Analysis of IMDB Data

We will analyze a subset of IMDB's actors, genres, movie actors, and movie ratings data. This dataset comes to us from Kaggle (https://www.kaggle.com/datasets/ashirwadsangwan/imdb-dataset) although we have taken steps to pull this data into a publis s3 bucket:

- s3://cis9760-lecture9-movieanalysis/name.basics.tsv ---> (actors)
- s3://cis9760-lecture9-movieanalysis/title.basics.tsv ---> (genres)
- s3://cis9760-lecture9-movieanalysis/title.principals.tsv ---> (movie actors)
- s3://cis9760-lecture9-movieanalysis/title.ratings.tsv ---> (movie ratings)

Content

name.basics.tsv.gz - Contains the following information for names:

nconst (string) - alphanumeric unique identifier of the name/person.

primaryName (string)- name by which the person is most often credited.

birthYear - in YYYY format.

deathYear – in YYYY format if applicable, else.

primaryProfession (array of strings)— the top-3 professions of the person.

knownForTitles (array of tconsts) – titles the person is known for.

title.basics.tsv.gz - Contains the following information for titles:

tconst (string) - alphanumeric unique identifier of the title.

titleType (string) – the type/format of the title (e.g. movie, short, tyseries, typeisode, video, etc). primaryTitle (string) – the more popular title / the title used by the filmmakers on promotional materials at the point of release.

originalTitle (string) - original title, in the original language.

isAdult (boolean) - 0: non-adult title; 1: adult title.

startYear (YYYY) – represents the release year of a title. In the case of TV Series, it is the series start year.

endYear (YYYY) – TV Series end year. for all other title types.

runtimeMinutes – primary runtime of the title, in minutes.

genres (string array) – includes up to three genres associated with the title.

title.principals.tsv - Contains the principal cast/crew for titles:

tconst (string) - alphanumeric unique identifier of the title.

ordering (integer) – a number to uniquely identify rows for a given titleld.

nconst (string) - alphanumeric unique identifier of the name/person.

category (string) - the category of job that person was in. job (string) - the specific job title if applicable, else.

characters (string) - the name of the character played if applicable, else.

title.ratings.tsv.gz - Contains the IMDb rating and votes information for titles:

tconst (string) - alphanumeric unique identifier of the title.

averageRating – weighted average of all the individual user ratings.

numVotes - number of votes the title has received.

PART 1 - Installation and Initial Setup

Begin by installing the necessary libraries that you may need to conduct your analysis. At the very least, you must install pandas and matplotlib

In [1]:	<pre>sc.list_packages()</pre>							
	VBox() Starting Spark application							
	ID	YARN Application	ID Kind	State	Spark UI	Driver log	Current session?	
	1 application	_1668909776641_00	002 pyspark	idle	Link	Link	✓	
	FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'), SparkSession available as 'spark'. FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),							
	Package		Version					
	beautifulsou	ip4	4.9.1					
	boto	•	2.49.0					
	click		7.1.2					
	jmespath		0.10.0					
	joblib		0.16.0					
	lxml		4.5.2					
	mysqlclient		1.4.2					
	nltk		3.5					
	nose		1.3.4					
	numpy		1.16.5					
	pip		9.0.1					
	py-dateutil		2.2					
	python37-sag	emaker-pyspark	1.4.0					
	pytz		2020.1					
	PyYAML		5.3.1					
	regex		2020.7.14					
	setuptools		28.8.0					
	six		1.13.0					
	soupsieve		1.9.5					
	tqdm		4.48.2					
	wheel		0.29.0					
	windmill		1.6					

