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 values 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 values 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 values 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 values 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

- 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

- 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

Output:

- 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

- 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

- 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

Output:

- 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

Output:

- 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