LAB 1 - Oleskevych Sofia, CS-314

Analysis part

I loaded "movies_metadata.csv" and "ratings.csv" (or ratings_small.csv) as spark dataframes https://www.kaggle.com/rounakbanik/the-movies-dataset/data

Saved them to Google Drive, mounted GD to my colab notebook.

Using Python + Pyspark performed basic investigation and printed out summary (are there any missing values; How many records; How many unique users are there)

Preprocessing part

Using pyspark DataFrame API transform data loaded on a previous step to create MovieProfile and UserProfile entities with following schemas:

MovieProfile entity schema:

- · Movield id of the movie
- MovieName name of the movie
- MovieRating average from all the user ratings for the movie
- Genres a string with the comma separated genres for the movie

UserProfile entity schema:

- · UserId id of the user
- NumberOfMarks how many marks user left
- YearsSpent how many years has passed since the first mark
- AvgMark average user mark
- AvgTimeBetweenMarks how often user watches movies how many time in average user spends between the marks in days (if I have 3 marks on the next dates: 21.07, 23.07 and 27.07, I will have (2 + 4) / 2 = 3 days in average)
- FavoriteGenre favorite genre of a user (my own implementation)

```
!pip install pyspark

Collecting pyspark

Downloading pyspark-3.2.0.tar.gz (281.3 MB)

| 281.3 MB 38 kB/s

Collecting py4j==0.10.9.2

Downloading py4j-0.10.9.2-py2.py3-none-any.whl (198 kB)
```

```
198 kB 54.0 MB/s
```

Building wheels for collected packages: pyspark

Building wheel for pyspark (setup.py) ... done

Created wheel for pyspark: filename=pyspark-3.2.0-py2.py3-none-any.whl size=281805911 Stored in directory: /root/.cache/pip/wheels/0b/de/d2/9be5d59d7331c6c2a7c1b6d1a4f463ce

Successfully built pyspark

Installing collected packages: py4j, pyspark

Successfully installed py4j-0.10.9.2 pyspark-3.2.0

→

from pyspark.sql import SparkSession

spark = SparkSession.builder.getOrCreate()

from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

raw_df = spark.read.option("header", "true").option("encoding", "UTF-8").csv("/content/drive/
raw_df2 = spark.read.option("header", "true").csv("/content/drive/MyDrive/ColabNotebooks/rati

raw_df.show()

+				+				++	- -
adult be	longs	_to_coll	lection	budget			genres	homepage id	
++				+					
	.1d.:							http://toystory.d 862	
False			null	65000000	[{'id':	12,	'name	null 8844 †	t١
False {	'id':	119050,	'n	0	[{'id':	1074	49, 'n	null 15602 †	t١
False			null	16000000	[{'id':	35,	'name	null 31357 †	t١
False {	'id':	96871,	'na	0	[{'id':	35,	'name	null 11862 †	t١
False			null	60000000	[{'id':	28,	'name	null 949 †	t١
False			null	58000000	[{'id':	35,	'name	null 11860 †	t١
False			null	0	[{'id':	28,	'name	null 45325 †	t١
False			null	35000000	[{'id':	28,	'name	null 9091 †	t١
False {	'id':	645, 'r	name	58000000	[{'id':	12,	'name	http://www.mgm.co 710	t١
False			null	62000000	[{'id':	35,	'name	null 9087 †	t١
False			null	0	[{'id':	35,	'name	null 12110 †	t١
False {	'id':	117693,	'n	0	[{'id':	1075	51, 'n	null 21032 †	t١
False			null	44000000	[{'id':	36,	'name	null 10858 ⁻	t١
False			null	98000000	[{'id':	28,	'name	null 1408 †	t١
False			null	52000000	[{'id':	18,	'name	null 524 †	t١
False			null	16500000	[{'id':	18,	'name	null 4584	t١
False			null	4000000	[{'id':	80,	'name	null 5 †	t١
False {	'id':	3167, '	nam	30000000	[{'id':	80,	'name	null 9273 †	t١
False			null	60000000	[{'id':	28,	'name	null 11517 1	t١
+				+				·	

