Final Project Submission

Please fill out:

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• Scheduled project review date/time: 15-10-2023

• Instructor name: Hardik Idnani

• Blog post URL:

In [2]:

▶ # First import the packages we will use in this project

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from matplotlib import style

Out[3]:

	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
146139	tt9916538	Kuambil Lagi Hatiku	Kuambil Lagi Hatiku	2019	123.0	Drama
146140	tt9916622	Rodolpho Teóphilo - O Legado de um Pioneiro	Rodolpho Teóphilo - O Legado de um Pioneiro	2015	NaN	Documentary
146141	tt9916706	Dankyavar Danka	Dankyavar Danka	2013	NaN	Comedy
146142	tt9916730	6 Gunn	6 Gunn	2017	116.0	NaN
146143	tt9916754	Chico Albuquerque - Revelações	Chico Albuquerque - Revelações	2013	NaN	Documentary

146144 rows × 6 columns

Out[4]:

	tconst	averagerating	numvotes
0	tt10356526	8.3	31
1	tt10384606	8.9	559
2	tt1042974	6.4	20
3	tt1043726	4.2	50352
4	tt1060240	6.5	21
73851	tt9805820	8.1	25
73852	tt9844256	7.5	24
73853	tt9851050	4.7	14
73854	tt9886934	7.0	5
73855	tt9894098	6.3	128

73856 rows × 3 columns

Out[5]:

	tconst	primary_title	original_title	start_year	runtime_minutes	genres	averagerating	numvotes
0	tt0063540	Sunghursh	Sunghursh	2013	175.0	Action,Crime,Drama	7.0	77
1	tt0066787	One Day Before the Rainy Season	Ashad Ka Ek Din	2019	114.0	Biography,Drama	7.2	43
2	tt0069049	The Other Side of the Wind	The Other Side of the Wind	2018	122.0	Drama	6.9	4517
3	tt0069204	Sabse Bada Sukh	Sabse Bada Sukh	2018	NaN	Comedy,Drama	6.1	13
4	tt0100275	The Wandering Soap Opera	La Telenovela Errante	2017	80.0	Comedy,Drama,Fantasy	6.5	119
73851	tt9913084	Diabolik sono io	Diabolik sono io	2019	75.0	Documentary	6.2	6
73852	tt9914286	Sokagin Çocuklari	Sokagin Çocuklari	2019	98.0	Drama,Family	8.7	136
73853	tt9914642	Albatross	Albatross	2017	NaN	Documentary	8.5	8
73854	tt9914942	La vida sense la Sara Amat	La vida sense la Sara Amat	2019	NaN	NaN	6.6	5
73855	tt9916160	Drømmeland	Drømmeland	2019	72.0	Documentary	6.5	11

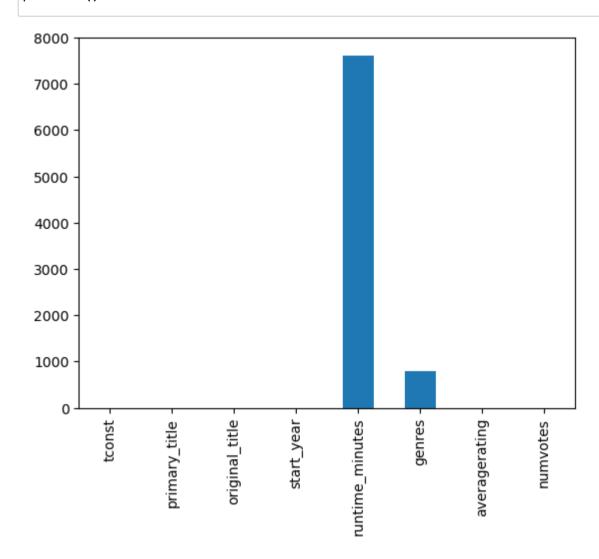
73856 rows × 8 columns

```
    data.info()

In [6]:
            <class 'pandas.core.frame.DataFrame'>
           Int64Index: 73856 entries, 0 to 73855
           Data columns (total 8 columns):
                Column
                                 Non-Null Count Dtype
               ----
                                 73856 non-null object
                tconst
                primary_title
                                 73856 non-null object
                original title 73856 non-null object
                start year
                                 73856 non-null int64
                runtime_minutes 66236 non-null float64
                genres
                                 73052 non-null object
                averagerating
                                 73856 non-null float64
                numvotes
                                 73856 non-null int64
           dtypes: float64(2), int64(2), object(4)
           memory usage: 5.1+ MB
        # We need to see if we have any missing data
In [7]:
           data.isna().any()
   Out[7]: tconst
                              False
           primary title
                              False
           original title
                              False
           start year
                              False
           runtime minutes
                               True
           genres
                               True
           averagerating
                              False
                              False
           numvotes
           dtype: bool
```

