

CS 595: Assignment 9

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1 Problem 1

Question

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.

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) 500 terms, this is *after* the criteria on p. 32 (slide 7) has been satisfied.

Create a histogram of how many pages each blog has (e.g., 30 blogs with just one page, 27 with two pages, 29 with 3 pages and so on).

1.1 Answer

1. The script shown in Listing 1 outputs a list of 100 random blogs in to a file .
2. The script show in Listing 2 outputs a list of atom feed for the respective blog. The atom feed grabbed will give only 25 entries by default.
3. The scripts were written based on the technique that is taught in the class.
4. To get the words from each feed I used *Segaran's generatefeedvector.py* from Program Collective Intelligence Text book.
5. I modified the *generatefeedvector.py* code to get the popular 500 words from all the atom feed.
6. I included a piece of code at line 31 in Listing 3 to eliminate the words which falls under the stop words, the stop word list used is the "MySQL full-text stopwords".
7. The code does not handle the entries if the entries for the blog is beyond 500, as the maximum limit of the entries for an atom feed is limited to only 500 entries at a time.
8. So I decided to visit each page and grab all the entries for a blog. This is implemented in lines 50 to 56 in Listing 3
9. To get the most popular words among all the blogs, firstly I am keeping track of the occurrence of each word in all the blogs.
10. So that I can pick only the top 500 words by sorting the words based on the occurrence value.
11. This is implemented at lines 106 -108 and 121-124 in Listing 3. And then I am limiting down the word to top 500 words only.
12. The script shown in Listing 3 outputs the matrix in expected format which is stored in *blogdata-500.txt* which is used for each subsequent question in this assignment, this file is uploaded in the github.
13. To get the number of pages in each blog, I wrote a small python code as shown in Listing 4 which outputs the file *noOfPages.txt*
14. The R script shown in the Listing 5 generates a Histogram shown in Figure 1 for Blogs Vs no of Pages.

15. From the Histogram in Figure 1, it's pretty clear that 60 blogs out of 100 have pages between 1 and 10, and only couple of blogs have pages between 280 and 290.

```
1  #!/bin/bash
2
3  echo ""
4
5  if [ $# -ne 1 ]
6  then
7      echo "Usage: $0 #"
8      exit
9  fi
10
11 num=$1
12
13 echo "http://f-measure.blogspot.com/" > blogList-$num.txt
14 echo "http://ws-dl.blogspot.com/" >> blogList-$num.txt
15
16 for ((i=0;i< $num;i++))
17 do
18     curl -I -L 'http://www.blogger.com/next-blog?navBar=true&blogID=3471633091411211117' |
19     grep Location | tail -n 1 | cut -d" " -f2 | cut -d"?" -f 1 >> blogList-$num.txt
20 done
```

Listing 1: Shell Program for getting 100 blogs

```
1  #!/bin/bash
2
3  if [ $# -ne 1 ]
4  then
5      echo "Usage: $0 #"
6      exit
7  fi
8
9  blogFile=$1
10
11 echo -n "" > 'basename $blogFile .txt'-atom.txt
12
13 for line in `cat $blogFile`
14 do
15     #curl $line | grep 'rel="alternate"' | grep atom | sed 's/.*href=//' | sed 's/\> //' |
16     sed 's/"//g' >> 'basename $blogFile .txt'-atom.txt
17
18     test=`curl $line | grep 'rel="alternate"' | grep atom | sed 's/.*href=//' | sed 's/
19     /\> //' | sed 's/"//g' | sed 's/ //g'`
20     echo $test >> 'basename $blogFile .txt'-atom.txt
21 done
```

Listing 2: Shell Program for getting atom feed for the blogs

```

1 import feedparser
2 import collections
3 import re
4 import operator
5
6
7 def getwords(html):
8     text = re.compile(r'<[^>]+>').sub(' ', html)
9     words = re.compile(r'[^A-z^a-z]+').split(text)
10    return [word.lower() for word in words if word]
11
12 def checkNextPage(url):
13     f = urllib.urlopen(url)
14     soup = BeautifulSoup(f.read(), from_encoding=f.info().getparam('charset'))
15
16     try:
17         link = soup.find('link', rel='next', href = True)['href']
18     except TypeError:
19         link = None
20     return link
21
22 def getwordcounts(url):
23
24     fd = feedparser.parse(url)
25     wc = collections.defaultdict(int)
26     stopwords = []
27
28     stopWordList = open('stopWordList.txt').readlines()
29     pages = len(fd['entries'])
30
31     for stopWord in stopWordList:
32         stopWord = stopWord.strip()
33         stopwords.append(stopWord)
34
35
36
37     for e in fd.entries:
38         if 'summary' in e:
39             summary = e.summary
40         else:
41             summary = e.description
42
43         words = getwords('%s %s' % (e.title, summary))
44
45         for word in words:
46             if word not in stopwords:
47                 wc[word] += 1
48
49
50     nextLink = checkNextPage( url )
51     while nextLink:
52         nextLink = checkNextPage( nextLink )
53         d = feedparser.parse(nextlink)
54         pages = len(d['entries'])
55         for e in d.entries:
56             if 'summary' in e:
57                 summary = e.summary
58             else:
59                 summary = e.description
60
61         words = getwords('%s %s' % (e.title, summary))
62
63         for word in words:
64             if word not in stopwords:
65                 #print word
66                 wc[word] += 1
67
68     if 'title' not in fd.feed:
69         print 'Invalid url', url
70         return 'bogus data', wc
71

