000.0	Irvin Kershner	Leigh Brackett	Mark Hamill	Uni Sta
3	Airplane!	PG	Comedy	1980	7.7	221000.0	Jim Abrahams	Jim Abrahams	Robert Hays	Uni Sta
4	Caddyshack	R	Comedy	1980	7.3	108000.0	Harold Ramis	Brian Doyle- Murray	Chevy Chase	Uni Sta
7648	Bad Boys for Life	R	Action	2020	6.6	140000.0	Adil El Arbi	Peter Craig	Will Smith	Uni Sta
7649	Sonic the Hedgehog	PG	Action	2020	6.5	102000.0	Jeff Fowler	Pat Casey	Ben Schwartz	Uni Sta
7650	Dolittle	PG	Adventure	2020	5.6	53000.0	Stephen Gaghan	Stephen Gaghan	Robert Downey Jr.	Uni Sta
7651	The Call of the Wild	PG	Adventure	2020	6.8	42000.0	Chris Sanders	Michael Green	Harrison Ford	Cana
7652	The Eight Hundred	Not Rated	Action	2020	6.8	3700.0	Hu Guan	Hu Guan	Zhi- zhong Huang	Ch

5421 rows × 16 columns

```
In [15]: # create a condensed dataframe that will be used to hold the release month
month_release = movies_df['release_date'].dt.month
month_release = pd.DataFrame(month_release)
```

In [16]: month_release

Out[16]:

	release_date
0	6
1	7
2	6
3	7
4	7
7648	1
7649	2
7650	1
7651	2
7652	8

5421 rows × 1 columns

```
In [17]: # create a 'gross' column in our condensed dataframe
# the 'gross' column will be the same as the original 'gross' column from t
month_release['gross'] = movies_df['gross']
month_release
```

Out[17]:

	release_date	gross
0	6	46998772.0
1	7	58853106.0
2	6	538375067.0
3	7	83453539.0
4	7	39846344.0
7648	1	426505244.0
7649	2	319715683.0
7650	1	245487753.0
7651	2	111105497.0
7652	8	461421559.0

5421 rows × 2 columns

```
In [18]: # use groupby method to get the average 'gross' by each month
    mean_revenue = month_release.groupby('release_date').mean()
    mean_revenue
```

Out[18]:

gross

release_date

- **1** 6.263414e+07
- 2 7.407725e+07
- 3 8.771098e+07
- **4** 7.240579e+07
- 5 1.697053e+08
- 6 1.672293e+08
- 7 1.519452e+08
- 8 6.736884e+07
- **9** 5.390261e+07
- 10 5.996056e+07
- 11 1.333690e+08
- 12 1.544489e+08

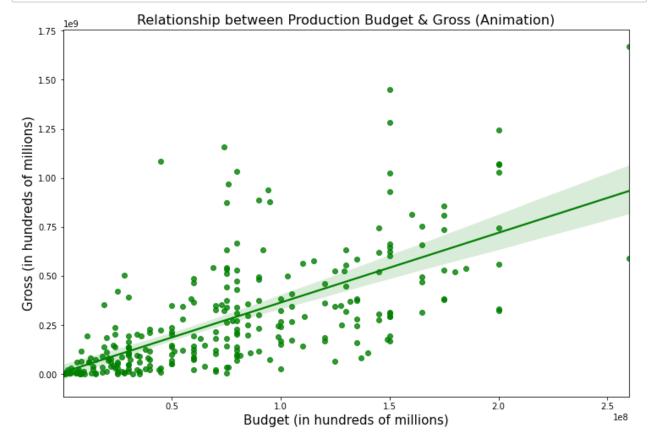
Visualizations

Production budget vs. Gross

```
In [19]: # using a regression plot to analyze relationship between production budget
plt.figure(figsize=(12,8))

# use seaborn to create the regression plot
sns.regplot(x=movies_df[movies_df['genre']=='Animation']['budget'],y=movies

# add title and axes labels
plt.title('Relationship between Production Budget & Gross (Animation)', fon
plt.xlabel('Budget (in hundreds of millions)', fontsize=15)
plt.ylabel('Gross (in hundreds of millions)', fontsize=15)
plt.savefig('Graphs/Animation_regplot.png')
```