dtype: int64

In [9]: data.isna().sum().plot.bar()
 plt.show()



```
#missing data in persentage
In [10]:
            missing data =data.isna().sum()*100/len(data)
            missing_data
   Out[10]: tconst
                                0.000000
            primary title
                                0.000000
            original title
                                0.000000
             start year
                                0.000000
             runtime_minutes
                               10.317374
                                1.088605
             genres
             averagerating
                                0.000000
             numvotes
                                0.000000
             dtype: float64
In [11]:
          #now we need to drop all missing values
             data.dropna(axis = 0, inplace = True)

    data.info()

In [12]:
             <class 'pandas.core.frame.DataFrame'>
            Int64Index: 65720 entries, 0 to 73855
             Data columns (total 8 columns):
                                  Non-Null Count Dtype
                 Column
                -----
                 tconst
                                  65720 non-null object
                 primary title
                                  65720 non-null object
                 original title 65720 non-null object
                 start year
                                  65720 non-null int64
                 runtime minutes 65720 non-null float64
                 genres
                                  65720 non-null object
                                  65720 non-null float64
                 averagerating
                  numvotes
                                  65720 non-null int64
             dtypes: float64(2), int64(2), object(4)
            memory usage: 4.5+ MB
```

```
    data.isna().sum()

In [13]:
   Out[13]: tconst
                                0
             primary_title
                                0
             original title
             start_year
                                0
             runtime_minutes
                                0
             genres
                                0
             averagerating
                                0
             numvotes
             dtype: int64
In [14]:
         # now we need to find duplicate values in data set
             data.duplicated()
   Out[14]: 0
                      False
             1
                      False
             2
                      False
                      False
                      False
                      . . .
             73849
                      False
             73850
                      False
             73851
                      False
             73852
                      False
             73855
                      False
             Length: 65720, dtype: bool
In [15]:
          data.duplicated().any()
   Out[15]: False
          ▶ data.columns
In [16]:
   Out[16]: Index(['tconst', 'primary_title', 'original_title', 'start_year',
                    'runtime_minutes', 'genres', 'averagerating', 'numvotes'],
                   dtype='object')
```

Out[20]:

	tconst	title	start_year	runtime_minutes	genres	averagerating	numvotes
0	tt0063540	Sunghursh	2013	175.0	Action,Crime,Drama	7.0	77
1	tt0066787	Ashad Ka Ek Din	2019	114.0	Biography,Drama	7.2	43
2	tt0069049	The Other Side of the Wind	2018	122.0	Drama	6.9	4517
4	tt0100275	La Telenovela Errante	2017	80.0	Comedy,Drama,Fantasy	6.5	119
6	tt0137204	Joe Finds Grace	2017	83.0	Adventure, Animation, Comedy	8.1	263
73849	tt9911774	Padmavyuhathile Abhimanyu	2019	130.0	Drama	8.4	365
73850	tt9913056	Swarm Season	2019	86.0	Documentary	6.2	5
73851	tt9913084	Diabolik sono io	2019	75.0	Documentary	6.2	6
73852	tt9914286	Sokagin Çocuklari	2019	98.0	Drama,Family	8.7	136
73855	tt9916160	Drømmeland	2019	72.0	Documentary	6.5	11

