# Technical test results for HEVA company

This notebook repeats the statement of the test. Under each activity you will find the code and the result produced. You will find all the requirements to run this notebook in the requirements.md file.

# Configuration

### 1. Importing packages

```
In [1]: # Import necessary modules

from pyspark.sql import SparkSession
import matplotlib.pyplot as plt
from pyspark.sql.functions import col, when, explode, split,\
    desc, from_unixtime, year
from pyspark.sql.types import DateType
import time
import sys
import contextlib
```

### 2. Settings

```
In [2]: # Definition of necessary parameters
    data_path = "../sources/data/movies.sqlite"
    output_log_path = "result.log"

In [3]: class Logger:

    def __init__(self, filename):
        self.console = sys.stdout
        self.file = open(filename, 'a')

    def write(self, message):
        self.console.write(message)
        self.file.write(message)

    def flush(self):
        self.console.flush()
        self.file.flush()
```

### 3. Reading data

```
spark = SparkSession.builder\
       .config(
           'spark.jars.packages',
           'org.xerial:sqlite-jdbc:3.34.0')\
       .getOrCreate()
   # Reading the movies table
   df_movies = spark.read.format('jdbc')\
       .options(
           driver='org.sqlite.JDBC',
           dbtable='movies',
           url=f'jdbc:sqlite:{data_path}')\
       .load()
   # Reading the ratings table
   df_ratings = spark.read.format('jdbc')\
       .options(
           driver='org.sqlite.JDBC',
           dbtable='ratings',
           url=f'jdbc:sqlite:{data_path}')\
    return df_movies, df_ratings
with contextlib.redirect_stdout(Logger(output_log_path)):
   df_movies, df_ratings = read_data(data_path)
22/07/21 20:51:21 WARN Utils: Your hostname, cornichon resolves to a loopback address: 1
27.0.1.1; using 192.168.1.156 instead (on interface wlp3s0)
22/07/21 20:51:21 WARN Utils: Set SPARK_LOCAL_IP if you need to bind to another address
:: loading settings :: url = jar:file:/home/matteo/anaconda3/envs/test_heva/lib/python3.
10/site-packages/pyspark/jars/ivy-2.5.0.jar!/org/apache/ivy/core/settings/ivysettings.xm
Ivy Default Cache set to: /home/matteo/.ivy2/cache
The jars for the packages stored in: /home/matteo/.ivy2/jars
org.xerial#sqlite-jdbc added as a dependency
:: resolving dependencies :: org.apache.spark#spark-submit-parent-b83cb341-a934-43d0-bac
7-7b64070cb6ad;1.0
       confs: [default]
       found org.xerial#sqlite-jdbc;3.34.0 in central
:: resolution report :: resolve 317ms :: artifacts dl 6ms
       :: modules in use:
       org.xerial#sqlite-jdbc;3.34.0 from central in [default]
       _____
            | modules || artifacts |
conf | number| search|dwnlded|evicted|| number|dwnlded|
       ______
       | default | 1 | 0 | 0 | 0 | 1 | 0 |
       :: retrieving :: org.apache.spark#spark-submit-parent-b83cb341-a934-43d0-bac7-7b64070cb6
ad
       confs: [default]
       O artifacts copied, 1 already retrieved (OkB/6ms)
22/07/21 20:51:23 WARN NativeCodeLoader: Unable to load native-hadoop library for your p
latform... using builtin-java classes where applicable
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLev
el).
```

#### 4. Data overview

```
Args:
    df_movies (Dataframe): Movies Dataframe
    df_ratings (Dataframe): Ratings Dataframe
"""

# Overview of movies table data
print("Movies table")
df_movies.show()

# Preview data from the ratings table
print("Ratings table")
df_ratings.show()

with contextlib.redirect_stdout(Logger(output_log_path)):
    preview_data(df_movies, df_ratings)
```

