DDA Exercise 05

Farjad Ahmed - 1747371 First Semester - Group 1

System Information

CPU op-mode(s): 32-bit, 64-bit

Address sizes: 39 bits physical, 48 bits virtual

Byte Order: Little Endian

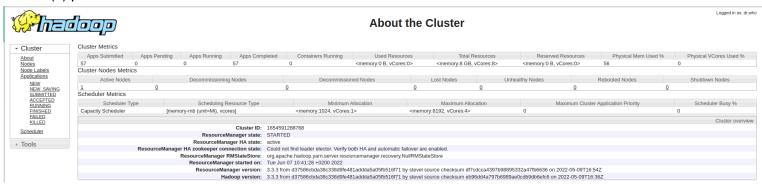
CPU(s): 8

On-line CPU(s) list: 0-7

Vendor ID: GenuineIntel

Model name: Intel(R) Core(TM) i5-8250U CPU @ 1.60GHz

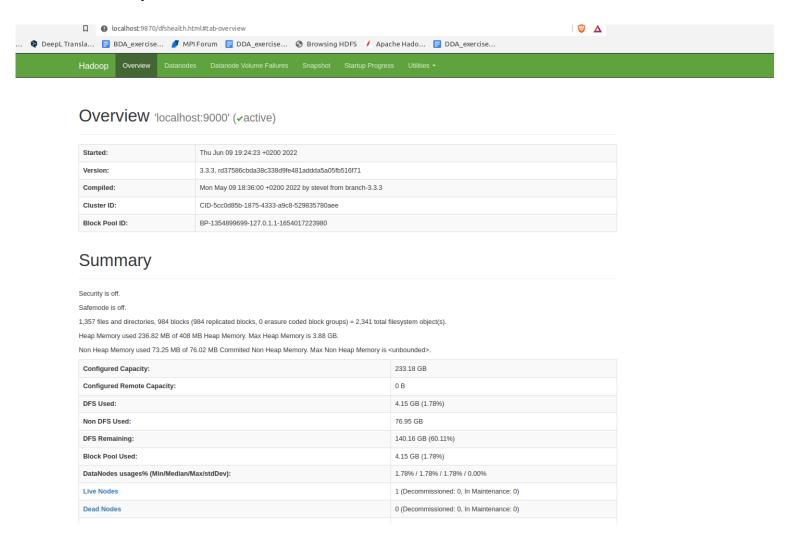
CPU family: 6 Model: 142 Thread(s) per core: 2 Core(s) per socket: 4



The report contains answers to questions, then code explanations and Performance Analysis at the very end.

MapReduce 1.3

Word Count Example:



 This example starts with downloading the gutenberg dataset and putting it into hadoop directory using the command.

hdfs dfs -put /mnt/Farjad_Ahmed/Masters/Distributed_Data_Analytics/Exercise_5/gutenberg.txt /temp/gutenberg.txt

Checking hadoop directory

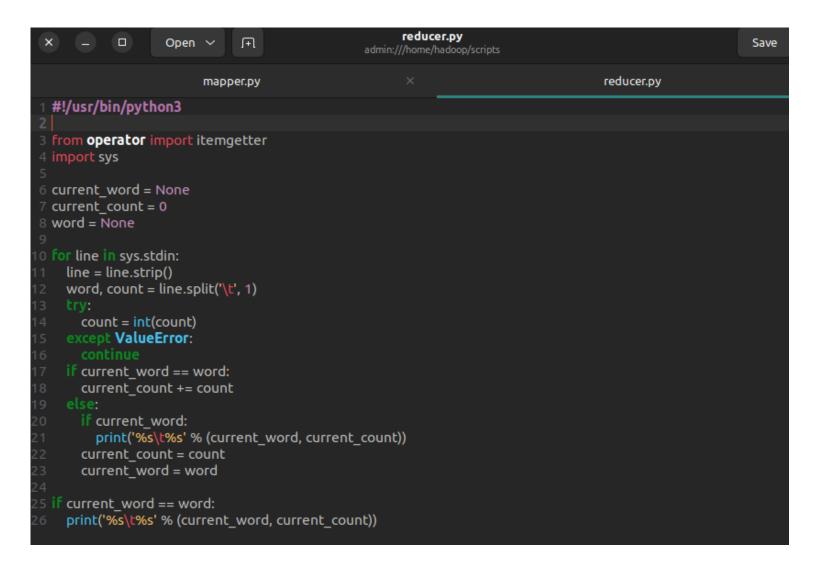
```
hadoop@joji-fish:~/hadoop-3.3.3$ hadoop fs -ls /temp
2022-06-09 19:47:33,986 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
Found 20 items
rw-r--r-- 1 hadoop supergroup
                                         726131 2022-06-04 15:21 /temp/10000merged.csv
rw-r--r--
             1 hadoop supergroup
                                          71659 2022-06-04 15:20 /temp/1000merged.csv
4876 2022-06-04 13:32 /temp/100movielense.csv
rw-r--r--
              1 hadoop supergroup
                                          38996 2022-06-02 01:07 /temp/100samples.csv
              1 hadoop supergroup
             1 hadoop supergroup 597349720 2022-06-04 18:45 /temp/10m_merged.csv
1 hadoop supergroup 956 2022-06-04 13:32 /temp/20movielense.csv
rw-r--r--
rw-r--r--
                                           9083 2022-06-02 01:07 /temp/20samples.csv
rw-r--r--
              1 hadoop supergroup
rw-r--r--
              1 hadoop supergroup
                                        1586336 2022-06-01 02:54 /temp/4300-0.txt
                                       1428843 2022-06-01 02:54 /temp/5000-8.txt
              1 hadoop supergroup
rw-r--r--
rw-r--r--
              1 hadoop supergroup
                                           1503 2022-06-01 15:40 /temp/Grades.csv
              1 hadoop supergroup 165693729 2022-06-04 01:12 /temp/cleansedData.csv
- rw - r - - r - -
              1 hadoop supergroup 203041987 2022-06-01 02:28 /temp/datafile.csv
rw-r--r--
              1 hadoop supergroup
                                             31 2022-06-01 02:08 /temp/foo.txt
гw-г--г--
                                         678064 2022-06-09 19:43 /temp/gutenberg.txt
-rw-r--r--
             1 hadoop supergroup
rw-r--r--
              1 hadoop supergroup 1549758993 2022-06-04 15:48 /temp/merged.csv
ΓW-Γ--Γ--
              1 hadoop supergroup 310355239 2022-06-04 13:32 /temp/movielense.csv
- rw - r - - r - -
              1 hadoop supergroup
                                       3038099 2022-06-04 15:02 /temp/movies.csv
                                         674570 2022-06-01 02:55 /temp/pg20417.txt
              1 hadoop supergroup
              1 hadoop supergroup 678266987 2022-06-04 15:02 /temp/ratings.csv
1 hadoop supergroup 522197 2022-06-04 17:15 /temp/testl.dat
rw-r--r--
hadoop@joji-fish:~/hadoop-3.3.3$
```