65720 rows × 7 columns

```
df3= pd.read_csv('zippedData/bom.movie_gross.csv.gz')
          df3.head()
          df3.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 3387 entries, 0 to 3386
          Data columns (total 5 columns):
              Column
                           Non-Null Count Dtype
           --- -----
                            _____
           0 title
                           3387 non-null object
           1 studio
                           3382 non-null object
           2 domestic_gross 3359 non-null float64
           3 foreign_gross 2037 non-null object
                            3387 non-null int64
              vear
           4
          dtypes: float64(1), int64(1), object(3)
          memory usage: 132.4+ KB
In [22]: ► #merge two datasets
          big_data = pd.merge(data,df3, on= 'title')
```

```
In [23]:  big_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2400 entries, 0 to 2399
Data columns (total 11 columns):
                     Non-Null Count Dtype
    Column
    -----
    tconst
                     2400 non-null
                                     object
 1
    title
                     2400 non-null
                                     object
                     2400 non-null
                                    int64
    start_year
     runtime_minutes 2400 non-null
                                    float64
    genres
                     2400 non-null
                                     object
                     2400 non-null
                                     float64
     averagerating
    numvotes
                     2400 non-null
                                    int64
    studio
                     2397 non-null
                                     obiect
    domestic gross 2382 non-null
                                    float64
    foreign_gross
                     1535 non-null
                                     object
                     2400 non-null
 10 year
                                     int64
```

dtypes: float64(3), int64(3), object(5)

memory usage: 225.0+ KB

In [24]: ► big_data

Out[24]:

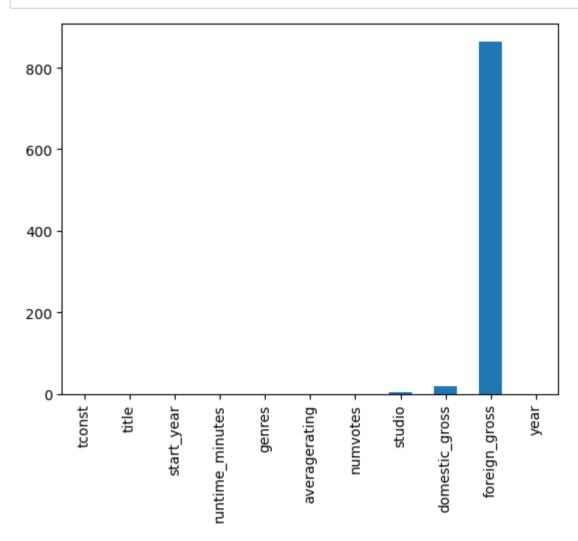
	tconst	title	start_year	runtime_minutes	genres	averagerating	numvotes	studio	domestic_gross	foreiç
0	tt0315642	Wazir	2016	103.0	Action,Crime,Drama	7.1	15378	Relbig.	1100000.0	
1	tt0337692	On the Road	2012	124.0	Adventure,Drama,Romance	6.1	37886	IFC	744000.0	
2	tt4339118	On the Road	2014	89.0	Drama	6.0	6	IFC	744000.0	
3	tt5647250	On the Road	2016	121.0	Drama	5.7	127	IFC	744000.0	
4	tt0359950	The Secret Life of Walter Mitty	2013	114.0	Adventure,Comedy,Drama	7.3	275300	Fox	58200000.0	12
									•••	
2395	tt8097306	Nobody's Fool	2018	110.0	Comedy,Drama,Romance	4.6	3618	Par.	31700000.0	
2396	tt8108198	Andhadhun	2018	139.0	Crime, Thriller	8.5	43409	Eros	1200000.0	
2397	tt8427036	Helicopter Eela	2018	135.0	Drama	5.4	673	Eros	72000.0	
2398	tt8549902	Oolong Courtyard	2018	103.0	Comedy	4.6	61	CL	37700.0	
2399	tt9151704	Burn the Stage: The Movie	2018	84.0	Documentary,Music	8.8	2067	Trafalgar	4200000.0	,