raw df2.show()

```
|userId|movieId|rating| timestamp|
      1
            110
                   1.0 1425941529
      11
            147
                   4.5 1425942435
      1
           858
                   5.0 | 1425941523 |
          1221
                   5.0 1425941546
      1 l
           1246
                   5.0 1425941556
                   4.0 | 1425942148 |
      1
          1968
      1
          2762
                   4.5 | 1425941300 |
      1
          2918
                   5.0 1425941593
      1
          2959
                  4.0 | 1425941601 |
      1
          4226
                  4.0 1425942228
          4878
      1
                   5.0 | 1425941434 |
      1
          5577
                   5.0 1425941397
      11
         33794
                  4.0 1425942005
      1
         54503
                   3.5 | 1425941313 |
      1
         58559
                  4.0 1425942007
      11
         59315
                   5.0 1425941502
      1
         68358
                   5.0 1425941464
         69844
                   5.0 1425942139
         73017
                   5.0 1425942699
      1
         81834
                   5.0 1425942133
only showing top 20 rows
```

investigate schema in tree format raw df.printSchema()

```
root
```

```
|-- adult: string (nullable = true)
|-- belongs_to_collection: string (nullable = true)
|-- budget: string (nullable = true)
|-- genres: string (nullable = true)
|-- homepage: string (nullable = true)
|-- id: string (nullable = true)
|-- imdb id: string (nullable = true)
-- original language: string (nullable = true)
|-- original title: string (nullable = true)
-- overview: string (nullable = true)
|-- popularity: string (nullable = true)
-- poster path: string (nullable = true)
|-- production_companies: string (nullable = true)
-- production_countries: string (nullable = true)
|-- release_date: string (nullable = true)
-- revenue: string (nullable = true)
|-- runtime: string (nullable = true)
|-- spoken_languages: string (nullable = true)
|-- status: string (nullable = true)
|-- tagline: string (nullable = true)
```

```
|-- title: string (nullable = true)
     |-- video: string (nullable = true)
     |-- vote average: string (nullable = true)
     |-- vote_count: string (nullable = true)
# count rows in dfs
raw_df.count(), raw_df2.count()
    (45572, 26024289)
# check unique users
from pyspark.sql.functions import countDistinct, when, count, col, isnull, avg, aggregate
raw_df2.select(countDistinct("userId")).show()
    +----+
    count(DISTINCT userId)
    +-----+
                 270896
    +----+
# check missing values in raw df in each column
raw df.select([count(when(isnull(c), c)).alias(c) for c in raw df.columns]).show()
        +-----
    |adult|belongs_to_collection|budget|genres|homepage| id|imdb_id|original_language|origin
     40981
                              17
                                   23
                                        37635 31
                                                   125
# rows of raw_df2 (ratings) where MovieId is determined
raw_df2.filter(raw_df2.movieId == 5).show()
    +----+
    |userId|movieId|rating| timestamp|
     ----+-----+
        2
               5
                  3.0 867039249
       40
               5
                  4.0 862515493
       54
               5
                  5.0 986221889
       56
               5
                  4.0 1410105373
       87
               5
                  2.0 867427886
       91 l
               5
                  3.0 838546731
```

3.0 835061814

4.0 850689925

93

101

5

5

```
118
              3.0 859364983
142
         5
              2.0 866388486
168
         5
              3.0 850139631
179
         5
              3.0 863713756
191
         5
              5.0 854204045
196
         5 l
              4.0 | 849688746 |
229
         5
              1.0 1037136141
250
         5
              3.0 995420603
261
         5 l
              3.0 864374769
296
         5
              3.5 | 1111987460 |
306
              3.0 846232643
317
              3.0 855147687
```

only showing top 20 rows

```
# 1 - MovieProfile creation
```

df m.show()

```
# cast the most important fields
mv_id_name = raw_df.select(raw_df.id.cast('int').alias('movieId'), raw_df.title.cast('string')

# basic preprocessing ( null values and valid movieName with regex)

def filtering(input_df):
    return input_df.filter(input_df.movieName.rlike('[a-zA-Z]+')) # one or more alpha symbols

def filtering2(input_df):
    return input_df.filter((input_df.movieName.startswith("[")==False)&(mv_id_name.movieId.is)

df_m = mv_id_name.transform(filtering).transform(filtering2)

df_m.count()

42842
```

```
movieId
                   movieName
                   Toy Story
    862
   8844
                     Jumanji|
           Grumpier Old Men
  15602
  11862 Father of the Bri...
    949
                        Heat
  11860
                     Sabrinal
  45325
               Tom and Huck
                Sudden Death
   9091
    710
                   GoldenEye
   9087 The American Pres...
  12110 Dracula: Dead and...
  21032
                       Baltol
  10858
                       Nixon
```

```
# select and cast from raw_df2 dataframe

mv_id_rate = raw_df2.select(raw_df2.movieId.cast('int').alias('movieId'), raw_df2.rating.cast

df_u = mv_id_rate.filter((mv_id_rate.movieId.isNotNull()) & (mv_id_rate.movieRating.isNotNull

# movie_id_name inner join movie_id_rating

mv_id_name_rate = df_m.join(df_u, df_m.movieId==df_u.movieId, 'inner')\
    .select(df_m.movieId.alias('movieId'), df_m.movieName.alias('movieName'), df_u.movieRating)
    .groupBy('movieId', 'movieName')\
    .agg(avg(df_u.movieRating).alias('movieRating'))
```

