F

1. Create a blog-term matrix. Start by grabbing 100 blogs; include:

http://f-measure.blogspot.com/

http://ws-dl.blogspot.com/

and grab 98 more as per the method shown in class. Note that this method randomly chooses blogs and each student will separately do this process, so it is unlikely that these 98 blogs will be shared among students. In other words, no sharing of blog data. Upload to github your code for grabbing the blogs and provide a list of blog URIs, both in the report and in github.

Use the blog title as the identifier for each blog (and row of the matrix). Use the terms from every item/title (RSS) or entry/title (Atom) for the columns of the matrix. The values are the frequency of occurrence. Essentially you are replicating the format of the "blogdata.txt" file included with the PCI book code. Limit the number of terms to the most "popular" (i.e., frequent) 1000 terms, this is \*after\* the criteria on p. 32 (slide 7) has been satisfied. Remember that blogs are paginated.

To solve this problem, I created getBlog.py, which is listed below.

```
1
    import requests
    import generatefeedvector.py
 2
 4
    def get100blogurls():
        f = open('100blogurls.txt', 'w')
 5
         url_set = set()
 7
         while (len(url_set) < 100):
 8
             url = "http://www.blogger.com/next-blog?navBar=true&blogID=3471633091411211117"
             generate url = requests.get(url)
             final_url = generate_url.url.strip('?expref=next-blog/')
             url_set.add(final_url)
         for element in url set:
13
             print element
14
             f.write(element + '\n')
15
         f.write('http://f-measure.blogspot.com' + '\n')
         f.write('http://ws-dl.blogspot.com')
16
17
19
    def completeURL():
         input = open('100blogurls.txt', 'r')
         output = open('100completeurls.txt', 'w')
21
         for element in input:
             add = "/feeds/posts/default?alt=rss"
24
            item = element.strip() + add
             output.write(item + '\n')
             print item
27
    #get100blogurls()
    completeURL()
```

I experimented with a few data structures in which to store the URLs, however a set gave me the least amount of trouble. I used the PowerPoints help in creating the file, as it gave several of the needed pieces, such as the blogger.com site I stored as my url variable, and a few others. This was also very similar to our earlier assignment in which we worked with URLs so I referenced that one as well in regards to handling them. I ran the two functions in succession, which took a few times because the first function would sometimes crash before it ran to completion. Once I was able to gather all 100 blog URLs, and completed them with the necessary suffix, '/feeds/post/default?alt=rss', I ran the newly generated file, 100completeurls.txt through generatefeedvector.py. I found this online at <a href="https://github.com/arthur-e/Programming-Collective-Intelligence/blob/master/chapter3/generatefeedvector.py">https://github.com/arthur-e/Programming-Collective-Intelligence/blob/master/chapter3/generatefeedvector.py</a> while searching for the PCI text. The code

<u>Intelligence/blob/master/chapter3/generatefeedvector.py</u> while searching for the PCI text. The code for generatevectorfee.py is found below.

```
#!/usr/bin/python
    # -*- coding: utf-8 -*-
    import feedparser
    import re
 6 def getwordcounts(url):
 7
        Returns title and dictionary of word counts for an RSS feed
8
        # Parse the feed
        d = feedparser.parse(url)
        WC = \{\}
13
14
        # Loop over all the entries
15
        for e in d.entries:
            if 'summary' in e:
17
                summary = e.summary
19
            else:
20
                summary = e.description
21
            # Extract a list of words
23
            words = getwords(e.title + ' ' + summary)
            for word in words:
25
                wc.setdefault(word, 0)
                wc[word] += 1
27
        return (d.feed.title, wc)
29
31 def getwords(html):
32
        # Remove all the HTML tags
        txt = re.compile(r'<[^>]+>').sub('', html)
        # Split words by all non-alpha characters
        words = re.compile(r'[^A-Z^a-z]+').split(txt)
        # Convert to lowercase
        return [word.lower() for word in words if word != '']
```

```
42
    apcount = {}
43 wordcounts = {}
44 feedlist = [line for line in file('feedlist.txt')]
45 for feedurl in feedlist:
        try:
47
            (title, wc) = getwordcounts(feedurl)
            wordcounts[title] = wc
48
            for (word, count) in wc.items():
49
                apcount.setdefault(word, 0)
                if count > 1:
51
52
                    apcount[word] += 1
53
        except:
54
            print 'Failed to parse feed %s' % feedurl
    wordlist = []
57 for (w, bc) in apcount.items():
58
        frac = float(bc) / len(feedlist)
        if frac > 0.1 and frac < 0.5:
            wordlist.append(w)
62 out = file('blogdata1.txt', 'w')
63 out.write('Blog')
64 for word in wordlist:
        out.write('\t%s' % word)
66 out.write('\n')
67 for (blog, wc) in wordcounts.items():
68
        print blog
        out.write(blog)
        for word in wordlist:
            if word in wc:
71
                out.write('\t%d' % wc[word])
            else:
74
                out.write('\t0')
        out.write('\n')
```