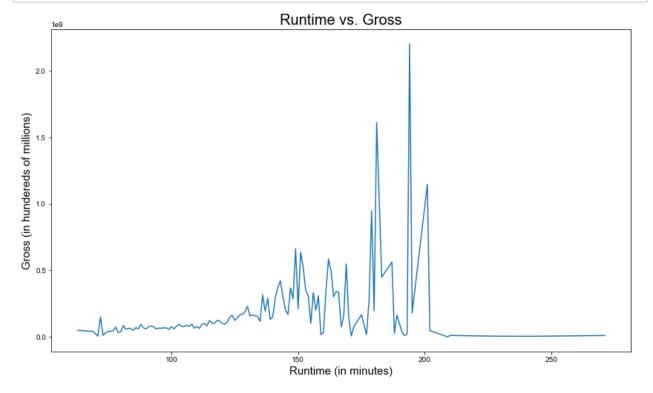


Runtime vs. Gross

```
In [20]: # create a lineplot using seaborn to visualize relationship between runtime
    plt.figure(figsize=(14, 8))
    sns.lineplot(x=movies_df['runtime'],y=movies_df['gross'], ci=None)

# title and axes labels
    sns.set_style('dark')
    plt.title('Runtime vs. Gross', fontsize=20)
    plt.xlabel('Runtime (in minutes)', fontsize=15)
    plt.ylabel('Gross (in hundereds of millions)', fontsize=15)

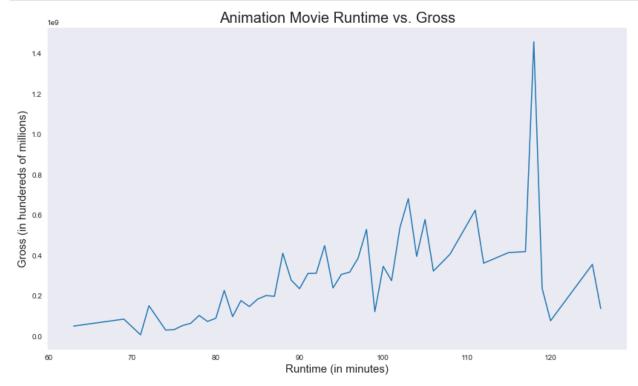
#plt.savefig('Runtime_Gross.png')
    plt.show()
```



Animation Runtime vs. Gross

```
In [21]: # create a lineplot using seaborn to visualize relationship between runtime
plt.figure(figsize=(14, 8))
sns.lineplot(x=movies_df[movies_df['genre']=='Animation']['runtime'],y=movi

# title and axes labels
sns.set_style('dark')
plt.title('Animation Movie Runtime vs. Gross', fontsize=20)
plt.xlabel('Runtime (in minutes)', fontsize=15)
plt.ylabel('Gross (in hundereds of millions)', fontsize=15)
plt.show()
```



Distribution of Runtime

```
In [22]: # distribution plot to visualize the movies by runtime (in minutes)
    plt.figure(figsize=(10,10))
    movie_runs = sns.distplot(movies_df['runtime'], kde=False, bins=25)

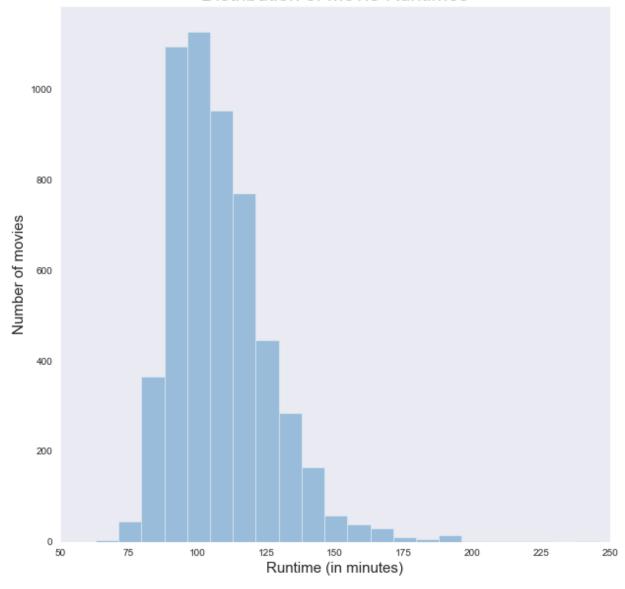
movie_runs.set_xlim(50, 250, 25)
    # add title and axes labels
    plt.title('Distribution of Movie Runtimes', fontsize=20)
    plt.xlabel('Runtime (in minutes)', fontsize=15)
    plt.ylabel('Number of movies', fontsize=15)

#plt.savefig('RuntimeDistribution.png')
    plt.show()
```