3. Mapper for this example is simple, it reads the designated file given in the hadoop command, The line is split by default by the whitespaces using the spli() function in the for loop which returns a list of tokens of the sentence. Another, nested for loop iterates over 'words' and print a key value combination, where each word

is the key while value is its count, since each token is just one word, hence value for each token would be 1, which is printed as <word><separator><value>. In this case the separator is a tab.



2. The reducer for this example starts with initiating 3 variables which will be used for this algorithm. It receives the input through the std.in, which was the printed information from the mapper. It starts with iterating over each print received and strips the received data and removes the leading and trailing spaces to avoid issues in splitting. The data is a key value pair hence it always breaks in a list of a key and a value when split based on the tab separator which it is in this case. In order to avoid invalid data a try except block is used by converting the count to int which it should be converted if the data is appropriate. If the count value is not invalid, it is ignored by continuing that line through 'continue' which skips that loop where it is called. The data received by reducer is sorted by key by hadoop, hence for the conditional each new word first goes into the else and if current_word is not false, it prints the output as the current_word and the current_count. It sets the current word as the last word received and the count to the count till this point. Now, if the same word comes in again, it goes to the first if condition in line 17 and the count is increased, and later printed when all similar instances are received before counting the new word. Lastly, for the last word it could be a singular instance, this case is handled separately in line 25.



3. Running this in hadoop.

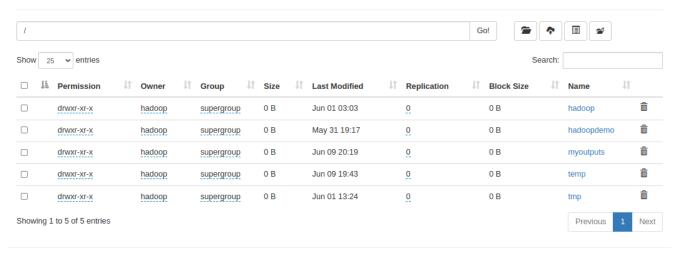
bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/mapper.py -mapper ~/scripts/mapper.py -file ~/scripts/reducer.py -reducer ~/scripts/reducer.py -input /temp/gutenberg.txt -output /myoutputs/GUTENBERG_OUT

```
A company of the comp
                                         ajoji-fish:~/hadoop-3.3.3$ bin/hadoop jar /home/hadoop/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/mapper.py -mapper ~/scripts/mapper.py -file ~/scripts/reducer.py -reducer ~/scripts/reducer.py -input /temp/gutenberg.txt -out/youtputs/fullTelBERG.0UT
```

g.StreamJob: Output directory: /mvoutputs/GUTENBERG OUT



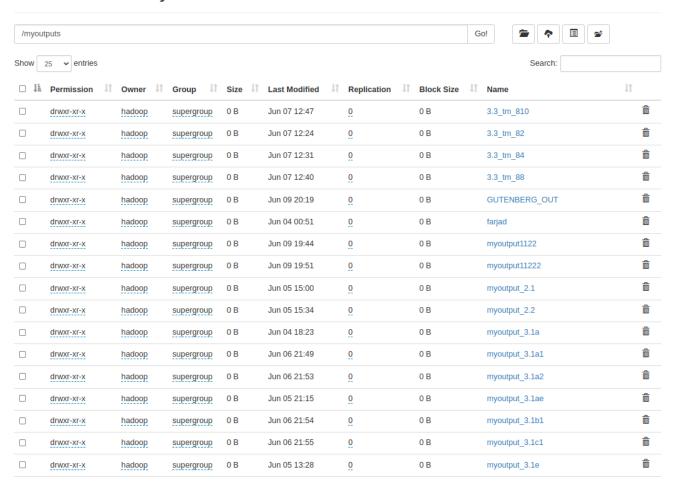
Browse Directory



Hadoop, 2022.