The output of this generatefeedvector.py was spat out into a .txt file entitled blogdata1.txt. A sample of the output is as follows, as the entire file is too large to display in one shot.

1	Blog woods	kids	golden	catchy	absolute	е	wrong	fit	songwri		service			feeling	singe
2	SPIN IT RECORDS	Moncton	467A Ma	in Stree	t Moncto	n NB CAN	ADA	3	4	8	0	0	0	0	0
3	Riley Haas' blog		0	0	0	0	0	0	0	0	0	0	0	0	0
4	Coyote Doc Music	Co-op	0	0	2	6	1	5	1	2	0	1	3	6	5
5	U-Rock Radio™	0	1	0	0	1	0	1	1	3	0	0	1	1	0
6	(Insert World Pr	oblem H	ere) Suc	ks.	0	0	0	0	0	0	0	0	0	0	1
7	Friday Night Dre	am	0	0	0	0	0	1	0	0	0	0	0	0	0
8	On Warmer Music	2	7	6	8	1	3	2	3	2	4	0	15	10	3
9	New Music Matter	S	3	1	1	3	1	0	0	12	0	3	0	1	23
10	She May Be Naked		2	3	0	0	0	2	2	0	0	8	1	9	0
11	Pithy Title Here		7	9	0	3	1	11	8	1	12	7	0	12	9
12	Spinitron Charts		0	0	0	0	0	0	0	0	1	0	0	0	0
13	THE HUB 0	2	1	0	0	0	0	0	1	0	0	0	0	0	0
.4	Anthems and Atle	ticos	0	7	0	0	2	0	1	0	5	2	3	7	1
15	Primitive Offeri	ngs	0	1	0	0	0	0	1	0	0	0	0	0	0
16	Web Science and	Digital	Librari	es Resea	rch Grou	p	0	2	0	0	1	5	2	0	33
7	Words 0	0	0	0	0	2	0	0	0	2	0	1	0	2	0
8	Steel City Rust	0	2	0	0	0	0	2	0	1	0	0	4	3	1
9	Fran Brighton	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0	Stereo Pills	0	0	0	7	0	0	0	2	0	1	0	0	2	0
1	MarkEOrtega's Jo	urnalis	m Portfo	lio	0	0	0	1	0	1	0	1	0	0	0
2	Oh Yes Jónsi!!	0	0	0	0	0	0	0	0	0	0	0	0	0	0
3	aubade 4	2	0	1	0	3	3	0	2	1	0	2	1	1	0
4	turnitup!	0	1	0	0	0	3	0	0	2	1	1	1	1	1
5	Stories From the	City,	Stories	From the	Sea	5	1	2	0	0	0	0	0	0	0
	Chemical Robert!		0	1	0	0	0	0	0	0	1	0	0	0	1
7	adrianoblog	0	0	0	0	0	0	0	0	0	0	0	0	0	0
8	Stephanie Veto P	hotogra	phy	0	0	0	0	0	0	0	0	0	0	0	0
	AHTAPOT	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	Diagnosis: No Ra	dio	0	9	0	0	0	3	5	1	0	2	2	11	1
1	Floorshime Zippe		0	0	0	1	3	0	0	1	0	0	0	0	1
32	Did Not Chart		0	0	0	0	0	0	0	3	0	0	2	0	0
3	Cast Iron Songs	0	0	0	0	0	0	0	0	0	2	0	0	0	0
4	[Tu queres ver		_	0	0	0	0	0	0	0	0	0	0	0	0
5	The Stearns Fami		0	12	0	0	0	4	0	0	0	2	0	1	0
6		0	0	0	0	0	0	0	0	0	0	0	0	0	0
7	DaveCromwell Wri	_	0	3	5	4	2	5	8	17	1	5	4	16	27
8		0	0	0	0	0	2	0	0	0	2	0	5	0	0
	A2 MEDIA COURSEW	_	_	0	0	0	0	0	0	1	0	1	2	0	0
	nonsense a la mo		0	9	1	0	0	1	0	0	4	2	1	0	0
	moniscuse a ra mo	uc	9	9	4	U	9	1	U	U	-	2	4	0	0