mv_id_name_rate.show()

```
+----+
                                movieRating
movieId
                 movieName
              Walk on Water 3.614058355437666
   3469 Far from the Madd... | 4.065651760228354
                    Tanguy | 3.463276836158192|
   2029
    681 Diamonds Are Forever 3.5798319327731094
   5927
              Junior Bonner 2.696526508226691
   8327
          The Holy Mountain 3.8284424379232505
   8094 | The Magdalene Sis... | 3.815525876460768 |
  81782
                Kurukshetra 3.3834776334776335
             The Storm Gate 2.3130841121495327
  26792
   5602
             The Common Man 3.7735602094240837
             The Short Game | 3.247887323943662
 169864
   4441
                     Candy 3.259450171821306
                  The Isle 2.58333333333333333
   8653
  32078 Panic in the Streets | 2.909448818897638 |
              Impy's Island 3.555555555555554
  26497
                 Unforgiven 3.05555555555554
     33 l
   8744
           Albino Alligator
  53883
                   Anjathe 3.992574257425743
   3073 Abbott and Costel... 3.3305555555555557
  70342 Beats Rhymes & Li... | 3.779220779220779
+----+
```

```
movie_id_genres = raw_df.select(raw_df.id.cast('int').alias('movieId'), raw_df.genres.alias('
def filter gen id(input df):
   return input_df.filter((input_df.movieId.isNotNull())&(input_df.genres.isNotNull()) )
movie_id_genres = movie_id_genres.transform(filter_gen_id)
movie_id_genres.count()
    45363
import pyspark.sql.functions as f
final df = movie id genres.select('movieId', f.get json object('genres', '$[*].name').alias('
final df.show()
    +----+
     movieId
                          genres
    +----+
         862 ["Animation", "Com...
        8844 ["Adventure", "Fan...
       15602 ["Romance", "Comedy"]
       31357|["Comedy","Drama"...|
       11862
                        "Comedy"
         949 [ "Action", "Crime"...
       11860|["Comedy", "Romance"]|
       45325 ["Action", "Advent...
        9091 ["Action", "Advent...
         710 [ "Adventure", "Act... |
        9087 ["Comedy", "Drama"...
       12110 ["Comedy", "Horror"]
       21032 ["Family", "Animat...|
       10858 ["History", "Drama"]
        1408 [ "Action", "Advent... |
         524 ["Drama", "Crime"]
        4584 ["Drama", "Romance"]
           5 ["Crime","Comedy"]|
        9273 ["Crime", "Comedy"...
       11517 ["Action", "Comedy...|
    +----+
    only showing top 20 rows
from pyspark.sql.functions import col, concat_ws, split, regexp_replace
movie_fin = final_df.withColumn("genres", regexp_replace(col("genres"), '[\\[\\]\\"]', ""))
```

movie fin.show()

```
+-----+
    862 Animation, Comedy,...
   8844 | Adventure, Fantasy... |
  15602
             Romance, Comedy
  31357 Comedy, Drama, Romance
  11862
                     Comedy
    949 Action, Crime, Dram...
  11860
             Comedy, Romance
  45325 Action, Adventure, ...
   9091 Action, Adventure, ...
    710 Adventure, Action, ...
   9087 Comedy, Drama, Romance
             Comedy, Horror
  21032 Family, Animation, ...
             History, Drama
  10858
          Action, Adventure
   1408
                Drama, Crime
    524
   4584
              Drama, Romance
      5
               Crime, Comedy
   9273 Crime, Comedy, Adve...
  11517 | Action, Comedy, Crime
+----+
only showing top 20 rows
```

final preprocessed MovieProfile dataframe

movie profile.show()