2400 rows × 11 columns

```
In [25]: ▶ # We need to see if we have any missing data
            big_data.isna().sum()
   Out[25]: tconst
                                 0
            title
                                 0
```

start_year 0 runtime_minutes 0 genres averagerating 0 0 numvotes studio 3 domestic_gross 18 foreign_gross 865 0 year dtype: int64

In [26]: #missing data in bar graph
big_data.isna().sum().plot.bar()
plt.show()



```
In [27]: ▶ #now we need to drop all missing values
          big_data.dropna(axis = 0, inplace = True)
          big_data.isna().sum()
   Out[27]: tconst
                          0
          title
                          0
          start_year
                          0
          runtime_minutes
                          0
                          0
          genres
          averagerating
          numvotes
                          0
          studio
                          0
          domestic_gross
                          0
          foreign_gross
                          0
          year
                          0
          dtype: int64
big_data.duplicated().any()
```

Out[28]: False

```
#to change genres object to list
In [29]:
             genres list = []
             for i in big_data['genres']:
                 genres list.append(i.split(','))
             genres list
   Out[29]: [['Adventure', 'Drama', 'Romance'],
              ['Drama'],
              ['Drama'],
              ['Adventure', 'Comedy', 'Drama'],
              ['Action', 'Crime', 'Drama'],
              ['Action', 'Adventure', 'Sci-Fi'],
              ['Action', 'Crime', 'Drama'],
              ['Action', 'Comedy', 'Crime'],
              ['Comedy', 'Drama'],
              ['Comedy', 'Family'],
              ['Adventure', 'Animation', 'Comedy'],
              ['Action', 'Adventure', 'Sci-Fi'],
              ['Action', 'Sci-Fi', 'Thriller'],
              ['Comedy'],
              ['Action', 'Adventure', 'Thriller'],
              ['Horror', 'Mystery', 'Thriller'],
              ['Action', 'Drama', 'Family'],
              ['Drama', 'Romance', 'Sci-Fi'],
              ['Adventure', 'Animation', 'Comedy'],
```

In [30]: big_data['genres'] = genres_list
big_data.head()

Out[30]:

	tconst	title	start_year	runtime_minutes	genres	averagerating	numvotes	studio	domestic_gross	foreign_gross	year
1	tt0337692	On the Road	2012	124.0	[Adventure, Drama, Romance]	6.1	37886	IFC	744000.0	8000000	2012
2	tt4339118	On the Road	2014	89.0	[Drama]	6.0	6	IFC	744000.0	8000000	2012
3	tt5647250	On the Road	2016	121.0	[Drama]	5.7	127	IFC	744000.0	8000000	2012
4	tt0359950	The Secret Life of Walter Mitty	2013	114.0	[Adventure, Comedy, Drama]	7.3	275300	Fox	58200000.0	129900000	2013
5	tt0365907	A Walk Among the Tombstones	2014	114.0	[Action, Crime, Drama]	6.5	105116	Uni.	26300000.0	26900000	2014

In [31]: #change each list of genres to individuals
big_data= big_data.explode('genres')
big_data

Out[31]:

On the Road Nobody's	2012 2012 2012 2014 2016 	124.0 124.0 124.0 89.0 121.0	Adventure Drama Romance Drama Drama	6.1 6.1 6.0 5.7	37886 37886 37886 6 127	IFC IFC IFC IFC	744000.0 744000.0 744000.0 744000.0	8000000 8000000 8000000 8000000	2012 2012 2012
Road On the Road On the Road On the Road On the Road Nobody's	2012 2014 2016 	124.0 89.0 121.0 	Romance Drama Drama	6.1 6.0 5.7	37886 6 127	IFC IFC	744000.0 744000.0 744000.0	8000000 8000000	2012 2012
Road On the Road On the Road Nobody's	2014 2016 	89.0 121.0 	Drama Drama	6.0 5.7	6	IFC	744000.0 744000.0	8000000	2012
Road On the Road Nobody's	2016	121.0 	Drama	5.7	127	IFC	744000.0		
Road Nobody's								8000000	2012
Nobody's									
	2018						•••		
Fool	2010	110.0	Comedy	4.6	3618	Par.	31700000.0	1800000	2018
Nobody's Fool	2018	110.0	Drama	4.6	3618	Par.	31700000.0	1800000	2018
Nobody's Fool	2018	110.0	Romance	4.6	3618	Par.	31700000.0	1800000	2018
Burn the Stage: The Movie	2018	84.0	Documentary	8.8	2067	Trafalgar	4200000.0	16100000	2018
Burn the	2018	84.0	Music	8.8	2067	Trafalgar	4200000.0	16100000	2018
	Stage: The Movie Burn the Stage: The	Stage: 2018 The Movie Burn the Stage: 2018	Stage: 2018 84.0 The Movie 84.0 Burn the Stage: 2018 84.0 The 84.0	Stage: 2018 84.0 Documentary Movie Burn the Stage: 7he 2018 84.0 Music	Stage: 2018 84.0 Documentary 8.8 Movie Burn the Stage: Music 8.8	Stage: 2018 84.0 Documentary 8.8 2067 Movie Burn the Stage: August 2018 Burn Music 8.8 2067	Stage: 2018 84.0 Documentary 8.8 2067 Trafalgar Movie Burn the Stage: The 2018 84.0 Music 8.8 2067 Trafalgar	Stage: 2018 84.0 Documentary 8.8 2067 Trafalgar 4200000.0 Burn the Stage: The 2018 84.0 Music 8.8 2067 Trafalgar 4200000.0	Stage: 2018 84.0 Documentary 8.8 2067 Trafalgar 4200000.0 16100000 Burn the Stage: The 2018 84.0 Music 8.8 2067 Trafalgar 4200000.0 16100000

3777 rows × 11 columns

In [32]: #to get overall statistics about data
big_data.describe()

Out[32]:

	start_year	runtime_minutes	averagerating	numvotes	domestic_gross	year
count	3777.000000	3777.000000	3777.000000	3.777000e+03	3.777000e+03	3777.000000
mean	2013.682023	108.950755	6.434525	1.186741e+05	5.926911e+07	2013.757744
std	2.638891	18.948308	0.977512	1.652693e+05	8.534017e+07	2.620473
min	2010.000000	3.000000	1.600000	5.000000e+00	7.000000e+02	2010.000000
25%	2011.000000	96.000000	5.900000	1.758500e+04	4.800000e+06	2011.000000
50%	2014.000000	106.000000	6.500000	6.104600e+04	3.100000e+07	2014.000000
75%	2016.000000	119.000000	7.100000	1.448210e+05	7.320000e+07	2016.000000
max	2019.000000	184.000000	9.200000	1.841066e+06	7.001000e+08	2018.000000

```
#to plot numvotes and genres
In [33]:
            round(big_data.groupby('genres')['numvotes'].mean().sort_values(ascending = False) , 2)
   Out[33]: genres
            Sci-Fi
                           284557.14
            Western
                           211835.67
            Adventure
                           197881.45
            Action
                           183216.10
            Fantasy
                           151937.22
            Mystery
                           127260.49
            Animation
                           123762.24
            Thriller
                           120452.02
            Crime
                           111856.59
            Biography
                           102643.64
            Comedy
                            92426.49
            Drama
                            90286.30
            History
                            80219.52
                            76821.69
            Horror
            Music
                            75045.27
            Romance
                            70723.04
            Sport
                            69910.83
            Family
                            68003.96
```