## 2. Create an ASCII and JPEG dendrogram that clusters (i.e., HAC)

the most similar blogs (see slides 12 & 13). Include the JPEG in your report and upload the ascii file to github (it will be too unwieldy for inclusion in the report).

My colleague Michelle talked me through how to construct the dendogram. I created dendogramCreator.py. that calls from clusters.py, <a href="https://github.com/arthur-e/Programming-Collective-Intelligence/blob/master/chapter3/clusters.py">https://github.com/arthur-e/Programming-Collective-Intelligence/blob/master/chapter3/clusters.py</a>

```
import clusters
import sys

blognames, words, data = clusters.readfile('blogdata1.txt')

clust = clusters.hcluster(data)

clusters.printclust(clust, labels=blognames)
sys.stdout = open('ascii.text', 'w+')

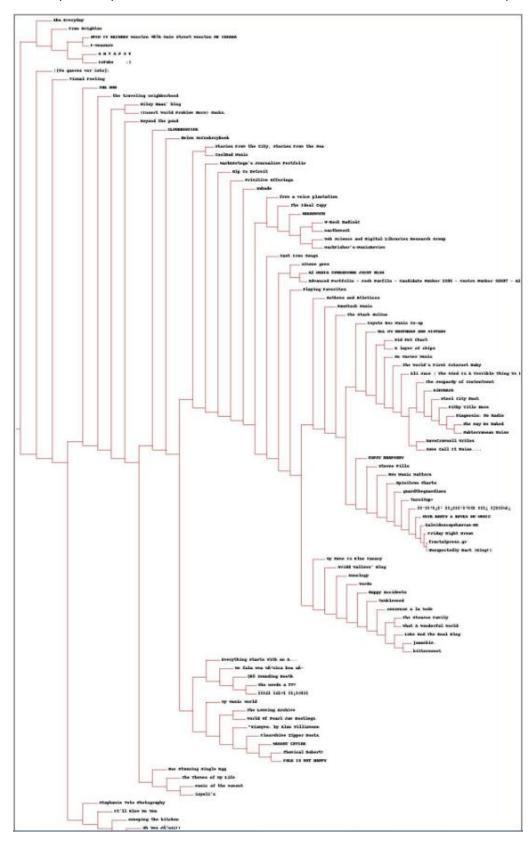
clusters.drawdendrogram(clust, blognames, jpeg='dendogram.jpg')
```