+	<u> </u>		L+
movieId	movieName	movieRating	genres
26	Walk on Water	3.614058355437666	Drama
3469	Far from the Madd	4.065651760228354	Drama,Romance
2029	Tanguy	3.463276836158192	Comedy
681	Diamonds Are Forever	3.5798319327731094	Adventure,Action,
5927	Junior Bonner	2.696526508226691	Action,Drama,Western
8327	The Holy Mountain	3.8284424379232505	Drama
8094	The Magdalene Sis	3.815525876460768	Drama
81782	Kurukshetra	3.3834776334776335	null
26792	The Storm Gate	2.3130841121495327	Drama,History
5602	The Common Man	3.7735602094240837	Drama,Thriller
169864	The Short Game	3.247887323943662	Documentary
4441	Candy	3.259450171821306	Drama,Romance
8653	The Isle	2.583333333333333	Drama,Thriller
32078	Panic in the Streets	2.909448818897638	Drama,Thriller,Crime

```
26497
               Impy's Island|3.5555555555555554|Animation,Family,...|
      33
                  Unforgiven 3.05555555555554
                                                             Western
    8744
            Albino Alligator
                                          3.125 Crime, Drama, Thriller
   53883
                     Anjathe 3.992574257425743
                                                        Action, Drama
    3073 Abbott and Costel... 3.330555555555557
                                                       Comedy, Horror
   70342 | Beats Rhymes & Li... | 3.779220779220779
                                                 Music,Documentary
only showing top 20 rows
```

2 - UserProfile

from pyspark.sql.functions import col, min as min_, max as max_, from_utc_timestamp, from_uni
raw_df2 = raw_df2.withColumn('timestamp', from_unixtime(raw_df2.timestamp).alias("timestamp")
raw_df2.show()

+- u	serId m	novieId r	++ timestamp		
+- I	+- 1	+- 110	1 0	+ 2015-03-09	+ ا ۲۲۰۶۵۰۵۵
-	- I				
-	1	147		2015-03-09	
ļ	1	858		2015-03-09	22:52:03
	1	1221		2015-03-09	22:52:26
	1	1246	5.0	2015-03-09	22:52:36
	1	1968	4.0	2015-03-09	23:02:28
	1	2762	4.5	2015-03-09	22:48:20
	1	2918	5.0	2015-03-09	22:53:13
j	1	2959	4.0	2015-03-09	22:53:21
ĺ	1	4226	4.0	2015-03-09	23:03:48
j	1	4878	5.0	2015-03-09	22:50:34
j	1	5577	5.0	2015-03-09	22:49:57
ĺ	1	33794	4.0	2015-03-09	23:00:05
j	1	54503	3.5	2015-03-09	22:48:33
j	1	58559	4.0	2015-03-09	23:00:07
	1	59315	5.0	2015-03-09	22:51:42
j	1	68358	5.0	2015-03-09	22:51:04
j	1	69844		2015-03-09	23:02:19
i	1	73017		2015-03-09	23:11:39
i	1	81834		2015-03-09	23:02:13
+-	+-	· ·+-			+

```
max (raw df2.timestamp).alias('lastMark') )
```

```
df_user_marks.count()
     270896
```

