Python MapReduce code

- We use HadoopStreaming
- It helps us by passing data between our Map and Reduce code via STDIN (standard input) and STDOUT (standard output)
- We use python's sys.stdin to read input data and print our own output to sys.stdout
- That's all we need to do because HadoopStreaming will take care of everything else!

Hadoop streaming

- Basically, write some code that runs like :
 - \$ cat data | mapper.py | sort | reducer.py

Ex. 1: wordcount

- We write a simple MapReduce program for Hadoop in Python
- Our program reads text files and counts how often words occur
- The input is text files and the output is text files, each line of which contains a word and the count of how often it occurred, separated by a tab

Mapper

- It will read data from STDIN, split it into words and output a list of lines mapping words to their (intermediate) counts to STDOUT
- The Map script will not compute an (intermediate) sum of a word's occurrences
- Instead, it will output "<word> 1" immediately even though the <word> might occur multiple times in the input – and just let the subsequent Reduce step do the final sum count

Reducer

It will read the results of the Mapper from STDIN, and sum the occurrences of each word to a final count, and output its results to STDOUT

pseudo-code

```
1: class Mapper
      method Map(docid a, doc d)
2:
          for all term t \in \text{doc } d do
3:
               Emit(term t, count 1)
4:
  class Reducer
       method Reduce(term t, counts [c_1, c_2, \ldots])
2:
          sum \leftarrow 0
3:
          for all count c \in \text{counts } [c_1, c_2, \ldots] do
4:
               sum \leftarrow sum + c
5:
          Emit(term t, count sum)
6:
```

Ex. 2: co-occurrences

- We want to build a word co-occurrence matrix from a corpus where
 - The co-occurrence matrix is a square n x n matrix where n is the number of unique words in the corpus (i.e, the vocabulary size)
 - A cell m_{i,j} contains the number of times word w_i co-occurs with w_j within a specific context – a certain window of m words in our case

Pseudo-code

```
1: class Mapper
      method Map(docid a, doc d)
          for all term w \in \text{doc } d do
             for all term u \in \text{Neighbors}(w) do
                 Emit (w, u), count 1) \triangleright Emit count for each co-occurrence
  class Reducer
      method Reduce(pair p, counts [c_1, c_2, \ldots])
          s \leftarrow 0
3:
         for all count c \in \text{counts } [c_1, c_2, \ldots] do
                                                            s \leftarrow s + c
          Emit(pair p, count s)
```

Mapper

- It will read data from STDIN, split it into words and output a list of lines mapping pair of words to their (intermediate) counts to STDOUT
- The Map script will not compute an (intermediate) sum of a word's co-occurrences
- Instead, it will output "<pair> 1" immediately even though the <pair> might occur multiple times in the input and just let the subsequent Reduce step do the final sum count

Reducer

It will read the results of the Mapper from STDIN, and sum the co-occurrences of each pair to a final count, and output its results to STDOUT

Run

- \$ cat data | mapper.py | sort | reducer.py
- If that works, run it live
 - Put the data into hadoops Distributed File System (DFS)
 - Run hadoop
 - Read the output data in the DFS

Run Hadoop

Stream data through these two files, saving the output back to HDFS:

```
$HADOOP_HOME/bin/hadoop jar \
$HADOOP_HOME/hadoop-streaming.jar \
-input input_dir \
-output output_dir \
-mapper mapper.py \
-reducer reducer.py \
-file mapper.py -file reducer.py
```

View output

- View output files:
 - \$ hadoop dfs -ls output_dir
- Note multiple output files ("part-00000", "part-00001", etc)
- View output file contents:
 - \$ hadoop dfs -cat output_dir/part-00000