DATASCI W261, Machine Learning at Scale

Assignement: week #2

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Due: 2016-09-13, 8AM PST

HW2.0.

- What is a race condition in the context of parallel computation? Give an example.
- What is MapReduce?
- How does it differ from Hadoop?
- Which programming paradigm is Hadoop based on? Explain and give a simple example in code and show the code running.
- A race condition refers to the ssituation where the output is dependent on the sequence or timing of other
 uncontrollable events. For example, when a few paralle threads have interaction with some common resource, due
 to the lack of coordination and synchronization, the execution can cause invalid state of the resource, which can
 randomly affect the result of each thread.

For example, consider the following logic, where B is a global counter variable:

temp = B temp = temp + 1 B = temp It will increase the value of B by one. If it is run twice sequentially, B will be increased by two. However, if two threads are running it at the same time, there is a chance that B will only be increased by one, not two.

- MapReduce is a programming framework for processing large data sets with a parallel, distributed algorithm on a cluster. It has two phases:
 - Map(): performs filtering and sorting
 - Reduce(): performs a summary operation
- Hadoop is an open source project which provides an implementation of the MapReduce functionality. It has two
 components:
 - HDFS: Hadoop distributed file system (Storage)
 - MapReduce implementation (Calculation)

Therefore, MapReduce is a calculating component of Hadoop

- Hadoop is based on Functional Programing languages. key features of functional languages include (1) such language can accept other functions as arguments; (2) it is a declarative programming paradigm, which means programming is done with expressions.
 - Example is given below:

```
In [17]: # function to count character in a word
    def char_count(word):
        return len(word)

# apply the function on a sentence
    sentence = 'Run both the Multinomial Naive Bayes and the Bernoulli Naive Bayes alg
    orithms'
    print('Count alphabeta numbers in each word of sentence: ')
    print({word:char_count(word) for word in sentence.split()})

Count alphabeta numbers in each word of sentence:
    {'and': 3, 'both': 4, 'Run': 3, 'Multinomial': 11, 'Bernoulli': 9, 'Naive': 5, '
        algorithms': 10, 'Bayes': 5, 'the': 3}
```

HW2.1.1 Lower case and upper case sort.

```
In [18]: !curl 'http://www.gutenberg.org/cache/epub/11/pg11.txt' -o alicesTExtFilename.txt

% Total % Received % Xferd Average Speed Time Time Current
Dload Upload Total Spent Left Speed
100 163k 100 163k 0 0 25509 0 0:00:06 0:00:06 --:--:- 107k
```

```
In [19]: %%writefile pWordCount n.sh
         #!/bin/bash
         ## pWordCount.sh
         ## Author: James G. Shanahan
         ## Usage: pWordCount.sh m wordlist testFile.txt
         ## Input:
                 m = number of processes (maps), e.g., 4
         ##
         ##
                  wordlist = a space-separated list of words in quotes, e.g., "the and of"
         ##
                  inputFile = a text input file
         ##
         ## Instructions: Read this script and its comments closely.
                          Do your best to understand the purpose of each command,
         ##
         ##
                          and focus on how arguments are supplied to mapper.py/reducer.py,
                          as this will determine how the python scripts take input.
         ##
         ##
                          When you are comfortable with the unix code below,
         ##
                          answer the questions on the LMS for HW1 about the starter code.
         if [ $# -eq 0 ]
           t.hen
             echo "No arguments supplied"
             echo "To run use"
                      pWordCount.sh m wordlist inputFile"
             echo "Input:"
             echo "
                     number of processes (maps), e.g., 4"
                        wordlist = a space-separated list of words in quotes, e.g., 'the an
         d of'"
             echo "
                        inputFile = a text input file"
             exit
         fi
         ## collect user input
         m=$1 ## the number of parallel processes (maps) to run
         wordlist=$2 ## if set to "*", then all words are used
         ## a text file
         data="alicesTExtFilename.txt"
         ## 'wc' determines the number of lines in the data
         ## 'perl -pe' regex strips the piped wc output to a number
         linesindata=`wc -l $data | perl -pe 's/^.*?(\d+).*?$/$1/'`
         ## determine the lines per chunk for the desired number of processes
         linesinchunk=`echo "$linesindata/$m+1" | bc`
         ## split the original file into chunks by line
         split -l $linesinchunk $data $data.chunk.
         ## assign python mappers (mapper.py) to the chunks of data
         ## and emit their output to temporary files
         for datachunk in $data.chunk.*; do
             ## feed word list to the python mapper here and redirect STDOUT to a temporary
         file on disk
             ####
             ####
             ./mapper_n.py $datachunk "$wordlist" > $datachunk.counts &
             ####
             ####
         done
         ## wait for the mappers to finish their work
```

Overwriting pWordCount_n.sh

```
In [20]: | %%writefile mapper_n.py
         #!/usr/bin/python
         ## mapper.py
         ## Given a file and list of words, read lines and count occurrences of words
         import sys
         import re
         WORD RE = re.compile(r"[\w']+")
         filename = sys.argv[1]
         ## Words in the word list are space delimited
         wordlist = sys.argv[2].split(' ')
         counts = {}
         with open (filename, "rU") as myfile:
             for text in myfile.readlines():
                 for word in WORD RE.findall(text):
                     if word in wordlist:
                         try:
                             counts[word] += 1
                          except:
                             counts[word] = 1
         for word in counts:
             sys.stdout.write('\{0\}\t\{1\}\n'.format(word, counts[word]))
```

Overwriting mapper n.py

```
In [21]: %%writefile reducer_n.py
         #!/usr/bin/python2
         import sys
         counts_U = {}
         counts_L = {}
         for intermediate file in sys.argv:
             with open(intermediate file, 'rU') as infile:
                 \# intermediate files are word <tab> count per line
                 for line in infile.readlines():
                     word count = line.split('\t')
                     # starting with an uppercase
                     word = word count[0]
                     if len(word count) == 2 and (word[0] == word[0].upper()):
                             counts_U[word_count[0]] += int(word_count[1])
                         except KeyError:
                             counts U[word count[0]] = int(word count[1])
                     elif len(word count) == 2:
                         try:
                             counts_L[word_count[0]] += int(word_count[1])
                         except KeyError:
                             counts_L[word_count[0]] = int(word_count[1])
         for word in counts U:
             sys.stdout.write('{0}\t{1}\n'.format(word, counts U[word]))
         for word in counts L:
             sys.stdout.write('{0}\t{1}\n'.format(word, counts_L[word]))
         Overwriting reducer n.py
```

For example, I input "Alice", "the", "we", "The", "We", "How", "dog" they are sorted as alphabetical order for Upper case begin (saved .outputU file), and Lower case begin (saved .outputL)

```
In [67]: |!chmod a+x mapper_n.py
         !chmod a+x reducer n.py
         !chmod a+x pWordCount_n.sh
         !./pWordCount_n.sh 4 "Alice the we The We How dog"
         print("Upper Case")
         !cat *.outputU
         print("Lower Case")
         !cat *.outputL
         Upper Case
         Alice
               398
        ALICE
        How
               27
        HOW
               1
         The
               119
         THE
               13
         W℮
                11
        WE
         Lower Case
        dog
                3
                44
        how
        the
                1686
         we
                31
```

HW2.1.2 TOTAL SORT using multiple reducers [OPTITIONAL for this week; will be covered in next live session]

HW2.1.3 How many times does the word alice occur in the book?

Write a MapReduce job to determine this. Please pay attention to what you use for a key and value as output from your mapper.

```
In [76]: %%writefile pWordCount 13.sh
         #!/bin/bash
         ## pWordCount.sh
         ## Author: James G. Shanahan
         ## Usage: pWordCount.sh m wordlist testFile.txt
         ## Input:
         ##
                  m = number of processes (maps), e.g., 4
         ##
                  wordlist = a space-separated list of words in quotes, e.g., "the and of"
         ##
                  inputFile = a text input file
         ##
         ## Instructions: Read this script and its comments closely.
                          Do your best to understand the purpose of each command,
         ##
         ##
                          and focus on how arguments are supplied to mapper.py/reducer.py,
                          as this will determine how the python scripts take input.
         ##
         ##
                          When you are comfortable with the unix code below,
         ##
                          answer the questions on the LMS for HW1 about the starter code.
         if [ $# -eq 0 ]
           t.hen
             echo "No arguments supplied"
             echo "To run use"
                      pWordCount.sh m wordlist inputFile"
             echo "Input:"
             echo "
                     number of processes (maps), e.g., 4"
                        wordlist = a space-separated list of words in quotes, e.g., 'the an
         d of'"
             echo "
                        inputFile = a text input file"
             exit
         fi
         ## collect user input
         m=$1 ## the number of parallel processes (maps) to run
         wordlist=$2 ## if set to "*", then all words are used
         ## a text file
         data="alicesTExtFilename.txt"
         ## 'wc' determines the number of lines in the data
         ## 'perl -pe' regex strips the piped wc output to a number
         linesindata=`wc -l $data | perl -pe 's/^.*?(\d+).*?$/$1/'`
         ## determine the lines per chunk for the desired number of processes
         linesinchunk=`echo "$linesindata/$m+1" | bc`
         ## split the original file into chunks by line
         split -l $linesinchunk $data $data.chunk.
         ## assign python mappers (mapper.py) to the chunks of data
         ## and emit their output to temporary files
         for datachunk in $data.chunk.*; do
             ## feed word list to the python mapper here and redirect STDOUT to a temporary
         file on disk
             ####
             ####
             ./mapper13.py $datachunk "$wordlist" > $datachunk.counts &
             ####
             ####
         done
         ## wait for the mappers to finish their work
```

Overwriting pWordCount 13.sh

```
In [69]: %%writefile mapper13.py
         #!/usr/bin/python
         import sys
         import re
         WORD_RE = re.compile(r"[\w']+")
         filename = sys.argv[1]
         ## Words in the word list are space delimited
         wordlist = sys.argv[2].lower().split(' ')
         counts = {}
         words = re.findall('\w+', open(filename).read())
         for word in words:
                     if word.lower() in wordlist:
                         try:
                              counts[word.lower()] += 1
                          except:
                             counts[word.lower()] = 1
         for word in counts:
             sys.stdout.write('{0}\t{1}\n'.format(word, counts[word]))
```