```

```

72     return fd.feed.title , wc
73
74 def main():
75
76     # XXX: break this up into smaller functions, write tests for them
77     apcount = collections.defaultdict(int)
78     wordcounts = {}
79     feedlist = open('blogList-120-atom.txt').readlines()
80     totalWordCount = {}
81
82     for url in feedlist:
83         title , wc = getwordcounts(url)
84         wordcounts[title] = wc
85
86         for word, count in wc.iteritems():
87             if count > 1:
88                 apcount[word] += 1
89
90                 try:
91                     totalWordCount[word] += count
92                 except KeyError:
93                     totalWordCount[word] = count
94
95
96     wordlist = []
97     for w, bc in apcount.iteritems():
98         frac = float(bc)/len(feedlist)
99         #print frac
100        if frac > 0.1 and frac < 0.5:
101            wordlist.append(w)
102
103     countOfWords = []
104
105     for word in wordlist:
106         countOfWords.append((word, totalWordCount[word]))
107
108     countOfWords.sort(key=lambda rating: rating[1], reverse = True )
109
110     for words in countOfWords[0:500]:
111         print words[0] , words[1]
112     out = file('blogdata-120-1500pages.txt', 'w')
113     out.write('Blog')
114
115     for w in countOfWords[0:500]:
116         #print w
117         out.write('\t' + w[0])
118     out.write('\n')
119
120     for blogname, counts in wordcounts.iteritems():
121         blogname = blogname.encode('UTF-8')
122         out.write(blogname)
123
124         for w in countOfWords[0:500]:
125             word = w[0]
126             if w in counts:
127                 out.write('\t%d' % counts[word])
128             else:
129                 out.write('\t0')
130         out.write('\n')
131
132     out.close()
133
134 if __name__ == '__main__':
135     main()

```

Listing 3: Python code for grabbing popular 500 words from 100 atom feeds

```

1  #!/usr/bin/env python
2  import os
3  import sys
4  import urllib
5  import time
6  import feedparser
7
8  from bs4 import BeautifulSoup
9
10 def checkNextPage(url):
11
12     f = urllib.urlopen(url)
13     soup = BeautifulSoup(f.read(), from_encoding=f.info().getparam('charset'))
14
15     try:
16         link = soup.find('link', rel='next', href = True)['href']
17     except TypeError:
18         link = None
19     return link
20
21
22 def main():
23     feedlist = open('blogList-120-atom.txt').readlines()
24
25     for url in feedlist:
26         try:
27             d = feedparser.parse(url)
28
29             title = d['feed']['title']
30
31             count = 1
32             nextLink = checkNextPage( url )
33
34             while nextLink:
35                 nextLink = checkNextPage( nextLink )
36                 count += 1
37
38             print u'|'.join((str(count), title)).encode('utf-8').strip())
39
40
41
42         except KeyError:
43             pass
44
45 if __name__ == '__main__':
46     main()

```

Listing 4: Python code for grabbing number of pages for each blog

```

1 data <- read.csv("noOfPages.txt", stringsAsFactors = F, header = FALSE, sep = "|")
2
3 incdata = data[, 1]
4
5 incdata <- as.numeric(incdata)
6
7 brk <- seq(0, 300, 10)
8
9 png("q1-histogram1.png")
10 hist(incdata, main = "Blogs vs. Number of Pages", breaks=brk, freq = T, xlab="Pages", ylab="
    Blogs")

```

Listing 5: R Script for generating a Histogram

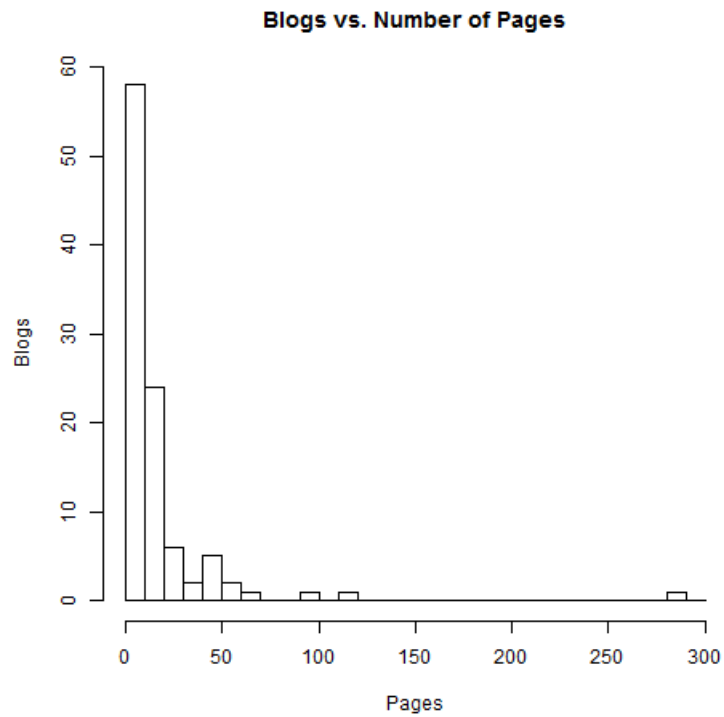


Figure 1: Histogram showing Blogs Vs No of Pages

2 Problem 2

Question

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).