Musical

Documentary

War

News

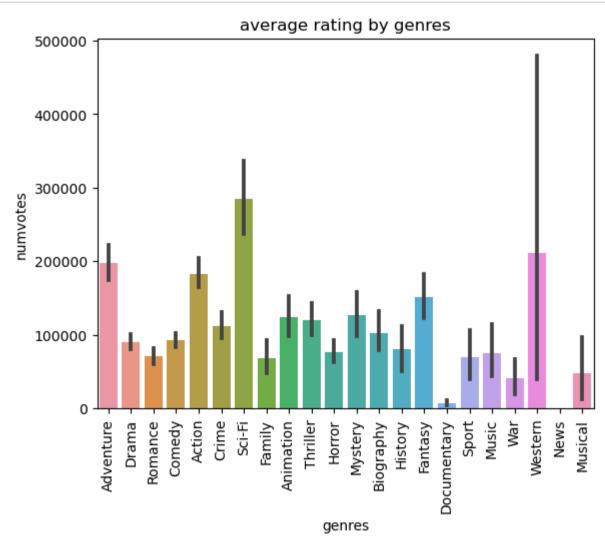
47826.33

41220.43

Name: numvotes, dtype: float64

7517.60 8.00

```
In [34]: N sns.barplot(y = 'numvotes' , x = 'genres', data = big_data)
plt.xticks(rotation='vertical')
plt.title('average rating by genres')
plt.show()
```



```
In [35]: ▶ #to plot average rating and genres
            round(big_data.groupby('genres')['averagerating'].mean().sort_values(ascending = False) , 2)
   Out[35]: genres
            Biography
                           7.04
            Documentary
                           7.03
            Sport
                           6.94
                           6.89
            War
                           6.87
            History
                           6.62
            Drama
            Animation
                           6.62
            Music
                           6.59
            Sci-Fi
                           6.54
            Adventure
                           6.51
            Crime
                           6.44
            News
                           6.40
            Western
                           6.38
            Romance
                           6.36
            Action
                           6.36
                           6.26
            Comedy
            Thriller
                           6.20
            Mystery
                           6.19
```

6.17

6.12

6.11

5.73 Name: averagerating, dtype: float64

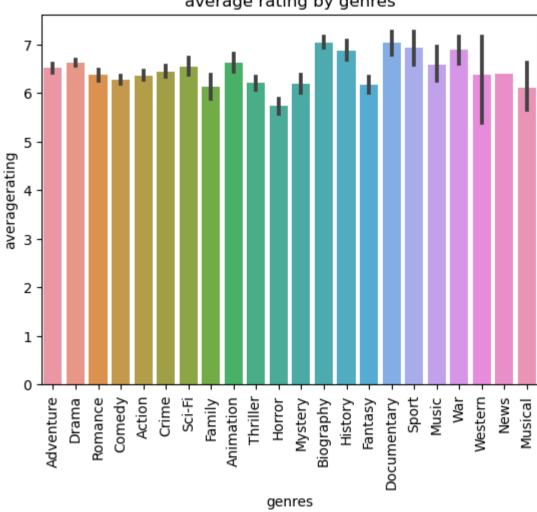
Fantasy

Family Musical

Horror

```
In [36]:
             sns.barplot(y = 'averagerating' , x = 'genres', data = big_data)
             plt.xticks(rotation='vertical')
             plt.title('average rating by genres')
             plt.show()
```





```
In [37]:
          # highest average gross as per genres
             big_data['foreign_gross'] = [float(str(i).replace(",", "")) for i in big_data['foreign_gross']]
             big_data['total_gross'] = big_data['foreign_gross'] + big_data['domestic_gross']

▶ big data.groupby('genres')['total gross'].mean().sort values(ascending=False)

In [38]:
   Out[38]: genres
             Animation
                           3.736899e+08
             Adventure
                            3.389146e+08
             Sci-Fi
                           3.287856e+08
             Action
                            2.384007e+08
             Fantasy
                           2.187723e+08
             Family
                           1.655285e+08
             Comedy
                           1.352298e+08
             Musical
                           1.191816e+08
             Thriller
                           1.186467e+08
             Sport
                           1.176932e+08
             Western
                           1.173332e+08
                           1.013936e+08
             Horror
             Mystery
                           9.139734e+07
             History
                           8.261405e+07
             Biography
                           8.155982e+07
             Music
                           7.877163e+07
             Crime
                           7.740127e+07
                           7.715433e+07
             Drama
             Documentary
                           7.493128e+07
```

