RDDs, Datasets and DataFrames

Spark - Exercises

- Log filtering
 - Input: a simplified log of a web server (i.e., a textual file)
 - Each line of the file is associated with a URL request
 - Output: the lines containing the word "google"
 - Store the output in an HDFS folder

Exercise #30 - Example

Input file

```
66.249.69.97 - - [24/Sep/2014:22:25:44 +0000] "GET http://www.google.com/bot.html" 66.249.69.97 - - [24/Sep/2014:22:26:44 +0000] "GET http://www.google.com/how.html" 66.249.69.97 - - [24/Sep/2014:22:28:44 +0000] "GET http://dbdmg.polito.it/course.html" 71.19.157.179 - - [24/Sep/2014:22:30:12 +0000] "GET http://www.google.com/faq.html" 66.249.69.97 - - [24/Sep/2014:31:28:44 +0000] "GET http://dbdmg.polito.it/thesis.html"
```

Output

```
66.249.69.97 - - [24/Sep/2014:22:25:44 +0000] "GET http://www.google.com/bot.html" 66.249.69.97 - - [24/Sep/2014:22:26:44 +0000] "GET http://www.google.com/how.html" 71.19.157.179 - - [24/Sep/2014:22:30:12 +0000] "GET http://www.google.com/faq.html"
```

- Log analysis
 - Input: log of a web server (i.e., a textual file)
 - Each line of the file is associated with a URL request
 - Output: the list of distinct IP addresses associated with the connections to a google page (i.e., connections to URLs containing the term "www.google.com")
 - Store the output in an HDFS folder

Exercise #31 - Example

Input file

```
66.249.69.97 - - [24/Sep/2014:22:25:44 +0000] "GET http://www.google.com/bot.html" 66.249.69.97 - - [24/Sep/2014:22:26:44 +0000] "GET http://www.google.com/how.html" 66.249.69.97 - - [24/Sep/2014:22:28:44 +0000] "GET http://dbdmg.polito.it/course.html" 71.19.157.179 - - [24/Sep/2014:22:30:12 +0000] "GET http://www.google.com/faq.html" 66.249.69.95 - - [24/Sep/2014:31:28:44 +0000] "GET http://dbdmg.polito.it/thesis.html" 66.249.69.97 - - [24/Sep/2014:56:26:44 +0000] "GET http://www.google.com/how.html" 56.249.69.97 - - [24/Sep/2014:56:26:44 +0000] "GET http://www.google.com/how.html"
```

Output

```
66.249.69.97
71.19.157.179
56.249.69.97
```

- Maximum value
 - Input: a collection of (structured) textual csv files containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: report the maximum value of PM10
 - Print the result on the standard output

Exercise #32 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

60.2

- Top-k maximum values
 - Input: a collection of (structured) textual csv files containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: report the top-3 maximum values of PM10
 - Print the result on the standard output

Exercise #33 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

60.2

55.5

52.5

- Readings associated with the maximum value
 - Input: a collection of (structured) textual csv files containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: the line(s) associated with the maximum value of PM10
 - Store the result in an HDFS folder

Exercise #34 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,60.2

$2,2016-01-03,52.5
```

Output

s1,2016-01-02,60.2
s1,2016-01-03,60.2

- Dates associated with the maximum value
 - Input: a collection of (structured) textual csv files containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: the date(s) associated with the maximum value of PM10
 - Store the result in an HDFS folder

Exercise #35 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,60.2

$2,2016-01-03,52.5
```

Output

2016-01-02 2016-01-03

- Average value
 - Input: a collection of (structured) textual csv files containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: compute the average PM10 value
 - Print the result on the standard output

Exercise #36 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

39.86

- Maximum values
 - Input: a textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: the maximum value of PM10 for each sensor
 - Store the result in an HDFS file

Exercise #37 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

(s1,60.2) (s2,52.5)

- Pollution analysis
 - Input: a textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: the sensors with at least 2 readings with a PM10 value greater than the critical threshold 50
 - Store in an HDFS file the sensorIds of the selected sensors and also the number of times each of those sensors is associated with a PM10 value greater than 50

Exercise #38 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output (\$1,2)

- Critical dates analysis
 - Input: a textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: an HDFS file containing one line for each sensor
 - Each line contains a sensorId and the list of dates with a PM10 value greater than 50 for that sensor

Exercise #39 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

```
($1, [2016-01-02, 2016-01-03])
($2, [2016-01-03])
```

Exercise #39 bis

- Critical dates analysis
 - Input: a textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: an HDFS file containing one line for each sensor
 - Each line contains a sensorId and the list of dates with a PM10 value greater than 50 for that sensor
 - Also the sensors which have never been associated with a PM10 values greater than 50 must be included in the result (with an empty set)