2.1 Answer

1. The script show in Listing 6 uses the *Toby Segaran's clusters.py* code in Listing 11 and produces the Dendrogram shown in Figure 2
2. The printclust function on line 8 in Listing 6 prints the dendrogram. The drawdendrogram function on line 11 saves a JPEG of the dendrogram.
3. The ascii file of the dendrogram is uploaded to github, and the file name is *ascii-dendrogram.txt*.
4. Unfortunately, it is difficult to see, but this dendrogram shows that the blogs calculated to be most like *F-Measure* are split into two clusters, the blogs in the first cluster are *YOUNGEST INDIE* and *Music Liberation*, the blogs in second cluster are *The Devils Music* and *McCrak's Juke*.
5. The blog calculated to be most like *Web Science and Digital Libraries Research Group* are *Koranteng's Toli* and *words of advance for young people*.

```
1 import clusters
2
3
4 blognames, words, data=clusters.readfile('blogdata120-atom-500.txt')
5 clust = clusters.hcluster(data)
6
7 # print ASCII dendrogram
8 clusters.printclust(clust, labels=blognames)
9
10 # save JPEG dendrogram
11 clusters.drawdendrogram(clust, blognames, jpeg='blogclust.jpg')
```

Listing 6: Python code for grabbing number of pages for each blog

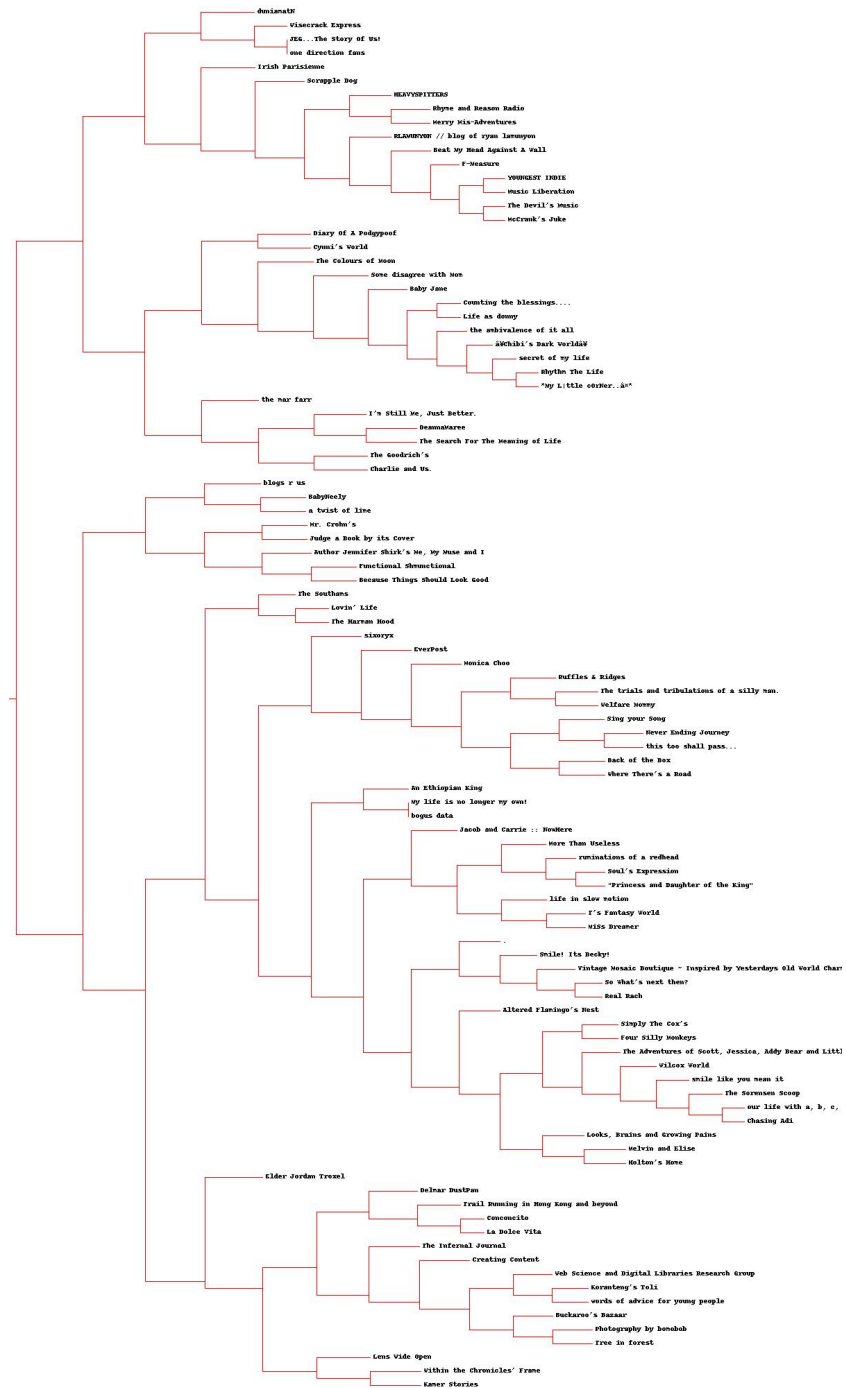


Figure 2: Dendrogram Produced by Program in Listing 6

3 Problem 3

Question

Cluster the blogs using K-Means, using $k=5,10,20$. (see slide 18).
How many iterations were required for each value of k ?

3.1 Answer

The blog clustering is performed by the script shown in Listing 7, which makes use of Toby Segaran's *clusters.py* using the function *kcluster* on lines 8, 12, and 16.

```
1  #!/usr/local/bin/python
2
3  import clusters
4
5  blognames, words, data=clusters.readfile('blogdata120-atom-500.txt')
6
7  print "For k=5"
8  kclust=clusters.kcluster(data, k=5)
9  print
10
11 print "For k=10"
12 kclust=clusters.kcluster(data, k=10)
13 print
14
15 print "For k=20"
16 kclust=clusters.kcluster(data, k=20)
17 print
```

Listing 7: Python script for clustering the blogs using K-means, using $k=5, 10$, and 20

From its output, we see how many iterations each value of k produces.

```
For k=5
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
Iteration 7
Iteration 8
Iteration 9
Iteration 10
Iteration 11

For k=10
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6

For k=20
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
```

Thus, for $k = 5$ we get 12 iterations, for $k = 10$ we get 7 iterations, and for $k = 20$ we get 6 iterations.

