CS246: Mining Massive Datasets

Assignment number: 1
Fill in and include this cover sheet with each of your assignments. Assignments and code are due at 5:00 PM on Scoryst and SNAP respectively. Failure to include the coversheet with you assignment will be penalized by 2 points. Each student will have a total of <i>two</i> free late periods. <i>One late period expires at the start of each class</i> . (Assignments are due on Thursdays, which means the first late period expires on the following Tuesday at 5:00 PM.) Once these late periods are exhausted, any assignments turned in late will be penalized 50% per late period. However, no assignment will be accepted more than <i>one</i> late period after its due date. (If an assignment is due on Thursday 5 PM then we will not accept it after the following Tuesday 5 PM.)
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I acknowledge and accept the Honor Code.
(Signed) Priyank Mathur

- 1) Source Code available towards the end of this file.
- 2) I solved this problem using a single map reduce job. The way I tackled this is for each line in the file, I outputted 2 sets of key value pairs from the mapper. First a list of direct relations to that user and the second was a set of distance 2 friends created from generating all pairs from the friends list of that line. For eg:

```
For line: 1 2,3
I generate the following pairs
1 1-2
1 1-3
2 2-3
```

Where the value is of the form distance-user2

In the reducer, for each user I kept track of others users distance to it and counted the number of people at distance of 2 from it. I sorted this counted list and outputted users who had most number of mutual friends. [1]

```
Sample output -
   38737,18591,27383,34211,337,352,1532,12143,12561,17880
   35621,44891,14150,15356,35630,13801,13889,14078,25228,13805
2
  41087,1,5,95,112,1085,1404,2411,3233,4875
  27679,1,10,16,29,30,38,82,83,85
3)
924
          439,2409,6995,11860,15416,43748,45881
8941
          8943.8944.8940
          8939,8940,8943,8944
8942
          9022,317,9023
9019
9020
          9021,9016,9017,9022,317,9023
9021
          9020,9016,9017,9022,317,9023
9022
          9019,9020,9021,317,9016,9017,9023
          13134,13478,13877,34299,34485,34642,37941
9990
9992
          9987,9989,35667,9991
9993
          9991,13134,13478,13877,34299,34485,34642,37941
```

[1] Cems.uwe.ac.uk, (2014). Paul Matthews Blog @ UWE | Teaching MapReduce. [online] Available at: http://www.cems.uwe.ac.uk/~pmatthew/blog/2014/03/04/teaching-mapreduce [Accessed 21 Jan. 2015].

```
Q2.
d)
Rule, Confidence
'DAI93865 -> FRO40251', 1.0
'GRO85051 -> FRO40251', 0.999176276771005
'GRO38636 -> FRO40251', 0.9906542056074766
'ELE12951 -> FRO40251', 0.9905660377358491
'DAI88079 -> FRO40251', 0.9867256637168141
e)
Rule, Confidence
```

Rule, Confidence
'DAI23334,ELE92920 -> DAI62779', 1.0
'DAI31081,GRO85051 -> FRO40251', 1.0
'DAI55911,GRO85051 -> FRO40251', 1.0
'DAI62779,DAI88079 -> FRO40251', 1.0
'DAI75645,GRO85051 -> FRO40251', 1.0

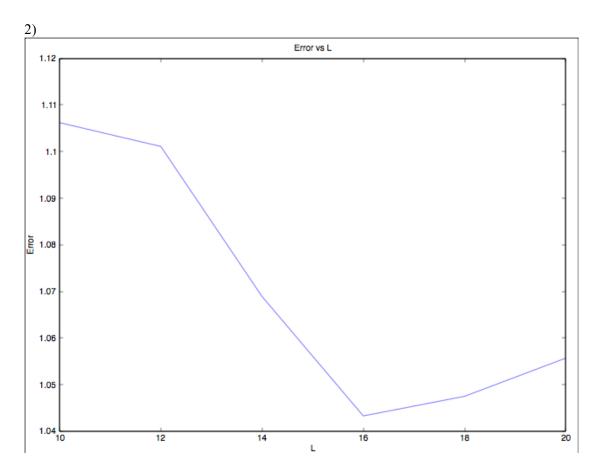
Source Code available towards the end of this file.

