For this project, we will create an anime recommendation system. A user will input his/hers genre of choice and the model will output the top 10 anime of that genre, along with their descriptions.

The Data used for this project consist of about 17.562 anime and the preference from 325.772 different users.

Datasets Description:

The Anime_df has 16214 rows and 5 columns:

- MAL_IDName: the anime id (Primary key)
- · Name: the title of the anime
- Score: the average rating (however we will drop that column)
- · Genres: the genres of each anime entry
- Synopsis: a brief description of the anime.

The Ratings_df has 57633278 rows and 3 columns:

- user_id: Id of the user who left a rating
- anime_id: anime id (foreigh key)
- rating: rating from 1 to 10

The Big Data problem surrounding this project is Volume, because of the large number of rows of the Ratings_df. Moreover, we will perform analysis on batch data.

Setting Up Pyspark

```
# innstall java
!apt-get install openjdk-8-jdk-headless -qq > /dev/null

# install spark (change the version number if needed)

https://colab.research.google.com/drive/1KG3WWZf6rJ8jFB55kDMdJcdGdQc7P8mY#scrollTo=vIYecBPig7id&printMode=true
```

```
!wget -q https://archive.apache.org/dist/spark/spark-3.0.0/spark-3.0.0-bin-hadoop3.2.tgz
# unzip the spark file to the current folder
!tar xf spark-3.0.0-bin-hadoop3.2.tgz
# set your spark folder to your system path environment.
import os
os.environ["JAVA HOME"] = "/usr/lib/jvm/java-8-openjdk-amd64"
os.environ["SPARK HOME"] = "/content/spark-3.0.0-bin-hadoop3.2"
# install findspark using pip
!pip install -q findspark
! pip install pyspark
    Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
     Collecting pyspark
       Downloading pyspark-3.3.1.tar.gz (281.4 MB)
                                                  - 281.4/281.4 MB 2.9 MB/s eta 0:00:00
       Preparing metadata (setup.py) ... done
     Collecting py4j==0.10.9.5
       Downloading py4j-0.10.9.5-py2.py3-none-any.whl (199 kB)
                                               --- 199.7/199.7 KB 13.8 MB/s eta 0:00:00
     Building wheels for collected packages: pyspark
       Building wheel for pyspark (setup.py) ... done
       Created wheel for pyspark: filename=pyspark-3.3.1-py2.py3-none-any.whl size=281845512 sha256=a2fb0778e2b37994376b88
       Stored in directory: /root/.cache/pip/wheels/43/dc/11/ec201cd671da62fa9c5cc77078235e40722170ceba231d7598
     Successfully built pyspark
     Installing collected packages: py4j, pyspark
     Successfully installed py4j-0.10.9.5 pyspark-3.3.1
import findspark
findspark.init()
findspark.find()
     '/content/spark-3.0.0-bin-hadoop3.2'
```

```
from pyspark.context import SparkContext, SparkConf
from pyspark.sql import *
from pyspark.sql import SparkSession
from pyspark.sql import Window
from pyspark.sql import Row
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from pyspark.sql import functions as F
from pyspark.sql.functions import explode
from pyspark.sql.functions import rank
from pyspark.sql.functions import col
from pyspark.sql.functions import avg
from pyspark.sql.functions import split
from pyspark.sql.functions import round
import time
from functools import reduce
spark = SparkSession.builder\
        .master('local[5]')\
        .appName('Anime recommendation system')\
        .config('spark.ui.port', '4050')\
        .getOrCreate()
```

Importing Data

```
Anime_df = spark.read.csv('/content/drive/MyDrive/Colab Notebooks/anime_with_synopsis.csv', header=True, inferSchema=True)
Ratings_df = spark.read.csv('/content/drive/MyDrive/Colab Notebooks/rating_complete.csv', header=True, inferSchema=True)

def withColumnRenamed(existingName: any, newName: any):
    DataFrame
Anime_df = Anime_df.withColumnRenamed("sypnopsis", "synopsis")
Anime_df = Anime_df.withColumnRenamed("sypnopsis", "synopsis")
```

+	+	Score	++ Genres synopsis +
1 5 6 7	Cowboy Bebop: Ten Trigun Witch Hunter Robin	8.39 8.24 7.27	Action, Adventure "In the year 2071 Action, Drama, My other day, anothe Action, Sci-Fi, A "Vash the Stamped Action, Mystery, ches are individu Adventure, Fantas It is the dark ce
only she	+ owing top 5 rows	H	++