The functions that were called are also listed below.

```
62 def hcluster(rows, distance=pearson):
      distances = {}
64
      currentclustid = -1
66 # Clusters are initially just the rows
      clust = [bicluster(rows[i], id=i) for i in range(len(rows))]
68
      while len(clust) > 1:
          lowestpair = (0, 1)
           closest = distance(clust[0].vec, clust[1].vec)
      # loop through every pair looking for the smallest distance
74
           for i in range(len(clust)):
               for j in range(i + 1, len(clust)):
           # distances is the cache of distance calculations
                   if (clust[i].id, clust[j].id) not in distances:
                       distances[(clust[i].id, clust[j].id)] = \
                           distance(clust[i].vec, clust[j].vec)
                    d = distances[(clust[i].id, clust[j].id)]
                    if d < closest:</pre>
                        closest = d
                       lowestpair = (i, j)
        # calculate the average of the two clusters
         mergevec = [(clust[lowestpair[0]].vec[i] + clust[lowestpair[1]].vec[i])
                       / 2.0 for i in range(len(clust[0].vec))]
        # create the new cluster
          newcluster = bicluster(mergevec, left=clust[lowestpair[0]],
                                  right=clust[lowestpair[1]], distance=closest,
                                  id=currentclustid)
      # cluster ids that weren't in the original set are negative
          currentclustid -= 1
           del clust[lowestpair[1]]
           del clust[lowestpair[0]]
           clust.append(newcluster)
      return clust[0]
```

```
def printclust(clust, labels=None, n=0):
      # indent to make a hierarchy layout
         for i in range(n):
              print ' ',
108
        if clust.id < 0:
         # negative id means that this is branch
              print '-'
112
         else:
        # positive id means that this is an endpoint
114
             if labels == None:
                  print clust.id
             else:
                  print labels[clust.id]
      # now print the right and left branches
        if clust.left != None:
              printclust(clust.left, labels=labels, n=n + 1)
         if clust.right != None:
              printclust(clust.right, labels=labels, n=n + 1)
124
146 def drawdendrogram(clust, labels, jpeg='clusters.jpg'):
     # height and width
      h = getheight(clust) * 20
       W = 1200
       depth = getdepth(clust)
     # width is fixed, so scale distances accordingly
scaling = float(w - 150) / depth
     # Create a new image with a white background
      img = Image.new('RGB', (w, h), (255, 255, 255))
       draw = ImageDraw.Draw(img)
      draw.line((0, h / 2, 10, h / 2), fill=(255, 0, 0))
     # Draw the first node
       drawnode(
           draw,
           clust,
           10,
            h / 2,
            scaling,
168
            labels,
           )
       img.save(jpeg, 'JPEG')
```

The output of my file is as follows, and the ascii.txt file is available on GitHub as requested.



- 3. Cluster the blogs using K-Means, using k=5,10,20. (see slide
- 18). Print the values in each centroid, for each value of k. How

many interations were required for each value of k?

To solve this problem I used another piece of functionality from clusters, to construct my k\_from\_clusters.py. My colleague also aided in the creation of this, I was instructed to create three identical calls to the kcluster function with the three specified increments of k.

```
import clusters

import clusters

print('k = 5')
    k = clusters.kcluster(data, k=5)

print str(k) + '\n'

print('k = 10')
    k = clusters.kcluster(data, k=10)
    print str(k) + '\n'

print('k = 20')
    k = clusters.kcluster(data, k=20)
    print str(k) + '\n'
```

Kcluster's declaration is as follows from clusters.py.

```
def kcluster(rows, distance=pearson, k=4):
      # Determine the minimum and maximum values for each point
         ranges = [(min([row[i] for row in rows]), max([row[i] for row in rows]))
                   for i in range(len(rows[0]))]
      # Create k randomly placed centroids
        clusters = [[random.random() * (ranges[i][1] - ranges[i][0]) + ranges[i][0]
234
                     for i in range(len(rows[0]))] for j in range(k)]
235
         lastmatches = None
        for t in range(100):
             print 'Iteration %d' % t
             bestmatches = [[] for i in range(k)]
241
        # Find which centroid is the closest for each row
             for j in range(len(rows)):
244
                 row = rows[j]
                 bestmatch = 0
                 for i in range(k):
247
                     d = distance(clusters[i], row)
                     if d < distance(clusters[bestmatch], row):</pre>
                         bestmatch = i
                 bestmatches[bestmatch].append(j)
        # If the results are the same as last time, this is complete
             if bestmatches == lastmatches:
                 break
254
             lastmatches = bestmatches
        # Move the centroids to the average of their members
            for i in range(k):
                 avgs = [0.0] * len(rows[0])
                 if len(bestmatches[i]) > 0:
                     for rowid in bestmatches[i]:
                         for m in range(len(rows[rowid])):
                             avgs[m] += rows[rowid][m]
                     for j in range(len(avgs)):
                         avgs[j] /= len(bestmatches[i])
                     clusters[i] = avgs
       return bestmatches
```

The output is as follows.

```
C:\Python27\python.exe Z:/CS432/Assignment8/Q3/k from clusters.py
k = 5
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
[[1, 2, 3, 11, 12, 14, 18, 19, 20, 23, 24, 29, 30, 31, 35, 37, 42, 44, 45,
k = 10
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
[[5, 10, 22, 32, 43, 46, 59, 63, 88, 97], [7, 18, 30, 35, 44, 49, 52, 56, 5
k = 20
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
Iteration 7
[[], [], [26], [87], [2, 12, 19, 28, 30, 51, 53, 61, 64, 72, 73, 74, 77, 90
Process finished with exit code 0
```

4. Use MDS to create a JPEG of the blogs similar to slide 29 of the

week 12 lecture. How many iterations were required?