Movies table

```
title|
|movie_id|
+----+
      8|Edison Kinetoscop...| Documentary|Short|
      10|La sortie des usi...| Documentary|Short|
      12|The Arrival of a ...| Documentary|Short|
      25|The Oxford and Ca...|
                                           nullI
      91|Le manoir du diab...| Short|Horror|
     131|Une nuit terrible...| Short|Comedy|Horror|
     417|A Trip to the Moo...|Short|Action|Adve...|
     439|The Great Train R...|Short|Action|Crim...|
     443|Hiawatha, the Mes...| null|
628|The Adventures of...| Action|Short|
833|The Country Docto...| Short|Drama|
    1223| Frankenstein (1910)| Short|Horror|Sci-Fi|
    1740|The Lonedale Oper...| Short|Drama|Romance|
            Cleopatra (1912)| Drama|History|
    2101|
    2130 l
            L'inferno (1911)|Adventure|Drama|F...|
    2354|Max et Jane veule...|Short|Comedy|Romance|
    2844|Fantômas - À l'om...| Crime|Drama|
    3740| Cabiria (1914)|Adventure|Drama|H...|
    3863|Dough and Dynamit...| Comedy|Short|
    4099|His Majesty, the ...|Family|Fantasy|Ad...|
+----+
only showing top 20 rows
```

#### Ratings table

```
3 | 2278388 |
                      8|
                               1597297732
      3 | 6199572 |
                      3|
                               1589482483
      3 | 6723592 |
                     8|
                              1599578941
       3 | 6751668 |
                     9|
                              1578955697
       3 | 7131622 |
                     8|
                              1579559244
only showing top 20 rows
```

## **Tasks**

#### 1. Counts

• 1.1 How many films are in the database?

```
In [6]:
        def activity_1_1(df_movies):
            """Counting the number of distinct film titles
                df_movies (Dataframe): Movies Dataframe
            Return:
                int: Number of movies
            return df_movies\
                .select("title")\
                 .distinct()\
                 .count()
        with contextlib.redirect_stdout(Logger(output_log_path)):
            result_1_1 = activity_1_1(df_movies)
            print("There are", result_1_1, "movies in the database")
        [Stage 2:>
                                                                              (0 + 1) / 1]
        There are 37947 movies in the database
```

• 1.2 How many different users are in the database?

```
In [7]: def activity_1_2(df_ratings):
    """Counting the number of distinct user id

Args:
    df_ratings (Dataframe): Ratings Dataframe

Return:
    int: Number of user id
    """

    return df_ratings\
        .select("user_id")\
        .distinct()\
        .count()

with contextlib.redirect_stdout(Logger(output_log_path)):
    result_1_2 = activity_1_2(df_ratings)
    print("There are", result_1_2, "user id in the database")
```

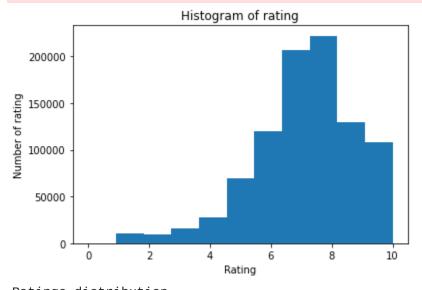
[Stage 8:> (0 + 1) / 1]

There are 71707 user id in the database

• 1.3 What is the distribution of the notes provided? Bonus: create a histogram.

```
def activity_1_3(df_ratings):
    """ Display rating distribution histogramme
        Counting the number of voters per rating
        Sorting based on ratings
    Args:
        df_ratings (Dataframe): Ratings Dataframe
    0.000
    # Creation of the histogram
    print("Converting Dataframe to Pandas...")
    plt.hist(
        df_ratings.select("rating").toPandas().squeeze(),
        bins=11) # [0 to 10] => 11 values
    plt.xlabel('Rating')
    plt.ylabel('Number of rating')
    plt.title('Histogram of rating')
    plt.show()
    print("Ratings distribution")
    df_ratings\
        .groupBy("rating")\
        .count()\
        .orderBy("rating")\
        .show()
with contextlib.redirect_stdout(Logger(output_log_path)):
    activity_1_3(df_ratings)
```