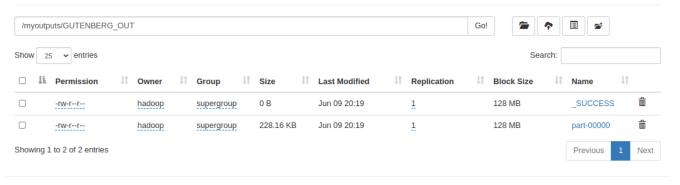


Browse Directory

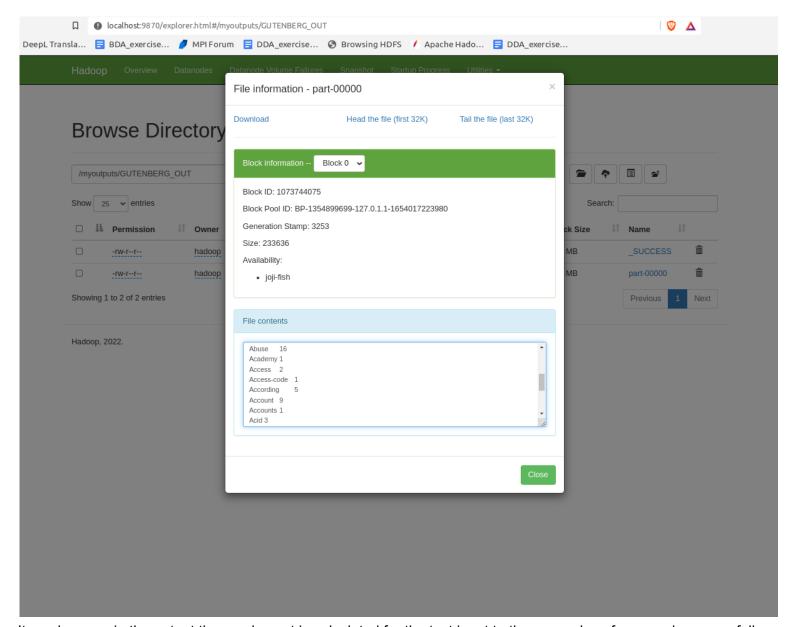




Browse Directory



Hadoop, 2022.

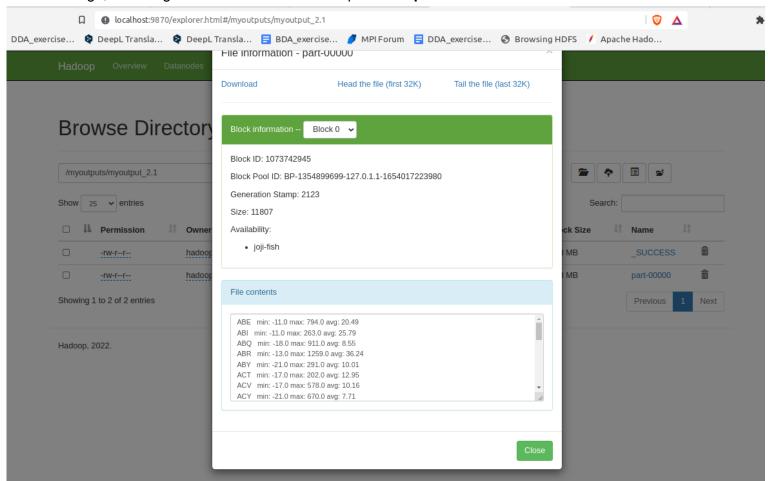


It can be seen in the output the word count is calculated for the text input to the mapreduce framework successfully.

Question 2

1. Computing the maximum, minimum, and average departure delay for each airport.[Hint: you are required to find max, min and avg in a single map reduce job]

The list is large, attaching the results from the Hadoop UI at http://localhost:9870/



2. Computing a ranking list that contains top 10 airports by their average Arrival delay. localhost:9870/explorer.html#/myoutputs/myoutput_2.2 ★ = □ = 🧧 the 🗧 DDA_exercise... 🗣 DeepL Transla... 👂 DeepL Transla... 🏮 BDA_exercise... 🥒 MPI Forum 🗧 DDA_exercise... 🚱 Browsing HDFS 🖊 Apache Hado... HIE Information - part-00000 Download Head the file (first 32K) Tail the file (last 32K) **Browse Directory** Block 0 ✓ Block ID: 1073742957 /myoutputs/myoutput_2.2 Block Pool ID: BP-1354899699-127.0.1.1-1654017223980 Generation Stamp: 2135 Show 25 ✓ entries Size: 119 1 Permission ↓↑ Owner Availability: joji-fish -rw-r--r---rw-r--r--File contents Showing 1 to 2 of 2 entries LAW 74.26 EKO 58.08 GGG 47.75 LWS 41.88 Hadoop, 2022. HDN 29.72 ASE 28.51 JAC 28.46 localhost:9870/explorer.html#/myoutputs/myoutput_2.2 ★ ■ □ = 🧝 the 🗧 DDA_exercise... 🗣 DeepL Transla... 👂 DeepL Transla... 🗧 BDA_exercise... 🥖 MPI Forum 🚦 DDA_exercise... 🚱 Browsing HDFS 🖊 Apache Hado... HIIE INTORMATION - part-UUUUU Download Head the file (first 32K) Tail the file (last 32K) **Browse Directory** Block 0 ✓ Block ID: 1073742957 /myoutputs/myoutput_2.2 Block Pool ID: BP-1354899699-127.0.1.1-1654017223980 Generation Stamp: 2135 Show 25 v entries Size: 119 **↓** Permission ↓↑ Owner Availability: joji-fish -rw-r--r-hadoo -rw-r--r--File contents Showing 1 to 2 of 2 entries LWS 41.88 ABR 33.9 HDN 29.72 Hadoop, 2022. ASE 28.51 JAC 28.46 ABI 27.93 ELM 27.93

- 3. What are your mapper.py and reduce.py solutions?
- 4. Describe step-by-step how you apply them and the outputs during this procedure.

Both of these questions are answered in the code explanation of 2.1 and 2.2 below.