df_user_marks.orderBy(df_user_marks.numberOfMarks.desc()).show()

```
luserId|numberOfMarks|
                              avgMark
                                              firstMark
           -----
 45811
              18276|3.1987579339023857|2015-12-15 06:59:27|2017-07-31 09:02:42|
  8659
              9279 3.2784243991809463 2001-08-05 18:10:38 2015-12-24 20:44:49
               7638 2.5974731605132235 2012-12-13 02:21:42 2017-08-04 03:03:56
270123
179792
               7515 3.2083166999334662 2006-10-10 06:41:32 2015-02-12 06:35:57
               7410 3.2201754385964914 2009-04-07 02:50:08 2017-08-04 00:28:21
228291
243443
               6320 | 1.5760284810126581 | 2000-04-07 | 13:13:16 | 2017-08-01 | 16:26:43 |
               6094 2.804972103708566 2002-07-05 04:57:01 2017-08-03 16:16:38
98415
               6024 3.498256972111554 2001-09-10 15:15:01 2017-08-04 02:08:42
229879
98787
               5814 2.4380804953560373 2016-06-25 04:55:26 2017-03-30 21:30:34
172224
               5701 3.7478512541659357 2000-08-03 08:06:12 2016-12-13 06:03:58
               5619 | 0.8632318917956932 | 2016-10-31 | 01:15:09 | 2017-08-02 | 02:42:08 |
230417
70648
               5356 3.011669156086632 2006-11-13 17:54:17 2011-10-04 01:46:59
194690
               5206 | 2.466192854398771 | 2006-01-09 07:59:42 | 2017-07-17 00:04:12 |
               107720
               4946 3.2052163364334816 2016-03-27 15:12:44 2016-03-27 16:10:44
24025
               4921 | 1.1382849014427963 | 2016-01-07 | 16:18:19 | 2017-03-27 | 07:13:06 |
165352
               4834 3.0393049234588334 1998-02-13 20:52:06 2017-08-02 02:43:20
243331
101276
               4834 2.863260239966901 2016-05-09 08:52:03 2016-05-09 09:36:29
74275
               4815 | 2.9912772585669782 | 2012-09-27 01:52:35 | 2017-08-03 22:08:35 |
               4785 3.2189132706374086 2008-09-07 22:07:23 2012-01-08 03:12:52
41190
```

only showing top 20 rows

```
# count years from today to firstMark
from pyspark.sql import functions as f
from pyspark.sql import types as t

df_min_max = df_user_marks.withColumn('yearsSpent', f.datediff(f.current_date(), df_user_mark

# count avgTimeBetweenMarks in days

df_min_max_avg = df_min_max.withColumn('avgTimeBetweenMarks', f.datediff(df_min_max.lastMark,

# Find favourite genre:
"""

My approach:
```

```
1. Find string of all genres of movies watched by certain user
2. convert string splitted by comma to list
3. map list {Genre:Count}
4. select the most frequent Genre by Count
# 1 step
from pyspark.sql.functions import sum as sum_
fav df = raw df2.join(movie profile, raw df2.movieId==movie profile.movieId)\
  .select(raw_df2.userId, movie_profile.genres)\
  .groupBy(raw df2.userId)\
  .agg(f.concat_ws(",", f.collect_list(movie_profile.genres)).alias('allGenres'))
fav_df.show()
     userId
                   allGenres
         100 Drama, Action, Thri...
       1000 Comedy, Family, Mus...
     | 10000|TV Movie,Drama,Hi...|
     |100000|Horror,Mystery,Ad...|
     |100004|Crime,Drama,Myste...|
     |100005|Drama,Thriller,Dr...|
     100006 Drama, Comedy, Dram...
     |100007|Drama,Thriller,Cr...|
     |100008|Crime,Drama,Thril...|
     |100009|Comedy,Thriller,C...|
     10001 Drama, Mystery, Act...
     |100010|Drama,History,War...|
     |100014|Drama,Action,Crim...|
     |100015|Horror,Mystery,Ad...|
     |100020|Drama,Drama,Drama...|
     100021 Drama, Thriller, Cr...
     |100022|Crime,Drama,Myste...|
     |100025|Drama,Romance,Adv...|
     |100029|Adventure,Drama,F...|
     | 10003|Drama,Crime,Drama...|
     +----+
    only showing top 20 rows
# 2 step
fav_df_list = fav_df.select(fav_df.userId, split(fav_df.allGenres, ",").alias("genresArray"))
fav df list.show()
                genresArray
     userId
```

```
100 [Drama, Action, T...]
       1000 [Comedy, Family, ...|
      10000 [TV Movie, Drama,...]
     100000 [Horror, Mystery,...
     |100004|[Crime, Drama, My...|
     |100005|[Drama, Thriller,...|
     |100006|[Drama, Comedy, D...|
     |100007|[Drama, Thriller,...|
     |100008|[Crime, Drama, Th...|
     |100009|[Comedy, Thriller...|
     10001 [Drama, Mystery, ...]
     |100010|[Drama, History, ...|
     |100014|[Drama, Action, C...|
     |100015|[Horror, Mystery,...|
     |100020|[Drama, Drama, Dr...|
     |100021|[Drama, Thriller,...|
     |100022|[Crime, Drama, My...|
     |100025|[Drama, Romance, ...|
     |100029|[Adventure, Drama...|
     | 10003|[Drama, Crime, Dr...|
     +----+
    only showing top 20 rows
fav df list.count()
    265049
# 3 step
temp = (fav_df_list.withColumn("Dist",f.array_distinct("genresArray"))
             .withColumn("Counts",f.expr("""transform(Dist,x->
                          aggregate(genresArray,0,(acc,y)-> IF (y=x, acc+1,acc))
             .withColumn("Map",f.arrays_zip("Dist","Counts")
             ))
out = temp.withColumn("favoriteGenre",
                   f.expr("""element_at(array_sort(Map,(first,second)->
        CASE WHEN first['Counts']>second['Counts'] THEN -1 ELSE 1 END),1)['Dist']"""))
out.show()
     userId
                     genresArray
                                                Dist
                                                                   Counts
                ------
        100|[Drama, Action, T...|[Drama, Action, T...| [2, 1, 1, 1, 1]|[{Drama, 2}, {Act
       1000 | [Comedy, Family, ... | [Comedy, Family, ... | [1, 2, 1, 3, 1, 1] | [{Comedy, 1}, {Family, ... |
     | 10000|[TV Movie, Drama,...|[TV Movie, Drama,...|       [1, 2, 1, 1]|[{TV Movie, 1}, √
     |100000|[Horror, Mystery,...|[Horror, Mystery,...|[2, 3, 5, 11, 1, ...|[{Horror, 2}, {My
     |100004|[Crime, Drama, My...|[Crime, Drama, My...|[27, 81, 15, 33, ...|[{Crime, 27}, {Dr
     |100005|[Drama, Thriller,...|[Drama, Thriller,...| [3, 1, 1, 1]|[{Drama, 3}, {Thr
     |100006|[Drama, Comedy, D...|[Drama, Comedy, R...|[6, 4, 1, 1, 1, 2...|[{Drama, 6}, {Con
     | 100007| [Drama, Thriller,...| [Drama, Thriller,...| [5, 4, 2, 3, 1, 1...| [{Drama, 5}, {Thr
     |100008||Crime, Drama, Th...||Crime, Drama, Th...||39, 99, 45, 5, 3...||{Crime, 39}, {Dr
```

```
| 100009 | Comedy, Thriller... | Comedy, Thriller... | [4, 2, 1, 6, 2, 2... | {Comedy, 4}, {The | 10001 | Comedy, Mystery, ... | Comedy, Thriller... | [4, 2, 1, 1, 3, 1... | {Comedy, 4}, {The | 10001 | Comedy, Mystery, ... | Comedy, Mystery, ... | [4, 2, 1, 1, 3, 1... | {Comedy, 4}, {Mystery, 100010 | Comedy, Mystery, ... | Comedy, Mystery, .
```