Ratings_df.show(5)

+	+	+_	+
user	_id aı	nime_id ra	ting
+	+	+_	+
	0	430	9
	0	1004	5
	0	3010	7
ĺ	0	570	7
ĺ	0	2762	9
+	+	+_	+
only	showi	ng top 5 r	ows

→ Data Processing

For the project, we will use the Anime_df as well as the Ratings_df. We will drop the 'Score' column for the Anime_df as we will create a model that will aggregate the ratings of the Ratings_df and then output the score.

```
Anime df = Anime df.drop('Score')
```

Then we will explode the genre column, so each row have just one gerne

Checking for Null values:

As we can see, Anime_df has 8 null values in the synopsis column, however, we will not drop these rows.

Printing the shapes of the above dataframes:

```
print(f'Anime_df has {Anime_df.count()} rows and {len(Anime_df.columns)} columns')
print(f'Ratings_df has {Ratings_df.count()} rows and {len(Ratings_df.columns)} columns')
Anime_df has 16214 rows and 4 columns
Ratings df has 57633278 rows and 3 columns
```

Because the genre column of the Anime_df consists of multiple genres per anime, we will split them, resulting in each row containing only one genre per anime.

```
# Converting each entry of the 'Genres' column to a list of strings, for them to be exploded
Anime_df = Anime_df.withColumn('Genres', split(Anime_df['Genres'], ', '))
Anime_df.show()
```

+	+		t+
MAL_ID +	Name	Genres	synopsis
1	Cowboy Bebop	[Action, Adventur	"In the year 2071
5	Cowboy Bebop: Ten	[Action, Drama, M	other day, anothe
6	Trigun	[Action, Sci-Fi,	"Vash the Stamped
7	Witch Hunter Robin	[Action, Mystery,	ches are individu
8	Bouken Ou Beet	[Adventure, Fanta	It is the dark ce
15	Eyeshield 21	[Action, Sports,	"Sena is like any
16	Hachimitsu to Clover	[Comedy, Drama, J	Yuuta Takemoto, a
17	Hungry Heart: Wil	[Slice of Life, C	Kyosuke Kano has
18	Initial D Fourth	[Action, Cars, Sp	"Takumi Fujiwara
19	Monster	[Drama, Horror, M	Dr. Kenzou Tenma,
20	Naruto	[Action, Adventur	oments prior to N
21	One Piece	[Action, Adventur	"Gol D. Roger was
22	Tennis no Ouji-sama	[Action, Comedy,	The world of tenn
23	Ring ni Kakero 1	[Action, Shounen,	In order to fulfi
24	School Rumble	[Comedy, Romance,	"Just the words "
25	Sunabouzu	[Action, Adventur	"The Great Kanto
26	Texhnolyze	[Action, Sci-Fi,	"Texhnolyze takes
27	Trinity Blood	[Action, Supernat	Following Armaged
28	Yakitate!! Japan	[Comedy, Shounen]	hile countries su

Explode the 'Genre' column

```
Anime df.printSchema()
    root.
      |-- MAL ID: integer (nullable = true)
      |-- Name: string (nullable = true)
      -- Genres: array (nullable = true)
         |-- element: string (containsNull = true)
      -- synopsis: string (nullable = true)
Anime df = Anime df.select( '*' , explode('Genres').alias('Genres2'))
Anime df = Anime df.drop('Genres')
Anime df = Anime df.withColumnRenamed('Genres2', 'Genres')
Anime df.show()
    +----+
    MAL ID
                          Name |
                                          synopsis Genres
    +----+
                   Cowboy Bebop | "In the year 2071... |
                                                     Action
                   Cowboy Bebop | "In the year 2071... | Adventure
         1 |
          1 |
                   Cowboy Bebop | "In the year 2071... |
                                                      Comedy
         1 |
                   Cowboy Bebop | "In the year 2071...
                                                      Drama
         1 |
                   Cowboy Bebop | "In the year 2071...
                                                      Sci-Fi
                   Cowboy Bebop | "In the year 2071...
          1 |
                                                       Space
         5 Cowboy Bebop: Ten... other day, anothe...
                                                      Action
         5 Cowboy Bebop: Ten... other day, anothe...
                                                       Drama
          5 Cowboy Bebop: Ten... other day, anothe...
                                                     Mystery
```

```
5 Cowboy Bebop: Ten... other day, anothe...
                                                 Sci-Fi
5 | Cowboy Bebop: Ten... other day, anothe...
                                                  Space
                Trigun | "Vash the Stamped...
6
                                                 Action
6
                Trigun | "Vash the Stamped...
                                                 Sci-Fi
                Trigun | "Vash the Stamped... | Adventure
6
6
                Trigun | "Vash the Stamped...
                                                 Comedy
                Trigun | "Vash the Stamped...
6
                                                  Drama
                Trigun | "Vash the Stamped... |
                                                Shounen
6
7 |
    Witch Hunter Robin ches are individu...
                                                 Action
   Witch Hunter Robin ches are individu...
7 l
                                                Mystery
   Witch Hunter Robin ches are individu...
                                                 Police
```