Q4. d)

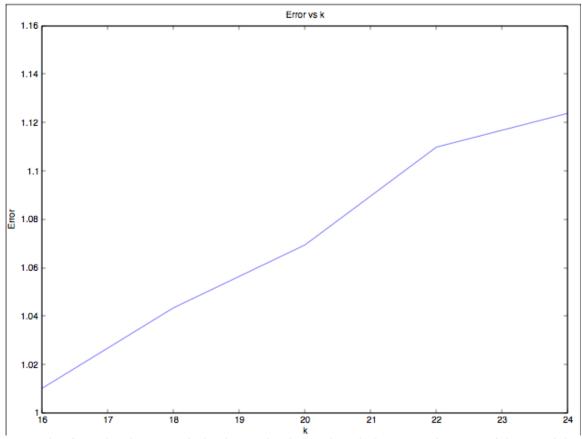
1)

Average search time for LSH - 0.019817 Average search time for Linear search - 0.25485

We see significant improvement in run time of LSH, which is approximately 10 times faster than linear search.



We notice from the above graph that increasing L (number of tables) generally should decrease the error of the approximate search of LSH. This may be because we are increasing the number of terms in the OR construct (buckets) and hence increasing the probability of similar things hashing together.

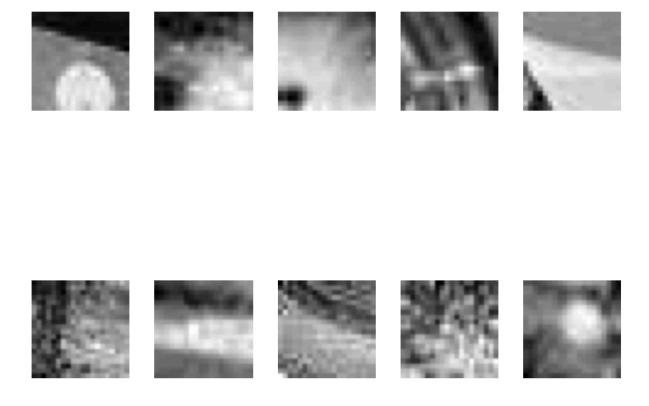


We notice from the above graph that increasing k (key length) increases the error of the search by LSH. This may be because we are increasing the number of terms in the AND construct (rows) and hence decreasing the probability of similar things hashing together.

3)

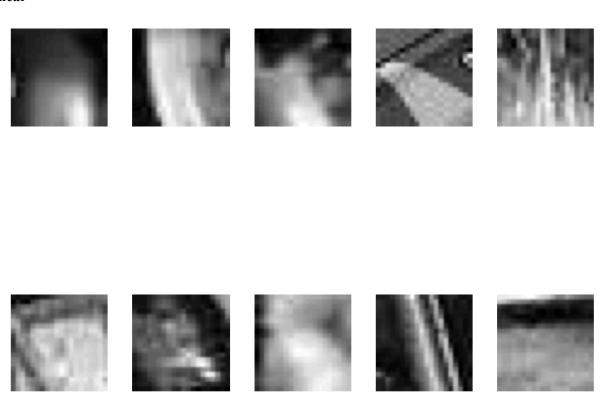
Query point (column 100) -





The nearest neighbors found look to be similar to query image above but some options towards the end (towards bottom and right) appear to further diverge away from the query.

Linear –



The result of linear search is much better than LSH. The query image contains a bright-elongated spot in the middle and is darker towards the edges. Most of the neighbors found by linear search have similar appearance. In contrast, some of the neighbors produced by LSH (eg. number 3 and 5 in top row) are different from this pattern.