Let's install the necessary packages here

```
In [2]: sc.install pypi package("pandas==1.0.3")
        sc.install pypi package("matplotlib==3.2.1")
        VBox()
        FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
        ght='25px', width='50%'),...
        Collecting pandas==1.0.3
          Using cached https://files.pythonhosted.org/packages/4a/6a/94b219b8ea0f2d580169e85e
        d1edc0163743f55aaeca8a44c2e8fc1e344e/pandas-1.0.3-cp37-cp37m-manylinux1 x86 64.whl
        Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.7/site-packages
        (from pandas==1.0.3)
        Requirement already satisfied: numpy>=1.13.3 in /usr/local/lib64/python3.7/site-packa
        ges (from pandas==1.0.3)
        Collecting python-dateutil>=2.6.1 (from pandas==1.0.3)
          Using cached https://files.pythonhosted.org/packages/36/7a/87837f39d0296e723bb9b62b
        bb257d0355c7f6128853c78955f57342a56d/python_dateutil-2.8.2-py2.py3-none-any.whl
        Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/site-packages (fr
        om python-dateutil>=2.6.1->pandas==1.0.3)
        Installing collected packages: python-dateutil, pandas
        Successfully installed pandas-1.0.3 python-dateutil-2.8.2
        Collecting matplotlib==3.2.1
          Using cached https://files.pythonhosted.org/packages/b2/c2/71fcf957710f3ba1f09088b3
        5776a799ba7dd95f7c2b195ec800933b276b/matplotlib-3.2.1-cp37-cp37m-manylinux1 x86 64.wh
        Requirement already satisfied: python-dateutil>=2.1 in /mnt/tmp/1668917922725-0/lib/p
        ython3.7/site-packages (from matplotlib==3.2.1)
        Collecting pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 (from matplotlib==3.2.1)
          Using cached https://files.pythonhosted.org/packages/6c/10/a7d0fa5baea8fe7b50f448ab
        742f26f52b80bfca85ac2be9d35cdd9a3246/pyparsing-3.0.9-py3-none-any.whl
        Collecting cycler>=0.10 (from matplotlib==3.2.1)
          Using cached https://files.pythonhosted.org/packages/5c/f9/695d6bedebd747e5eb0fe8fa
        d57b72fdf25411273a39791cde838d5a8f51/cycler-0.11.0-py3-none-any.whl
        Requirement already satisfied: numpy>=1.11 in /usr/local/lib64/python3.7/site-package
        s (from matplotlib==3.2.1)
        Collecting kiwisolver>=1.0.1 (from matplotlib==3.2.1)
          Using cached https://files.pythonhosted.org/packages/ab/8f/8dbe2d4efc4c0b08ec67d6ef
        b7cc31fbfd688c80afad85f65980633b0d37/kiwisolver-1.4.4-cp37-cp37m-manylinux_2_5_x86_6
        4.manylinux1 x86 64.whl
        Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/site-packages (fr
        om python-dateutil>=2.1->matplotlib==3.2.1)
        Collecting typing-extensions; python_version < "3.8" (from kiwisolver>=1.0.1->matplot
        lib==3.2.1)
          Using cached https://files.pythonhosted.org/packages/0b/8e/f1a0a5a76cfef77e1eb6004c
        b49e5f8d72634da638420b9ea492ce8305e8/typing_extensions-4.4.0-py3-none-any.whl
        Installing collected packages: pyparsing, cycler, typing-extensions, kiwisolver, matp
        lotlib
        Successfully installed cycler-0.11.0 kiwisolver-1.4.4 matplotlib-3.2.1 pyparsing-3.0.
        9 typing-extensions-4.4.0
        Now, import the installed packages from the previous block below.
```

```
import numpy as np
In [3]:
        import matplotlib.pyplot as plt
```

```
from pyspark.sql.functions import approx_count_distinct, avg, collect_set, countDistin
VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
ght='25px', width='50%'),...
```

Loading Data

Load all data from S3 into a Spark dataframe object

```
In [4]: actors = spark.read.csv('s3://cis9760-lecture9-movieanalysis/name.basics.tsv', sep=r'\
genres = spark.read.csv('s3://cis9760-lecture9-movieanalysis/title.basics.tsv', sep=r'\
movie_actors = spark.read.csv('s3://cis9760-lecture9-movieanalysis/title.principals.ts\
movie_ratings = spark.read.csv('s3://cis9760-lecture9-movieanalysis/title.ratings.tsv'\)

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

Actors

Display the schema below:

Display the first 5 rows with the following columns:

- primaryName
- birthYear
- deathYear
- knownForTitles

```
In [6]: actors.select("primaryName","birthYear","deathYear","knownForTitles").show(5)

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

Genres

Display the first 10 rows with the following columns:

- titleType
- primaryTitle
- genres

```
genres.select("titleType", "primaryTitle", "genres").show(10)
In [7]:
       VBox()
       FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
       ght='25px', width='50%'),...
       +----+
                    primaryTitle|
                                                  genres |
        |titleType|
                           Carmencita
                                       Documentary, Short
            short
            short|Le clown et ses c...|
                                         Animation, Short
                      Pauvre Pierrot Animation, Comedy,...
            short
                          Un bon bock
                                         Animation, Short
            short|
                    Blacksmith Scene
                                            Comedy, Short
            short|
            short
                    Chinese Opium Den
                                                   Short
            short | Corbett and Court... |
                                             Short, Sport
            short Edison Kinetoscop...
                                       Documentary, Short
                                                 Romance
            movie
                           Miss Jerry
                                        Documentary, Short
            short | Exiting the Factory
       only showing top 10 rows
```

Display the unique categories below:

```
In [8]: genres.select('titleType').distinct().show()

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

```
titleType|
tvSeries|
tvMiniSeries|
wovie|
videoGame|
tvSpecial|
video|
tvMovie|
tvEpisode|
tvShort|
short|
```

Display the schema below:

Movie Actors

|-- genres: string (nullable = true)

Display the schema below:

FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),... +----+ tconst|ordering| nconst| category job| characters| \N|["Herself"]| self| |tt0000001| 1|nm1588970| 2|nm0005690| \N| |tt0000001| director N|tt0000001| 3|nm0374658|cinematographer|director of photo...| \N| 1 nm0721526 director \N| |tt0000002| |tt0000002| 2|nm1335271| composer \N| N\N| |tt0000003| 1 nm0721526 director N|tt0000003| 2 nm5442194 producer producer \N| 3 nm1335271 |tt0000003| composer /N/ N|tt0000003| 4 nm5442200 editor \N| \N|

director

\N|

\N|

only showing top 10 rows

|tt0000004|

1|nm0721526|

+----+---+----

Movie Ratings

Display the schema below:

```
movie_ratings.orderBy(col('numVotes').desc()).show(10)
In [13]:
         VBox()
         FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
         ght='25px', width='50%'),...
         +----+
            tconst|averageRating|numVotes|
         |tt7430722|
                            6.8
                                    9999
         |tt4445154|
                            8.1
                                    9997
         |tt2229907|
                            6.3
                                    9996
         |tt0294097|
                            8.0
                                    9994
                            6.5
         |tt0264734|
                                    9993
         |tt8860450|
                            6.3
                                    9991
         |tt2032572|
                            5.2
                                    9991
                            8.4
                                     999
         |tt0664505|
         |tt7508752|
                            7.9
                                     999
         |tt1077089|
                            7.3
                                     999
```

+----+

only showing top 10 rows

Overview of Data

Display the number of rows and columns in each dataFrame object.

```
actors cols = len(actors.columns)
In [14]:
         print(f"Number of columns in Actors table: {actors_cols}")
         actors rows = actors.count()
         print(f"Number of rows in Actors table: {actors rows}", "\n")
         genres cols = len(genres.columns)
         print(f"Number of columns in Genres table: {genres cols}")
         genres_rows = genres.count()
         print(f"Number of rows in Genres table: {genres_rows}", "\n")
         movieactors_cols = len(movie_actors.columns)
         print(f"Number of columns in Movie Actors table: {movieactors_cols}")
         movieactors rows = movie actors.count()
         print(f"Number of rows in Movie Actors table: {movieactors_rows}", "\n")
         movieratings cols = len(movie ratings.columns)
         print(f"Number of columns in Movie Ratings table: {movieratings cols}")
         movieratings_rows = movie_ratings.count()
         print(f"Number of rows in Movie Ratings table: {movieratings rows}", "\n")
         VBox()
         FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
         ght='25px', width='50%'),...
         Number of columns in Actors table: 6
         Number of rows in Actors table: 9706922
         Number of columns in Genres table: 9
         Number of rows in Genres table: 6321302
         Number of columns in Movie Actors table: 6
         Number of rows in Movie Actors table: 36468817
         Number of columns in Movie Ratings table: 3
         Number of rows in Movie Ratings table: 993153
```

PART 2 - Analyzing Genres

Let's now answer this question: how many unique genres are represented in this dataset?

Essentially, we have the genres per movie as a list - this is useful to quickly see what each movie might be represented as but it is difficult to easily answer questions such as:

- How many movies are categorized as Comedy, for instance?
- What are the top 20 most popular genres available?

Association Table

We need to "break out" these genres from the tconst? One common approach to take is to build an association table mapping a single tconst multiple times to each distinct genre.

For instance, given the following:

tconst	titleType	genres	
abcd123	XXX	a,b,c	

We would like to derive something like:

tconst	titleType	genre
abcd123	XXX	a
abcd123	XXX	b
abcd123	XXX	С

What this does is allow us to then perform a myriad of rollups and other analysis on this association table which can aid us in answering the questions asked above.

Implement the code necessary to derive the table described from the data set