/Users/nobletang/opt/anaconda3/envs/learn-env/lib/python3.8/site-package s/seaborn/distributions.py:2551: FutureWarning: `distplot` is a deprecate d function and will be removed in a future version. Please adapt your cod e to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Distribution of Movie Runtimes



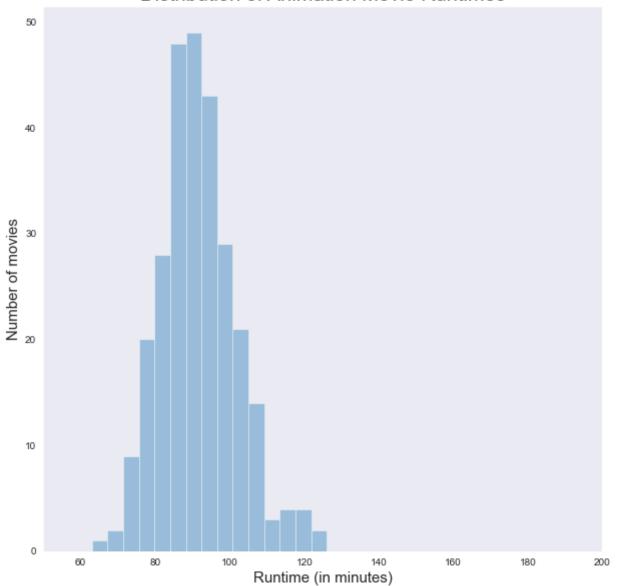
Distribution of Runtime for Animation Movies

```
In [23]: # distribution plot to visualize the animation movies by runtime (in minute
    plt.figure(figsize=(10,10))
    animation = sns.distplot(movies_df[movies_df['genre']=='Animation']['runtim

    animation.set_xlim(50, 200)
    # add title and axes labels
    plt.title('Distribution of Animation Movie Runtimes', fontsize=20)
    plt.xlabel('Runtime (in minutes)', fontsize=15)
    plt.ylabel('Number of movies', fontsize=15)

# save image
#plt.savefig('Graphs/animation_distribution.png')
    plt.show()
```

Distribution of Animation Movie Runtimes

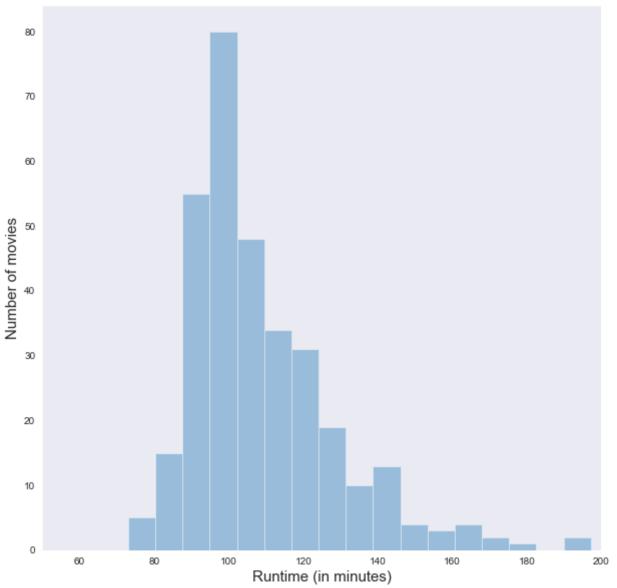


Distribution of Runtime for Adventure Movies

```
In [24]: # distribution plot to visualize the adventure movies by runtime (in minute
    plt.figure(figsize=(10,10))
    adventure = sns.distplot(movies_df[movies_df['genre']=='Adventure']['runtim
    adventure.set_xlim(50, 200)
    # add title and axes labels
    plt.title('Distribution of Adventure Movie Runtimes', fontsize=20)
    plt.xlabel('Runtime (in minutes)', fontsize=15)
    plt.ylabel('Number of movies', fontsize=15)

# save image
#plt.savefig('Graphs/adventure_distribution.png')
    plt.show()
```