We can conclude from this experiment that even though LSH is significantly faster, it might not produce very accurate results.

Q1 Source code package edu.stanford.cs246.friendrecommender; import java.io.IOException; import java.util.ArrayList; import java.util.Arrays; import java.util.Collection; import java.util.Comparator; import java.util.HashMap; import java.util.Iterator; import java.util.Map.Entry; import java.util.PriorityQueue; import java.util.Stack; import java.util.regex.Matcher; import java.util.regex.Pattern; import org.apache.hadoop.conf.Configuration; import org.apache.hadoop.conf.Configured; import org.apache.hadoop.fs.Path; import org.apache.hadoop.io.IntWritable; import org.apache.hadoop.io.LongWritable; import org.apache.hadoop.io.Text; import org.apache.hadoop.mapreduce.Job; import org.apache.hadoop.mapreduce.Mapper; import org.apache.hadoop.mapreduce.Reducer; import org.apache.hadoop.mapreduce.lib.input.FileInputFormat; import org.apache.hadoop.mapreduce.lib.input.TextInputFormat; import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat; import org.apache.hadoop.mapreduce.lib.output.TextOutputFormat; import org.apache.hadoop.util.Tool; import org.apache.hadoop.util.ToolRunner; public class FriendRecommender extends Configured implements Tool { public static void main(String[] args) throws Exception { System.out.println(Arrays.toString(args)); int res = ToolRunner.run(new Configuration(), new FriendRecommender(), args); System.exit(res); @Override public int run(String[] args) throws Exception { System.out.println(Arrays.toString(args)); Job job = new Job(getConf(), "FriendRecommender"); job.setJarByClass(FriendRecommender.class); iob.setOutputKeyClass(IntWritable.class); job.setOutputValueClass(Text.class); job.setMapperClass(Map.class); job.setReducerClass(Reduce.class); job.setInputFormatClass(TextInputFormat.class); job.setOutputFormatClass(TextOutputFormat.class);