only showing top 20 rows

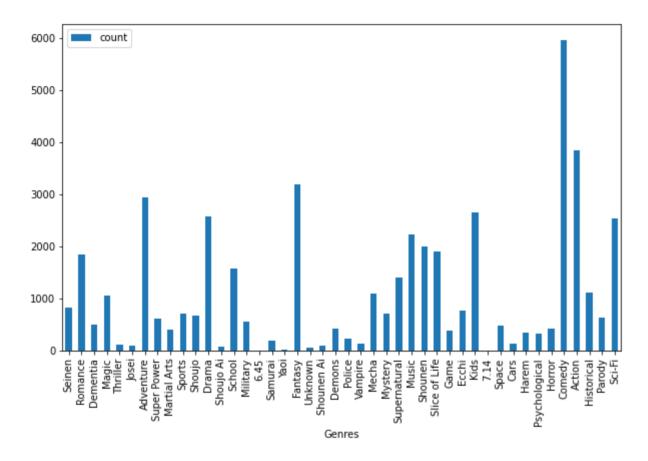
```
Ratings = Ratings df.select(['anime id' , 'rating'])
Ratings.show(5)
```

```
+----+
|anime id|rating|
+----+
     430
             9 |
    1004
             5 |
    3010
     570
             7 |
    2762
             9 |
only showing top 5 rows
```

▼ Visualization of the Genre column

```
# Group column by distinct values and count occurrences
df count = Anime df.groupBy('Genres').count()
# Convert DataFrame to Pandas
df pd = df count.toPandas()
```

```
# Create bar plot
df_pd.plot(kind='bar', x='Genres', y='count' , figsize=(10,6))
plt.show()
```



As we can observe, the Genre column is unbalanced as the counts differ. Comedy appears to be the most dominant genre of the dataframe contrastly with '7.14' and '6.45' to be the least popular genres.

First Model Design

For the first phase, we will process the 'Ratings' Dataframe, to compute the average rating of each anime. To achieve this, we will first turn the dataframe into an RDD. Then, with a Map function, we will create key-value pairs, where keys will be the anime ids and values their corresponding ratings. Afterward, the key-value pairs will be grouped by key creating key-lists of value pairs that then be input into a reduce function. Lastly, the reduce function will take the key-lists of values and output the average rating of each key.

```
def average ratings(df):
    start time = time.time()
    # Turning the dataframe into an RDD
    RDD = df.rdd
# Map function:
    # Performing the map function in parallel on each chunk, in which it will create key-value pairs, key: anime id, value: rating:
    Map = RDD.map(lambda x : (x[0], x[1]))
#Group-By key function:
    # Using the groupByKey() function to group the key-value pairs by key, creating key-lists of values pairs:
    GroupBy = Map.groupByKey()
#Reduce function:
    # Computing the average rating for each key:
    Reduce = GroupBy.mapValues(lambda x : sum(x) / len(x))
    end time = time.time()
    print('Map-Reduce time: ', end time - start time)
    # Converting the RDD back to a DataFrame
    Avrg df = Reduce.toDF(['anime id', 'score'])
    end time = time.time()
    print('Processing time: ', end time - start time)
    return Avrg df
```

```
Average_Ratings = average_ratings(Ratings)
Average_Ratings.show()
```

```
Map-Reduce time: 0.15545392036437988
Processing time: 282.65165519714355
+----+
anime id
    3010 | 7.172523961661342
     448 7.083866465045946
   36456 8.239965483879402
   32935 | 8.859160636758322
    9919 7.6643402810525325
    1575 8.768003729304906
   38731 8.159376708583926
   35847 7.182853504986913
   37779 | 8.566153479646228
   18753 7.4640287769784175
    4053 8.08744394618834
   16870 7.640721115780498
    8246 7.317592254293022
   14749 | 7.143337558527563
     154 7.710264577780667
   37450 | 8.264643210684385
   37086 7.235680913652881
     189 7.303943499936086
   29575 | 7.4541772151898735
   35581 6.993574297188755
+----+
```