News

War

Romance

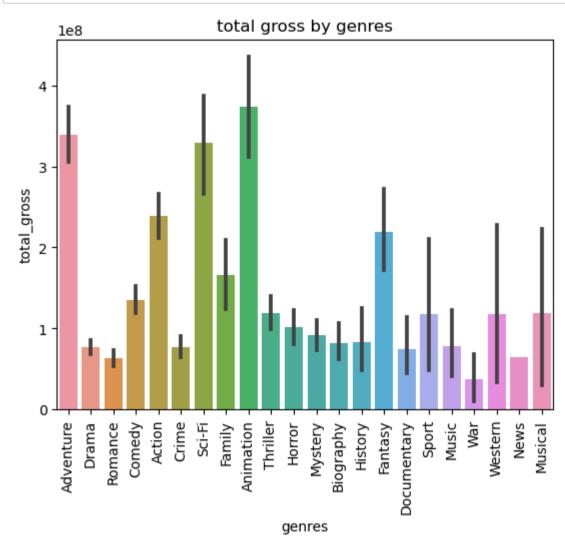
6.460000e+07

6.299393e+07

3.772676e+07

Name: total gross, dtype: float64

```
In [39]: N sns.barplot(x='genres' , y= 'total_gross', data = big_data)
plt.xticks(rotation='vertical')
plt.title('total gross by genres')
plt.show()
```



```
In [75]: #using groupby method create numvotes, totla_gross, averagerating using genres
grouped = big_data.groupby('genres')['numvotes', 'total_gross', 'averagerating'].mean()
grouped.reset_index()
```

```
C:\Users\vasun\AppData\Local\Temp\ipykernel_20664\2137210240.py:2: FutureWarning: Indexing with multiple keys (i
mplicitly converted to a tuple of keys) will be deprecated, use a list instead.
  grouped = big_data.groupby('genres')['numvotes', 'total_gross', 'averagerating'].mean()
```

Out[75]:

	genres	numvotes	total_gross	averagerating
0	Action	183216.101010	2.384007e+08	6.361364
1	Adventure	197881.454545	3.389146e+08	6.506667
2	Animation	123762.244898	3.736899e+08	6.618367
3	Biography	102643.641892	8.155982e+07	7.036486
4	Comedy	92426.491429	1.352298e+08	6.263619
5	Crime	111856.591928	7.740127e+07	6.435874
6	Documentary	7517.602564	7.493128e+07	7.029487
7	Drama	90286.300261	7.715433e+07	6.618668
8	Family	68003.962963	1.655285e+08	6.120988
9	Fantasy	151937.222222	2.187723e+08	6.168519
10	History	80219.516667	8.261405e+07	6.871667
11	Horror	76821.687075	1.013936e+08	5.727211
12	Music	75045.272727	7.877163e+07	6.586364
13	Musical	47826.333333	1.191816e+08	6.111111
14	Mystery	127260.491935	9.139734e+07	6.187903
15	News	8.000000	6.460000e+07	6.400000
16	Romance	70723.037915	6.299393e+07	6.363507
17	Sci-Fi	284557.137615	3.287856e+08	6.544954
18	Sport	69910.828571	1.176932e+08	6.937143
19	Thriller	120452.015326	1.186467e+08	6.197701
20	War	41220.428571	3.772676e+07	6.892857
21	Western	211835.666667	1.173332e+08	6.377778

In [52]: ▶ grouped.describe(include='all')

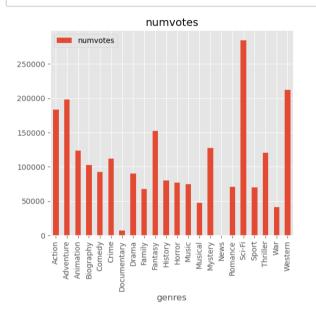
Out[52]:

