# ELEN 4020 Lab 3

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#### I. Introduction

The purpose of this lab is to expose the students to the programming model of MapReduce. MapReduce makes use of underlying principles of parallelization, allowing for large datasets to be processed in a scalable and fault-tolerant way.

MapReduce is ordinarily comprised of two types of functions, namely mapping functions and reducing functions (sometimes there are combine functions). The map function operates on every member of an input list, returning as output two linked-pieces of data - a key and a value. These key-value pairs are then grouped according to the keys generated.

The output of the map function is then passed on to the reduce function. The reduce function takes in as an argument a key and all values associated with that key, to perform a function set out by the user.

For the purposes of this lab, students are required to develop two algorithms which make use of the MapReduce programming model, in order to perform matrix multiplication. Students have opted to make use of Python and the MrJob MapReduce framework, to implement the two algorithms.

## II. ALGORITHM A: NAIVE MATRIX MULTIPLICATION

## A. Map Function

In order to perform matrix multiplication, the algorithm is required to obtain element data for two matrices (say Matrix A and B) of appropriate size. Hence, when the algorithm is run in the terminal, two input file names are supplied as arguments. Each input file, states the dimensions of a matrix in its first line. This is followed by every single matrix element's coordinate-position as well as value (where the information of each element is on its own newline). This is known as Matrix Market format and is useful for large sparse matrices as there is no need to store elements with value zero.

The map function is designed to read every line of two inputted files. It first determines which input file it has read data from; this information is necessary, since specific logic is employed when generating key value pairs that inserts an 'M' or an 'N' to the value of each key value pair depending on whether the text file is the first text file or the second. For more details about the logic employed, please refer to Figure 3. The map function is completed, when a key-value pair is generated for each element of the input files.

For Matrix M its elements have coordinates (i, j) where i refers to the rows and j refers to the columns. Matrix M's values are given by  $m_{ij}$ . For Matrix N its elements have

coordinates (j,k) where j refers to the rows and k refers to the columns. Matrix M's values are given by  $n_{jk}$ . The keys are given by (i,k) where for M i is the i coordinate of element  $m_{ij}$  being mapped and k is a value that goes from 0 to the number of columns of the N matrix. For N i is a value that goes from 0 to the number of rows of the M matrix and k is the k coordinate of element  $n_{jk}$  being mapped. The values for the M matrix are generated in this format  $('M', j, m_{ij})$ . The values for the N matrix are generated in this format  $('N', j, n_{jk})$ .

Fig. 1: Excerpt code

#### B. Reduce Function

Once the map function has generated all the key-value pairs for the two matrices, the reduce function is defined to create two lists (one for each matrix) of all the values for each key. These two lists are then sorted according to their j values.

```
def reducer(self, key, values):
    oldlistM = []
    oldlistN = []

for i in values:
        if i[0]=='M':
              oldlistM.append([i[1],i[2]])
        else:
                    oldlistN.append([i[1],i[2]])
    #print oldlistM
    #print oldlistN
    listM = sorted(oldlistM, key=lambda x:x[1])
    listN = sorted(oldlistN, key=lambda x:x[1])
```

The matrix multiplication is then performed by calculating the sum of the multiplication of all values  $m_{ij}$  and  $n_{jk}$  in the lists that have the same j value. This is described by the equation below:

Fig. 2: List formation

$$P(i,k) = \sum_{i=1}^{n} (m_{ij} * n_{jk})$$

Where P(i,k) is the element in the resulting matrix with coordinates (i,k). The code implementing the multiplication is shown below:

Fig. 3: Matrix Multiplication

TABLE I: Table of time comparisons

	Time taken (s)		
	A1*B1	A2*B2	A3*B3
Algorithm A	24.95679		
Algorithm B			

#### III. COMPARISON

### IV. DIRECTED GRAPH PROBLEM

A problem is presented in which an unweighted graph G =(v, e) with |v| nodes and |e| edges (of length 1) must be represented in some data structure and then the nodes that are connected by paths of length 3 must be calculated. The most straightforward solution would be to represent the graph as a  $v \times v$  matrix with  $2 \times |e|$  non-zero entries. The each row and each column represent a node and each entry determines connections between the two nodes. A one would mean a direct connection between the two nodes and a zero would mean there is either no connection or it is the same node. In order to see the nodes which are connected by length 2, we multiply the matrix by itself and the non-zero entries are connections of length 2 between the nodes. Therefore, the formula for finding connections of length n between nodes in the graph represented by matrix G is  $G^n$ . Therefore, to find the connections between nodes of length 3, we need to multiply the matrix by itself 3 times.

## V. CONCLUSION