Second Approach to the First Model:

only showing top 20 rows

Another approach to compute the average rating of each anime in the Ratings Dataframe, would be to first group by the anime_id column. This is a similar procedure to the Map and GroupByKey functions. The output is key-lists of values pairs, where the keys are the anime ids paired

with lists of their corresponding ratings. Then, we will apply a reduction function by computing the average rating of each anime

```
def average ratings(df):
   # We will also compute the processing time
   start time = time.time()
   # Grouping by the anime id
   df grouped = df.groupBy('anime id')
   # we will aggregate the results computing the average rating of each transaction
   df avg = df grouped.agg(avg('rating').alias('Score'))
   end time = time.time()
   print('Processing time: ', end time - start time)
   return df avg
Average Ratings = average ratings(Ratings)
Average Ratings.show()
     Processing time: 0.03953242301940918
     +----+
     anime id
                           Score
     +----+
         1829 | 7.035614035087719
         5300 | 8.622606525227313
         8086 | 7.698882018100659
        22097 7.786092786849789
        30654 8.531057608714937
        36538 8.389559360089812
           496 6.663366336633663
         38422 7.987347170884276
          6336 | 8.23187528378992
         1088 7.97648330855521
          463 7.287537257824143
          2142 6.377174486030575
          9465 7.95027027027027
          3918 | 7.183235867446394
```

```
| 25517|7.883919202518363|
| 30903|6.673501577287066|
| 11033|7.133634602969658|
| 17389|8.448859847979731|
| 2122|6.754385964912281|
| 40515|6.264705882352941|
+-----+
```

Next step is to join the two tables, Average_Ratings and Anime_df, on anime_id.

```
Anime_DF = Anime_df.join(Average_Ratings, Anime_df.MAL_ID == Average_Ratings.anime_id, 'inner')
Anime_DF = Anime_DF.drop('MAL_ID')
Anime_DF.show()
```

+	+	+	+	++
Name	synopsis	Genres	anime_id	Score
Ged Senki	Calamities are pl	Fantasy	1829	7.035614035087719
Ged Senki	Calamities are pl	Magic	1829	7.035614035087719
Ged Senki	Calamities are pl	Adventure	1829	7.035614035087719
Zoku Natsume Yuuj	"s with its prequ	Shoujo	5300	8.622606525227313
Zoku Natsume Yuuj	"s with its prequ	Drama	5300	8.622606525227313
Zoku Natsume Yuuj	"s with its prequ	Supernatural	5300	8.622606525227313
Zoku Natsume Yuuj	"s with its prequ	Demons	5300	8.622606525227313
Zoku Natsume Yuuj	"s with its prequ	Slice of Life	5300	8.622606525227313
Densetsu no Yuush	"""Alpha Stigma""	Shounen	8086	7.698882018100659
Densetsu no Yuush	"""Alpha Stigma""	Fantasy	8086	7.698882018100659
Densetsu no Yuush	"""Alpha Stigma""	Magic	8086	7.698882018100659
Densetsu no Yuush	"""Alpha Stigma""	Adventure	8086	7.698882018100659
Densetsu no Yuush	"""Alpha Stigma""	Action	8086	7.698882018100659
Magi: Sinbad no B	"Not so long ago,	Shounen	22097	7.786092786849789
Magi: Sinbad no B	"Not so long ago,	Magic	22097	7.786092786849789
Magi: Sinbad no B	"Not so long ago,	Fantasy	22097	7.786092786849789
Magi: Sinbad no B	"Not so long ago,	Adventure	22097	7.786092786849789
Magi: Sinbad no B	"Not so long ago,	Action	22097	7.786092786849789
Ansatsu Kyoushits	The students retu	Shounen	30654	8.531057608714937
Ansatsu Kyoushits	The students retu	School	30654	8.531057608714937