4 Problem 4

Question

Use MDS to create a JPEG of the blogs similar to slide 29.
How many iterations were required?

4.1 Answer

1. The blog space is generated using multidimensional scaling from the script `makeMDS.py` shown in Listing 8, which makes use of Toby Segaran's *clusters.py* using the functions *scaledown* on line 7 and *draw2d* on line 9.

```
1  #!/usr/local/bin/python
2
3  import clusters
4
5  blognames, words, data=clusters.readfile('blogdata120-atom-500.txt')
6
7  coords = clusters.scaledown(data)
8
9  clusters.draw2d(coords, blognames, jpeg='blogs2d.jpg')
```

Listing 8: Python script for generating a MDS from the blog data

2. unfortunately the blog space produced does not fit well on a letter-sized page shown in Figure 3.
3. Listing 9 shows the output from running this script, which took 253 iterations on this run.

1	4532.41130905
2	3366.98206519
3	3326.3306574
4	3308.09464169
5	3297.94258112
6	3291.51602489
7	3287.28793801
8	3283.77516396
9	3280.99776391
10	3278.44993199
11	3275.81243841
12	3273.31397104
13	3270.8684595
14	3268.42055835
15	3266.16460163
16	3264.23476083
17	3262.42227912
18	3260.51325928
19	3258.82854152
20	3257.18175939
21	3255.37923243
22	3253.30711541
23	3251.00612794
24	3248.77308262
25	3246.70118667
26	3244.47115104
27	3241.94493708
28	3239.40739168
29	3236.99558481
30	3234.81656727
31	3232.88203014
32	3231.04567171
33	3229.23063872
34	3227.47967223
35	3225.91794645
36	3224.58649843
37	3223.20000436
38	3222.0867435
39	3221.01271989
40	3219.95473094
41	3218.75956428
42	3217.59979619
43	3216.4522233
44	3215.34629148
45	3214.36905778
46	3213.26833436
47	3212.24470346
48	3211.31994368
49	3210.29982355
50	3209.20246315
51	3207.95132503
52	3206.82241072
53	3205.91909053
54	3204.92220274
55	3204.31947561
56	3203.74640843
57	3203.27606811
58	3202.67555472
59	3201.89979232
60	3200.88578203
61	3200.01531056
62	3199.2966446
63	3198.45478403
64	3197.6201472
65	3196.73438036
66	3196.02119032
67	3195.41196943
68	3194.72376496
69	3194.0329727
70	3193.41355901
71	3192.78060731

72	3191.90770762
73	3190.80437498
74	3189.55398117
75	3188.23142174
76	3186.89288539
77	3185.71338432
78	3184.5789406
79	3183.5206291
80	3182.51360921
81	3181.42690675
82	3180.41702115
83	3179.54898356
84	3178.76281891
85	3177.99771823
86	3177.24097774
87	3176.58456935
88	3176.0328065
89	3175.38333701
90	3174.71747303
91	3174.01886829
92	3173.32737481
93	3172.70330603
94	3172.1138461
95	3171.52419342
96	3170.90758333
97	3170.4179592
98	3170.02282149
99	3169.5821675
100	3169.07816383
101	3168.448593
102	3167.75862833
103	3167.05280662
104	3166.42332043
105	3165.9904937
106	3165.73529802
107	3165.54265976
108	3165.33739243
109	3165.07807778
110	3164.91141097
111	3164.62059016
112	3164.17905363
113	3163.65023788
114	3163.14226209
115	3162.70328486
116	3162.36024226
117	3161.99817832
118	3161.70317464
119	3161.37377857
120	3160.9434041
121	3160.44914717
122	3159.86024781
123	3159.29974867
124	3158.58691893
125	3157.78851118
126	3156.84328215
127	3155.8879416
128	3155.00788822
129	3154.23167922
130	3153.44215741
131	3152.76590863
132	3152.13154349
133	3151.49922048
134	3150.91731722
135	3150.33609096
136	3149.74067727
137	3149.15263975
138	3148.52983426
139	3147.83979099
140	3147.25183635
141	3146.72688685
142	3146.10697938
143	3145.55140026

144	3145.038229
145	3144.50839712
146	3143.94311783
147	3143.40783945
148	3142.95445563
149	3142.52189469
150	3141.99941939
151	3141.35987167
152	3140.6610068
153	3140.01716194
154	3139.47786778
155	3138.91063005
156	3138.29379482
157	3137.61144599
158	3136.89591297
159	3136.14533364
160	3135.38865056
161	3134.64653422
162	3133.91168296
163	3133.23112402
164	3132.52038824
165	3131.88007921
166	3131.30351041
167	3130.66122515
168	3129.90421203
169	3129.1405021
170	3128.32782048
171	3127.47536739
172	3126.54560283
173	3125.62757639
174	3124.67176012
175	3123.72838126
176	3122.76127432
177	3121.76341232
178	3120.77006527
179	3119.75759981
180	3118.76498364
181	3117.77794841
182	3116.78570912
183	3115.80880423
184	3114.85906684
185	3113.95665663
186	3113.0751775
187	3112.2783084
188	3111.51872253
189	3110.81819246
190	3110.12121549
191	3109.49131097
192	3108.87630713
193	3108.24307894
194	3107.61218582
195	3106.9346598
196	3106.17407566
197	3105.30253203
198	3104.46828824
199	3103.70960692
200	3103.04853569
201	3102.374827
202	3101.76925634
203	3101.10225207
204	3100.39181475
205	3099.80313972
206	3099.32543062
207	3098.83756084
208	3098.39524227
209	3097.89054287
210	3097.45857954
211	3097.16391173
212	3096.85125512
213	3096.55032349
214	3096.25660576
215	3096.00089054


```
216 3095.71194597
217 3095.44859148
218 3095.20946085
219 3095.07182068
220 3095.06452066
221 3095.04654607
222 3094.9754195
223 3094.85516015
224 3094.79234814
225 3094.67410977
226 3094.52095848
227 3094.33078357
228 3094.08903464
229 3093.79059454
230 3093.51468176
231 3093.20791956
232 3092.91243999
233 3092.65113648
234 3092.34912442
235 3092.05447449
236 3091.80452138
237 3091.55424259
238 3091.27376913
239 3090.97576404
240 3090.65628056
241 3090.30346956
242 3089.97015589
243 3089.57549888
244 3089.19007073
245 3088.82871046
246 3088.44864717
247 3088.11602282
248 3087.91481431
249 3087.80681744
250 3087.74255397
251 3087.66780485
252 3087.65822646
253 3087.66845007
```