Converting Dataframe to Pandas...



```
Ratings distribution
+----+
|rating| count|
+----+
| 0| 281|
| 1| 10814|
| 2| 9223|
```

```
| 3| 15487|
| 4| 28193|
| 5| 69747|
| 6|120370|
| 7|206680|
| 8|222146|
| 9|130106|
| 10|108351|
```

1.4 Finally, we want to obtain a table of frequencies to express the distribution of notes as a
percentage.

```
In [9]: def activity_1_4(df_ratings):
               """ Added column count which represents the number of voters
                   by notes.
                   Added a percentage column,
                   which is a transformation of the count column into a percentage.
                   Selection of rating and percentage columns.
                   Sort by rating column.
               Args:
                   df_ratings (Dataframe): Rating Dataframe
               df_ratings.groupBy("rating")\
                   .count()\
                    .withColumn(
                         'percentage',
                     (\text{col}(\text{"count"})*100)/\text{float}(\text{df\_ratings.count}())) \\ . \text{select}(\text{"rating"}, \text{"percentage"}) \\ \\ \\ \end{aligned}
                    .orderBy("rating")\
                    .show()
          with contextlib.redirect_stdout(Logger(output_log_path)):
               print("Ratings frequencies")
               activity_1_4(df_ratings)
```

```
Ratings frequencies
+----+
|rating|
              percentage|
+----+
     0|0.03049713587396543|
    1 | 1.1736513428507551 |
    2 | 1.0009789472084811 |
     3 | 1.6808154565128208 |
     4 | 3.0598069455327663 |
     5 | 7.569693009969633 |
     6 | 13.063844288787255 |
    7 | 22.431131823598488 |
    8 | 24.10966813472571 |
    9 | 14.120499501843938 |
    10 | 11.759413413096187 |
```

### 2. Data selection and enrichment

• 2.1 In order to set up a certain statistical model, we must transform the rating note into two modalities: did the user like the film or not? Create a new liked column in the ratings table with

the following values: 0 for ratings [0-6] and 1 for ratings [7-10].

```
In [10]:
         def activity_2_1(df_ratings):
              """ Added a liked column.
                  Depending on the rating column,
                  the liked column takes the value 0 or 1
             Args:
                  df_ratings (Dataframe): Ratings Dataframe
              Returns:
                  Dataframe: Updated ratings Dataframe
              df_ratings = df_ratings\
                  .withColumn(
                      'liked',
                      when(df_ratings.rating < 7, 0)</pre>
                      .when(df_ratings.rating >= 7, 1))
              df_ratings.show()
              return df_ratings
         with contextlib.redirect_stdout(Logger(output_log_path)):
              print("Updated ratings Dataframe")
              df_ratings = activity_2_1(df_ratings)
```

```
Updated ratings Dataframe
+----+
|user_id|movie_id|rating|rating_timestamp|liked|
   +-----
                         11
                         1|
                         0 |
                         1|
                         1|
                         1|
                         0 l
                         1|
                         0 |
                         0 l
                         0 l
                         1|
                         1|
                         1|
only showing top 20 rows
```

• 2.2 Which genres are rated highest by users? We want to get the **top 10** movie genres liked by users (using the new liked column).

```
In [11]: def activity_2_2(df_movies, df_ratings):
    """ Separation of genres in an array with the split function.
    Extract genre arrays with the explode function alias explode_genre.
```

```
Selection of the movie_id and explode_genre column.
        Joining with ratings table on movie_id columns.
        Sum of the liked column by grouping on the explode_genre column.
        Rename sum(liked) column to sum_liked.
        Rename explode_genre column to genre.
        Sort in descending order based on the sum_liked column.
        Limitation to the first 10 records.
    Args:
        df_movies (Dataframe): Movies Dataframe
        df_ratings (Dataframe): Ratings Dataframe
    df_movies.select(
        "movie_id",
        explode(
            split(
                col("genre"),
                "\|"))
        .alias("explode_genre"))\
        .join(
            df_ratings,
            df_ratings.movie_id == df_movies.movie_id,
            "inner")\
        .groupBy("explode_genre")\
        .sum("liked")\
        .withColumnRenamed("sum(liked)", "sum_liked")\
        .withColumnRenamed("explode_genre", "genre")\
        .sort(desc("sum_liked"))\
        .limit(10)\
        .show()
with contextlib.redirect_stdout(Logger(output_log_path)):
    print("Top 10 genres")
    activity_2_2(df_movies, df_ratings)
Top 10 genres
```

```
[Stage 30:>
                                                               (0 + 4) / 4
+-----+
   genre|sum_liked|
+----+
   Drama| 397116|
| Thriller| 220867|
| Action| 199212|
   Comedy| 169233|
|Adventure| 169194|
   Crime| 136140|
Sci-Fi| 131233|
  Romance| 92047|
  Mystery|
            83036|
  Fantasy|
             80337|
  -----+
```