Question 2.1 Code Explanation

1. Question 2 starts with downloading the required data and performing some preprocessing on it. The data contains missing values hence, for simplicity we replace missing values with 0 in the columns of interest.

```
mnt > Farjad_Ahmed > Masters > Distributed_Data_Analytics > Exercise_5 > scripts > 2.1 > 🔮 datacleanse.py > ...
      import pandas as pd
  1
      def read file(file):
  2
           df = pd.read csv(file, low memory=False)
  3
  5
      df = read file('/mnt/Farjad Ahmed/Masters/Distributed Data Analytics/Exercise 5/datafile.csv')
  6
      df['DepDelayMinutes'] = df['DepDelayMinutes'].fillna(0)
  7
      df['ArrDelayMinutes'] = df['ArrDelayMinutes'].fillna(0)
  8
      df['ArrDelay'] = df['ArrDelay'].fillna(0)
  9
      df['DepDelay'] = df['DepDelay'].fillna(0)
 10
      df.to_csv('cleansedData')
 11
```

- 2. This csv is saved and then exported to hadoop using the command below.\
 hdfs dfs -put -f /mnt/Farjad_Ahmed/Masters/Distributed_Data_Analytics/Exercise_5/scripts/2.1/cleansedData.csv
 /temp/cleansedData.csv
 - 3. Now, we define the mapper script for this question. The script is simple: since the first line contains headers, hence next(sys.stdin) iterates to the next line skipping the header. Next, it takes the input from stdin and iterates over each line printing the desired columns to the output. In my case the columns used for this task, 2.1 were the 'Origin' and 'DepDelay'. Based on this the column numbers are printed from the data list that contains stripped and split rows by ',' since it is a comma separated file.

```
amapper2.1.py

Open 
Farjad_Ahmed/mnt/Farjad_Ahmed/Masters/Dist...buted_Data_Analytics/Ex...

1 #!/usr/bin/python3

import sys
import re

next(sys.stdin)

data = line.strip().split(',')
print(data[15],'\t',data[34])
```

4. Next, the reducer for this task is defined. The reducer receives stream from the mapper, it works by creating a dictionary and storing each new origin, while the dep_delay goes into the value as a list. For each line incoming the dictionary is checked if that key already exists, if it does then the dep_delay is appended. Else, a new list is created with dep_delay as the first element, and the value for that key is set as this list.

```
1 #!/usr/bin/python3
2 # -*- coding: utf-8 -*-
3 import sys
4
5 mydict1 = {}
6 for line in sys.stdin:
7    origin, dep_delay = line.strip().split('\t')
8    dep_delay = float(dep_delay)
9    #print(origin, dep_delay)
10
11    if origin in mydict1:
12         mydict1[origin].append(dep_delay)
13    else:
14    ddelay_list = [dep_delay]
15    mydict1[origin] = ddelay_list
16    for key in mydict1:
17    print( key,' min:', min(mydict1[key]), 'max:', max(mydict1[key]), 'avg:', round(sum(mydict1[key])/len(mydict1[key]), 2))
18
```

- 5. Next, the dictionary contains each unique origin value as the key and all the associated dep_delay as the 'value' in the list container. Now, we can iterate over all elements in the dictionary using a for loop and use python's min and max functions to find the min and maximum departure delay respectively and for the avg departure delay, we sum this list and divide by its length. Lastly, for the ease of readability, the values are rounded to 2 decimal places using the round() function.
- 6. Now that the scripts are ready we can execute them. First we do it locally using cat command, then in hadoop.
- 7. Running through cat command we get the results as shown below.

```
hadoop@joji-fish:~$ hdfs dfs -cat /temp/cleansedData.csv | ~/scripts/amapper2.1.py | sort | ~/scripts/areducer2.1.py
2022-06-05 14:59:12,504 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
     min: -11.0 max: 794.0 avg: 20.49
     min: -11.0 max: 263.0 avg: 25.79
     min: -18.0 max: 911.0 avg: 8.55
     min: -13.0 max: 1259.0 avg: 36.24
     min: -21.0 max: 291.0 avg: 10.01
     min: -17.0 max: 202.0 avg: 12.95
ACT
     min: -17.0 max: 578.0 avg: 10.16
     min: -21.0 max: 670.0 avg: 7.71
     min: -34.0 max: 41.0 avg: 0.67
     min: -28.0 max: 73.0 avg: -6.07
     min: -20.0 max: 354.0 avg: 11.55
     min: -11.0 max: 1130.0 avg: 20.97
ALB
     min: -21.0 max: 981.0 avg: 11.75
     min: -15.0 max: 298.0 avg: 4.42
     min: -48.0 max: 1195.0 avg: 3.62
     min: -19.0 max: 278.0 avg: 19.04
     min: -20.0 max: 1110.0 avg: 31.64
     min: -22.0 max: 1470.0 avg: 15.07
     min: -16.0 max: 1065.0 avg: 24.85
ATW
AUS
     min: -22.0 max: 1246.0 avg: 8.72
AVL
     min: -14.0 max: 425.0 avg: 14.63
     min: -18.0 max: 783.0 avg: 32.04
AVP
     min: -20.0 max: 268.0 avg: 9.33
AZ0
     min: -20.0 max: 549.0 avg: 6.65
     min: -27.0 max: 72.0 avg: -2.48
     min: -20.0 max: 290.0 avg:
     min: -15.0 max: 216.0 avg: 9.02
     min: -18.0 max: 1068.0 avg: 13.75
BIL
     min: -19.0 max: 340.0 avg: 7.03
     min: -18.0 max: 189.0 avg: 10.41
     min: -15.0 max: 65.0 avg: 0.37
     min: -21.0 max: 324.0 avg: 8.17
     min: -18.0 max: 1458.0 avg: 29.52
     min: -20.0 max: 1447.0 avg: 8.7
ROT
     min: -22.0 max: 889.0 avg: 14.53
     min: -28.0 max: 1545.0 avg: 9.02
     min: -6.0 max: 129.0 avg: 22.86
     min: -9.0 max: 343.0 avg: 23.52
вок
     min: -23.0 max: 351.0 avg: 9.96
```

8. Running this in mapreduce in hadoop using the command.

bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/amapper2.1.py -mapper ~/scripts/amapper2.1.py -file ~/scripts/areducer2.1.py -reducer ~/scripts/areducer2.1.py -input /temp/cleansedData.csv -output /myoutputs/myoutput_2.1