```
genres.select('tconst', 'titleType', 'genres').show(5)
In [15]:
       VBox()
       FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
       ght='25px', width='50%'),...
       +----+
          tconst|titleType|
                                   genres
       +----+
       |tt0000001|
                  short| Documentary, Short|
       |tt0000002|
                   short | Animation, Short
       |tt0000003| short|Animation,Comedy,...|
       |tt0000004| short| Animation, Short|
                              Comedy, Short
       |tt0000005|
                   short
       +----+
       only showing top 5 rows
```

Display the first 10 rows of your association table below

```
In [16]: association_table = genres.select("tconst","titleType",split(col("genres"),",").alias(
    association_table2=association_table.select("tconst","titleType",explode(association_t
    association_table2.show(10)

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
    ght='25px', width='50%'),...
```

```
+----+
                      genres
   tconst|titleType|
|tt0000001|
            short | Documentary |
|tt0000001|
            short
                       Short
            short | Animation
|tt0000002|
|tt0000002|
            short
                       Short
|tt0000003|
            short | Animation
|tt0000003|
            short
                     Comedy
|tt0000003|
            short| Romance|
            short | Animation
|tt0000004|
|tt0000004|
            short
                       Short
|tt0000005|
            short
                      Comedy
+-----+
only showing top 10 rows
```

Total Unique Genres

What is the total number of unique genres available in the movie category?

What are the unique genres available?

```
In [18]: association_table2.select('genres').distinct().show()

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

```
genres
    Mystery|
    Musical|
      Sport
     Action|
  Talk-Show
    Romance
   Thriller|
         \N|
 Reality-TV
     Family
    Fantasy
    History|
  Animation|
  Film-Noir
      Short
     Sci-Fi|
       News
      Drama
|Documentary|
    Western
+----+
only showing top 20 rows
```

Oops! Something is off!

```
In [19]: nll = '\\N'
   association_table3 = association_table2.select('genres').filter(col('genres') != nll).
   association_table3.show()

VBox()
   FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

In [20]:

```
+----+
      genres
     Mystery|
     Musical|
       Sport|
      Action|
   Talk-Show
     Romance
    Thriller
  Reality-TV
      Family|
     Fantasy|
     History|
   Animation|
       Short
   Film-Noir
      Sci-Fi
        News
       Drama
|Documentary|
     Western
      Comedy
+----+
only showing top 20 rows
association table3.select(countDistinct("genres")).show()
VBox()
ght='25px', width='50%'),...
```

```
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
+----+
```

|count(DISTINCT genres)| +----+ 28 +----+

Top Genres by Movies

Now let's find the highest rated genres in this dataset by rolling up genres.

Average Rating / Genre

So now, let's unroll our distinct count a bit and display the per average rating value of per genre.

The expected output should be:

genre	averageRating
a	8.5
b	6.3
С	7.2

Or something to that effect.

First, let's join our two dataframes (movie ratings and genres) by tconst

```
In [21]: nll = ' \ N'
        association table4 = association table2.filter(col('genres') != nll)
        top_genres = association_table4.join(movie_ratings).where(association_table4['tconst']
        top_genres2 = top_genres.select('genres','averageRating')
        top genres2.show(10)
        VBox()
        FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
        ght='25px', width='50%'),...
        +----+
             genres|averageRating|
        +----+
        |Documentary|
                             5.1
                             5.1
              Short|
        |Documentary|
                             5.2
              Short
                             5.2
              Comedy
                             5.2
              Short
                             5.2
                             6.0
              Comedy
              Horror
                             6.0
              Short
                             6.0
        |Documentary|
                             4.9
        +----+
        only showing top 10 rows
```

Now, let's aggregate along the averageRating column to get a resultant dataframe that displays average rating per genre.