final UserProfile dataframe

user profile.show()

+	+	+		
userId numberOfMarks	yearsSpent	avgMark	avgTimeBetweenMarks	favor
+	+	+	+-	
100	12.714579055441478	3.333333333333333	0.0	
1000 6	21.284052019164957	3.5	0.0	
10000 10	4.725530458590007	4.25	0.0	
100005 5	6.176591375770021	4.4	0.0	
100006 17	9.949349760438055	3.676470588235294	0.0	
100007 18	15.84394250513347	3.25	0.0	
100008 307	24.18069815195072	3.996742671009772	0.08496732026143791	
100009 21	13.905544147843942	3.333333333333333	0.0	
10001 11	6.316221765913758	4.27272727272725	0.0	
100010 36	24.908966461327857	3.80555555555554	0.0	
100014 15	11.520876112251882	2.766666666666666	0.0	
100020 100	6.12457221081451	3.425	0.010101010101010102	
100021 2216	16.520191649555098	3.1656137184115525	0.009932279909706547	
100022 292	19.802874743326488	3.914383561643836	0.0	
100025 126	21.56605065023956	3.5317460317460316	0.0	
100029 20	17.727583846680357	3.5	0.0	
10003 50	14.858316221765914	2.53	0.0	
100031 53	21.782340862423	4.018867924528302	0.0	
100039 50	16.134154688569474	3.42	0.0	
100040	13.708418891170432	4.0	0.0 S	cience
+	+	+	·	

user_profile.count()

```
# !pip install pyspark==<compatible-spark-version>
# !pyspark --packages io.delta:delta-core_2.12:1.1.0 --conf "spark.sql.extensions=io.delta.sq

# user_df.write.format("delta").mode("append").save("/tmp/delta/UserProfile")
# user_df.write.format("delta").mode("append").saveAsTable("default.UserProfile")
```

×