Question 2.2 Code Explanation

1. For this question the logic remains the same for the mapper, the columns considered here are 'Origin' and 'ArrDelay'.

```
1 #!/usr/bin/python3
2
3 import sys
4 import re
5
6 next(sys.stdin)
7
8 for line in sys.stdin:
9  data = line.strip().split(',')
10  print(data[15],'\t',data[45])
```

2. For this question the reducer is similar to the one in 2.1 but some parts are different. The reducer receives stream from the mapper, it works by creating a dictionary and storing each new origin, while the arr_delay goes into the value as a list. For each line incoming the dictionary is checked if that key already exists, if it does then the arr_delay is appended. Else, a new list is created with arr_delay as the first element, and the value for that key is set as this list. Next, when the dictionary is populated fully, a food loop is used to iterate over each key of the dictionary and replace the value which was a list, by the average of that list, as a result we get average arrival delay times for each origin value. This is rounded to 2 decimal places for better

readability. Now since we are supposed to rank the airports, we need to sort them. Dictionaries cannot be sorted, so a tuple is created using the key. Value pairs with the tuple containing (key, value), and this is appended to a tuple list called tuplist. This list can be sorted using the sort() function of python by feeding the lambda parameter and specifying the tuple element index to be considered for sorting, through y:y[1]. The list is sorted in descending order by specifying reverse = True, this was the largest arrival delay will be the first element of this list and the smallest being the last one. Lastly, a for loop prints the top 10 elements of this list.

```
#!/usr/bin/python3
  # -*- coding: utf-8 -*-
  import sys
  mydict1 = {}
  tuplist = []
  for line in sys.stdin:
    origin, arr_delay = line.strip().split('\t')
    arr_delay = float(arr_delay)
    if origin in mydict1:
       mydict1[origin].append(arr_delay)
       ddelay_list = []
       ddelay_list.append(arr_delay)
       mydict1[origin] = ddelay_list
  for key in mydict1:
    mydict1[key] = round(sum(mydict1[key])/len(mydict1[key]), 2)
    tuplist.append(tuple((key, mydict1[key])))
    tuplist.sort(key=lambda y: y[1], reverse=True)
  for i in range(10):
    print(tuplist[i][0], tuplist[i][1])
Saving file "admin:///home/hadoop/scripts/areducer2.2.py"...
                                                                                     Python V Tab Width: 4 V
                                                                                                                     Ln 8, Col 23
                                                                                                                                          INS
```

3. Running this locally using the command below.

hdfs dfs -cat /temp/cleansedData.csv | ~/scripts/amapper2.2.py | sort | ~/scripts/areducer2.2.py

```
hadoop@joji-fish:~$ hdfs dfs -cat /temp/cleansedData.csv | ~/scripts/amapper2.2.py | sort | ~/scripts/areducer2.2.py
2022-06-05 15:42:33,665 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
LAW 74.26
EKO 58.08
6GG 47.75
LWS 41.88
ABR 33.9
HDN 29.72
ASE 28.51
JAC 28.46
ABI 27.93
ELM 27.93
ELM 27.93
```

4. Running this in hadoop, using the command below.

bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/amapper2.2.py -mapper ~/scripts/amapper2.2.py -file ~/scripts/areducer2.2.py -reducer ~/scripts/areducer2.2.py -input /temp/cleansedData.csv -output /myoutputs/myoutput_2.2

Question 3

This question will contain the answers to the question along with results screenshots, followed by a detailed explanation of mapper and reducer scripts for each part.

Since the data contained two files that needed to be merged, movies.dat and ratings.dat. I will first describe the preprocessing on this data. The .dat files were read using pd.read_csv. Columns of interest for this exercise are selected as a subset of the dataframe in df1 and df2 for movies and ratings respectively.

Next, pd.merge is used to join both the data frames using the 'inner' option in order to remove redundant columns with missing keys. The key used is 'movield'. This dataframe is then written to a csv as '10m_merged.csv'. This data will be used in all parts of this question.

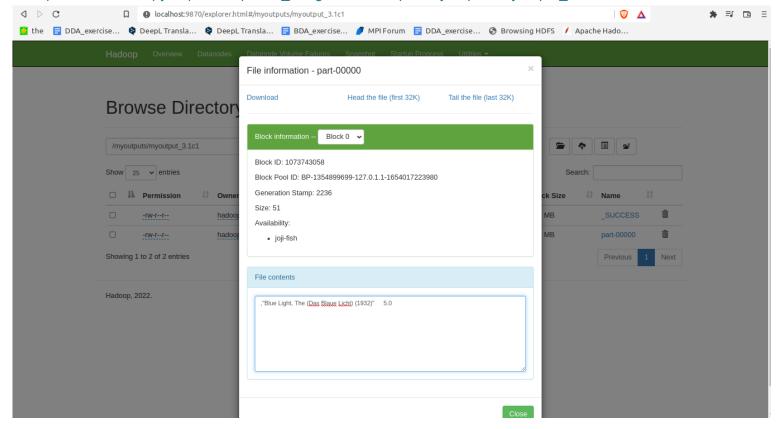
From this the acquired data set contains repeating movie titles for each user's rating, which would be required later in a question to be grouped.