```
In [22]: top_genres3 = top_genres2.withColumn('averageRating', col('averageRating').cast('Float
top_genres4 = top_genres3.groupBy('genres').agg(avg("averageRating").alias("averageRat
top_genres4.show()

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
ght='25px', width='50%'),...
```

```
genres
                  averageRating
    Mystery | 7.215679898056961 |
     Musical | 6.544660195495522|
      Action | 6.951029447217718 |
       Sport | 6.995047307520108 |
  Talk-Show | 6.598412160498378
     Romance | 6.784248168750174 |
    Thriller 6.312686083353854
 Reality-TV | 6.8388670073012765 |
      Family | 6.989731263527516
     Fantasy | 7.093731237046438 |
     History | 7.304633665280264 |
  Animation | 7.046786062925208 |
  Film-Noir 6.636246780503378
       Short | 6.799363502678704
      Sci-Fi | 6.747496253901592
        News | 6.467539497212473 |
       Drama 7.018453637663789
|Documentary| 7.241740584659479|
     Western 7.109783420299982
      Comedy | 6.919198969550161 |
+-----+
only showing top 20 rows
```

Horizontal Bar Chart of Top Genres

With this data available, let us now build a barchart of all genres

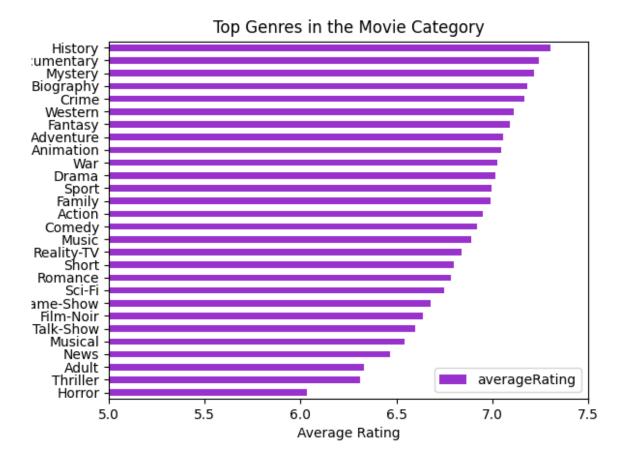
HINT: don't forget about the matplotlib magic!

%matplot plt

```
In [23]: top_genres4.sort(col('averageRating').desc()).show()

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

```
genres
                         averageRating
         +----+
              History | 7.304633665280264 |
         |Documentary| 7.241740584659479|
              Mystery | 7.215679898056961|
            Biography | 7.180114836990299 |
                Crime 7.165008360689229
              Western | 7.109783420299982 |
              Fantasy | 7.093731237046438 |
            Adventure 7.0567301815847685
            Animation | 7.046786062925208 |
                  War | 7.026155360390018 |
                Drama 7.018453637663789
                Sport | 6.995047307520108 |
               Family | 6.989731263527516
               Action | 6.951029447217718 |
               Comedy | 6.919198969550161 |
                Music | 6.890572244550894
           Reality-TV | 6.8388670073012765 |
                Short | 6.799363502678704 |
              Romance 6.784248168750174
               Sci-Fi | 6.747496253901592
         +----+
         only showing top 20 rows
        hbchart=top genres4.select('genres','averageRating').sort(col('averageRating')).toPand
In [24]:
         hbchart.plot.barh(x='genres', y='averageRating', color='darkorchid')
         plt.xlim([5.0,7.5])
         plt.xlabel('Average Rating')
         plt.ylabel('Genre')
         plt.title('Top Genres in the Movie Category')
         plt.legend(loc=4)
         %matplot plt
         VBox()
         FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
         ght='25px', width='50%'),...
```



PART 3 - Analyzing Job Categories

Total Unique Job Categories

What is the total number of unique job categories?

```
movie_actors.select('tconst','category').show(5)
In [25]:
        VBox()
        FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
        ght='25px', width='50%'),...
         +----+
            tconst
                         category
         |tt0000001|
                             self
         |tt0000001|
                         director
         |tt0000001|cinematographer|
         |tt0000002|
                         director|
         |tt0000002|
                         composer
         +----+
        only showing top 5 rows
        movie_actors.select(countDistinct('category')).show()
In [26]:
        VBox()
```

What are the unique job categories available?

```
movie actors.select('category').distinct().show()
FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
ght='25px', width='50%'),...
  -----+
           category
            actress
           producer
|production_designer|
             writer
              actor
    cinematographer|
      archive sound
    archive footage
               self|
             editor
           composer|
           director|
           ----+
```

Top Job Categories

Now let's find the top job categories in this dataset by rolling up categories.