 $File Input Format. add Input Path(job, new Path(args[0])); \\ File Output Format. set Output Path(job, new Path(args[1])); \\$

```
job.waitForCompletion(true);
 return 0;
public static class Map extends Mapper<LongWritable, Text, IntWritable, Text> {
        Text dist1Text = new Text();
        Text dist2Text = new Text():
 @Override
 public void map(LongWritable key, Text value, Context context)
       throws IOException, InterruptedException {
       // Get the information from the line
       String line = value.toString();
       String[] users = line.split("\\t");
       String friendString = users.length > 1 ? users[1] : "";
       String[] friends = friendString.split(",");
       /* Loop through friends list to generate pairs of following types -
        * For line: 1 2,3
        * we generate the following pairs
        * 1
            1-2
        * 1
              1-3
        * 2
              2-3
        * Where the value is of the form distance-user2
       for (int i=0; !friendString.equals("") && i<friends.length; i++) {
               String dist1Str = "1-" + friends[i];
               dist1Text.set(dist1Str);
               context.write(new IntWritable(Integer.parseInt(users[0])), dist1Text);
               // Generate pairs with distance 2
               for (int j=i+1; j < friends.length; j++) {
                        String dist2Str = "2-" + friends[j];
                        dist2Text.set(dist2Str);
                        context.write(new IntWritable(Integer.parseInt(friends[i])), dist2Text);
                        dist2Str = "2-" + friends[i];
                        dist2Text.set(dist2Str);
                        context.write(new IntWritable(Integer.parseInt(friends[i])), dist2Text);
               }
       }
public static class Reduce extends Reducer<IntWritable, Text, IntWritable, Text> {
        * For each user user1, keep track of how many common friends does he
        * have with another user user2. Then remove direct relations and
        * output the distance 2 users sorted by the number of mutual friends.
```

```
@Override
   public void reduce(IntWritable key, Iterable<Text> values, Context context)
        throws IOException, InterruptedException {
     HashMap<Integer, Integer> map = new HashMap<Integer, Integer>();
     for (Text val : values) {
       String association = val.toString();
       int length = Integer.parseInt(association.split("-")[0]);
       int friend = Integer.parseInt(association.split("-")[1]);
       // Keep track of other users and their distance
       if (length == 1)
       map.put(friend, 0);
       else if (length == 2) {
       if (map.containsKey(friend)) {
                if (map.get(friend)!=0)
                        map.put(friend, map.get(friend)+1);
        }
       else {
                map.put(friend, 1);
     PriorityQueue<Entry<Integer, Integer>> queue = new PriorityQueue<Entry<Integer, Integer>>(10, new
EntryComparator());
     // Put them in a priority queue to get the 10 users
     // with most number of mutual friends
     for (Entry<Integer, Integer> e: map.entrySet()) {
        if (e.getValue()==0)
                continue;
        if (queue.size() < 10)
                queue.add(e);
        else {
                int tempValue = ((Entry<Integer, Integer>) queue.peek()).getValue();
                int tempKey = ((Entry<Integer, Integer>) queue.peek()).getKey();
                if ((tempValue < e.getValue()) || (tempValue == e.getValue() && tempKey > e.getKey())){
                        queue.poll();
                        queue.add(e);
                }
     String recommendations = "";
     Stack<Entry<Integer, Integer>> reverser = new Stack<Entry<Integer, Integer>>();
     // Reverse the order
     while (queue.size() > 0) {
        reverser.push(queue.poll());
     }
     if (reverser.size() > 0) {
```

```
Entry<Integer, Integer> element = reverser.pop();
        recommendations += element.getKey();
     }
     while (reverser.size() > 0) {
        Entry<Integer, Integer> element = reverser.pop();
        recommendations += "," + element.getKey();
     Text recommendationsText = new Text(recommendations);
     // Write out the output
     context.write(key, recommendationsText);
* Comparator to sort the entries in the priority queue
* first by value and then by ascending numerical order
* of the keys
class EntryComparator implements Comparator<Entry<Integer, Integer>>
  public int compare(Entry<Integer, Integer> e1, Entry<Integer, Integer> e2)
       if (e1.getValue() != e2.getValue())
               return e1.getValue().compareTo(e2.getValue());
       else
               return e2.getKey().compareTo(e1.getKey());
  }
```