Listing 9: Output from script *makeMDS.py*



Figure 3: Blog space produced by the *makeMDS.py* script

5 Problem 5

Question

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 500 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 500 terms and instead come up with TFIDF scores for all the terms and then choose the top 500 from that list, but I'm trying to limit the amount of work necessary.

5.1 Answer

1. I used the python which is used to get the matrix for the questions 1 but added few lines of code to calculate the TFIDF value for each word
2. IDF value is calculated at line 149 in Listing 10
3. The TF value for each word with respect to each blog is calculated at line 166 which is used to calculate TFIDF value at line 167 in Listing 10
4. The matrix with blog name and the words along with their TFIDF value is calculated by The program Listed in Listing 10. The output is uploaded in the github as *bloglist-500-matrix.txt*.
5. I used the same program which I used in Problem 2 to generate the dendrogram shown in Figure 4
6. Unfortunately, it is difficult to see, but this dendrogram shows that the blogs calculated to be most like *F-Measure* are split into two clusters, the blogs in the first cluster is *Music Liberation* contrast to the out from the dendrogram from the second question the blog */YOUNGEST INDIE*
7. The blogs in second cluster are *YOUNGEST INDIE* and *McCrak's Juke* where as the the blog *The Devil's Music* is in the second cluster instead of *YOUNGEST INDIE* in the dendrogram produced for second question .
8. Unlike in the dendrogram produced for the second question The blog calculated to be most like *Web Science and Digital Libraries Research Group* is *Buckaroo's Bazar*
9. From my observation of both the dendrograms I found that the hierarchy of the clusters are changing but somehow that clusters are around the same blogs with different hierarchy.
10. The accuracy of matching the most alike blogs is changing.

5.2 Output for 5th problem

```

1 import feedparser
2 import collections
3 import re
4 import operator
5 import math
6
7 INPUT_FILE = "blogList-120-atom.txt"
8 OUTPUT_FILE = "blogList-120-500-matrix.txt"
9
10 def getwords(html):
11     text = re.compile(r'<[^>]+>').sub('', html)
12     words = re.compile(r'^A-Za-z+').split(text)
13
14     return [word.lower() for word in words if word]
15
16 def getwordcounts(url):
17
18     fd = feedparser.parse(url)
19     wc = collections.defaultdict(int)
20     stopwords = []
21
22     stopWordList = open('stopWordList.txt').readlines()
23     pages = len(fd['entries'])
24
25     for stopWord in stopWordList:
26         stopWord = stopWord.strip()
27         stopwords.append(stopWord)
28
29     for e in fd.entries:
30         if 'summary' in e:
31             summary = e.summary
32         else:
33             summary = e.description
34         words = getwords('%s %s' % (e.title, summary))
35
36         for word in words:
37             if word not in stopwords:
38                 wc[word] += 1
39
40     if pages == 500:
41         next_link = url + "?start-index=501"
42         d = feedparser.parse(next_link)
43         pages = len(d['entries'])
44         for e in d.entries:
45             if 'summary' in e:
46                 summary = e.summary
47             else:
48                 summary = e.description
49
50             words = getwords('%s %s' % (e.title, summary))
51
52             for word in words:
53                 if word not in stopwords:
54                     #print word
55
56                     wc[word] += 1
57
58     if pages == 500:
59         next_link = url + "?start-index=1001"
60         for e in d.entries:
61             if 'summary' in e:
62                 summary = e.summary
63             else:
64                 summary = e.description
65
66             words = getwords('%s %s' % (e.title, summary))
67
68             for word in words:
69                 if word not in stopwords:
70                     #print word
71