The dataset looks like:

A	В	C	D	E	F
1 title	genres	movield	rating	userld	
2 Jumanji (1995)	Adventure Children Fantasy	2	2.5		
3 Jumanji (1995)	Adventure Children Fantasy	2	3	13	
4 Jumanji (1995)	Adventure Children Fantasy	2	3	14	
5 Jumanji (1995)	Adventure Children Fantasy	2	3	18	
6 Jumanji (1995)	Adventure Children Fantasy	2	3	34	
7 Jumanji (1995)	Adventure Children Fantasy	2	3	36	
8 Jumanji (1995)	Adventure Children Fantasy	2	3.5	47	
9 Jumanji (1995)	Adventure Children Fantasy	2	2.5	56	
10 Jumanji (1995)	Adventure Children Fantasy	2	3	65	
11 Jumanji (1995)	Adventure Children Fantasy	2	3	77	
12 Jumanji (1995)	Adventure Children Fantasy	2	4	91	
13 Jumanji (1995)	Adventure Children Fantasy	2	4	101	
14 Jumanji (1995)	Adventure Children Fantasy	2	5	116	
15 Jumanji (1995)	Adventure Children Fantasy	2	2.5	126	

1. Find the movie title which has the maximum average rating?

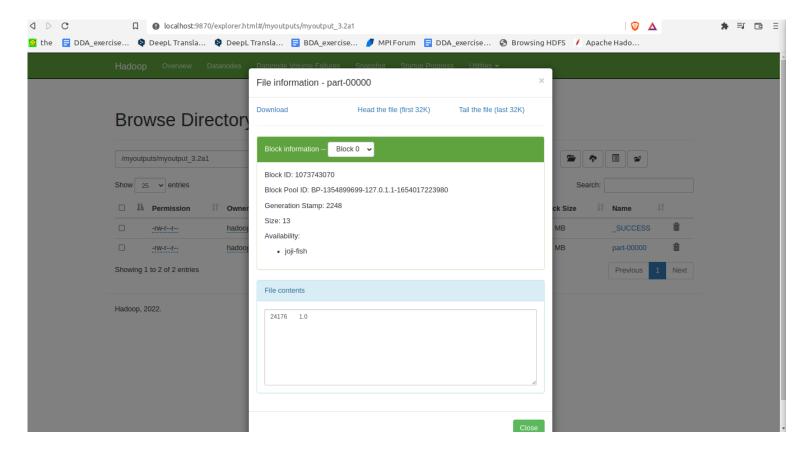
bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/amapper3.1.py -mapper ~/scripts/amapper3.1.py -file ~/scripts/areducer3.1.py -reducer ~/scripts/areducer3.1.py -input /temp/10m_merged.csv -output /myoutputs/myoutput_3.1a1



```
| Mary | Company | Company
```

2. Find the user who has assign lowest average rating among all the users who rated more than 40 Times?

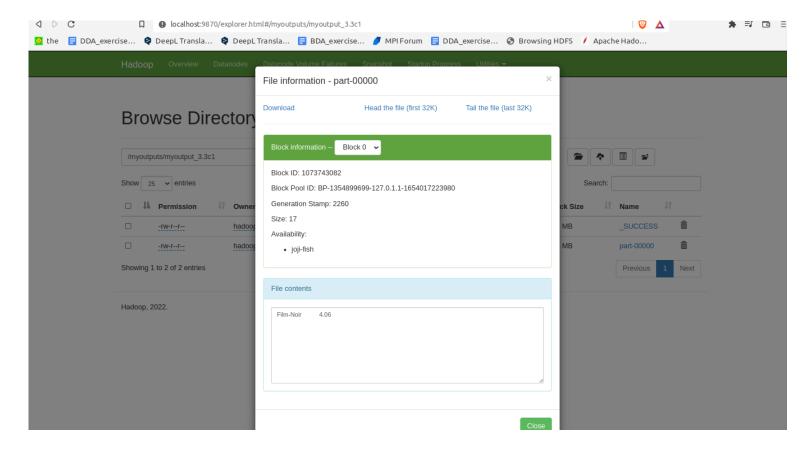
bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/amapper3.2.py -mapper ~/scripts/amapper3.2.py -file ~/scripts/areducer3.2.py -reducer ~/scripts/areducer3.2.py -input /temp/10m_merged.csv -output /myoutputs/myoutput_3.2a1



```
### Comment of the Co
```

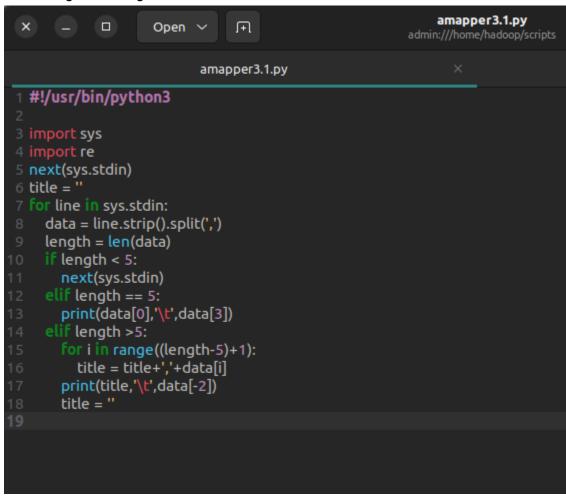
3. Find the highest average rated movie genre? [Hint: you may need to merge/combine movie.dat and rating.dat in a preprocessing step.]

bin/hadoop jar /home/hadoop/hadoop-3.3.3/share/hadoop/tools/lib/hadoop-streaming-3.3.3.jar -file ~/scripts/amapper3.3.py -mapper ~/scripts/amapper3.3.py -file ~/scripts/areducer3.3.py -reducer ~/scripts/areducer3.3.py -input /temp/10m_merged.csv -output /myoutputs/myoutput_3.3c1



```
| Column | C
```

1. For this question the movie titles and their respective rating were required. This dataset contained an interesting challenge in the movie titles. First the header is skipped using next(sys.stdin). Then we need to read each title and its associated rating. Since it is a csv and we are splitting based on ',' the title containing ',' would also be separated. As per the file it contains 5 columns, wherever ','s exist the length of 'data' will be more than 5. This case only exists for the title column. Hence, we will use negative indexing for rating for such cases while for data being of length 5 we can use either. While in case any data length is less than 5, it means there is a missing value so we skip it. For rows containing length > 5, we concatenate these titles and print the title along with rating at -2 index.