My MDS.py file was very simple and called from two functions form the clusters.py file- scaledown & draw2d. My file is as follows,

```
import clusters

blognames, words, data = clusters.readfile('blogdata1.txt')

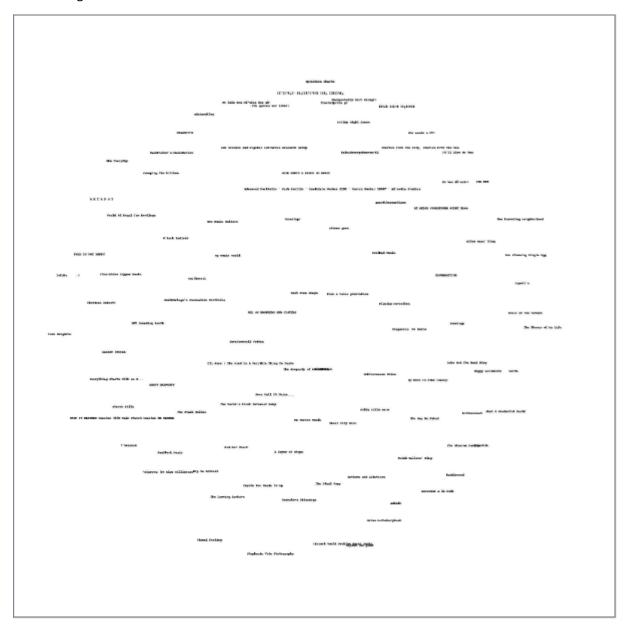
c = clusters.scaledown(data)

clusters.draw2d(c, blognames, jpeg='mds.jpg')
```

And the borrowed functionality are

```
def scaledown(data, distance=pearson, rate=0.01):
     n = len(data)
     # The real distances between every pair of items
       realdist = [[distance(data[i], data[j]) for j in range(n)] for i in
                    range(0, n)]
    def draw2d(data, labels, jpeg='mds2d.jpg'):
         img = Image.new('RGB', (2000, 2000), (255, 255, 255))
         draw = ImageDraw.Draw(img)
341
342
        for i in range(len(data)):
             x = (data[i][0] + 0.5) * 1000
344
             y = (data[i][1] + 0.5) * 1000
             draw.text((x, y), labels[i], (0, 0, 0))
         img.save(jpeg, 'JPEG')
```

There was more to the scaledown function, however it would not fit in one screenshot.



The MDS.jpeg I generated. It took a total of 29 interations.

```
C:\Python27\python.exe Z:/CS432/Assignment8/Q4/MDS.py
4233.62320823
3283.64398847
3228.77840884
3188.93674699
3174.08189037
3161.24926018
3137.88980421
3127.71465533
3119.15387249
3112.55859396
3106.80427737
3095.92650838
3091.64319906
3088.48428465
3085.87100794
3083.34331957
3079.74051901
3078.73751234
3077.6126892
Process finished with exit code 0
```

======The questions below is for 3 points extra credit=======

5. Re-run question 2, but this time with proper TFIDF calculations instead of the hack discussed on slide 7 (p. 32). Use the same 1000 words, but this time replace their frequency count with TFIDF scores as computed in assignment #3. Document the code, techniques, methods, etc. used to generate these TFIDF values. Upload the new data file to github.

Compare and contrast the resulting dendrogram with the dendrogram from question #2.

Note: ideally you would not reuse the same 1000 terms and instead come up with TFIDF scores for all the terms and then choose the top 1000 from that list, but I'm trying to limit the amount of work necessary.

=====The questions below is for 5 points extra credit=======

\_\_\_\_\_

6. Re-run questions 1-4, but this time instead of using the 98 "random" blogs, use 98 blogs that should be "similar" to:

http://f-measure.blogspot.com/ http://ws-dl.blogspot.com/

Choose approximately equal numbers for both blog sets (it doesn't have to be a perfect 49-49 split, but it should be close).

Explain in detail your strategy for locating these blogs.

Compare and contrast the results from the 98 "random" blogs and the 98 "targeted" blogs.