# EECS E6893 Big Data Analytic HW0

# Chong Hu ch3467

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# Problem 1. Warm-up exercises

1. Screenshots

Initialize the cluster using command line:

huchong/huchong:-/Huchong/columbis/EEC\_E6993\_Big\_Data\_AmalyticsS\_gcloud beta dataproc clusters create big\_data --optional-components-part and the components of the components



# (a) "Pi calculation"



#### Result in GUI:

```
Live Conference Confer
```

### Output in command line interface:

```
Control of the Contro
```

(b) "word count" job:

Result in command line:





#### Result in GUI:



- 2. Transformations and actions involved in each exercise:
  - (a) "Pi calculation"

First, we parallelize the data that randomly generated to create a RDD. Then we use "map" and "reduce", those two transformations to count how many points are inside the circle. In the end, we can simulate the the "Pi" value.

(b) "word count"

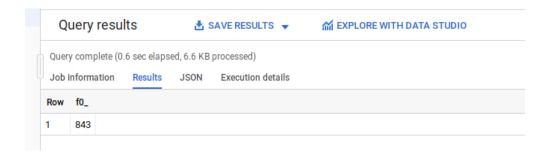
First, we use "textFile" function to create a RDD from text file. Then, we apply three transformations, "flatMap" to split word from lines, "map" to generate key-value pairs, and "reduceByKey" to count the frequency of each word. In the end, use "saveAsTextFile" (action) to save the "wordcount" result.

#### Problem 2. NYC Bike expert

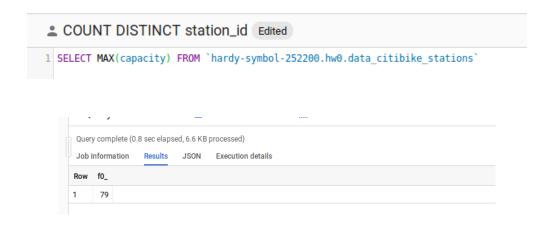
1. How many unique station\_ids are there in the dataset?

```
    COUNT DISTINCT station_id

1 SELECT COUNT(DISTINCT station_id) FROM `hardy-symbol-252200.hw0.data_citibike_stations`
```



2. Whats the largest *capacity* for a station?



List all the *station\_id* of the stations that have the largest capacity?



3. Whats the total number of bikes available in region\_id 70?

1 SELECT SUM( num bikes available ) FROM `hardy-symbol-252200.hw0.data\_citibike\_stations` WHERE region\_id = 70



#### Problem 3. Understanding William Shakespeare

1. Without any text preprocessing, here are the result for top 5 frequent words:

```
[('the', 620), ('and', 427), ('of', 396), ('to', 367), ('I', 326)]
```

```
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```

Here is my source code.

```
import pyspark
   import sys
   if __name__ == '__main__':
       inputUri = sys.argv[1]
5
       sc = pyspark.SparkContext()
       lines = sc.textFile(sys.argv[1])
       words = lines.flatMap(lambda line: line.split())
       wordCounts = words.map(lambda word: (word, 1)).reduceByKey(lambda count1, count2: count1 + count2
10
            lambda x: x[1], False)
11
       res = wordCounts.take(5)
12
       print(res)
13
```

2. With NLTK text preprocessing, here are the result for top 5 frequent words.

```
[('macb', 137), ('haue', 122), ('thou', 90), ('enter', 81), ('shall', 68)]
```

```
Note of the control o
```

Here is my source code.

```
import pyspark
   import sys
   import nltk
   stop_words = set(nltk.corpus.stopwords.words('english'))
   tokenizer = nltk.tokenize.RegexpTokenizer(r'\w+')
6
8
   def stop_words_filter(line):
9
        text = tokenizer.tokenize(line)
10
        words = []
11
        for word in text:
12
            word = word.lower()
13
            if word not in stop_words:
14
                words.append(word)
15
        return words
16
17
18
   if __name__ == '__main__':
19
        inputUri = sys.argv[1]
20
^{21}
        sc = pyspark.SparkContext()
22
        lines = sc.textFile(sys.argv[1])
23
        words = lines.flatMap(stop_words_filter)
24
        wordCounts = words.map(lambda word: (word, 1)).reduceByKey(lambda count1, count2: count1 + count2
25
            lambda x: x[1], False)
26
       res = wordCounts.take(5)
27
        print(res)
```