Overwriting mapper13.py

```
In [70]: %%writefile reducer13.py
         #!/usr/bin/python
         import sys
         counts = {}
         for intermediate file in sys.argv:
             with open(intermediate_file, 'rU') as infile:
                 # intermediate files are word <tab> count per line
                 for line in infile.readlines():
                     word count = line.split('\t')
                     if len(word count) == 2:
                         trv:
                              counts[word count[0]] += int(word count[1])
                         except KeyError:
                              counts[word count[0]] = int(word count[1])
         for word in counts:
             sys.stdout.write('{0}\t{1}\n'.format(word, counts[word]))
```

Overwriting reducer13.py

```
In [77]: !chmod a+x mapper13.py
!chmod a+x reducer13.py
!chmod a+x pWordCount_13.sh
!./pWordCount_13.sh 4 "Alice"
!cat output13
```

alice 403

HW2.2.1 WORDCOUNT

Using the Enron data from and Hadoop MapReduce streaming, write the mapper/reducer job that will determine the word count (number of occurrences) of each white-space delimitted token (assume spaces, fullstops, comma as delimiters). Examine the word "assistance" and report its word count in both SPAM and HAM classes.

```
In [27]: %%writefile mapper21.py
         #!/usr/bin/python
         import re
         import sys
         WORD RE = re.compile(r''[\w']+")
         # extract words to count from first positional argument, making them all lower case
         wordlist = []
         if len(sys.argv) > 1:
             for word in sys.argv[1].strip().split():
                 wordlist.append(word.lower())
         for line in sys.stdin:
             try:
                 # Get the id, label, subject and body, respectively
                 email_id, label, subject, body = line.split('\t')
             except ValueError:
                 # if there are only 3 fields in the input, field 3 is default as body
                 email_id, label, body = line.split('\t')
                 subject = ''
             # extract words from the combined subject and body text
             for word in WORD_RE.findall(subject + ' ' + body):
                 if len(wordlist) > 0:
                     if word.lower() in wordlist:
                         print('{0}\t{1}'.format(word.lower(), 1))
                 else:
                     # otherwise count all words
                     print('{0}\t{1}'.format(word.lower(), 1))
```

Overwriting mapper21.py

```
In [82]: %%writefile reducer21.py
         #!/usr/bin/python
         from operator import itemgetter
         import sys
         tmp word = None
         tmp count = 0
         word = None
         # input comes from STDIN
         for line in sys.stdin:
             # remove leading and trailing whitespace
             line = line.strip()
             # parse the input we got from mapper.py
             word, count = line.split('\t', 1)
             # convert count to int
             try:
                 count = int(count)
             except ValueError:
                 continue
             if tmp_word == word:
                 tmp count += count
                 if tmp_word:
                     # write result to STDOUT
                     print '%s\t%s' % (tmp_word, tmp_count)
                 # update current word
                 tmp count = count
                 tmp word = word
         # do not forget to output the last word if required
         if tmp_word == word:
             print '%s\t%s' % (tmp_word, tmp_count)
         Overwriting reducer21.py
In [83]: !chmod a+x mapper21.py
         !chmod a+x reducer21.py
         !cat enronemail 1h.txt | ./mapper21.py "assistance" | sort -k1,1 | ./reducer21.py
         assistance
                         10
In [84]: !hdfs dfs -mkdir -p /user/shihyu
         !hdfs dfs -put enronemail_1h.txt /user/shihyu
```

put: `/user/shihyu/enronemail 1h.txt': File exists

```
In [86]: !hdfs dfs -rm -r results21
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-mapper "/home/cloudera/mapper21.py 'assistance'" \
-reducer /home/cloudera/reducer21.py \
-input /user/shihyu/enronemail_1h.txt \
-output results21
```

```
Deleted results21
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob6452960282347316161.jar tmpDir=null
16/09/09 21:27:53 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 21:27:53 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/09 21:27:54 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 21:27:54 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 21:27:54 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0016
16/09/09 21:27:54 INFO impl.YarnClientImpl: Submitted application application_14
73444507507 0016
16/09/09 21:27:54 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0016/
16/09/09 21:27:54 INFO mapreduce.Job: Running job: job_1473444507507_0016
16/09/09 21:28:01 INFO mapreduce. Job: Job job 1473444507507 0016 running in uber
mode : false
16/09/09 21:28:01 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 21:28:09 INFO mapreduce.Job: map 50% reduce 0%
16/09/09 21:28:10 INFO mapreduce.Job: map 100% reduce 0% 16/09/09 21:28:15 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 21:28:15 INFO mapreduce. Job: Job job 1473444507507 0016 completed succe
ssfully
16/09/09 21:28:15 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=156
                FILE: Number of bytes written=355991
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=14
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=11199
                Total time spent by all reduces in occupied slots (ms)=3594
                Total time spent by all map tasks (ms) =11199
                Total time spent by all reduce tasks (ms)=3594
                Total vcore-seconds taken by all map tasks=11199
                Total vcore-seconds taken by all reduce tasks=3594
                Total megabyte-seconds taken by all map tasks=11467776
                Total megabyte-seconds taken by all reduce tasks=3680256
        Map-Reduce Framework
                Map input records=100
                Map output records=10
                Map output bytes=130
                Map output materialized bytes=162
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=1
                Reduce shuffle bytes=162
                Reduce input records=10
                Reduce output records=1
                Spilled Records=20
                Shuffled Maps =2
                Failed Shuffles=0
                Merged Map outputs=2
```

```
In [88]: !hdfs dfs -cat /user/cloudera/results21/part-00000 ### Should be 10
assistance 10
```

HW2.2.2

Using Hadoop MapReduce and your wordcount job (from HW2.2.1) determine the top-10 occurring tokens (most frequent tokens) using a single reducer for the SPAM class and for the HAM class.

```
In [89]: |%%writefile reducer22.py
         #!/usr/bin/python
         from operator import itemgetter
         import sys, operator
         tmp word = None
         tmp count = 0
         word = None
         wordcount = {}
         # input comes from STDIN
         for line in sys.stdin:
             # remove leading and trailing whitespace
             line = line.strip()
             # parse the input we got from mapper.py
             word, count = line.split('\t', 1)
             # convert count (currently a string) to int
             try:
                 count = int(count)
             except ValueError:
                 continue
             if tmp_word == word:
                 tmp_count += count
             else:
                 if tmp_word:
                     # save count
                     wordcount[tmp_word] = tmp_count
                 tmp count = count
                 tmp_word = word
         if tmp_word == word:
             wordcount[tmp_word] = tmp_count
         # sort count top get top n counts:
         sorted count = sorted(wordcount.items(), key=operator.itemgetter(1))
         print 'Top %d counts from %d words:' %(n, len(sorted_count))
         for i in range(n):
             print '%s\t%d' %(sorted count[-i-1])
```

Overwriting reducer22.py

Top 10 tokens of SPAM messages

```
!chmod a+x mapper21.py
!chmod a+x reducer22.py
!hdfs dfs -put spam_enronemail_1h.txt /user/shihyu
!hdfs dfs -rm -r results22_spam
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-mapper /home/cloudera/mapper21.py \
-reducer /home/cloudera/reducer22.py \
-input /user/shihyu/spam_enronemail_1h.txt \
-output results22 spam
```

```
put: `/user/shihyu/spam enronemail 1h.txt': File exists
Deleted results22 spam
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob9139697224943270396.jar tmpDir=null
16/09/09 21:45:50 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 21:45:51 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 21:45:51 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 21:45:51 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 21:45:51 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507\_0021
16/09/09 21:45:52 INFO impl. YarnClientImpl: Submitted application application 14
73444507507 0021
16/09/09 21:45:52 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0021/
16/09/09 21:45:52 INFO mapreduce. Job: Running job: job_1473444507507 0021
16/09/09 21:45:58 INFO mapreduce. Job: Job job 1473444507507 0021 running in uber
mode : false
16/09/09 21:45:58 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 21:46:07 INFO mapreduce.Job: map 50% reduce 0% 16/09/09 21:46:08 INFO mapreduce.Job: map 100% reduce 0%
16/09/09 21:46:13 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 21:46:14 INFO mapreduce. Job: Job job 1473444507507 0021 completed succe
16/09/09 21:46:14 INFO mapreduce.Job: Counters: 50
        File System Counters
                FILE: Number of bytes read=180515
                FILE: Number of bytes written=716688
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=139858
                HDFS: Number of bytes written=108
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Killed map tasks=1
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=13432
                Total time spent by all reduces in occupied slots (ms)=4041
                Total time spent by all map tasks (ms) = 13432
                Total time spent by all reduce tasks (ms)=4041
                Total vcore-seconds taken by all map tasks=13432
                Total vcore-seconds taken by all reduce tasks=4041
                Total megabyte-seconds taken by all map tasks=13754368
                Total megabyte-seconds taken by all reduce tasks=4137984
        Map-Reduce Framework
                Map input records=44
                Map output records=18690
                Map output bytes=143129
                Map output materialized bytes=180521
                Input split bytes=236
                Combine input records=0
                Combine output records=0
                Reduce input groups=3787
                Reduce shuffle bytes=180521
                Reduce input records=18690
                Reduce output records=11
                Spilled Records=37380
                Shuffled Maps =2
```

```
In [97]: !hdfs dfs -cat /user/cloudera/results22_spam/part-00000
        Top 10 counts from 3787 words:
        the
             698
                565
        to
        and
               390
        your 356
                346
        of
                336
                332
        you
                235
                203
        for
        com
                151
```