The reducer is similar to as explained in question 2, except that instead of printing the list of top ranked n items, the algorithm takes in the ratings and movie titles and does the same calculations and sorting on the list of tuples but at the end, prints the 0th element of the list.

```
#!/usr/bin/python3
2 # -*- coding: utf-8 -*-
3 import sys
5 last movie = None
6 mydict1 = {}
7 tuplist = []
9 for line in sys.stdin:
    movie, rating = line.strip().split('\t')
    rating = float(rating)
    if movie in mydict1:
       mydict1[movie].append(rating)
       rating_list = []
       rating_list.append(rating)
       mydict1[movie] = rating list
18 for key in mydict1:
    mydict1[key] = round(sum(mydict1[key])/len(mydict1[key]), 2)
    tuplist.append(tuple((key, mydict1[key])))
1 tuplist.sort(key=lambda y: y[1], reverse=True)
23 print(tuplist[0][0], '\t', tuplist[0][1])
```

3.2

1. The mapper for this question is simpler since it requires only the userId and the ratings, we can simply do this by using negative indices -1 and -2 for these items in the line, for the data list generated as a result of splitting based on ','.

```
areducer3.1.py × amapper3.2.py ×

1 #!/usr/bin/python3

2

4 import sys
5 import re
6 next(sys.stdin)

7

8 for line in sys.stdin:
9 data = line.strip().split(',')
10 print(data[-1],'\t',data[-2])

11

12
```

2. For this reducer the dictionary creation remains the same as preceding questions, i.e userId as key and ratings as a list in the value of the associated key. Next, we need to take only those users which have rated movies more than 40 times. Hence we condition out tuple creation and appending process for userId and

ratings based on the the len(<dictionary_value>) > 40, where dictionary value is a list containing ratings of that user. So all users with more than 40 ratings are filtered and shortlisted_users dictionary is created for the user key with the average of their ratings. This is then appended to the tuplist, which is then sorted in the ascending order. Lastly, the first element of this list is printed which is the lowest average rating of all the users who rated for more than 40 times.

```
areducer3.1.py
                                                               amapper3.2.py
                                                                                                               areducer3.2.py
5 last_user = None
6 mydict1 = {}
7 tuplist = []
8 shortlisted users = {}
o for line in sys.stdin:
    user, rating = line.strip().split('\t')
    rating = float(rating)
    user = str(user)
   if user in mydict1:
      mydict1[user].append(rating)
      rating_list = []
      rating list.append(rating)
      mydict1[user] = rating list
0 for key in mydict1:
    getlist = mydict1[key]
    if len(getlist) > 40:
      shortlisted users[key] = round(sum(mydict1[key])/len(mydict1[key]), 2)
       tuplist.append(tuple((key, shortlisted users[key])))
 tuplist.sort(key=lambda y: y[1], reverse=False)
  print(tuplist[0][0], '\t', tuplist[0][1])
```

3.3

1. In this case mapper requires the genre column and the rating column, this can simply be taken using the negative indexing in order to prevent the indexing complexity if we count from 0, due to inconsistency of title splitting due to ',', negative indexing solves this very simply.

2. In this case again, the dictionary creation logic remains the same but there is one major difference, the genre contains multiple genres separated by '|', hence, these genres will be split and each will be counted with the cumulative rating that was assigned to the combined genre combination. For this we first attain the rating and the genre string from the sys.stdin. Then we split the genre string with '|'. A for loop loops over each genre after the split and creates a dictionary with key as the genre and value as the rating, as explained earlier.

When all lines are read and the dictionary is populated, a for loop is used to iterate over each key of this dictionary called mydict1, and the values of each genre is replaced by the average of the ratings for that genre. This is then appended to the tuplist. After this loop the tuplist contains all genres and their average ratings, this is sorted in descending order and the first element is printed which represents the highest rated genre.

```
amapper3.3.py
                                                                                            areducer3.3.py
1 #!/usr/bin/python3
2 # -*- coding: utf-8 -*-
3 import sys
5 last_genre = None
6 mydict1 = {}
7 tuplist = []
9 for line in sys.stdin:
    genre, rating = line.strip().split('\t')
    rating = float(rating)
    parts = genre.split('|')
    for part in parts:
      genre = part
      if genre in mydict1:
         mydict1[genre].append(rating)
         rating list = []
         rating_list.append(rating)
         mydict1[genre] = rating list
23 for key in mydict1:
    mydict1[key] = round(sum(mydict1[key])/len(mydict1[key]), 2)
    tuplist.append(tuple((key, mydict1[key])))
27 tuplist.sort(key=lambda y: y[1], reverse=True)
```

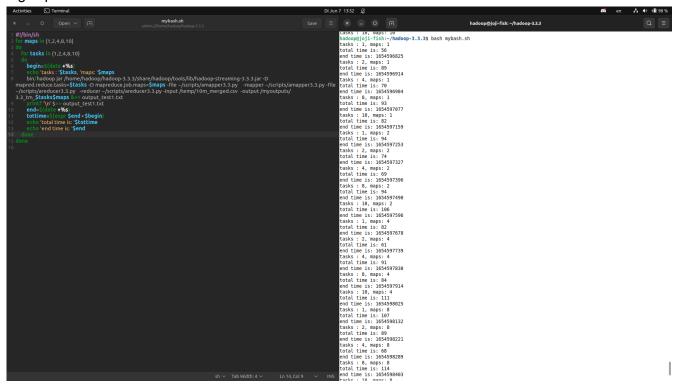
Performance

For the performance, I used a range of mappers and reducers to record running times on hadoop. **The factor under observation is how varying the number of mappers and reducers affect the performance.** This was done by automating the hadoop execution using a bash script. I'll explain the script first then share my conclusion on the performance.

