

School of Information Technologies

Faculty of Engineering & IT

ASSIGNMENT/PROJECT COVERSHEET - GROUP ASSESSMENT

Unit of Study:	COMP 5349 Cloud Computing		
Assignment name:	Hadoop MapReduce Programming		
Tutorial time: Thursday 4p.	m-6p.m Tutor name: Waiho Wong		

DECLARATION

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We realise that we may be asked to identify those portions of the work contributed by each of us and required to demonstrate our individual knowledge of the relevant material by answering oral questions or by undertaking supplementary work, either written or in the laboratory, in order to arrive at the final assessment mark.

Project team members					
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Assignment 1 Cloud Computing

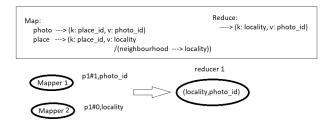
Xuhong Guo 450489321 Yuchen Zhao 460134794

Abstract. In this assignment We will write a series of Hadoop jobs to analyze a data set from flickr.com. The analysis is structured into three separate tasks which build on each other. We will practice basic Hadoop programming features and also observe the ease of use of the Hadoop framework. The data set is the same as those we have used in the labs.

1 Number of photos taken per locality.

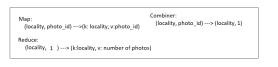
1.1 Mapreduce 1

In order to find every locality photo numbers, we would use two map reduce in this question. In the first part map reduce, we would Implementing and partitioner in Hadoop, because Hadoop supports multiple inputs as well as different mappers, so we use different values from inputs. For the photo record, which comes from n0*.txt. We use place-id as the key followed by #1 to represent the source from and values are photo-id(key:place-id, value: photo-id). For the place record, we set place-id as the key with #0 and value is place-name from place.txt file(key:place-id,value:place-name).Use the reduce-side join by the key place-id, and final output is the values which is locality name and photo-is. We also set placetype-id 22 belongs to the place-type-id7 as the question requested. By the way, the characteristic are shorten of attributes' name in files.



1.2 Mapreduce 2

The second part map reduce in question one, we use the map to read result from previous map reduce. Then we use combiner function to get the photo number for each locality, reduce is used to sum all these photo numbers by each locality which can be shown in diagrams.





1.3 Total running time

The total running time in for question 1 is 5 minutes.

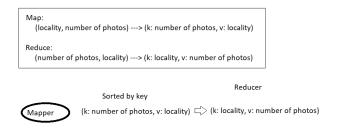
1.4 Result

```
4 'Ain Taya, Boumerdes, Algeria 7
5 'Ain al 'Awda, Rabat-Sale, Morocco 1
6 'Ali Shah 'avaz, Tehran, Iran 7
7 's-Graveland, North Holland, Netherlands 13
8 's-Gravenbrakel, Hainault, Belgium 33
9 's-Gravendeel, South Holland, Netherlands 4
10 's-Gravenpolder, Zeeland, Netherlands 11
1 's-Gravenweg, South Holland, Netherlands 19
12 's-Gravenwezel, Antwerp, Belgium 8
13 's-Gravenzande, South Holland, Netherlands 56
14 's-Heer Hendrikskinderen, Zeeland, Netherlands 16
15 's-Heer-Abtskerke, Zeeland, Netherlands 3
16 's-Heer-Arendskerke, Zeeland, Netherlands 3
```

2 The top 50 locality level places based on the number of photos taken in this locality.

2.1 Mapreduce

We could get the number of photos in each locality from the sector 1. Firstly we use map to get these source, which can be shown in table. For the reduce part, we set the a total list sized by 50, using reduce to list the top 50. Moreover, we use a reserve function here by the key. The reduce table could be seen in diagram as well.



2.2 Running Time

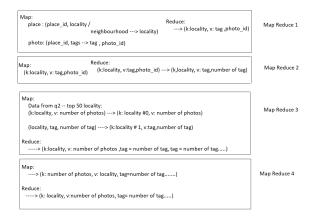
This running time is 1 minute.

2.3 Result

```
New York, NY, US, United States 1188956
Seattle, WA, US, United States 698078
NY, US, United States
                               613379
Chicago, IL, US, United States
Paris, Ile-de-France, FR, France
Washington, DC, US, United States
Tokyo, Tokyo Prefecture, JP, Japan
Taipei City, Taipei City, TW, Taiwan
                                                     494227
Sydney, NSW, AU, Australia 489161
Los Angeles, CA, US, United States
Toronto, ON, CA, Canada 445834
                                               456481
Vancouver, BC, CA, Canada 42
Melbourne, VIC, AU, Australia
                                     429842
Tokyo, Tokyo Prefecture, Japan
New York, NY, United States 318895
Austin, TX, US, United States
                                          315899
Portland, OR, US, United States 281729
San Diego, CA, US, United States 27
                                                272381
Barcelona, Catalonia, ES, Spain 265809
San Francisco, California, United States
                                                          24255
```

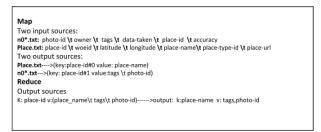
3 Ten most popular tags for each of the top-50 localities.