Top 10 tokens of HAM messages

```
!chmod a+x mapper21.py
!chmod a+x reducer22.py
!hdfs dfs -put ham_enronemail_1h.txt /user/shihyu
!hdfs dfs -rm -r results22_ham
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-mapper /home/cloudera/mapper21.py \
-reducer /home/cloudera/reducer22.py \
-input /user/shihyu/ham_enronemail_1h.txt \
-output results22 ham
```

```
put: `/user/shihyu/ham enronemail 1h.txt': File exists
rm: `results22 ham': No such file or directory
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob6025724084762220367.jar tmpDir=null
16/09/09 21:50:37 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 21:50:37 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 21:50:38 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 21:50:38 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 21:50:38 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0022
16/09/09 21:50:38 INFO impl. YarnClientImpl: Submitted application application 14
73444507507 0022
16/09/09 21:50:38 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0022/
16/09/09 21:50:38 INFO mapreduce. Job: Running job: job_1473444507507 0022
16/09/09 21:50:46 INFO mapreduce. Job: Job job 1473444507507 0022 running in uber
mode : false
16/09/09 21:50:46 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 21:50:54 INFO mapreduce.Job: map 50% reduce 0% 16/09/09 21:50:55 INFO mapreduce.Job: map 100% reduce 0%
16/09/09 21:51:00 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 21:51:01 INFO mapreduce. Job: Job job 1473444507507 0022 completed succe
16/09/09 21:51:01 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=137520
                FILE: Number of bytes written=630692
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=93864
                HDFS: Number of bytes written=106
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=12500
                Total time spent by all reduces in occupied slots (ms)=4040
                Total time spent by all map tasks (ms)=12500
                Total time spent by all reduce tasks (ms)=4040
                Total vcore-seconds taken by all map tasks=12500
                Total vcore-seconds taken by all reduce tasks=4040
                Total megabyte-seconds taken by all map tasks=12800000
                Total megabyte-seconds taken by all reduce tasks=4136960
        Map-Reduce Framework
                Map input records=56
                Map output records=14221
                Map output bytes=109072
                Map output materialized bytes=137526
                Input split bytes=234
                Combine input records=0
                Combine output records=0
                Reduce input groups=2768
                Reduce shuffle bytes=137526
                Reduce input records=14221
                Reduce output records=11
                Spilled Records=28442
                Shuffled Maps =2
                Failed Shuffles=0
```

```
In [100]: !hdfs dfs -cat /user/cloudera/results22_ham/part-00000
          Top 10 counts from 2768 words:
                  549
          the
                  398
          to
                  382
          ect
                  278
          and
          of
                  230
                 206
         hou
                 196
          а
          in
                 182
          for
                  170
                  135
          on
```

HW2.2.3 (Optional)

Using Hadoop MapReduce and your wordcount job (from HW2.2.1) determine the top-10 occurring tokens (most frequent tokens) using multiple reducers.

```
In [107]: %%writefile mapper23.py
          #!/usr/bin/python
          import sys
          import re
          from csv import reader
          WORD RE = re.compile(r''[\w']+")
          count = 0 #Running total of occurrances for the chosen word
          sys.stderr.write("reporter:counter:Mapper,Script Count,1\n")
          for line in reader(sys.stdin):
              sys.stderr.write("reporter:counter:Mapper,Line Count,1\n")
                  int(line[0]) #check if the ID field is an integer, skip the record if not
              except ValueError:
                  continue
              words = re.findall(WORD RE, line[3])
              for word in words:
                  print word.lower()+'\t1'
```

Overwriting mapper23.py

```
In [108]: | %%writefile reducer23.py
          #!/usr/bin/python
          import sys
          tmp_word=''
          count = 0 #Running total of occurrances for the chosen word
          sys.stderr.write("reporter:counter:Reducer,Script Count,1\n")
          for line in sys.stdin:
              line=line.strip().split('\t') #Parse line into a list of fields
              word, sub count=line
              if tmp word==word:
                  count+=int(sub count) #Extract chunk count from the second field of each i
          ncoming line
              else:
                  if tmp word:
                      sys.stderr.write("reporter:counter:Reducer,Line Count,1\n")
                     print tmp_word+'\t'+str(count)
                  tmp word=word
                  count=int(sub count)
          if tmp_word:
              sys.stderr.write("reporter:counter:Reducer, Line Count, 1 \n")
              print tmp word+'\t'+str(count)
```

Overwriting reducer23.py

```
In [109]: !chmod a+x mapper23.py
!chmod a+x reducer23.py

!hdfs dfs -rm -r results23
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-D mapred.map.tasks=2 \\
-D mapred.reduce.tasks=2 \\
-mapper /home/cloudera/mapper23.py \\
-reducer /home/cloudera/reducer23.py \\
-input /user/shihyu/enronemail_lh.txt \\
-output results23
!hdfs dfs -cat results23/* | head -10
```

```
Deleted results23
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob6434113247504200456.jar tmpDir=null
16/09/09 22:01:44 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 22:01:44 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/09 22:01:45 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 22:01:45 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 22:01:45 INFO Configuration.deprecation: mapred.reduce.tasks is depreca
ted. Instead, use mapreduce.job.reduces
16/09/09 22:01:45 INFO Configuration.deprecation: mapred.map.tasks is deprecated
. Instead, use mapreduce.job.maps
16/09/09 22:01:45 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0024
16/09/09 22:01:45 INFO impl.YarnClientImpl: Submitted application application 14
73444507507 0024
16/09/09 22:01:45 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0024/
16/09/09 22:01:45 INFO mapreduce. Job: Running job: job 1473444507507 0024
16/09/09 22:01:53 INFO mapreduce. Job: Job job 1473444507507 0024 running in uber
mode : false
16/09/09 22:01:53 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 22:02:01 INFO mapreduce.Job: map 50% reduce 0%
16/09/09 22:02:02 INFO mapreduce.Job: map 100% reduce 0%
16/09/09 22:02:10 INFO mapreduce.Job: map 100% reduce 50%
16/09/09 22:02:11 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 22:02:11 INFO mapreduce.Job: Job job_1473444507507_0024 completed succe
ssfully
16/09/09 22:02:11 INFO mapreduce. Job: Counters: 51
       File System Counters
               FILE: Number of bytes read=12
               FILE: Number of bytes written=474206
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
               FILE: Number of write operations=0
               HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=0
                HDFS: Number of read operations=12
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=4
        Job Counters
               Launched map tasks=2
                Launched reduce tasks=2
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=11754
                Total time spent by all reduces in occupied slots (ms)=11915
                Total time spent by all map tasks (ms) = 11754
                Total time spent by all reduce tasks (ms)=11915
                Total vcore-seconds taken by all map tasks=11754
                Total vcore-seconds taken by all reduce tasks=11915
                Total megabyte-seconds taken by all map tasks=12036096
                Total megabyte-seconds taken by all reduce tasks=12200960
        Map-Reduce Framework
                Map input records=100
                Map output records=0
                Map output bytes=0
                Map output materialized bytes=24
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=0
                Reduce shuffle bytes=24
                Reduce input records=0
```

HW2.3.1 Learn a Multinomial Naive Bayes model

Chinese Data Set

Overwriting ChineseExample Train.txt

```
In [112]: | %%writefile mapper_t_ch.py
          #!/usr/bin/python
          import sys, re, string
          # define regex for punctuation removal
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          # input comes from STDIN (standard input)
          for line in sys.stdin:
              # use contenty
              msg = line.split('\t', 2)
              if len(msg) < 3:
                  continue
              msgID, isChina = msg[0], msg[1]
              # remove punctuations, only have white-space as delimiter
              words = regex.sub(' ', msg[-1].lower())
              # split the line into words
              words = words.split()
              # increase counters
              for word in words:
                  # write the results to STDOUT (standard output);
                  print '%s\t%d\t%s\t%s' % (word, 1, isChina, msgID)
```

Overwriting mapper_t_ch.py

```
In [117]: | %%writefile reducer_t_ch.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator
          import numpy as np
          tmp word = None
          smooth factor = 0 # no smoothing
          \# because no smoothing, so count from zero for isChina and nonChina cases
          tmp count = [smooth factor, smooth factor]
          msgIDs = {}
          word = None
          # key is words, values are word probability at isChina class and nonChina class, r
          espectively
          wordcount = {}
          # input comes from STDIN
          for line in sys.stdin:
              # remove leading and trailing whitespace
              line = line.strip()
              # parse the input we got from mapper
              word, count, isChina, msgID = line.split('\t', 3)
              # convert to int
              try:
                  count = int(count)
                  isChina = int(isChina)
              except ValueError:
                  continue
              # Used to count how many training items are at China category
              if msgID not in msgIDs:
                  msgIDs[msgID] = isChina
              if tmp word == word:
                  tmp_count[isChina] += count
              else:
                  if tmp_word:
                      # count finish so save it
                      wordcount[tmp_word] = tmp_count
                  # begin new count for new word
                  tmp_count = [smooth_factor, smooth_factor]
                  tmp count[isChina] = count
                  tmp_word = word
          # Last word count!
          if tmp word == word:
              wordcount[tmp_word] = tmp_count
          # calculate NB parameters, and write to a file for the classification job
          # prior probabilities
          n msg = len(msgIDs)
          n_China = sum(msgIDs.values())
          n not China = n msg - n China
          print '%s\t%s' %('prior prob', 1.0*n not China/n msg, 1.0*n China/n msg)
          # conditional probabilities for each class
          n_total = np.sum(wordcount.values(), 0)
          for (key, value) in zip(wordcount.keys(), wordcount.values()/(1.0*n_total)):
              print '%s\t%s' %(key, value[0], value[1])
```

Overwriting reducer_t_ch.py

Systems test for Chinese data set

HW2.3.2 Learn a multinomial naive Bayes model (with no smoothing) by hand

From formula

$$P(C|D) = \frac{P(D|C)P(C)}{P(D)}$$
,

where C represents category variable and D represents observation. Since our goal is to obtain P(C|D), we have to evaluate P(D|C) and P(C) first.

Priori and conditional probabilities tables for Chinese Data Set:

Prob	c_1 = China	c_0 = not China	
Prior, $P(C=c_i)$	3/4	1/4	
\$P(Chinese	C = c_i)\$	5/8	1/3
\$P(Beijing	C = c_i)\$	1/8	0
\$P(Shanghai	C = c_i)\$	1/8	0
\$P(Macao	C = c_i)\$	1/8	0
\$P(Tokyo	C = c_i)\$	0	1/3
\$P(Japan	C = c_i)\$	0	1/3

Classifier:

Given D_5 as observation, it can be classfied by comparing probabilities of $P(D_5|C=c_0)$ and $P(D_5|C=c_1)$. Since

%%latex

$$P(D_5|C=c_0) = P(Chinese|C=c_0)^3 P(Tokyo|C=c_0) P(Japan|C=c_0) = 1/243$$

and

%%latex

$$P(D_5|C=c_1)=P(Chinese|C=c_1)^3P(Tokyo|C=c_1)P(Japan|C=c_1)=0$$

Therefore,

%%latex \begin{align} $P(C = c_0 \mid D_5) = \alpha P(D_5 \mid C = c_0) P(C = c_0) = \alpha 1/243 * 1/4 \end{cases}$

and

%%latex \begin{align} P(C = c_1 | D_5) = \alpha $P(D_5 | C = c_1) P(C = c_1) = \lambda d = 0 * 3/4 \end{align}$

Finally, since $P(C = c_0|D_5)$ is greter then $P(C = c_1|D_5)$, D_5 will be classified as non China. This classification result is totally different from smoothing case that presented at attached file.