```

```

72         wc[word] += 1
73     if pages == 500:
74         next_link = url + "?start-index=1501"
75         for e in d.entries:
76             if 'summary' in e:
77                 summary = e.summary
78             else:
79                 summary = e.description
80
81         words = getwords('%s %s' % (e.title, summary))
82
83         for word in words:
84             if word not in stopwords:
85                 wc[word] += 1
86
87 if 'title' not in fd.feed:
88     print 'Invalid url', url
89     return 'bogus data', wc
90
91 return fd.feed.title, wc
92
93 def main():
94
95     # XXX: break this up into smaller functions, write tests for them
96
97     apcount = collections.defaultdict(int)
98     wordcounts = {}
99     feedlist = open( INPUT_FILE ).readlines()
100     totalWordCount = {}
101
102     for url in feedlist:
103         title, wc = getwordcounts(url)
104         wordcounts[title] = wc
105
106         for word, count in wc.iteritems():
107             if count > 1:
108                 apcount[word] += 1
109
110                 try:
111                     totalWordCount[word] += count
112                 except KeyError:
113                     totalWordCount[word] = count
114
115     wordlist = []
116
117
118     for w, bc in apcount.iteritems():
119         frac = float(bc)/len(feedlist)
120         #print frac
121         if frac > 0.1 and frac < 0.5:
122             wordlist.append(w)
123
124     countOfWords = []
125
126     for word in wordlist:
127         countOfWords.append((word, totalWordCount[word]))
128
129     countOfWords.sort(key=lambda rating: rating[1], reverse = True )
130
131     countOfWords = countOfWords[0:500]
132
133     out = file(OUTPUT_FILE, 'w')
134     out.write('Blog')
135
136     idfWordCount = {}
137
138     for w in countOfWords:
139         word = w[0]
140         noOfBlogs = 0
141         for blogname, counts in wordcounts.iteritems():
142
143             if word in counts:

```

```

144         noOfBlogs += 1
145         #print noOfBlogs
146         idf = math.log( 100.0 / noOfBlogs , 2 )
147
148
149         idfWordCount[word] = idf
150
151
152     for w in countOfWords:
153         #print w
154         out.write( '\t' + w[0] )
155
156     out.write( '\n' )
157
158     for blogname, counts in wordcounts.iteritems():
159         blogname = blogname.encode( 'UTF-8' )
160         out.write( blogname )
161
162         for w in countOfWords:
163             word = w[0]
164             occurrence = w[1]
165
166             tf = float( counts[word] ) / occurrence
167             tfidf = tf * idfWordCount[word]
168
169             # d is integer, f is float
170             out.write( '\t%f' % tfidf )
171
172         out.write( '\n' )
173
174     out.close()
175
176 if __name__ == '__main__':
177     main()

```

Listing 10: Python code for grabbing popular 500 words from 100 atom feeds and their TFIDF values

```

1  #!/usr/bin/python
2  # -*- coding: utf-8 -*-
3  from PIL import Image, ImageDraw
4  from math import sqrt
5  import random
6
7  def readfile(filename):
8      lines = [line for line in file(filename)]
9
10     # First line is the column titles
11     colnames = lines[0].strip().split('\t')[1:]
12     rownames = []
13     data = []
14     for line in lines[1:]:
15         p = line.strip().split('\t')
16         # First column in each row is the rowname
17         rownames.append(p[0])
18         # The data for this row is the remainder of the row
19         data.append([float(x) for x in p[1:]])
20     return (rownames, colnames, data)
21
22
23 def pearson(v1, v2):
24     # Simple sums
25     sum1 = sum(v1)
26     sum2 = sum(v2)
27
28     # Sums of the squares
29     sum1Sq = sum([pow(v, 2) for v in v1])
30     sum2Sq = sum([pow(v, 2) for v in v2])
31
32     # Sum of the products
33     pSum = sum([v1[i] * v2[i] for i in range(len(v1))])
34
35     # Calculate r (Pearson score)
36     num = pSum - sum1 * sum2 / len(v1)
37     den = sqrt((sum1Sq - pow(sum1, 2) / len(v1)) * (sum2Sq - pow(sum2, 2)
38         / len(v1)))
39     if den == 0:
40         return 0
41
42     return 1.0 - num / den
43
44
45 class bicluster:
46
47     def __init__(
48         self,
49         vec,
50         left=None,
51         right=None,
52         distance=0.0,
53         id=None,
54     ):
55         self.left = left
56         self.right = right
57         self.vec = vec
58         self.id = id
59         self.distance = distance
60
61
62 def hcluster(rows, distance=pearson):
63     distances = {}
64     currentclustid = -1
65
66     # Clusters are initially just the rows
67     clust = [bicluster(rows[i], id=i) for i in range(len(rows))]
68
69     while len(clust) > 1:
70         lowestpair = (0, 1)
71         closest = distance(clust[0].vec, clust[1].vec)

```

```

72
73     # loop through every pair looking for the smallest distance
74     for i in range(len(clust)):
75         for j in range(i + 1, len(clust)):
76             # distances is the cache of distance calculations
77             if (clust[i].id, clust[j].id) not in distances:
78                 distances[(clust[i].id, clust[j].id)] = \
79                     distance(clust[i].vec, clust[j].vec)
80
81                 d = distances[(clust[i].id, clust[j].id)]
82
83                 if d < closest:
84                     closest = d
85                     lowestpair = (i, j)
86
87     # calculate the average of the two clusters
88     mergevec = [(clust[lowestpair[0]].vec[i] + clust[lowestpair[1]].vec[i])
89                 / 2.0 for i in range(len(clust[0].vec))]
90
91     # create the new cluster
92     newcluster = bicluster(mergevec, left=clust[lowestpair[0]],
93                           right=clust[lowestpair[1]], distance=closest,
94                           id=currentclustid)
95
96     # cluster ids that weren't in the original set are negative
97     currentclustid -= 1
98     del clust[lowestpair[1]]
99     del clust[lowestpair[0]]
100    clust.append(newcluster)
101
102    return clust[0]
103
104
105 def printclust(clust, labels=None, n=0):
106     # indent to make a hierarchy layout
107     for i in range(n):
108         print ' ',
109     if clust.id < 0:
110         # negative id means that this is branch
111         print '-'
112     else:
113         # positive id means that this is an endpoint
114         if labels == None:
115             print clust.id
116         else:
117             print labels[clust.id]
118
119     # now print the right and left branches
120     if clust.left != None:
121         printclust(clust.left, labels=labels, n=n + 1)
122     if clust.right != None:
123         printclust(clust.right, labels=labels, n=n + 1)
124
125
126 def getheight(clust):
127     # Is this an endpoint? Then the height is just 1
128     if clust.left == None and clust.right == None:
129         return 1
130
131     # Otherwise the height is the same of the heights of
132     # each branch
133     return getheight(clust.left) + getheight(clust.right)
134
135
136 def getdepth(clust):
137     # The distance of an endpoint is 0.0
138     if clust.left == None and clust.right == None:
139         return 0
140
141     # The distance of a branch is the greater of its two sides
142     # plus its own distance
143     return max(getdepth(clust.left), getdepth(clust.right)) + clust.distance