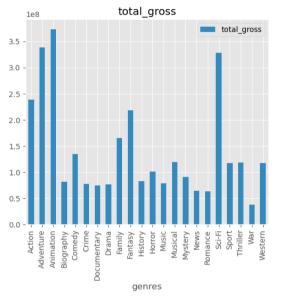
	numvotes	total_gross	averagerating
count	22.000000	2.200000e+01	22.000000
mean	106155.092278	1.410327e+08	6.470829
std	67622.765755	9.679249e+07	0.336945
min	8.000000	3.772676e+07	5.727211
25%	70113.880907	7.774386e+07	6.214181
50%	91356.395845	1.093634e+08	6.417937
75%	126385.930176	1.579538e+08	6.618593
max	284557.137615	3.736899e+08	7.036486

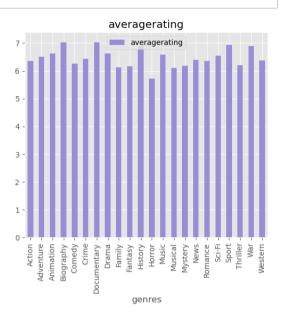
In [41]:

#create plots style.use('ggg

style.use('ggplot')
grouped.plot(subplots=True, layout=(1, 3), figsize=(20, 5), sharex=True , kind= 'bar')
plt.show()







```
In [45]: #to disply highest rating genres with titles
big_data.nlargest(10, 'averagerating')[['title' , 'genres']]
```

Out[45]:

		title	genres
16	32	The Runaways	Adventure
57	75	The Wall	Documentary
45	52	Inception	Action
45	52	Inception	Adventure
45	52	Inception	Sci-Fi
239	9	Burn the Stage: The Movie	Documentary
239	9	Burn the Stage: The Movie	Music
92	28	Eyes Wide Open	Documentary
92	28	Eyes Wide Open	History
9	94	Interstellar	Adventure

```
#to disply highest numvotes genres with titles
In [61]:
              big data.nlargest(10, 'numvotes')[['title' , 'numvotes']]
    Out[61]:
                                   title numvotes
                452
                                          1841066
                               Inception
                452
                                          1841066
                               Inception
                452
                               Inception
                                          1841066
                    The Dark Knight Rises
                                          1387769
                429 The Dark Knight Rises
                                          1387769
                 94
                              Interstellar
                                          1299334
                 94
                              Interstellar
                                          1299334
                 94
                              Interstellar
                                          1299334
                        Django Unchained
                1127
                                          1211405
                        Django Unchained
                1127
                                          1211405
           #to disply highest total gross genres with titles
In [87]:
              big data[big data['total gross'].max() == big data['total gross']][['title', 'genres']]
    Out[87]:
                                    title
                                            genres
               1563 Avengers: Age of Ultron
                                            Action
               1563 Avengers: Age of Ultron Adventure
               1563 Avengers: Age of Ultron
                                             Sci-Fi
           #to calculate the number of votes garnered by the 70% movie.
In [81]:
              big_data['numvotes'].quantile(0.70)
```

Out[81]: 119789.0

```
#to display the movies (title, runtime) longer than 30 minutes or shorter than 360 minutes.
 In [97]:
               big data[(big data['runtime minutes'] <30) | (big data['runtime minutes']>360)][['title' , 'genres']]
     Out[97]:
                         title
                                  genres
                293 Limitless
                                Biography
                293 Limitless Documentary
In [104]:
            #to display the movies (title, number of votes) that received specified number of votes.
               n = 5000
               big data[big data['numvotes'] >= n][['title', 'numvotes']].sort values(by='numvotes', ascending=False)
    Out[104]:
                                       title numvotes
                                             1841066
                 452
                                   Inception
                 452
                                             1841066
                                   Inception
                 452
                                   Inception
                                             1841066
                        The Dark Knight Rises
                 429
                                             1387769
                         The Dark Knight Rises
                 429
                                             1387769
                                                  ...
                2057 My Little Pony: The Movie
                                                5133
                      My Little Pony: The Movie
                                                5133
                2057 My Little Pony: The Movie
                                                5133
                1378
                               I Still See You
                                                5010
                1378
                               I Still See You
                                                5010
               3297 rows × 2 columns
  In [ ]:
            H
```

In []: 🕨	1