HW2.3.3 Learn a multinomial naive Bayes model (with no smoothing) for SPAM filtering

Save model as SPAM_Model_MNB.tsv.

```
In [120]: | %%writefile mapper_t_em.py
          #!/usr/bin/python
          import sys, re, string
          # define regex for punctuation removal
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          # input comes from STDIN (standard input)
          for line in sys.stdin:
              # use email subject and body
              msg = line.split('\t', 2)
              if len(msq) \le 2:
                  continue
              msgID, isSpam = msg[0], msg[1]
              # remove punctuations, only have white-space as delimiter
              words = regex.sub(' ', msg[-1].lower())
              # split the line into words
              words = words.split()
              # increase counters
              for word in words:
                  # write the results to STDOUT (standard output);
                  print '%s\t%d\t%s\t%s' % (word, 1, isSpam, msgID)
```

Overwriting mapper_t_em.py

```
In [121]: | %%writefile reducer_t_em.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator
          import numpy as np
          tmp word = None
          smooth factor = 0 # no smoothing
          \# because no smoothing, so count from zero for isChina and nonChina cases
          tmp count = [smooth factor, smooth factor]
          msgIDs = {}
          word = None
          # key is words, values are word probability at isChina class and nonChina class, r
          espectively
          wordcount = {}
          # input comes from STDIN
          for line in sys.stdin:
               # remove leading and trailing whitespace
              line = line.strip()
               # parse the input we got from mapper
              word, count, isSpam, msgID = line.split('\t', 3)
               # convert to int
              try:
                  count = int(count)
                  isSpam = int(isSpam)
               except ValueError:
                  continue
               # Used to count how many training items are at SPAM emails category
              if msgID not in msgIDs:
                  msgIDs[msgID] = isSpam
               if tmp word == word:
                  tmp_count[isSpam] += count
              else:
                   if tmp_word:
                       # count finish so save it
                       wordcount[tmp_word] = tmp_count
                   # begin new count for new word
                   tmp count = [smooth factor, smooth factor]
                   tmp count[isSpam] = count
                  tmp_word = word
          # Last word count!
          if tmp word == word:
               wordcount[tmp_word] = tmp_count
          # calculate NB parameters, and write to a file for the classification job
          # prior probabilities
          n msg = len(msgIDs)
          n Spam = sum(msgIDs.values())
          n \text{ Ham} = n \text{ msg} - n \text{ Spam}
          print '%s\t%s\t%s' %('prior prob', 1.0*n Ham/n msg, 1.0*n Spam/n msg)
          # conditional probabilities for each class
          n_total = np.sum(wordcount.values(), 0)
          for (key, value) in zip(wordcount.keys(), wordcount.values()/(1.0*n_total)):
              print '%s\t%s' %(key, value[0], value[1])
```

Overwriting reducer_t_em.py

```
In [123]: !chmod a+x mapper_t_em.py
!chmod a+x reducer_t_em.py
!hdfs dfs -rm -r prob_em
!rm SPAM_Model_MNB.tsv
# run training job
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[ \]
-mapper /home/cloudera/mapper_t_em.py \
-reducer /home/cloudera/reducer_t_em.py \
-input /user/shihyu/enronemail_lh.txt \
-output prob_em
!hadoop fs -mv prob_em/part-00000 /user/shihyu/SPAM_Model_MNB.tsv
!hadoop fs -get /user/shihyu/SPAM_Model_MNB.tsv
### output prob will be write in a file SPAM_Model_MNB.tsv.###
```

```
Deleted prob em
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob1804159925030180330.jar tmpDir=null
16/09/09 22:31:31 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 22:31:31 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/09 22:31:32 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 22:31:32 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 22:31:32 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0026
16/09/09 22:31:33 INFO impl.YarnClientImpl: Submitted application application_14
73444507507 0026
16/09/09 22:31:33 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0026/
16/09/09 22:31:33 INFO mapreduce.Job: Running job: job_1473444507507_0026
16/09/09 22:31:40 INFO mapreduce. Job: Job job 1473444507507 0026 running in uber
mode : false
16/09/09 22:31:40 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 22:31:48 INFO mapreduce.Job: map 50% reduce 0%
16/09/09 22:31:49 INFO mapreduce.Job: map 100% reduce 0% 16/09/09 22:31:55 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 22:31:55 INFO mapreduce. Job: Job job 1473444507507 0026 completed succe
ssfully
16/09/09 22:31:55 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=1172006
                FILE: Number of bytes written=2699652
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=173923
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=12593
                Total time spent by all reduces in occupied slots (ms)=4298
                Total time spent by all map tasks (ms) = 12593
                Total time spent by all reduce tasks (ms)=4298
                Total vcore-seconds taken by all map tasks=12593
                Total vcore-seconds taken by all reduce tasks=4298
                Total megabyte-seconds taken by all map tasks=12895232
                Total megabyte-seconds taken by all reduce tasks=4401152
        Map-Reduce Framework
                Map input records=100
                Map output records=33203
                Map output bytes=1105594
                Map output materialized bytes=1172012
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=5408
                Reduce shuffle bytes=1172012
                Reduce input records=33203
                Reduce output records=5409
                Spilled Records=66406
                Shuffled Maps =2
                Failed Shuffles=0
                Merged Map outputs=2
```

Top 10 and bottom 10 model entries

```
In [134]: ### Buttom 10
           !cat SPAM Model MNB.tsv | sort -k1,1-r | head -n 10 > a
           !cat a
          sort: write failed: standard output: Broken pipe
          sort: write error
          zxs 0.0 5.29464711177e-05
          zolam 0.0 0.000105892942235
zo 0.0 0.000105892942235
zk 0.0 5.29464711177e-05
zinc 0.0 5.29464711177e-05
          zimin 0.000349259569712 0.0
          zesto 0.0 0.000529464711177
          zero 0.000209555741827 5.29464711177e-05
          zadorozhny 0.00027940765577 0.0
zac 0.0 0.000105892942235
In [135]: ### Top 10
           !cat SPAM Model MNB.tsv | sort -k1,1 | head -n 10 > a
           !cat a
          sort: write failed: standard output: Broken pipe
          sort: write error
                  0.00146689019279 0.0
                  0.0 0.000317678826706
                  0.001327186364910.0007412505956480.001746297848560.00142955472018
          00
          001 0.000139703827885 5.29464711177e-05
0011 6.98519139424e-05 0.0
          00450 0.0 5.29464711177e-05
0080 0.0 5.29464711177e-05
                 012
```

HW 2.3.4 Classify Documents using the learnt Multinomial Naive Bayes model using Hadoop Streaming

Unit test your classifier map reduce job using the Chinese example. Input D_5 , output 0, not China.

```
In [141]: %%writefile ChineseExample Test.txt
         D1 1 Chinese Beijing Chinese
                1
                        Chinese Chinese Shanghai
                1
         D3
                        Chinese Macao
                     Tokyo Japan Chinese
                0
         D4
         D5
                0
                      Chinese Chinese Chinese Tokyo Japan
         Overwriting ChineseExample Test.txt
In [146]: !hdfs dfs -put Chinese probabilities.txt /user/shihyu
         put: `/user/shihyu/Chinese probabilities.txt': File exists
```

```
In [152]: | %%writefile mapper_c_ch.py
          #!/usr/bin/python
          import sys, re, string, subprocess
          # read the probabilities from Chinese_probabilities.txt
          prob = \{\}
          cat = subprocess.Popen(["hadoop", "fs", "-cat", "/user/shihyu/Chinese probabilitie
          s.txt"], stdout=subprocess.PIPE)
          ### Prepare key (word) values (conditional prob at each class) pair
          for line in cat.stdout:
                  word, p0, p1 = line.split()
                  prob[word] = [p0, p1]
          # get prior probability
          prior = prob['prior prob']
          # define regex for punctuation removal
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          # input comes from STDIN (standard input)
          for line in sys.stdin:
              # use content
              msg = line.split('\t', 2)
              # skip strange message
              if len(msq) < 3:
                  continue
              msgID, isChina = msg[0], msg[1]
              # remove punctuations, only have white-space as delimiter
              words = regex.sub(' ', msg[-1].lower())
              # split the line into words
              words = words.split()
              # increase counters
              for word in words:
                  # write the results to STDOUT (standard output);
                  print '%s\t%s\t%s\t%s\t%s' % (msgID, prob[word][0], prob[word][1], isC
          hina, prior[0], prior[1])
```

Overwriting mapper c ch.py

```
In [160]: %%writefile reducer c ch.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator, math
          import numpy as np
          tmp msg = None
          tmp prob = [0, 0]
          tmp truth = 0
          msgID = None
          n = rror = 0
          n msg = 0
          n zero = [0, 0]
          print '%s\t%s' %('ID', 'TRUTH', 'PREDICTION')
          # input comes from STDIN
          for line in sys.stdin:
              #print line
              # remove leading and trailing whitespace
              line = line.strip()
              # parse the input we got from mapper c.py
              msgID, p0, p1, isChina, pr0, pr1 = line.split('\t')
              prob = [float(p0), float(p1)]
              # convert count and spam flag (currently a string) to int
              trv:
                  isChina = int(isChina)
              except ValueError:
                  continue
              # get posterior prob assume independence
              if tmp msg == msgID:
                  # by considering zero prob
                  tmp prob = np.sum([[math.log(x) if x>0 else float('-inf') for x in prob],
          tmp prob], 0)
              else:
                  if tmp msg:
                      # count finish for current word, predict and print result
                      pred = np.argmax(tmp prob)
                      n error += pred != tmp truth
                      n msg += 1
                      n zero[tmp truth] += float('-inf') in tmp prob
                      print '%s\t%s\t%s' %(tmp_msg, tmp_truth, pred)
                  # initialize new count for new word
                  prior = [math.log(float(pr0)), math.log(float(pr1))]
                  tmp prob = np.sum([[math.log(x) if x>0 else float('-inf') for x in prob],
          prior], 0)
                  tmp msg = msgID
                  tmp truth = isChina
          # do not forget to print the last msg
          if tmp msg == msgID:
              pred = np.argmax(tmp prob)
              n error += pred != isChina
              n msg += 1
              n_zero[tmp_truth] += float('-inf') in tmp_prob
              print '%s\t%s' %(msgID, tmp_truth, pred)
          # calculate the overall error rate
          print 'Error rate is: %.f' %(1.0*n_error/n_msg)
          print 'Number of messages with zero probability: China(%d). non China(%d)' %(n zer
```