Q2 Source Code -

from collections import defaultdict from itertools import chain, combinations

```
# Globals
data file = 'browsing.txt'
support = 100
item counter = 1
def getNewId():
  global item counter
  retVal = item counter
  item counter += 1
  return retVal
# Containers
name hash = defaultdict(getNewId)
id hash = \{\}
C1 ctr = defaultdict(int)
C2 ctr = defaultdict(int)
C3 ctr = defaultdict(int)
R2 conf = defaultdict(int)
R3 conf = defaultdict(int)
def readFileGenerator(filename=data file):
  Return lines of data
  file obj = open(data file)
  for line in file_obj:
     yield line
def count singletons():
  Count single items
  for line in readFileGenerator():
     items = set(line.strip().split(" "))
     for item in items:
       key = name hash[item] # maintain name to id mapping
       id hash[key] = item # maintain id to name mapping
       C1 ctr[key] = C1 ctr[key] + 1
def isFrequentItem(item):
  Check if an item has support > min support
  key = name hash[item] if item in name hash else None
  return C1 ctr[key] > support if key else False
def isFrequentPair(pair):
  Check if a pair has support > min support
  key1 = name_hash[pair[0]] if pair[0] in name_hash else None
  key2 = name hash[pair[1]] if pair[1] in name hash else None
```

```
if key1==None or key2==None:
     return False
  dict key = tuple(sorted((key1, key2)))
  return C2 ctr[dict key] > support if dict key in C2 ctr else False
def prune storage(storage):
  Remove counts of items where support < min support
  remove keys = [key for key in storage.keys() if storage[key] < support]
  for k in remove keys: del storage[k]
def prune_items(items, k=2):
  Returns a list of items to be used for creation of
  higher order item sets
  freq items = {item:0 for item in items if isFrequentItem(item)}
  if k == 3:
     candidate sets = combinations(freq items, 2)
     freq candidate sets = [pair for pair in candidate sets if isFrequentPair(pair)]
     Check each of the items appears in at least 2 freq pairs
     for it in freq items:
       for c in freq candidate sets:
          if it in c:
            freq items[it] = freq items[it] + 1
     freq items = {item:cnt for item, cnt in freq items.items() if cnt \geq 2}
  return freq items
def makePassK(k=2):
  Create and filter item sets of size k
  cnt = 0
  storage = C2_ctr if k==2 else C3_ctr
  storage.clear()
  for line in readFileGenerator():
     items = set(line.strip().split(" "))
     freq_items = prune_items(items, k)
     #print freq items
     candidate sets = combinations(freq items, k)
     for c in candidate_sets:
```

```
key1 = name hash[c[0]]
       key2 = name hash[c[1]]
       key3 = name hash[c[2]] if k==3 else None
       dict key = (key1, key2, key3) if k==3 else (key1, key2)
       dict key = tuple(sorted(dict key))
       storage[dict key] = storage[dict key] + 1
  prune_storage(storage)
def get3Rules(k):
  Generate rules from triples of form
  item1, item2 -> item3
  ks = set(k)
  for t in combinations(k, 2):
     C2 key = tuple(sorted(t))
     #print C2_key
     n1 = id hash[t[0]]
     n2 = id hash[t[1]]
     srt = sorted([n1, n2])
     lhs = srt[0] + "," + srt[1]
     ts = set(t)
     rhs = ks.difference(ts)
     rhs = rhs.pop()
     representation = lhs + " -> " + id hash[rhs]
     num = C3 \text{ ctr}[k] * 1.0 \text{ if } k \text{ in } C3 \text{ ctr else None}
     denom = C2_ctr[C2_key] if C2_key in C2_ctr else None
     #print num, denom
     if num==None or denom==None:
       yield (representation, 0)
     else:
       conf = num / denom
       yield (representation, conf)
def get2Rules(k):
  Generate rules from pairs of form
  item1 -> item2
  ks = set(k)
  for t in combinations(k, 1):
     C1 key = t[0]
     #print C1 key
     lhs = str(C1 key)
     ts = set(t)
     rhs = ks.difference(ts)
     rhs = rhs.pop()
     representation = id hash[C1 key] + " -> " + id hash[rhs]
     num = C2_{ctr}[k] * 1.0 if k in C2_{ctr} else None
     denom = C1 ctr[C1 key] if C1 key in C1 ctr else None
```

```
#print num, denom
     if num==None or denom==None:
       yield (representation, 0)
     else:
       conf = num / denom
       yield (representation, conf)
def getRules(k):
  Create and store rules of length k
  if k==3:
     for triplet in C3 ctr:
       #print triplet
       for rule in get3Rules(triplet):
          R3\_conf[rule[0]] = rule[1]
  elif k==2:
     for pair in C2 ctr:
       for rule in get2Rules(pair):
          R2\_conf[rule[0]] = rule[1]
# Execute commands
count singletons()
makePassK(2)
makePassK(3)
getRules(2)
getRules(3)
# Display results
#sorted(R2_conf.items(), key=lambda x: (x[1],x[0]), reverse=True )[:20]
#sorted(R3 conf.items(), key=lambda x: (x[1],x[0]), reverse=True)[:20]
```