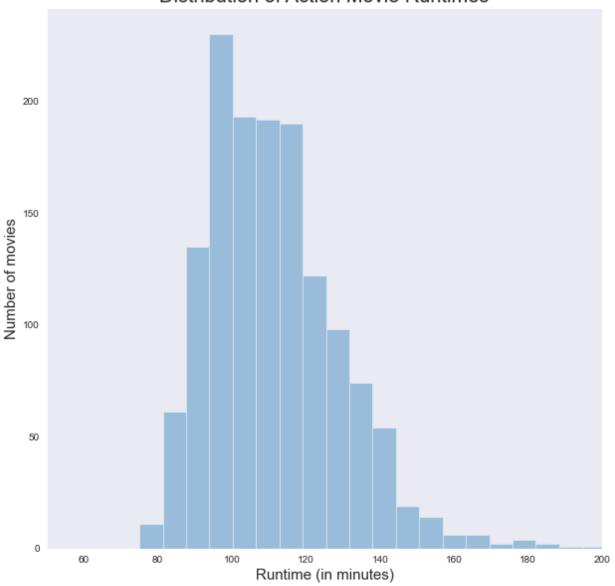


Distribution of Runtime for Action Movies

```
In [25]: # distribution plot to visualize the animation movies by runtime (in minute
    plt.figure(figsize=(10,10))
    action = sns.distplot(movies_df[movies_df['genre']=='Action']['runtime'], k
    action.set_xlim(50,200)
    # add title and axes labels
    plt.title('Distribution of Action Movie Runtimes', fontsize=20)
    plt.xlabel('Runtime (in minutes)', fontsize=15)
    plt.ylabel('Number of movies', fontsize=15)

# save fig
#plt.savefig('Graphs/action_distribution.png')
    plt.show()
```

Distribution of Action Movie Runtimes



Genres by Avg. Gross

```
In [26]: # check the value counts of each genre
         # genres with less than 10 movies may be misleading on a visualization
         movies_df['genre'].value_counts()
Out[26]: Comedy
                       1496
         Action
                       1415
         Drama
                       863
         Crime
                        399
         Adventure
                        327
         Biography
                        311
         Animation
                       277
         Horror
                        251
         Fantasy
                        41
         Mystery
                        17
                         7
         Thriller
         Sci-Fi
                         6
         Romance
                         5
                         4
         Family
                          2
         Western
         Name: genre, dtype: int64
In [27]: # group the dataset by genre and store into a separate variable
         genre groupby = movies df.groupby('genre')
         genre_groupby = genre_groupby.describe()['gross'].sort_values('mean', ascen
         genre_groupby
Out[27]: genre
         Animation
                      2.811044e+08
         Action
                      1.680232e+08
         Adventure
                      1.332682e+08
         Mystery
                      1.178877e+08
         Biography
                      6.139527e+07
         Drama
                      6.036914e+07
         Comedy
                      5.916766e+07
         Horror
                      5.681695e+07
         Crime
                      5.016958e+07
         Thriller
                      4.248814e+07
         Fantasy
                      3.987870e+07
         Sci-Fi
                      3.761010e+07
         Romance
                      3.357538e+07
                      1.001294e+07
         Western
         Name: mean, dtype: float64
In [28]: x = genre groupby.index[:11]
Out[28]: Index(['Animation', 'Action', 'Adventure', 'Mystery', 'Biography', 'Dram
         a',
                 'Comedy', 'Horror', 'Crime', 'Thriller', 'Fantasy'],
               dtype='object', name='genre')
```

```
In [29]: |y = genre_groupby.values[:11]
Out[29]: array([2.81104365e+08, 1.68023229e+08, 1.33268232e+08, 1.17887733e+08,
                6.13952743e+07, 6.03691365e+07, 5.91676588e+07, 5.68169523e+07,
                5.01695794e+07, 4.24881411e+07, 3.98786978e+07])
In [30]: # check the length of x to make sure they match for the barchart
         len(x)
Out[30]: 11
In [31]: # check the length of y to make sure they match the barchart
         len(y)
Out[31]: 11
In [32]: # delete the thriller and its value from both lists because it only has 7 m
         x = list(x)
         x.pop(-2)
Out[32]: 'Thriller'
In [33]: len(x)
Out[33]: 10
In [34]: # do the same for y
         y = list(y)
         y.pop(-2)
Out[34]: 42488141.14285714
In [35]: |len(y)
Out[35]: 10
```