Overwriting reducer_c_ch.py

```
In [161]: # clean up HDFS
!chmod a+x mapper_c_ch.py
!chmod a+x reducer_c_ch.py
!hdfs dfs -put ChineseExample_Test.txt /user/shihyu

!hdfs dfs -rm -r Chinese_classification_results
# run classification job
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[ \]
-mapper /home/cloudera/mapper_c_ch.py \
-reducer /home/cloudera/reducer_c_ch.py \
-input /user/shihyu/ChineseExample_Test.txt \
-output Chinese_classification_results
```

```
put: `/user/shihyu/ChineseExample Test.txt': File exists
Deleted Chinese classification results
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob2950033205685806109.jar tmpDir=null
16/09/09 23:30:22 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 23:30:23 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/09 23:30:24 INFO mapred. File Input Format: Total input paths to process: 1
16/09/09 23:30:24 INFO mapreduce. JobSubmitter: number of splits:2
16/09/09 23:30:24 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0035
16/09/09 23:30:25 INFO impl. YarnClientImpl: Submitted application application 14
73444507507 0035
16/09/09 23:30:25 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0035/
16/09/09 23:30:25 INFO mapreduce. Job: Running job: job_1473444507507 0035
16/09/09 23:30:33 INFO mapreduce. Job: Job job 1473444507507 0035 running in uber
mode : false
16/09/09 23:30:33 INFO mapreduce.Job: map 0% reduce 0%
16/09/09 23:30:47 INFO mapreduce.Job: map 33% reduce 0% 16/09/09 23:30:48 INFO mapreduce.Job: map 100% reduce 0%
16/09/09 23:30:55 INFO mapreduce.Job: map 100% reduce 100%
16/09/09 23:30:55 INFO mapreduce. Job: Job job 1473444507507 0035 completed succe
16/09/09 23:30:56 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=573
                FILE: Number of bytes written=356873
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=453
                HDFS: Number of bytes written=139
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms) = 26415
                Total time spent by all reduces in occupied slots (ms)=4321
                Total time spent by all map tasks (ms) = 26415
                Total time spent by all reduce tasks (ms)=4321
                Total vcore-seconds taken by all map tasks=26415
                Total vcore-seconds taken by all reduce tasks=4321
                Total megabyte-seconds taken by all map tasks=27048960
                Total megabyte-seconds taken by all reduce tasks=4424704
        Map-Reduce Framework
                Map input records=5
                Map output records=16
                Map output bytes=535
                Map output materialized bytes=579
                Input split bytes=238
                Combine input records=0
                Combine output records=0
                Reduce input groups=5
                Reduce shuffle bytes=579
                Reduce input records=16
                Reduce output records=8
                Spilled Records=32
                Shuffled Maps =2
                Failed Shuffles=0
```

```
In [162]: !hdfs dfs -cat Chinese_classification_results/part-00000
         ΙD
                 TRUTH PREDICTION
         D1
                 1
                         1
         D2
                         1
                 1
                         1
         D3
                 1
         D4
                        0
                 0
         D5
                 0
                         0
         Error rate is: 0
         Number of messages with zero probability: China(3), non China(2)
```

Classification for emails

```
In [180]: %%writefile mapper c em.py
          #!/usr/bin/python
          import sys, re, string, subprocess
          # read the probability prob em from HDFS
          prob = \{\}
          cat = subprocess.Popen(["hadoop", "fs", "-cat", "/user/cloudera/prob_em/part-00000
          "], stdout=subprocess.PIPE)
          ### Prepare key (word) values (conditional prob at each class) pair
          for line in cat.stdout:
              word, p0, p1 = line.split()
              prob[word] = [p0, p1]
              print '%s\n' % (word)
          # get prior probability
          prior = prob['prior_prob']
          # define regex for punctuation removal
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          # input comes from STDIN (standard input)
          for line in sys.stdin:
              # use subject and body
              msg = line.split('\t', 2)
              # skip bad message
              if len(msg) < 3:
                  continue
              msgID, isSpam = msg[0], msg[1]
              # remove punctuations, only have white-space as delimiter
              words = regex.sub(' ', msg[-1].lower())
              # split the line into words
              words = words.split()
              # increase counters
              for word in words:
                  # write the results to STDOUT (standard output);
                  print '%s\t%s\t%s\t%s\t%s\t%s\t%s\ (msgID, prob[word][0], prob[word][1], isS
          pam, prior[0], prior[1])
```

Overwriting mapper_c_em.py

```
In [192]: | %%writefile reducer_c_em.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator, math
          import numpy as np
          tmp msg = None
          tmp prob = [0, 0]
          tmp truth = 0
          msgID = None
          n = rror = 0
          n msg = 0
          n zero = [0, 0]
          print '%s\t%s\t%s' %('EMAIL ID', 'TRUTH', 'PREDICTION')
          # input comes from STDIN
          for line in sys.stdin:
              #print line
              # remove leading and trailing whitespace
              line = line.strip()
                  msgID, p0, p1, isSpam, pr0, pr1 = line.split('\t')
              except ValueError:
                  continue
              prob = [float(p0), float(p1)]
              # convert count and spam flag (currently a string) to int
              trv:
                  isSpam = int(isSpam)
              except ValueError:
                  continue
              if tmp msg == msgID:
                  tmp prob = np.sum([[math.log(x) if x>0 else float('-inf') for x in prob],
          tmp_prob], 0)
              else:
                  if tmp msg:
                      # count finish for current word, predict and print result
                      pred = np.argmax(tmp prob)
                      n error += pred != tmp truth
                      n msg += 1
                      n zero[tmp truth] += float('-inf') in tmp prob
                      print '%s\t%s\t%s' %(tmp msg, tmp truth, pred)
                  # initialize new count for new word
                  prior = [math.log(float(pr0)), math.log(float(pr1))]
                  tmp prob = np.sum([[math.log(x) if x>0 else float('-inf') for x in prob],
          prior], 0)
                  tmp msg = msgID
                  tmp truth = isSpam
          # do not forget to print the last msg result if needed!
          if tmp msg == msgID:
              pred = np.argmax(tmp_prob)
              n_error += pred != isSpam
              n msg += 1
              n_zero[tmp_truth] += float('-inf') in tmp_prob
              print '%s\t%s' %(msgID, tmp_truth, pred)
```

Overwriting reducer_c_em.py

```
In [186]: |!chmod a+x mapper_t_em.py
          !chmod a+x reducer_t_em.py
          !hdfs dfs -rm -r prob_em
          # run training job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \
          -mapper /home/cloudera/mapper_t_em.py \
          -reducer /home/cloudera/reducer_t_em.py \
          -input /user/shihyu/enronemail 1h.txt \
          -output prob em
          # clean up HDFS
          !chmod a+x mapper_c_em.py
          !chmod a+x reducer_c_em.py
          !hdfs dfs -rm -r Emails_classification_results
          # run classification job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \
          -mapper /home/cloudera/mapper c em.py \
          -reducer /home/cloudera/reducer_c_em.py \
          -input /user/shihyu/enronemail_1h.txt \
          -output Emails_classification_results
```

```
Deleted prob em
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob4653791537771942229.jar tmpDir=null
16/09/10 05:47:18 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/10 05:47:18 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/10 05:47:19 INFO mapred. File Input Format: Total input paths to process: 1
16/09/10 05:47:19 INFO mapreduce. JobSubmitter: number of splits:2
16/09/10 05:47:19 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0049
16/09/10 05:47:19 INFO impl.YarnClientImpl: Submitted application application_14
73444507507 0049
16/09/10 05:47:19 INFO mapreduce.Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0049/
16/09/10 05:47:19 INFO mapreduce. Job: Running job: job 1473444507507 0049
16/09/10 05:47:27 INFO mapreduce. Job: Job job 1473444507507 0049 running in uber
mode : false
16/09/10 05:47:27 INFO mapreduce.Job: map 0% reduce 0%
16/09/10 05:47:36 INFO mapreduce.Job: map 100% reduce 0%
16/09/10 05:47:43 INFO mapreduce.Job: map 100% reduce 100%
16/09/10 05:47:44 INFO mapreduce.Job: Job job_1473444507507_0049 completed succe
ssfully
16/09/10 05:47:44 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=1172006
                FILE: Number of bytes written=2699652
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=173923
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=13838
                Total time spent by all reduces in occupied slots (ms)=4599
                Total time spent by all map tasks (ms)=13838
                Total time spent by all reduce tasks (ms)=4599
                Total vcore-seconds taken by all map tasks=13838
                Total vcore-seconds taken by all reduce tasks=4599
                Total megabyte-seconds taken by all map tasks=14170112
                Total megabyte-seconds taken by all reduce tasks=4709376
        Map-Reduce Framework
                Map input records=100
                Map output records=33203
                Map output bytes=1105594
                Map output materialized bytes=1172012
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=5408
                Reduce shuffle bytes=1172012
                Reduce input records=33203
                Reduce output records=5409
                Spilled Records=66406
                Shuffled Maps =2
                Failed Shuffles=0
                Merged Map outputs=2
                GC time elapsed (ms) = 144
```

In [187]: !hdfs dfs -cat Emails_classification_results/part-00000

EMAIL ID	TRUTH	PREDICTI		
0001.1999-12-10.		0	0	^
0001.1999-12-10.			0	0
0001.2000-01-17.		0	0	
0001.2000-06-06.	-	0	0	
0001.2001-02-07.		0	0	0
0001.2001-04-02.			0	0
0002.1999-12-13. 0002.2001-02-07.		0	0	
0002.2001-02-07.		() D	0	1
0002.2001 03 23.		1	1	_
0002.2004-08-01.		1	1	
0003.1999-12-10.			0	0
0003.1999-12-14.		0	0	
0003.2000-01-17.	beck	0	0	
0003.2001-02-08.	kitchen	0	0	
0003.2003-12-18.		1	1	
0003.2004-08-01.	BG	1	1	
0004.1999-12-10.	kaminski		0	0
0004.1999-12-14.	farmer	0	0	
0004.2001-04-02.	williams		0	0
0004.2001-06-12.		P	1	1
0004.2004-08-01.	BG	1	1	
0005.1999-12-12.			0	0
0005.1999-12-14.		0	0	
0005.2000-06-06.	_	0	0	
0005.2001-02-08.		0	0	
0005.2001-06-23.			1	1
0005.2003-12-18.		1	1	_
0006.1999-12-13.			0	0
0006.2001-02-08.			0	^
0006.2001-04-03.			0	0
0006.2001-06-25. 0006.2003-12-18.		1	1	1
0006.2003-12-18.		1	1	
0000.2004 00 01.			0	0
0007.1999-12-14.		0	0	0
0007.2000-01-17.		0	0	
0007.2001-02-09.		0	0	
0007.2003-12-18.	GP	1	1	
0007.2004-08-01.		1	1	
0008.2001-02-09.	kitchen	0	0	
0008.2001-06-12.	SA_and_H	ΙP	1	1
0008.2001-06-25.	SA_and_H	ΙP	1	1
0008.2003-12-18.	GP	1	1	
0008.2004-08-01.	BG	1	1	
0009.1999-12-13.			0	0
0009.1999-12-14.		0	0	
0009.2000-06-07.		0	0	
0009.2001-02-09.		0	0	
0009.2001-06-26.			1	1
0009.2003-12-18.		1	1	
0010.1999-12-14.		0	0	^
0010.1999-12-14.			0	0
0010.2001-02-09.			0	1
0010.2001-06-28. 0010.2003-12-18.		1	1	1
0010.2003-12-18.		1	1	
0010.2004-08-01.		0	0	
0011.1999-12-14.			1	1
0011.2001-06-29.			1	1
0011.2003-12-18.		1	1	_
0011.2004-08-01.		1	1	
0012.1999-12-14.		0	0	