Q4 Source Code -

```
%% One time initialization
load patches;
% Perform linear search and return nearest neighbors
function nnlinind = linearsearch(query, data, num)
        d = sum(abs(bsxfun(@minus, query, data)));
        [ignore,ind]=sort(d);
        cand = ind(1:num+1);
        nnlinind = cand;
end
%% Calculate error ratio
function [err] = calcerror(linn, lshnn, data)
        err=0;
        for i=1:10,
                 qcol = data(:, lshnn(i, 1));
                 lshids = lshnn(i, 2:4);
                 lshids = lshids(lshids \sim = 0);
                 lshdcols = data(:, lshids);
                linids = linn(i, 2:4);
                 linids = linids(linids \sim= 0);
                 lindcols = data(:, linids);
                 lshdist = sum(sum(abs(bsxfun(@minus, lshdcols, qcol))));
                 lindist = sum(sum(abs(bsxfun(@minus, lindcols, qcol))));
                 % disp(sprintf('%s%d%s%d%s%f', 'lshdist-', lshdist, '; lindist-', lindist, '; error-',
lshdist/lindist));
                err = err + (lshdist * 1.0 /lindist);
        end
        err = 10;
end
%% driver function to calulate error
function [errors] = driver(data)
        errors = [];
        % Vary L and get nearest neighbors
        for i=10:2:20,
                 T1=lsh('lsh',i,24,size(data,1),data,'range',255);
                 linn = findnn(data, 'linearsearch', T1, 3);
                lshnn = findnn(data, 'lsh', T1, 4);
                 err = calcerror(linn, lshnn, data);
                 errors = [errors err];
```

```
% Vary k and get nearest neighbors
        for i=16:2:24,
                T1=lsh('lsh',10,i,size(data,1),data,'range',255);
                linn = findnn(data, 'linearsearch', T1, 3);
                lshnn = findnn(data, 'lsh', T1, 4);
                err = calcerror(linn, lshnn, data);
                errors = [errors err];
        end
end
% find nearest neighbors based on search type
function op = findnn(patches, searchtype, T1, num)
        neigbours = [];
        disp(searchtype);
        tic;
        for i=1:10,
                colno = i*100;
                query = patches(:, colno);
                if strcmp(searchtype, 'linearsearch'),
                        nn=linearsearch(query, patches, num);
                elseif strcmp(searchtype, 'lsh'),
                        [nn,numcand]=lshlookup(query, patches, T1, 'k', num, 'distfun', 'lpnorm', 'distargs',
{1});
                        if numcand < num,
                                 pads = zeros(1, num-numcand);
                                 disp('-- Padding --');
                                 nn = [nn pads]
                        end
                end
                disp(sprintf('%s%d%s%d', 'neigbours-', size(neigbours,2), '; numcand-', size(nn,2)));
                neigbours = [neigbours; nn];
        end
        op = neigbours;
        time = toc;
        disp(time/10);
end
%% plot nearest neighbors
function plotnn(data, query, nn, numcand)
        % plot the query point
        figure(1); clf;
        imagesc(reshape(query,20,20));
        colormap gray;
```

end

axis image;

```
set(gca,'YTickLabel', sprintf(",[]));
        set(gca,'XTickLabel', sprintf(",[]));
        % plot neigbours
        figure(2);clf;
        for k=1:numcand,
                subplot(2,5,k);
                imagesc(reshape(data(:,nn(k+1)),20,20));
                colormap gray;
                axis image;
                set(gca,'YTickLabel', sprintf(",[]));
                set(gca,'XTickLabel', sprintf(",[]));
        end
end
% Plot the 10 nearest neighbors for both types
query = patches(:, 100);
T1=lsh('lsh',10,24,size(patches,1),patches,'range',255);
[nn,numcand]=lshlookup(query, patches, T1, 'k', 11, 'distfun', 'lpnorm', 'distargs', {1});
plotnn(patches, query, nn, 10);
nn=linearsearch(query, patches, 10);
plotnn(patches, query, nn, 10);
```