The bash script iterates through two nested for loops, one for reducers and one for mappers. The params used for mappers and reducers are mapreduce.job.maps and mapred.reduce.tasks respectively. In the inner loop each loop starts with noting the time and echo-ing the mapper and reducers numbers. Next, the hadoop command is executed with dynamic food loop variables to generate the output files as well as to record the output in a text file through &>> parameter, that not only suppresses printing the output but also appends to the output_test1.txt file. After this, a line is separated through '\n' and the end time is noted. Total time is calculated by taking a difference of begin with end times. This is shown for 3.3, but changing the file name and output file names it was performed for 3.1 and 3.2 exactly the same way.



A glimpse of the bash execution:



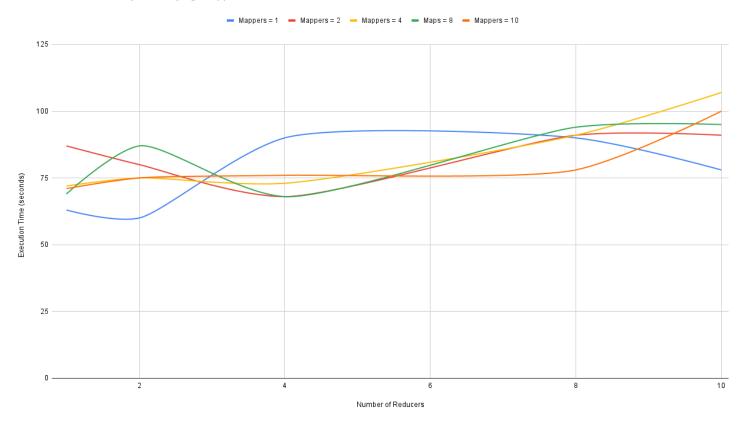
Conclusion

As far as the performance is concerned, I observed an increase in time with increasing reducers. However it is evident from the graph, as we increase the number of mappers, the execution time decreases and has a positive performance improvement impact on the algorithm's performance. This is evident by the orange line in the graph that bulges downward to the lowest point at reducer = 2. It does move upward with increasing reducers but shows the best performance time for a specific number of reducers. I think with data and the application this could be better determined so as to how many mappers and reducers can optimally improve the performance, keeping in mind the nodes available and the computational problem size at hand. This could also be due to processing required to share data among an increasing number of mappers and then combining data from multiple reducers. In an actual distributed system, network delays can play a significant role in delays.

3.1 Performance Graphs

Farjad Ahmed, DDA 05, Performance Analysis									
Maps									
1		2		4		8		10	
Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time
1	63	1	87	1	72	1	69	1	71
2	60	2	80	2	75	2	87	2	75
4	90	4	68	4	73	4	68	4	76
8	90	8	91	8	91	8	94	8	78
10	78	10	91	10	107	10	95	10	100

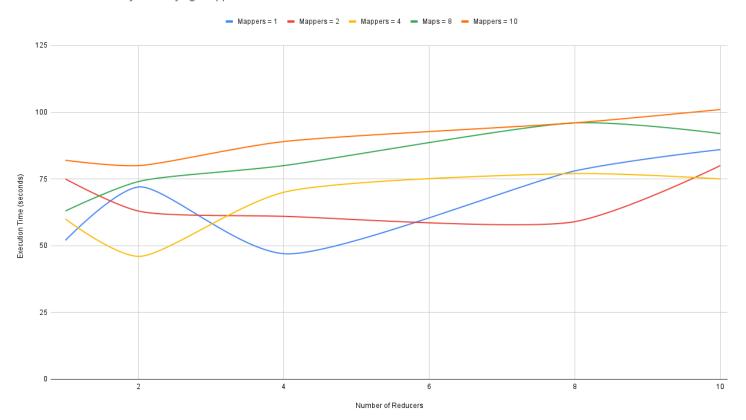
3.1 Performance Analysis: Varying Mappers and Reducers



3.2 Performance Graphs

Farjad Ahmed, DDA 05, Performance Analysis									
Maps									
1			2	4		8		10	
Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time
1	52	1	75	1	60	1	63	1	82
2	72	2	63	2	46	2	74	2	80
4	47	4	61	4	70	4	80	4	89
8	78	8	59	8	77	8	96	8	96
10	86	10	80	10	75	10	92	10	101

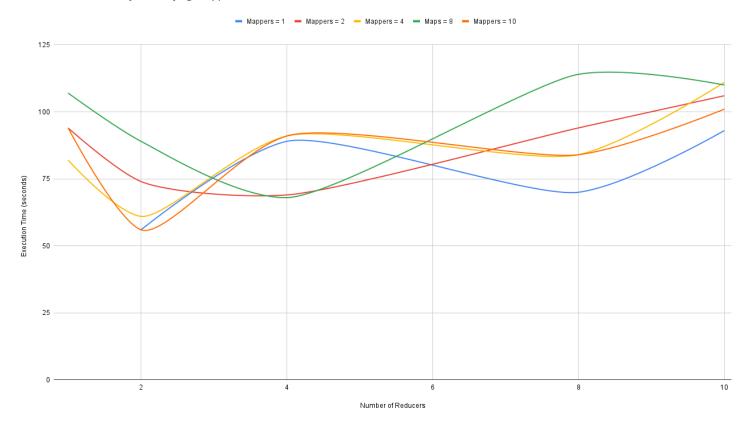
3.2 Performance Analysis: Varying Mappers and Reducers



3.3 Performance Graphs

Farjad Ahmed, DDA 05, Performance Analysis									
Maps									
1		2		4		8		10	
Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time	Tasks	Time
1	56	1	94	1	82	1	107	1	94
2	89	2	74	2	61	2	89	2	56
4	70	4	69	4	91	4	68	4	91
8	93	8	94	8	84	8	114	8	84
10	82	10	106	10	111	10	110	10	101

3.3 Performance Analysis: Varying Mappers and Reducers



References:

- [1] Apache Hadoop Documentation: https://hadoop.apache.org/docs/stable/
- [2] https://www.michael-noll.com/tutorials/writing-an-hadoop-mapreduce-program-in-python/