Plot a histogram of the posterior probabilities. Since the NB posterior probability is:

$$P(C=c_j \mid X_1, X_2, \dots, X_n) = rac{P(C=c_j) \prod_i P(X_i \mid C=c_i)}{\sum_k P(C=c_k) \prod_i P(X_i \mid C=c_k)}$$

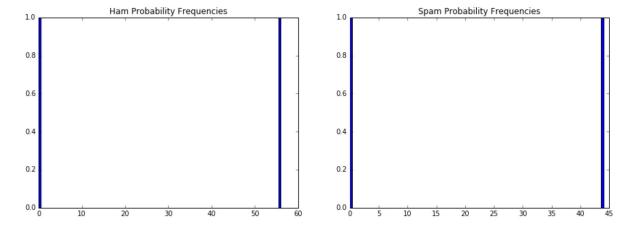
where C is email category and X_i are wrods. The conditional probability of those words for the opposite category, $P(X_i \mid C = c_k)$, are zero, and this will make one term on the denominator become zero. Therefore, posterior probability is 1 for the category that contains all the words from same category, and 0 for the other category. Hisogram is shown below

```
In [218]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt

# 44 spam and 56 ham all has posterior probability of 1
spam = (0,44)
ham = (0,56)
plt.figure(figsize=(15,5))
p = plt.subplot(1, 2, 1)
p.hist(ham,100)
plt.title('Ham Probability Frequencies')

p = plt.subplot(1, 2, 2)
p.hist(spam,100)
plt.title('Spam Probability Frequencies')
```

Out[218]: <matplotlib.text.Text at 0x7f78f1abf090>



HW2.4 Use Laplace plus-one smoothing

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Repeat HW2.3 with the following modification: use Laplace plus-one smoothing.

In addition, compare the misclassification error rates for 2.3 versus 2.4 and explain the differences.

```
In [193]: |%%writefile reducer_t_em_s.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator
          import numpy as np
          tmp word = None
          smooth factor = 1 # because smoothing
          # because smoothing, so count from ONE
          tmp count = [smooth factor, smooth factor]
          msgIDs = {}
          word = None
          # key is words, values are word probability at isChina class and nonChina class, r
          espectively
          wordcount = {}
          # input comes from STDIN
          for line in sys.stdin:
               # remove leading and trailing whitespace
              line = line.strip()
               # parse the input we got from mapper
              word, count, isSpam, msgID = line.split('\t', 3)
               # convert to int
              try:
                  count = int(count)
                  isSpam = int(isSpam)
               except ValueError:
                  continue
               # Used to count how many training items are at SPAM emails category
              if msgID not in msgIDs:
                  msgIDs[msgID] = isSpam
               if tmp word == word:
                  tmp_count[isSpam] += count
              else:
                  if tmp_word:
                       # count finish so save it
                       wordcount[tmp_word] = tmp_count
                   # begin new count for new word
                  tmp count = [smooth factor, smooth factor]
                  tmp count[isSpam] = count
                  tmp_word = word
          # Last word count!
          if tmp word == word:
               wordcount[tmp_word] = tmp_count
          # calculate NB parameters, and write to a file for the classification job
          # prior probabilities
          n msg = len(msgIDs)
          n Spam = sum(msgIDs.values())
          n \text{ Ham} = n \text{ msg} - n \text{ Spam}
          print '%s\t%s\t%s' %('prior prob', 1.0*n Ham/n msg, 1.0*n Spam/n msg)
          # conditional probabilities for each class
          n_total = np.sum(wordcount.values(), 0)
          for (key, value) in zip(wordcount.keys(), wordcount.values()/(1.0*n_total)):
              print '%s\t%s' %(key, value[0], value[1])
          Overwriting reducer_t_em_s.py
```

```
In [194]: |!chmod a+x mapper_t_em.py
          !chmod a+x reducer_t_em_s.py
          !hdfs dfs -rm -r prob_em
          # run training job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \overline{\mathbb{N}}
          -mapper /home/cloudera/mapper t em.py \
          -reducer /home/cloudera/reducer_t_em_s.py \
          -input /user/shihyu/enronemail 1h.txt \
          -output prob em
          # clean up HDFS
          !chmod a+x mapper_c_em.py
          !chmod a+x reducer_c_em.py
          !hdfs dfs -rm -r Emails_classification_results_s
          # run classification job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[\]
          -mapper /home/cloudera/mapper c em.py \
          -reducer /home/cloudera/reducer_c_em.py \
          -input /user/shihyu/enronemail_1h.txt \
          -output Emails_classification_results_s
```

```
Deleted prob em
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob5515545378449372430.jar tmpDir=null
16/09/10 06:03:36 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/10 06:03:36 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/10 06:03:37 INFO mapred. File Input Format: Total input paths to process: 1
16/09/10 06:03:38 INFO mapreduce. JobSubmitter: number of splits:2
16/09/10 06:03:38 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0053
16/09/10 06:03:38 INFO impl.YarnClientImpl: Submitted application application 14
73444507507 0053
16/09/10 06:03:38 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0053/
16/09/10 06:03:38 INFO mapreduce.Job: Running job: job_1473444507507_0053
16/09/10 06:03:44 INFO mapreduce. Job: Job job 1473444507507 0053 running in uber
mode : false
16/09/10 06:03:44 INFO mapreduce.Job: map 0% reduce 0%
16/09/10 06:03:52 INFO mapreduce.Job: map 50% reduce 0%
16/09/10 06:03:53 INFO mapreduce.Job: map 100% reduce 0% 16/09/10 06:04:01 INFO mapreduce.Job: map 100% reduce 100%
16/09/10 06:04:01 INFO mapreduce. Job: Job job 1473444507507 0053 completed succe
ssfully
16/09/10 06:04:01 INFO mapreduce.Job: Counters: 50
        File System Counters
                FILE: Number of bytes read=1172006
                FILE: Number of bytes written=2699658
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=234964
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Killed map tasks=1
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=11916
                Total time spent by all reduces in occupied slots (ms)=4755
                Total time spent by all map tasks (ms)=11916
                Total time spent by all reduce tasks (ms) = 4755
                Total vcore-seconds taken by all map tasks=11916
                Total vcore-seconds taken by all reduce tasks=4755
                Total megabyte-seconds taken by all map tasks=12201984
                Total megabyte-seconds taken by all reduce tasks=4869120
        Map-Reduce Framework
                Map input records=100
                Map output records=33203
                Map output bytes=1105594
                Map output materialized bytes=1172012
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=5408
                Reduce shuffle bytes=1172012
                Reduce input records=33203
                Reduce output records=5409
                Spilled Records=66406
                Shuffled Maps =2
                Failed Shuffles=0
```

In [195]: !hdfs dfs -cat Emails_classification_results_s/part-00000

EMAIL ID	TRUTH	PREDICTI	ON	
0001.1999-12-10.	farmer	0	0	
0001.1999-12-10.	kaminski		0	0
0001.2000-01-17.	beck	0	0	
0001.2000-06-06.	lokay	0	0	
0001.2001-02-07.	kitchen	0	0	
0001.2001-04-02.	williams		0	0
0002.1999-12-13.	farmer	0	0	
0002.2001-02-07.	kitchen	0	0	
0002.2001-05-25.		P	1	1
0002.2003-12-18.		1	1	_
0002.2004-08-01.		1	1	
0003.1999-12-10.			0	0
				U
0003.1999-12-14.		0	0	
0003.2000-01-17.		0	0	
0003.2001-02-08.		0	0	
0003.2003-12-18.		1	1	
0003.2004-08-01.	BG	1	1	
0004.1999-12-10.	kaminski		0	0
0004.1999-12-14.	farmer	0	0	
0004.2001-04-02.	williams		0	0
0004.2001-06-12.	SA and H	P	1	1
0004.2004-08-01.	BG	1	1	
0005.1999-12-12.			0	0
0005.1999-12-14.		0	0	
0005.2000-06-06.		0	0	
0005.2001-02-08.	_	_	0	
		0		1
0005.2001-06-23.			1	1
0005.2003-12-18.		1	1	_
0006.1999-12-13.			0	0
0006.2001-02-08.			0	
0006.2001-04-03.			0	0
0006.2001-06-25.	SA_and_H	P	1	1
0006.2003-12-18.	GP	1	1	
0006.2004-08-01.	BG	1	1	
0007.1999-12-13.	kaminski		0	0
0007.1999-12-14.	farmer	0	0	
0007.2000-01-17.	beck	0	0	
0007.2001-02-09.	kitchen	0	0	
0007.2003-12-18.	GP	1	1	
0007.2004-08-01.		1	1	
0008.2001-02-09.		0	0	
0008.2001-06-12.			1	1
0008.2001-06-25.			1	1
				Τ.
0008.2003-12-18.		1	1	
0008.2004-08-01.			1	^
0009.1999-12-13.			0	0
0009.1999-12-14.		0	0	
0009.2000-06-07.		0	0	
0009.2001-02-09.		0	0	
0009.2001-06-26.	SA_and_H	P	1	1
0009.2003-12-18.	GP	1	1	
0010.1999-12-14.	farmer	0	0	
0010.1999-12-14.	kaminski		0	0
0010.2001-02-09.	kitchen	0	0	
0010.2001-06-28.	SA and H	ΙP	1	1
0010.2003-12-18.	GP -	1	0	
0010.2004-08-01.		1	1	
0011.1999-12-14.		0	0	
0011.2001-06-28.			1	1
0011.2001-06-29.			1	1
0011.2001 00 23.		1	1	_
0011.2003-12-18.		1	1	
0012.1999-12-14.		0	0	
0012.1333 12-14.	TATILET	V	•	

HW2.4 Discussion: Since smoothing essentially add noise to the data, and thus introduce some misclassification on the training set.