Counts of Titles / Job Category

The expected output should be:

category	count
а	15
b	2
С	45

Or something to that effect.

```
In [28]: movie_actors.groupBy('category').count().show()

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei ght='25px', width='50%'),...
```

Bar Chart of Top Job Categories

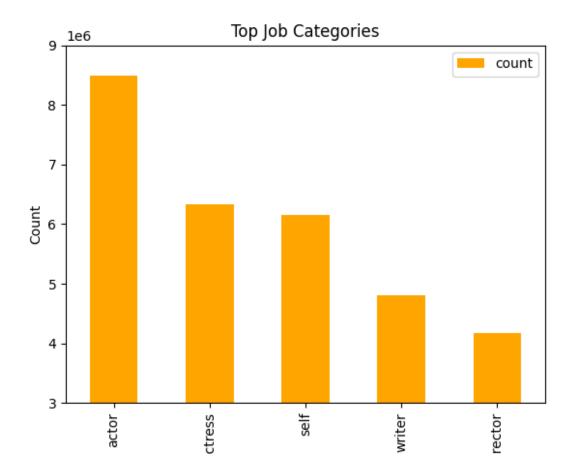
With this data available, let us now build a barchart of the top 5 categories.

HINT: don't forget about the matplotlib magic!

%matplot plt

```
movie_actors2 = movie_actors.groupBy('category').count().sort(col('count').desc())
In [29]:
         movie actors2.show()
         VBox()
         FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
         ght='25px', width='50%'),...
         +----+
                    category | count |
         +----+
                       actor | 8493701 |
                     actress | 6325097 |
                        self | 6153089 |
                      writer | 4811596 |
                    director 4179106
                    producer | 2197866 |
                    composer | 1313187 |
              cinematographer | 1300404 |
                      editor | 1197669 |
         |production designer| 285924|
              archive footage | 209035|
               archive_sound 2143
            ----+
        bctjc=movie actors2.limit(5).toPandas()
In [30]:
         bctjc.plot.bar(x='category', y='count', color='orange')
         plt.ylim([3000000 , 9000000])
         plt.xlabel('Job Categories')
         plt.ylabel('Count')
         plt.title('Top Job Categories')
         plt.legend(loc=1)
         %matplot plt
```

VBox()
FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
ght='25px', width='50%'),...



PART 4 - Answer to the following questions:

1) Find all the "movies" featuring "Johnny Depp" and "Helena Bonham Carter".

First join actors, genres, and movie actors on each other

```
In [31]: jd_hbc = actors.join(movie_actors).where(actors['nconst'] == movie_actors['nconst'])
    jd_hbc2 = jd_hbc.join(genres).where(jd_hbc['tconst'] == genres['tconst']).select('prim
    jd_hbc3 = jd_hbc2.select('primaryTitle','primaryName','titleType').filter("titleType =
    jd_hbc4 = jd_hbc3.select('primaryTitle','primaryName').filter("primaryName=='Johnny De
    jd_hbc4.groupBy('primaryTitle').count().where("count>1").drop('count').show(truncate=F
    VBox()
    FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
```

ght='25px', width='50%'),...

2) Find all the "movies" featuring "Brad Pitt" after 2010.

```
In [32]: nll = ' \ N'
        bp = actors.join(movie actors).where(actors['nconst'] == movie actors['nconst'])
        bp2 = bp.join(genres).where(bp['tconst'] == genres['tconst']).select('primaryTitle',
        bp2.select('primaryTitle','startYear').sort(col('startYear').desc()).show(truncate=Fal
        VBox()
        FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
        ght='25px', width='50%'),...
        +----+
        primaryTitle
                                       startYear
        +----+
        Babylon
                                       2021
         Kajillionaire
                                       2020
         |Irresistible
                                       2020
         Ad Astra
                                       2019
         Once Upon a Time ... in Hollywood 2019
         The King
                                       2019
         Vice
                                       2018
         |War Machine
                                       2017
         Allied
                                       2016
         |Voyage of Time: Life's Journey
                                       2016
         |The Big Short
                                       2015
         By the Sea
                                       2015
         |Hitting the Apex
                                       2015
         Fury
                                       2014
         |World War Z
                                       2013
         |Kick-Ass 2
                                       2013
         |12 Years a Slave
                                       2013
         |Killing Them Softly
                                       2012
         The Tree of Life
                                       2011
         Moneyball
                                       2011
```