Exercise #39 bis - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5

$3,2016-01-03,12.5
```

Output

```
($1, [2016-01-02, 2016-01-03])
($2, [2016-01-03])
($3, [])
```

- Order sensors by number of critical days
 - Input: a textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - Output: an HDFS file containing the sensors ordered by the number of critical days
 - Each line of the output file contains the number of days with a PM10 value greater than 50 for a sensor s and the sensorId of sensor s

Exercise #40 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

Output

2, S1 1, S2

- Top-k most critical sensors
 - Input:
 - A textual csv file containing the daily value of PM10 for a set of sensors
 - Each line of the files has the following format sensorId, date, PM10 value (μg/m³)\n
 - The value of k
 - It is an argument of the application

- Top-k most critical sensors
 - Output:
 - An HDFS file containing the top-k critical sensors
 - The "criticality" of a sensor is given by the number of days with a PM10 value greater than 50
 - Each line contains the number of critical days and the sensorId

Exercise #41 - Example

Input file

```
$1,2016-01-01,20.5

$2,2016-01-01,30.1

$1,2016-01-02,60.2

$2,2016-01-02,20.4

$1,2016-01-03,55.5

$2,2016-01-03,52.5
```

- k = 1
- Output

2, S1

- Mapping Question-Answer(s)
 - Input:
 - A large textual file containing a set of questions
 - Each line contains one question
 - Fach line has the format
 - QuestionId, Timestamp, TextOfTheQuestion
 - A large textual file containing a set of answers
 - Each line contains one answer
 - Each line has the format
 - AnswerId, QuestionId, Timestamp, TextOfThe Answer

Output:

- A file containing one line for each question
- Each line contains a question and the list of answers to that question
 - QuestionId, TextOfTheQuestion, list of Answers

Exercise #42 - Example

Questions

Q1,2015-01-01,What is ..? Q2,2015-01-03,Who invented ..

Answers

A1,Q1,2015-01-02,lt is .. A2,Q2,2015-01-03,John Smith A3,Q1,2015-01-05,Ithink it is ..

Exercise #42 - Example

Output

```
(Q1,([What is ..?],[It is .., I think it is ..]))
(Q2,([Who invented ..],[John Smith]))
```

Exercise #43 - 1

- Critical bike sharing station analysis
- Input:
 - A textual csv file containing the occupancy of the stations of a bike sharing system
 - The sampling rate is 5 minutes
 - Each line of the file contains one sensor reading/sample has the following format
 - stationId,date,hour,minute,num_of_bikes,num_of_free_slots
 - Some readings are missing due to temporarily malfunctions of the stations
 - Hence, the number of samplings is not exactly the same for all stations
 - The number of distinct stations is 100

Exercise #43 – 2

Input:

- A second textual csv file containing the list of neighbors of each station
 - Each line of the file has the following format stationId_x, list of neighbors of stationId_x
 - E.g.,
 s1,s2 s3
 means that s2 and s3 are neighbors of s1

Exercise #43 - 3

Outputs:

- Compute the percentage of critical situations for each station
 - A station is in a critical situation if the number of free slots is below a user provided threshold (e.g., 3 slots)
 - The percentage of critical situations for a station Si is defined as (number of critical readings associated with Si)/(total number of readings associated with Si)

Exercise #43 – 4

- Store in an HDFS file the stations with a percentage of critical situations higher than 80% (i.e., stations that are almost always in a critical situation and need to be extended)
 - Each line of the output file is associated with one of the selected stations and contains the percentage of critical situations and the stationId
 - Sort the stored stations by percentage of critical situations

Exercise #43 – 5

- Compute the percentage of critical situations for each pair (timeslot, station)
 - Timeslot can assume the following 6 values
 - [0-3]
 - **[**4-7]
 - **8-11**
 - **[**12-15]
 - **•** [16-19]
 - **[**20-23]

Exercise #43 – 6

- Store in an HDFS file the pairs (timeslot, station)
 with a percentage of critical situations higher than
 80% (i.e., stations that need rebalancing
 operations in specific timeslots)
 - Each line of the output file is associated with one of the selected pairs (timeslot, station) and contains the percentage of critical situations and the pair (timeslot, stationId)
 - Sort the result by percentage of critical situations

Exercise #43 - 7

- Select a reading (i.e., a line) of the first input file if and only if the following constraints are true
 - The line is associated with a full station situation
 - i.e., the station Si associated with the current line has a number of free slots equal to o
 - All the neighbor stations of the station Si are full in the time stamp associated with the current line
 - i.e., bikers cannot leave the bike at Station Si and also all the neighbor stations are full in the same time stamp
- Store the selected readings/lines in an HDFS file and print on the standard output the total number of such lines

- Misleading profile selection
- Input:
 - A textual file containing the list of movies watched by the users of a video on demand service
 - Each line of the file contains the information about one visualization
 - userid, movieid, start-timestamp, end-timestamp
 - The user with id userid watched the movie with id movieid from start-timestamp to end-timestamp

Input:

- A second textual file containing the list of preferences for each user
 - Each line of the file contains the information about one preference userid, movie-genre
 - The user with id userid liked the movie of type moviegenre

Input:

- A third textual file containing the list of movies with the associated information
 - Each line of the file contains the information about one movie
 - movieid, title, movie-genre
 - There is only one line for each movie
 - i.e., each movie has one single genre

- Select the userids of the list of users with a misleading profile
 - A user has a misleading profile if more than threshold% of the movies he/she watched are not associated with a movie genre he/she likes
 - threshold is an argument/parameter of the application and it is specified by the user
- Store the result in an HDFS file

- Profile update
- Input:
 - A textual file containing the list of movies watched by the users of a video on demand service
 - Each line of the file contains the information about one visualization
 - userid, movieid, start-timestamp, end-timestamp
 - The user with id userid watched the movie with id movieid from start-timestamp to end-timestamp

Input:

- A second textual file containing the list of preferences for each user
 - Each line of the file contains the information about one preference userid, movie-genre
 - The user with id userid liked the movie of type moviegenre

Input:

- A third textual file containing the list of movies with the associated information
 - Each line of the file contains the information about one movie
 - movieid, title, movie-genre
 - There is only one line for each movie
 - i.e., each movie has one single genre

- Select for each user with a misleading profile
 (according to the same definition of Exercise #44) the
 list of movie genres that are not in his/her preferred
 genres and are associated with at least 5 movies
 watched by the user
- Store the result in an HDFS file
 - Each line of the output file is associated with one pair (user, selected misleading genre) associated with him/her
 - The format is userid, selected (misleading) genre
 - Users associated with a list of selected genres are associated with multiple lines of the output file

- Time series analysis
- Input:
 - A textual file containing a set of temperature readings
 - Each line of the file contains one timestamp and the associated temperature reading
 - timestamp, temperature
 - The format of the timestamp is the Unix timestamp that is defined as the number of seconds that have elapsed since oo:oo:oo Coordinated Universal Time (UTC), Thursday, 1 January 1970
 - The sample rate is 1 minute
 - i.e., the difference between the timestamps of two consecutive readings is 60 seconds

- Consider all the windows containing 3 consecutive temperature readings and
 - Select the windows characterized by an increasing trend
 - A window is characterized by an increasing trend if for all the temperature readings in it temperature(t)>temperature(t-60 seconds)
 - Store the result into an HDFS file

Exercise #46 - Example

Input file