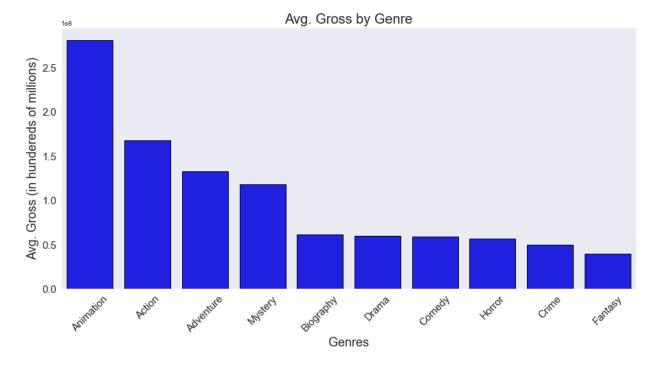
```
In [36]: # barchart to show avg. gross by genre
    # using seaborn module
    plt.figure(figsize=(15,7))
    sns.barplot(x, y, color='blue', edgecolor='black')

# add title and axes labels
    # rotating x-tick labels and editing fontsize so that is is readable
    plt.title('Avg. Gross by Genre', fontsize=20)
    plt.xlabel('Genres', fontsize=18)
    plt.ylabel('Avg. Gross (in hundereds of millions)', fontsize=18)
    plt.xticks(rotation=45, fontsize=15)
    plt.yticks(fontsize=15)

plt.savefig('Graphs/Gross_by_Genre2.png')
    plt.show()
```

/Users/nobletang/opt/anaconda3/envs/learn-env/lib/python3.8/site-package s/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argum ent will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

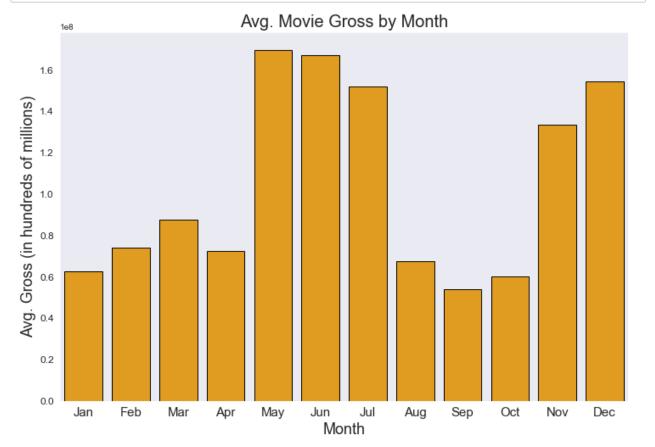


Mean Revenue by Month

```
In [37]: # month column is currently stored as numbers
# hard-coded labels to add as x-tick labels
x_tick_labels = ['Jan','Feb','Mar','Apr','May','Jun','Jul','Aug','Sep','Oct
# using seaborn library to create barchart of avg. gross by month
plt.figure(figsize=(12,8))
sns.barplot(x=x_tick_labels, y=mean_revenue['gross'], color='orange', edgec

# add title and axes labels
plt.title('Avg. Movie Gross by Month', fontsize=20)
plt.xlabel('Month', fontsize=18)
plt.ylabel('Avg. Gross (in hundreds of millions)',fontsize=18)
plt.xticks(fontsize=15)
plt.yticks(fontsize=12)

#plt.savefig('AvgGross_Month.png')
plt.show()
```

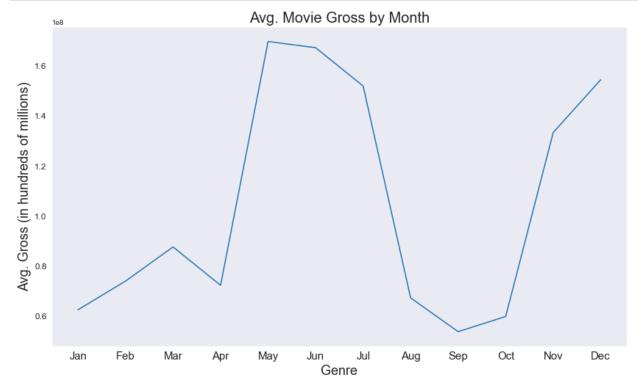


Alternative visualization to the barchart above. Instead, visualizing the same data using a lineplot with seaborn

```
In [38]: # lineplot using seaborn
plt.figure(figsize=(14, 8))
sns.lineplot(x=x_tick_labels,y=mean_revenue['gross'], ci=None)

# title and axes labels
sns.set_style('darkgrid')
plt.title('Avg. Movie Gross by Month', fontsize=20)
plt.xlabel('Genre', fontsize=18)
plt.ylabel('Avg. Gross (in hundreds of millions)',fontsize=18)
plt.xticks(fontsize=15)
plt.yticks(fontsize=12)

#plt.savefig('AvgGross_Month_line.png')
plt.show()
```



```
In [39]: movies_df[movies_df['genre']=='Animation']['runtime'].mean()
```

Out[39]: 91.57039711191335