HW2.5 Ignore rare words

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Repeat HW2.4. This time when modeling and classification ignore tokens with a frequency of less than three (3) in the training set. How does it affect the misclassification error of learnt naive multinomial Bayesian Classifier on the training dataset. Report the error and the change in error.

```
In [196]: | %%writefile reducer_t_em_I.py
          #!/usr/bin/python
          from operator import itemgetter
          import sys, operator
          import numpy as np
          tmp word = None
          smooth factor = 0 # no smoothing
          # because no smoothing, so count from zero for isChina and nonChina cases
          tmp count = [smooth factor, smooth factor]
          msgIDs = {}
          word = None
          # key is words, values are word probability at isChina class and nonChina class, r
          espectively
          wordcount = {}
          # input comes from STDIN
          for line in sys.stdin:
               # remove leading and trailing whitespace
               line = line.strip()
               # parse the input we got from mapper
              word, count, isSpam, msgID = line.split('\t', 3)
               # convert to int
              try:
                  count = int(count)
                  isSpam = int(isSpam)
               except ValueError:
                  continue
               # Used to count how many training items are at SPAM emails category
              if msgID not in msgIDs:
                  msgIDs[msgID] = isSpam
               if tmp word == word:
                  tmp_count[isSpam] += count
              else:
                   if tmp_word:
                       # count finish so save it
                       wordcount[tmp_word] = tmp_count
                   # begin new count for new word
                   tmp_count = [smooth_factor, smooth_factor]
                   tmp count[isSpam] = count
                  tmp_word = word
          # Last word count!
          if tmp word == word:
              wordcount[tmp_word] = tmp_count
          # calculate NB parameters, and write to a file for the classification job
          # prior probabilities
          n msg = len(msgIDs)
          n Spam = sum(msgIDs.values())
          n \text{ Ham} = n \text{ msg} - n \text{ Spam}
          print '%s\t%s\t%s' %('prior prob', 1.0*n Ham/n msg, 1.0*n Spam/n msg)
          # conditional probabilities for each class
          n_total = np.sum(wordcount.values(), 0)
          for (key, value) in zip(wordcount.keys(), wordcount.values()/(1.0*n_total)):
               # only emit probability when the count (spam and ham together) is greater or e
          qual than 3
               if sum(wordcount[key]) >= 3:
                   print '%s\t%s\t%s' %(kev. value[0]. value[1])
```

Overwriting reducer t em I.py

```
In [200]: %%writefile mapper_c_em_I.py
          #!/usr/bin/python
          import sys, re, string, subprocess
          # read the probability prob em from HDFS
          prob = {}
          cat = subprocess.Popen(["hadoop", "fs", "-cat", "/user/cloudera/prob em/part-00000
          "], stdout=subprocess.PIPE)
          ### Prepare key (word) values (conditional prob at each class) pair
          for line in cat.stdout:
              word, p0, p1 = line.split()
              prob[word] = [p0, p1]
              print '%s\n' % (word)
          # get prior probability
          prior = prob['prior prob']
          # define regex for punctuation removal
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          # input comes from STDIN (standard input)
          for line in sys.stdin:
              # use subject and body
              msg = line.split('\t', 2)
              # skip bad message
              if len(msg) < 3:
                  continue
              msgID, isSpam = msg[0], msg[1]
              # remove punctuations, only have white-space as delimiter
              words = regex.sub(' ', msg[-1].lower())
              # split the line into words
              words = words.split()
              # increase counters
              for word in words:
                  # write the results to STDOUT (standard output)
                  # which has probability from training step
                  if word in prob:
                      print '%s\t%s\t%s\t%s\t%s' % (msgID, prob[word][0], prob[word][1],
           isSpam, prior[0], prior[1])
```

Overwriting mapper c em I.py

```
In [201]: |!chmod a+x mapper_t_em.py
          !chmod a+x reducer_t_em_I.py
          !hdfs dfs -rm -r prob_em
          # run training job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \overline{\mathbb{N}}
          -mapper /home/cloudera/mapper t em.py \
          -reducer /home/cloudera/reducer_t_em_I.py \
          -input /user/shihyu/enronemail 1h.txt \
          -output prob em
          # clean up HDFS
          !chmod a+x mapper_c_em_I.py
          !chmod a+x reducer_c_em.py
          !hdfs dfs -rm -r Emails_classification_results_I
          # run classification job
          !hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \
          -mapper /home/cloudera/mapper c em I.py \
          -reducer /home/cloudera/reducer_c_em.py \
          -input /user/shihyu/enronemail_1h.txt \
          -output Emails_classification_results_I
```

```
Deleted prob em
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob2597624802238306615.jar tmpDir=null
16/09/10 06:39:40 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/10 06:39:41 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
:8032
16/09/10 06:39:41 INFO mapred. File Input Format: Total input paths to process: 1
16/09/10 06:39:41 INFO mapreduce. JobSubmitter: number of splits:2
16/09/10 06:39:41 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507 0059
16/09/10 06:39:42 INFO impl.YarnClientImpl: Submitted application application_14
73444507507 0059
16/09/10 06:39:42 INFO mapreduce.Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0059/
16/09/10 06:39:42 INFO mapreduce. Job: Running job: job 1473444507507 0059
16/09/10 06:39:48 INFO mapreduce. Job: Job job 1473444507507 0059 running in uber
mode : false
16/09/10 06:39:48 INFO mapreduce.Job: map 0% reduce 0%
16/09/10 06:39:57 INFO mapreduce.Job: map 100% reduce 0%
16/09/10 06:40:03 INFO mapreduce.Job: map 100% reduce 100%
16/09/10 06:40:04 INFO mapreduce.Job: Job job_1473444507507_0059 completed succe
ssfully
16/09/10 06:40:04 INFO mapreduce.Job: Counters: 49
        File System Counters
                FILE: Number of bytes read=1172006
                FILE: Number of bytes written=2699658
                FILE: Number of read operations=0
                FILE: Number of large read operations=0
                FILE: Number of write operations=0
                HDFS: Number of bytes read=217093
                HDFS: Number of bytes written=66986
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=12983
                Total time spent by all reduces in occupied slots (ms)=4557
                Total time spent by all map tasks (ms) = 12983
                Total time spent by all reduce tasks (ms)=4557
                Total vcore-seconds taken by all map tasks=12983
                Total vcore-seconds taken by all reduce tasks=4557
                Total megabyte-seconds taken by all map tasks=13294592
                Total megabyte-seconds taken by all reduce tasks=4666368
        Map-Reduce Framework
                Map input records=100
                Map output records=33203
                Map output bytes=1105594
                Map output materialized bytes=1172012
                Input split bytes=226
                Combine input records=0
                Combine output records=0
                Reduce input groups=5408
                Reduce shuffle bytes=1172012
                Reduce input records=33203
                Reduce output records=1877
                Spilled Records=66406
                Shuffled Maps =2
                Failed Shuffles=0
                Merged Map outputs=2
                GC time elapsed (ms) = 206
```

In [203]: !hdfs dfs -cat Emails_classification_results_I/part-00000

EMAIL ID	TRUTH	PREDICTI		
0001.1999-12-10.		0	0	
0001.1999-12-10.			0	0
0001.2000-01-17.		0	0	
0001.2000-06-06.	_	0	0	
0001.2001-02-07.		0	0	_
0001.2001-04-02.			0	0
0002.1999-12-13.		0	0	
0002.2001-02-07.		0	0	1
0002.2001-05-25.			1	1
0002.2003-12-18.		1	1	
0002.2004-08-01.		1	1	0
0003.1999-12-10. 0003.1999-12-14.		0	0	0
0003.1999-12-14.		0		
0003.2000-01-17.		0	0	
0003.2001-02-08.		1	1	
0003.2003 12 10.		1	1	
0004.1999-12-10.	_	_	0	0
0004.1999-12-14.		0	0	O
0004.1999 12 14.			0	0
0004.2001-04-02.			1	1
0004.2001 00 12.		1	1	_
0005.1999-12-12.			0	0
0005.1999-12-14.		0	0	Ü
0005.2000-06-06.		0	0	
0005.2001-02-08.	_	0	0	
0005.2001-06-23.			1	1
0005.2003-12-18.		1	1	
0006.1999-12-13.			0	0
0006.2001-02-08.			0	
0006.2001-04-03.	williams		0	0
0006.2001-06-25.	SA and H	ΙP	1	1
0006.2003-12-18.	GP	1	1	
0006.2004-08-01.	BG	1	1	
0007.1999-12-13.	kaminski		0	0
0007.1999-12-14.	farmer	0	0	
0007.2000-01-17.	beck	0	0	
0007.2001-02-09.	kitchen	0	0	
0007.2003-12-18.	GP	1	1	
0007.2004-08-01.		1	1	
0008.2001-02-09.		0	0	
0008.2001-06-12.			1	1
0008.2001-06-25.			1	1
0008.2003-12-18.		1	1	
0008.2004-08-01.		1	1	_
0009.1999-12-13.			0	0
0009.1999-12-14.		0	0	
0009.2000-06-07.	_	0	0	
0009.2001-02-09.		0	0	1
0009.2001-06-26.			1	1
0009.2003-12-18.		1	1	
0010.1999-12-14. 0010.1999-12-14.		0	0	0
0010.1999-12-14.			0	U
0010.2001-02-09.			1	1
0010.2001-00-28.		1	0	_
0010.2003-12-18.		1	1	
0011.1999-12-14.		0	0	
0011.2001-06-28.			1	1
0011.2001-06-29.			1	1
0011.2003-12-18.		1	1	
0011.2004-08-01.		1	1	
0012.1999-12-14.	farmer	0	0	

HW2.5 Discussion:

Similar to smoothing, ignoring small frequency words will remove some that only belong to one category. Hence, the training error increases a bit. Compared to smoothing case, training error is smaller.