```



```

144
145
146 def drawdendrogram(clust, labels, jpeg='clusters.jpg'):
147     # height and width
148     h = getheight(clust) * 20
149     w = 1200
150     depth = getdepth(clust)
151
152     # width is fixed, so scale distances accordingly
153     scaling = float(w - 150) / depth
154
155     # Create a new image with a white background
156     img = Image.new('RGB', (w, h), (255, 255, 255))
157     draw = ImageDraw.Draw(img)
158
159     draw.line((0, h / 2, 10, h / 2), fill=(255, 0, 0))
160
161     # Draw the first node
162     drawnode(
163         draw,
164         clust,
165         10,
166         h / 2,
167         scaling,
168         labels,
169     )
170     img.save(jpeg, 'JPEG')
171
172
173 def drawnode(
174     draw,
175     clust,
176     x,
177     y,
178     scaling,
179     labels,
180 ):
181     if clust.id < 0:
182         h1 = getheight(clust.left) * 20
183         h2 = getheight(clust.right) * 20
184         top = y - (h1 + h2) / 2
185         bottom = y + (h1 + h2) / 2
186         # Line length
187         ll = clust.distance * scaling
188         # Vertical line from this cluster to children
189         draw.line((x, top + h1 / 2, x, bottom - h2 / 2), fill=(255, 0, 0))
190
191         # Horizontal line to left item
192         draw.line((x, top + h1 / 2, x + ll, top + h1 / 2), fill=(255, 0, 0))
193
194         # Horizontal line to right item
195         draw.line((x, bottom - h2 / 2, x + ll, bottom - h2 / 2), fill=(255, 0,
196             0))
197
198         # Call the function to draw the left and right nodes
199         drawnode(
200             draw,
201             clust.left,
202             x + ll,
203             top + h1 / 2,
204             scaling,
205             labels,
206         )
207         drawnode(
208             draw,
209             clust.right,
210             x + ll,
211             bottom - h2 / 2,
212             scaling,
213             labels,
214         )
215     else:

```

```

216     # If this is an endpoint, draw the item label
217     draw.text((x + 5, y - 7), labels[clust.id], (0, 0, 0))
218
219
220 def rotatematrix(data):
221     newdata = []
222     for i in range(len(data[0])):
223         newrow = [data[j][i] for j in range(len(data))]
224         newdata.append(newrow)
225     return newdata
226
227
228 def kcluster(rows, distance=pearson, k=4):
229     # Determine the minimum and maximum values for each point
230     ranges = [(min([row[i] for row in rows]), max([row[i] for row in rows]))
231               for i in range(len(rows[0]))]
232
233     # Create k randomly placed centroids
234     clusters = [[random.random() * (ranges[i][1] - ranges[i][0]) + ranges[i][0]
235                 for i in range(len(rows[0]))] for j in range(k)]
236
237     lastmatches = None
238     for t in range(100):
239         print 'Iteration %d' % t
240         bestmatches = [[] for i in range(k)]
241
242         # Find which centroid is the closest for each row
243         for j in range(len(rows)):
244             row = rows[j]
245             bestmatch = 0
246             for i in range(k):
247                 d = distance(clusters[i], row)
248                 if d < distance(clusters[bestmatch], row):
249                     bestmatch = i
250             bestmatches[bestmatch].append(j)
251
252         # If the results are the same as last time, this is complete
253         if bestmatches == lastmatches:
254             break
255         lastmatches = bestmatches
256
257         # Move the centroids to the average of their members
258         for i in range(k):
259             avgs = [0.0] * len(rows[0])
260             if len(bestmatches[i]) > 0:
261                 for rowid in bestmatches[i]:
262                     for m in range(len(rows[rowid])):
263                         avgs[m] += rows[rowid][m]
264             for j in range(len(avgs)):
265                 avgs[j] /= len(bestmatches[i])
266             clusters[i] = avgs
267
268     return bestmatches
269
270
271 def tanamoto(v1, v2):
272     (c1, c2, shr) = (0, 0, 0)
273
274     for i in range(len(v1)):
275         if v1[i] != 0: # in v1
276             c1 += 1
277         if v2[i] != 0: # in v2
278             c2 += 1
279         if v1[i] != 0 and v2[i] != 0: # in both
280             shr += 1
281
282     return 1.0 - float(shr) / (c1 + c2 - shr)
283
284
285 def scaledown(data, distance=pearson, rate=0.01):
286     n = len(data)
287