3) What is the number of "movies" "acted" by "Zendaya" per year?

```
In [33]: nll = '\\N'
zendaya = actors.join(movie_actors).where(actors['nconst'] == movie_actors['nconst'])
zendaya2 = zendaya.join(genres).where(zendaya['tconst'] == genres['tconst']).select('partyear').count().show(truncate=False)
```

4) What are the "movies" by average rating greater than "9.7" and released in "2019"?

```
movies = genres.join(movie_ratings).where(genres['tconst'] == movie_ratings['tconst'])
movies2 = movies.select('primaryTitle', 'averageRating').filter((movies.titleType=='movies2')
movies3 = movies2.withColumn('averageRating', col('averageRating').cast('Float'))
movies3.sort(col('averageRating').desc()).show(truncate=False)
VBox()
FloatProgress(value=0.0, bar style='info', description='Progress:', layout=Layout(hei
ght='25px', width='50%'),...
+-----
primaryTitle
                                             |averageRating|
+-----
Our Scripted Life
|The Twilight Zone: A 60th Anniversary Celebration | 10.0
A Grunt's Life
|Bu Can Var Oldugu Sürece
                                             10.0
|L'Enfant Terrible
                                             10.0
|Kirket
                                             10.0
A Medicine for the Mind
                                             10.0
|The Butcher Baronet
                                             10.0
Love in Kilnerry
                                             10.0
|Puritan: All of Life to The Glory of God
                                             9.9
|The Cardinal
                                             9.9
Superhombre
                                             19.9
|Kamen Rider Zi-O: Over Quartzer
                                             9.8
                                             9.8
|Square One
Time and motion
                                             9.8
We Shall Not Die Now
                                             9.8
|From Shock to Awe
                                             9.8
Randhawa
                                             9.8
|Gini Helida Kathe
                                             9.8
```

Extra Credit - Analysis of your choice

Try and analyze some interesting dimension to this data. You should specify the question in your Project2_Analysis.ipynb.

You must join at least two datasets.

Top 20 Movie Directors with the highest Weighted Average Ratings released in 2017-2022.

```
In [35]: nll = ' \setminus N'
         actor = genres.join(movie_actors).where(genres['tconst'] == movie_actors['tconst'])
         actor2 = actor.join(actors).where(actor['nconst'] == actors['nconst']).select('primary
         actor3 = actor2.join(movie ratings).where(actor2['tconst'] == movie ratings['tconst'])
         actor4 = actor3.filter((genres.titleType=='movie')&(genres.startYear != nll)&(genres.s
         actor5 = actor4.withColumn('averageRating', col('averageRating').cast('Float')).withCo
         actor6 = actor5.groupby('primaryName').sum('Weighted Score').withColumnRenamed("sum(We
         VBox()
         FloatProgress(value=0.0, bar_style='info', description='Progress:', layout=Layout(hei
         ght='25px', width='50%'),...
         +----+
                         |Weighted Score
         |primaryName
         +----
         Joe Russo
                          1.1287869E7
         Anthony Russo
                         1.1286742E7
         |Jon Watts
                          5174425.25
         David Leitch
                         5025152.625
         |James Mangold
                          4872498.5
         Jordan Peele
                          4410769.125
         |Todd Phillips
                          4108825.75
         |Taika Waititi
                          4096711.0
         Ryan Coogler
                          4024088.5
         James Gunn
                          3883387.25
         |Chad Stahelski | 3872830.125
         Christopher Nolan 3851313.25
         Patty Jenkins
                          3753087.75
         Andy Muschietti | 3746174.5
         Rian Johnson
                          |3439359.900390625|
         |Steven Spielberg | 3220537.5
         |Denis Villeneuve |3199704.0
                         3140792.0
         Bryan Singer
         |Martin McDonagh | 3016148.5
         |Edgar Wright
                          2941025.25
```

only showing top 20 rows