In this question, we plan to use four map reduce in Hadoop. The graph below shows the simple inputs and outputs of each map reduce process.



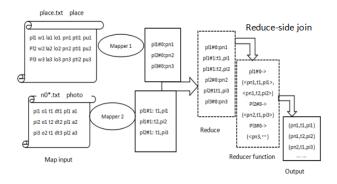
3.1 Mapreduce 1

At first, we will explain the first map reduce structure. we could know the input and output of the first map reduce process.



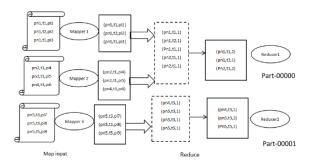
In the first map reduce, we use the function partitioner to differ the data comes from two separate input sources based on the original key place-id. The key and value comes from place information is marked as #0, key and value about photo which comes from n0*.txt is marked as #1. Using reduce-side join by the same key which is place-id. we use reduce function to get the same key with different values. We get the key with place-id, value pairs with locality, tags and photo-id. Furthermore, we set the output of reduce is place-name, tags and photo-id. This changing is convenient for further using. The following diagram would show this process design well, however, the diagram shows simple example, due to large data-set, we set 7 map partitoners and 6 reduce partitoners.

The important point is that we filter out place or year tags in the reduce process here in void of useless tags, which including the lower characteristics' problems, same locality problems and/or data-time problems.



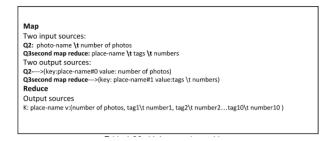
3.2 Mapreduce 2

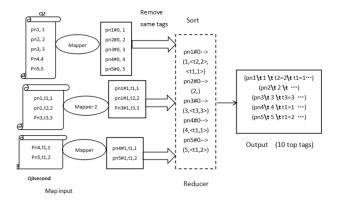
The second map reduce focuses on tags' processing. We use this map reduce to get each tags' number. In this process, map is used to read the information to lines. Reduce is used to count the same tags number under the same key. The output of this map reduce is locality, tags and number of this tag.



3.3 Mapreduce 3

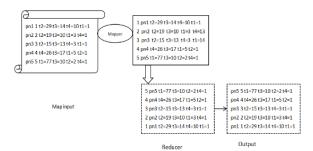
Map is used to get the previous output and input sources. In reduce process, we use keypartitioner to reduce-side join by placename. We set a dictionary to reduce the repeated tags. We set a total list sized by 10 to get the top ten tags and reversed by the tags' number decreasing.





3.4 Mapreduce 4

At last, we use a map reduce to process reserve, use -nr in reduce to make the results sorted by each places' photo numbers decreasing. Using map to read lines and using reduce to accomplish sorting. The following diagram shows the structure of fourth map reduce in this question.



3.5 Running Time

This running time is nearly 20 minutes.

3.6 Tags Cleaning

- Tag with year is deleted.
- Tag include london, uk when the place contains London, UK is deleted.
- Tag like sanfrancisco but the place is San Francisco is deleted.
- Tag list new or york, but the place is New York we keep it to prevent not the same meaning.

3.7 Result

Tokyo 498360 cat-7787, iwalking-1714, asia-7092, sandiegozoo-5343, walking-5026, zoo-4725, panda-4586, q Taipei City 484227 f;#8-41630, wedding-26390, canon-25258, 204-21338, f;tl:-20034, silkpyix-11868, 304-9705, panda-4586, q Taipei City 484227 f;#8-41630, wedding-26390, canon-18103, 4006-4265, thepstenda-2331, sonlipror-2258, bill-2164, Toronto 448384 ontario-78538, city-12582, canon-1807, tygonstillams-1229, wetrysonstillams-0-18090, arr vancouver 49942 Tolandama[caphto-187912], camezaphore-5316, shoru-34146, britishoushia-29880, normal solution and solu

4 Appendix

The HDFS location of our final output files from various executions.

4.1 Question 1

- /user/xguo8788/q1
- /user/yzha3522/q1

4.2 Question 2

- /user/xguo8788/q2
- /user/yzha3522/q2

4.3 Question 3

- /user/xguo8788/q3
- /user/yzha3522/q3