#### 3. Advanced Selections

• 3.1 What are the titles of the films most popular with Internet users? We are looking for the 10 films with the best ratings on average by users, with a minimum of 5 ratings for the measurement to be relevant.

```
In [12]:
         def activity_3_1(df_movies, df_ratings):
             """ Join between movies and ratings tables,
```

```
on movie_id columns, alias movies_ratings.
        Join with subtable alias title_count,
        which represents the number of votes per film.
        Filter on movies that have at least 5 ratings.
        Average ratings per movie title.
        Renamed avg(rating) column to mean_rating.
        Descending sort based on mean_rating column.
        Limitation to the first 10 records.
    Args:
        df_movies (Dataframe): Movies Dataframe
        df_ratings (Dataframe): Ratings Dataframe
    df_movies.join(
        df_ratings,
        df_movies.movie_id == df_ratings.movie_id,
        "inner").alias("movies_ratings")\
        .join(
            (df_movies.join(
                df_ratings,
                df_movies.movie_id == df_ratings.movie_id,
                "inner")
                .groupBy("title")
                .count()).alias("title_count"),
            col("movies_ratings.title") == col("title_count.title"),
            "inner")\
        .filter(col("count") >= 5)\
        .groupBy("movies_ratings.title")\
        .mean("rating")\
        .withColumnRenamed("avg(rating)", "mean_rating")\
        .sort(desc("mean_rating"))\
        .limit(10)\
        .show()
with contextlib.redirect_stdout(Logger(output_log_path)):
    print("Top 10 movies")
    activity_3_1(df_movies, df_ratings)
```

Top 10 movies

```
+-----+

| title|mean_rating|
+-----+
| Five Minutes (2017)| 10.0|
|Crawl Bitch Crawl...| 10.0|
| Selam (2013)| 10.0|
| Romeo Juliet (2009)| 10.0|
|Third Contact (2011)| 10.0|
|Let There Be Ligh...| 10.0|
|Chasing Happiness...| 10.0|
|Avengers: Age of ...| 10.0|
|Make Like a Dog (...| 10.0|
|Quiet Riot: Well ...| 10.0|
```

• 3.2 What is the most rated film in 2020? **Note**: the rating\_timestamp column is provided in the database as Unix time.

```
""" Adding a rating_year column,
        which corresponds to the year in which the vote was recorded.
        Join movies and ratings tables on movie_id columns.
        Counting the number of votes per film title.
        Sort descending in order of count.
        Rename column count to rating_count.
        Limitation to the first record.
    Args:
        df_movies (Dataframe): Movies Dataframe
        df_ratings (Dataframe): Ratings Dataframe
    df_ratings\
        .withColumn(
            'rating_year',
            year(
                from_unixtime('rating_timestamp')
                .cast(DateType())))\
        .join(
            df_movies,
            df_ratings.movie_id == df_movies.movie_id)\
        .filter(col("rating_year") == 2020)\
        .groupBy("title")\
        .count()\
        .sort(desc("count"))\
        .withColumnRenamed("count", "rating_count")\
        .limit(1)\
        .show()
with contextlib.redirect_stdout(Logger(output_log_path)):
    print("Best film of the year 2020")
    activity_3_2(df_movies, df_ratings)
Best film of the year 2020
                                                                     (0 + 1) / 1]
```

```
[Stage 50:> (0 + 1) / 1]
+-----+
| title|rating_count|
+-----+
|1917 (2019)| 2858|
+-----+
```

## 4. Data management

• 4.1 In order to find the notes of a particular user more quickly, we want to set up an index on the user ids. Do you see a performance difference when looking up the ratings given by user 255 ?