```
1451606400,12.1
1451606460,12.2
1451606520,13.5
1451606580,14.0
1451606640,14.0
1451606700,15.5
1451606760,15.0
```

Output file

1451606400,12.1,1451606460,12.2,1451606520,13.5 1451606460,12.2,1451606520,13.5,1451606580,14.0

Input:

- A CSV file containing a list of user profiles
 - Header
 - name,age,gender
 - Each line of the file contains the information about one user

- Select male users (gender="male"), increase by one their age, and store in the output folder name and age of these users sorted by decreasing age and ascending name (if the age value is the same)
- The output does not contain the header line

Example of input data:

```
name,age,gender
Paul,40,male
John,40,male
David,15,male
Susan,40,female
Karen,34,female
```

Example of expected output:

```
John,41
Paul,41
David,16
```

- Implement two different solutions for this exercise
 - A solution based only on DataFrames
 - A solution based on SQL like queries executed on a temporary table associated with the input data

Input:

- A CSV file containing a list of user profiles
 - Header
 - name,age,gender
 - Each line of the file contains the information about one user

- Select the names occurring at least two times and store in the output folder name and average(age) of the selected names
- The output does not contain the header line

Example of input data:

```
name, age, gender
```

Paul, 40, male

Paul, 38, male

David, 15, male

Susan, 40, female

Susan, 34, female

Example of expected output:

Paul,39

Susan, 37

- Implement two different solutions for this exercise
 - A solution based only on DataFrames
 - A solution based on SQL like queries executed on a temporary table associated with the input data

Input:

- A csv file containing a list of profiles
 - Header: name, surname, age
 - Each line of the file contains one profile
 - name,surname,age

- A csv file containing one line for each profile. The original age attribute is substituted with a new attributed called rangeage of type String
 - rangeage = "[" + (age/10)*10 + "-" + (age/10)*10 +9"]"

Input:

```
name, surname, age
Paolo, Garza, 42
Luca, Boccia, 41
Maura, Bianchi, 16
```

Expected output:

name, surname, rangeage Paolo, Garza, [40-49] Luca, Boccia, [40-49] Maura, Bianchi, [10-19]

Input:

- A csv file containing a list of profiles
 - Header: name, surname, age
 - Each line of the file contains one profile
 - name,surname,age

- A csv file containing one single column called "name_surname" of type String
 - name_surname = name+" "+surname

Input:

```
name, surname, age
```

Paolo, Garza, 42

Luca, Boccia, 41

Maura, Bianchi, 16

Expected output:

name_surname

Paolo Garza

Luca Boccia

Maura Bianchi