HW2.6 Benchmark your code with the Python SciKit-Learn

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```
In [214]: | from sklearn.feature_extraction.text import TfidfVectorizer
          from sklearn.naive_bayes import MultinomialNB, BernoulliNB
          X train = []
          Y train = []
          # read file
          with open('enronemail 1h.txt', 'rU') as infile:
              for line in infile.readlines():
                  try:
                      email_id, label, subject, body = line.split('\t')
                      X_train.append(subject + ' ' + body)
                  except ValueError:
                      email id, label, body = line.split('\t')
                      X_train.append(body)
                  # extract only words from the combined subject and body text
                  Y_train.append(int(label))
          # Use the TfidVectorizer to create the feature vectors
          vectorizer = TfidfVectorizer(token pattern = "[\w']+")
          vf = vectorizer.fit(X train, Y train)
          clf M = MultinomialNB()
          clf M.fit(vf.fit transform(X train), Y train)
          training error mnb = 1.0 - clf M.score(vf.fit transform(X train), Y train)
          clf B = BernoulliNB()
          clf B.fit(vf.fit transform(X train), Y train)
          training error bnb = 1.0 - clf B.score(vf.fit transform(X train), Y train)
          print 'SK- multinomial NB error rate is : %.4f' %training error mnb
          print 'SK- Bernoulli NB error rate is : %.4f' %training error bnb
          SK- multinomial NB error rate is : 0.0000
          SK- Bernoulli NB error rate is: 0.1800
```

HW 2.6.1 Bernoulli Naive Bayes (OPTIONAL: note this exercise is a stretch HW and optional)

From SK-Learn, we have Multinomial NB with 0% error rate, and Bernoulli NB with 18% error rate. The difference is made by the properties of classifiers. Bernoulli NB classifier will focuses on the appearance of word, while Multinomial NB classifier will reflect the frequency of the word. At this example, Bernoulli classifier is better since, as for SPAM detection, the appearance of some words (keywords) is more important than the frequency.

HW2.6.1 Results: See execution above, performance summary:

	Training Error
SK-Learn Multinomial NB	0%
SK-Learn Bernoulli NB	18%
Multinomial NB (HW2.4)	2%
Multinomial NB (HW2.5)	1%

HW2.6 Discussion:

Since Bernoulli NB classifier emphasized on the appearance of word, while Multinomial NB classifier emphasizes on the frequency of the word, one would like to choose Bernoulli classifier for SPAM detection by focusing keywords appearing in SPAM emails.

HW2.7 Preprocess the Entire Spam Dataset (OPTIONAL)

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```
In [179]: | # Put the download folder at the same place of this notebook
          from os import listdir
          from os.path import isfile, join
          # function to read one folder into text
          def readmail(path, isSpam, handler):
              files = [f for f in listdir(path) if isfile(join(path, f))]
              for file_name in files:
                  id = '.'.join(file_name.split('.')[:-2])
                  f = open(path+file name, 'r')
                  mail = f.read().split('\n', 1)
                  subject = mail[0].strip('Subject:').strip()
                  body = ' '.join(mail[1].split()).strip()
                  row = '%s\t%s\t%s' %(id, isSpam, subject, 'NA' if not body else body)
                  f.close()
                  handler.write(row + '\n')
          # read two folder separately
          text file = open("train-Enron-1.txt", "w")
          readmail('./enron1-Training-Data-RAW/ham/', '0', text file)
          readmail('./enron1-Training-Data-RAW/spam/', '1', text file)
          text file.close()
          print 'File train-Enron-1.txt was generated!'
          !hdfs dfs -rm /user/shihyu/train-Enron-1.txt
          !hdfs dfs -put train-Enron-1.txt /user/shihyu
          print 'train-Enron-1.txt is uploaded to hdfs.'
```

File train-Enron-1.txt was generated! rm: `/user/shihyu/train-Enron-1.txt': No such file or directory train-Enron-1.txt is uploaded to hdfs.

HW2.8 Build and evaluate a NB classifier on the Entire Spam Dataset (OPTIONAL)

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```
In [211]: # clean up HDFS
!hdfs dfs -rm -r Emails_8_classification_results
!hdfs dfs -rm -r prob_em

# run trainning job
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-mapper /home/cloudera/mapper_t_em.py \\
-reducer /home/cloudera/reducer_t_em_I.py \\
-input /user/shihyu/train-Enron-1.txt \\
-output prob_em

# run classification job
!hadoop jar /usr/lib/hadoop-mapreduce/hadoop-streaming.jar \[
-mapper /home/cloudera/mapper_c_em_I.py \\
-reducer /home/cloudera/reducer_c_em.py \\
-input /user/shihyu/train-Enron-1.txt \\
-output Emails_8_classification_results
```

```
rm: `Emails 8 classification results': No such file or directory
Deleted prob em
packageJobJar: [] [/usr/lib/hadoop-mapreduce/hadoop-streaming-2.6.0-cdh5.8.0.jar
] /tmp/streamjob2182033911936835284.jar tmpDir=null
16/09/10 07:03:56 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/10 07:03:56 INFO client.RMProxy: Connecting to ResourceManager at /0.0.0.0
16/09/10 07:03:56 INFO mapred. File Input Format: Total input paths to process: 1
16/09/10 07:03:57 WARN hdfs.DFSClient: Caught exception
java.lang.InterruptedException
        at java.lang.Object.wait(Native Method)
        at java.lang.Thread.join(Thread.java:1281)
        at java.lang.Thread.join(Thread.java:1355)
        at org.apache.hadoop.hdfs.DFSOutputStream$DataStreamer.closeResponder(DF
SOutputStream.java:862)
       at org.apache.hadoop.hdfs.DFSOutputStream$DataStreamer.endBlock(DFSOutpu
tStream.java:600)
        at org.apache.hadoop.hdfs.DFSOutputStream$DataStreamer.run(DFSOutputStre
am.java:789)
16/09/10 07:03:57 INFO mapreduce. JobSubmitter: number of splits:2
16/09/10 07:03:57 INFO mapreduce. JobSubmitter: Submitting tokens for job: job 14
73444507507_0061
16/09/10 07:03:57 INFO impl.YarnClientImpl: Submitted application application 14
73444507507 0061
16/09/10 07:03:57 INFO mapreduce. Job: The url to track the job: http://quickstar
t.cloudera:8088/proxy/application 1473444507507 0061/
16/09/10 07:03:57 INFO mapreduce. Job: Running job: job_1473444507507_0061
16/09/10 07:04:05 INFO mapreduce. Job: Job job 1473444507507 0061 running in uber
mode : false
16/09/10 07:04:05 INFO mapreduce.Job: map 0% reduce 0%
16/09/10 07:04:15 INFO mapreduce.Job: map 50% reduce 0%
16/09/10 07:04:16 INFO mapreduce.Job: map 100% reduce 0%
16/09/10 07:04:26 INFO mapreduce.Job: map 100% reduce 100%
16/09/10 07:04:26 INFO mapreduce. Job: Job job 1473444507507 0061 completed succe
16/09/10 07:04:27 INFO mapreduce. Job: Counters: 50
        File System Counters
                FILE: Number of bytes read=28116899
                FILE: Number of bytes written=56589444
                FILE: Number of read operations=0
               FILE: Number of large read operations=0
               FILE: Number of write operations=0
               HDFS: Number of bytes read=5413593
                HDFS: Number of bytes written=576332
                HDFS: Number of read operations=9
                HDFS: Number of large read operations=0
                HDFS: Number of write operations=2
        Job Counters
               Killed map tasks=1
                Launched map tasks=2
                Launched reduce tasks=1
                Data-local map tasks=2
                Total time spent by all maps in occupied slots (ms)=16575
                Total time spent by all reduces in occupied slots (ms)=8133
                Total time spent by all map tasks (ms) = 16575
                Total time spent by all reduce tasks (ms)=8133
                Total vcore-seconds taken by all map tasks=16575
                Total vcore-seconds taken by all reduce tasks=8133
                Total megabyte-seconds taken by all map tasks=16972800
                Total megabyte-seconds taken by all reduce tasks=8328192
        Map-Reduce Framework
                Map input records=5172
                Map output records=848224
```

```
In [ ]: !hdfs dfs -cat Emails_8_classification_results/part-00000
```

Since the result is too long, so I do not show total result about entries. Since Hadoop is executed successfully, for this train-Enron-1.txt data set, rrror rate is: 0.008509, and the Number of messages with zero probability are spam(1360), ham(3590).

HW2.8.1 OPTIONAL

```
In [217]: from sklearn.naive_bayes import BernoulliNB
          from sklearn.naive bayes import MultinomialNB
          from sklearn.feature extraction.text import *
          import csv, re, string
          import numpy as np
          # read email message, and organize training data
          f = open('train-Enron-1.txt', 'r')
          txt = f.read().strip()
          f.close()
          emails = txt.split('\n')
          train_label = [msg.strip().split('\t', 2)[1] for msg in emails]
          train data = [msg.strip().split('\t', 2)[-1] for msg in emails]
          # removing the funky characters
          regex = re.compile('[%s]' % re.escape(string.punctuation))
          train data = [' '.join(regex.sub(' ', msg).split()).decode('latin-1') for msg in t
          rain data]
          # read test data
          with open ('enronemail 1h.txt', 'r') as f:
              reader = csv.reader(f, delimiter="\t")
              emails = list(reader)
          test label = [msg[1] for msg in emails]
          test data = [msg[2]+msg[3] if len(msg)==4 else msg[2] for msg in emails
          # feature vectorization
          Vectorizeration = CountVectorizer()
          Train = Vectorizeration.fit_transform(train_data)
          Test = Vectorizeration.transform(test_data)
          # multinomial Naive Bayes Classifier from sklearn
          clf M = MultinomialNB()
          clf_M.fit(Train, train_label)
          pred mnb = clf_M.predict(Test)
          training_error_mnb = 1.0*sum(pred_mnb != test_label) / len(test_label)
          # Bernoulli Naive Bayes Classifier from sklearn
          clf B = BernoulliNB()
          clf B.fit(Train, train label)
          pred bnb = clf B.predict(Test)
          training error bnb = 1.0*sum(pred bnb != test label) / len(test label)
          print 'SK- multinomial NB error rate is : %.4f' %training error mnb
          print 'SK- Bernoulli NB error rate is : %.4f' %training error bnb
          SK- multinomial NB error rate is: 0.0200
          SK- Bernoulli NB error rate is: 0.1900
```

HW2.8.1 Discussion: Performance comparison shown by following table.

	Training Error
SK-Learn Multinomial NB	0.02
SK-Learn Bernoulli NB	0.19
Multinomial NB (HW2.4)	0.02
Multinomial NB (HW2.5)	0.01

In []:	
---------	--