+-----tonly showing top 20 rows

Second Model Design

In the second phase, we will collect the top 10 animes of each genre based on their score and store them in a Dataframe. This will be achieved by first turning the Anime_DF into an RDD and then applying a map function emitting key-value pairs, the key will be the genre and the values will be the name, synopsis, and score. Then, we will group by genre, creating key-lists of values pairs. Lastly, we will sort the key-lists of values pairs by score and save the top 10 animes in a new dataframe.

```
def top 10 by genre(df):
    start time = time.time()
    # Converting DataFrame to RDD
    rdd = df.rdd
# Map function:
    # Performing the map function, in which it will creat key-value pairs, key: genre , value: ( Name, Synopsis , Score ):
   Map = rdd.map(lambda x: (x[2], (x[0], x[1], x[4])))
# Group by key:
    # Grouping by key, and creating key-lists of value pairs
   List of values = Map.groupByKey().mapValues(lambda x: list(x))
# Sorting phase:
    # We will create an empty list to append the top 10 animes of each genre and then use it to create a dataframe:
    top 10 list = []
    # Creating a for loop, firstly sorting the values of each genre by their score, collecting the top 10 and lastly appending the results in
    for genre, values in List of values.collect():
        top10 = sorted( values, key = lambda x: x[2] , reverse=True )[:10]
```

Processing time: 90.39276504516602

+	+	tt	+
Genre	Name	Synopsis Sco	re
+	+	++	+
Fantasy	Momotarou no Kout	omotarou themed t 10	.0
Fantasy	Nendjuugyouji Ani	$ $ animation teachin $\ldots $ 10	.0
Fantasy	Tatoeba Last Dung	$ $ "long time ago, t \ldots $ $ 10	.0
Fantasy	Oz no Mahoutsukai	ducational film a 9.	33
Fantasy	Fullmetal Alchemi	"""In order for s $ $ 9.	24
Fantasy	Hunter x Hunter (Hunter x Hunter i 9.	17
Fantasy	Shingeki no Kyoji	"Seeking to resto 9.	03
Fantasy	Aria the Crepuscolo	The season is fal $ $ 9	.0
Fantasy	Re:Zero kara Haji	Second half of Re 9	.0
Fantasy	Mushishi	"""Mushi"": the m 8.	84
Shoujo	Natsume Yuujincho	"Takashi Natsume, 8.	76
Shoujo	Natsume Yuujincho	Takashi Natsume h 8	.7
Shoujo	Natsume Yuujincho	"Natsume Yuujinch 8.	66
Shoujo	Nana	Nana Komatsu is a 8.	65
Shoujo	Zoku Natsume Yuuj	"s with its prequ $ $ 8.	62
Shoujo	Natsume Yuujincho	Season 5 of Natsu 8.	61

Now we will create a function, that will asks a user for genres and then return the top animes that belong to those genres.

```
def Anime Recommendation System(df):
 print('-----
 print('Welcome to the Anime Recommendation System!\n')
 print('You will be asked to input your genre of choice and the system will return the top 10 animes that correspond to that genre')
 print('Lets get started!\n')
 print('-----
 Genres = input( 'What genre you are interested in? \n')
 df.filter(df.Genre == Genres).select('Name', 'Synopsis').show()
Anime Recommendation System(Top10)
   Welcome to the Anime Recommendation System!
   You will be asked to input your genre of choice and the system will return the top 10 animes that correspond to that
   Lets get started!
   What genre you are interested in?
   Comedy
   +----+
                Name | Synopsis |
   +----+
```

```
Gintama° | Gintoki, Shinpach... |
| Fullmetal Alchemi... | """In order for s... |
| Gintama' | fter a one-year h... |
| Gintama: The Final | New Gintama movie. |
| Gintama | The Amanto, alien... |
| Clannad: After Story | Clannad: After St... |
| Gintama': Enchousen | hile Gintoki Saka... |
| Gintama Movie 2: ... | "hen Gintoki appr... |
| Owarimonogatari 2... | Following an enco... |
| Gintama. | fter joining the ... |
```

→ Bibliography

Setting up Pyspark on google collab:

https://www.analyticsvidhya.com/blog/2020/11/a-must-read-guide-on-how-to-work-with-pyspark-on-google-colab-for-data-scientists/

https://stackoverflow.com/questions/55240940/error-while-installing-spark-on-google-colab

Datasets:

https://www.kaggle.com/datasets/hernan4444/anime-recommendation-database-2020

Google Cloud Dataproc:

https://holowczak.com/getting-started-with-pyspark-on-google-cloud-platform-dataproc/9/

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