```

```

288 # The real distances between every pair of items
289     realdist = [[distance(data[i], data[j]) for j in range(n)] for i in
290                 range(0, n)]
291
292 # Randomly initialize the starting points of the locations in 2D
293     loc = [[random.random(), random.random()] for i in range(n)]
294     fakedist = [[0.0 for j in range(n)] for i in range(n)]
295
296     lasterror = None
297     for m in range(0, 1000):
298         # Find projected distances
299         for i in range(n):
300             for j in range(n):
301                 fakedist[i][j] = sqrt(sum([pow(loc[i][x] - loc[j][x], 2)
302                                             for x in range(len(loc[i]))]))
303
304         # Move points
305         grad = [[0.0, 0.0] for i in range(n)]
306
307         totalerror = 0
308         for k in range(n):
309             for j in range(n):
310                 if j == k:
311                     continue
312                 # The error is percent difference between the distances
313                 errorterm = (fakedist[j][k] - realdist[j][k]) / realdist[j][k]
314
315                 # Each point needs to be moved away from or towards the other
316                 # point in proportion to how much error it has
317                 grad[k][0] += (loc[k][0] - loc[j][0]) / fakedist[j][k] \
318                     * errorterm
319                 grad[k][1] += (loc[k][1] - loc[j][1]) / fakedist[j][k] \
320                     * errorterm
321
322                 # Keep track of the total error
323                 totalerror += abs(errorterm)
324             print totalerror
325
326         # If the answer got worse by moving the points, we are done
327         if lasterror and lasterror < totalerror:
328             break
329         lasterror = totalerror
330
331         # Move each of the points by the learning rate times the gradient
332         for k in range(n):
333             loc[k][0] -= rate * grad[k][0]
334             loc[k][1] -= rate * grad[k][1]
335
336     return loc
337
338
339 def draw2d(data, labels, jpeg='mds2d.jpg'):
340     img = Image.new('RGB', (2000, 2000), (255, 255, 255))
341     draw = ImageDraw.Draw(img)
342     for i in range(len(data)):
343         x = (data[i][0] + 0.5) * 1000
344         y = (data[i][1] + 0.5) * 1000
345         draw.text((x, y), labels[i], (0, 0, 0))
346     img.save(jpeg, 'JPEG')

```

Listing 11: Segaran's *clusters.py*

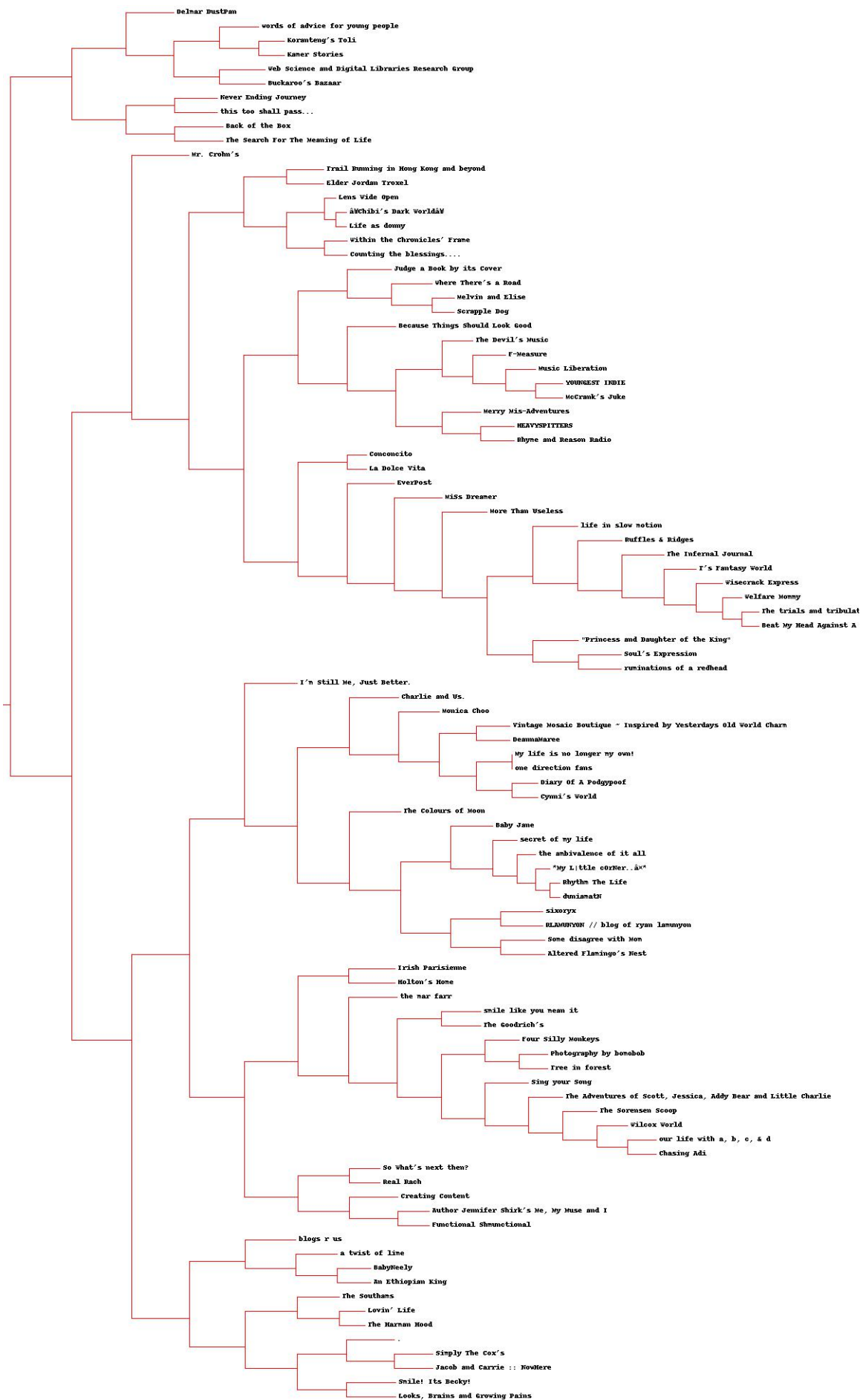


Figure 4: Dendrogram produced based on word's TFIDF values script

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