Spark DataFrames are inherently unordered and do not support random access. (There is no built-in index concept like there is in pandas). Each row is treated as an independent collection of structured data, and this is what enables distributed parallel processing. So any executor can take any block of data and process it regardless of row order. More info here

Instead we can order the pyspark ratings dataframe according to the 'user\_id' column. Otherwise the koalas package can be an alternative. Because Koala supports indexes and can be used for big data. Also, pandas cannot be scaled for big data oriented use.

To check performance, I created the function time test which print the execution time of a function.

```
func (function): A function name
             time_list = []
             for i in range(100):
                 start_time = time.time()
                 # beginning of the code to test
                 func()
                 # end of the code to test
                 time_list.append(time.time() - start_time)
             mean_time = sum(time_list) / len(time_list)
             max_time = max(time_list)
             min_time = min(time_list)
             print("min:", min_time, "mean:", mean_time, "max:", max_time, end="\n\n")
In [15]:
         def activity_4_1(df_ratings):
             """Compare time perfomance for indexed and not indexed Dataframe
             Args:
                 df_ratings (Dataframe): Ratings Dataframe
             df_ratings_indexed = df_ratings.orderBy("user_id")
             print("Converting Dataframe to Pandas...")
             pandas_df_ratings = df_ratings.toPandas()
             pandas_df_ratings_indexed = pandas_df_ratings.set_index("user_id")
             print("Execution time for unindexed PYSPARK dataframe")
             time_test(lambda: df_ratings.filter(col("user_id") == 255))
             print("Execution time for PYSPARK dataframe indexed by 'user_id'")
             time_test(lambda: df_ratings_indexed.filter(col("user_id") == 255))
             print("Execution time for unindexed PANDAS dataframe")
             time_test(
                  lambda: pandas_df_ratings
                  .loc[pandas_df_ratings.loc[:, "user_id"] == 255])
             print("Execution time for PANDAS dataframe indexed by 'user_id'")
             time_test(lambda: pandas_df_ratings_indexed.loc[255])
         with contextlib.redirect_stdout(Logger(output_log_path)):
             activity_4_1(df_ratings)
         Converting Dataframe to Pandas...
```

In [14]: def time\_test(func):

Args:

""" Check function time performance.

min: 0.0039598941802978516 mean: 0.008981218338012695 max: 0.02070164680480957

Execution time for PYSPARK dataframe indexed by 'user\_id'
min: 0.003888368606567383 mean: 0.009223527908325195 max: 0.0683746337890625

Execution time for unindexed PANDAS dataframe
min: 0.001016855239868164 mean: 0.00222933292388916 max: 0.021737098693847656

Execution time for PANDAS dataframe indexed by 'user\_id'

Execution time for unindexed PYSPARK dataframe

min: 5.841255187988281e-05 mean: 0.0004265451431274414 max: 0.03124547004699707

#### Ranking:

- 1. Indexed Pandas Dataframe
- 2. Unindexed Pandas Dataframe
- 3. Indexed Pyspark Dataframe / Unindexed Pyspark Dataframe

# Code quality check

```
In [16]: !flake8-nb result.ipynb
result.ipynb#In[11]:22:18: W605 invalid escape sequence '\|'
result.ipynb#In[]:1:80: E501 line too long (92 > 79 characters)
```

# Safe notebook versioning

```
In [17]: !jupyter nbconvert result.ipynb --to="python"

[NbConvertApp] Converting notebook result.ipynb to python
[NbConvertApp] Writing 14008 bytes to result.py
```

# PDF export

```
In [18]: !jupyter nbconvert --to webpdf --allow-chromium-download result.ipynb
```