

# CS595 Intro to Web Science, Assignment #9

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## Question 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.

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.

## Answer to Question 1

Use the book here [1]

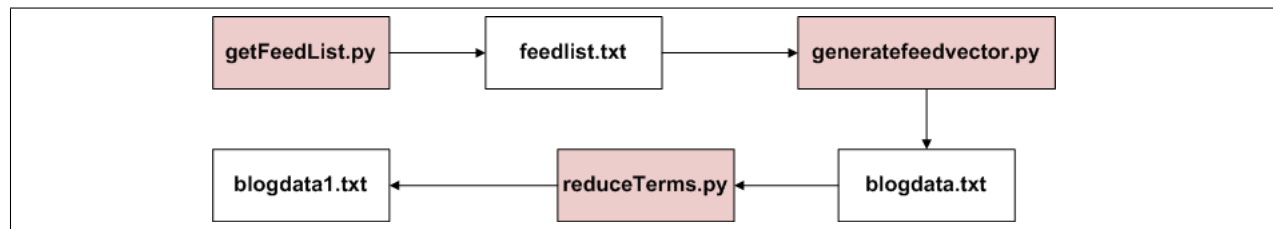


Figure 1: Creation of the Blog-Term Matrix

## Question 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.)

## Answer to Question 2

Listing 1: generateImages.py

```
import sys

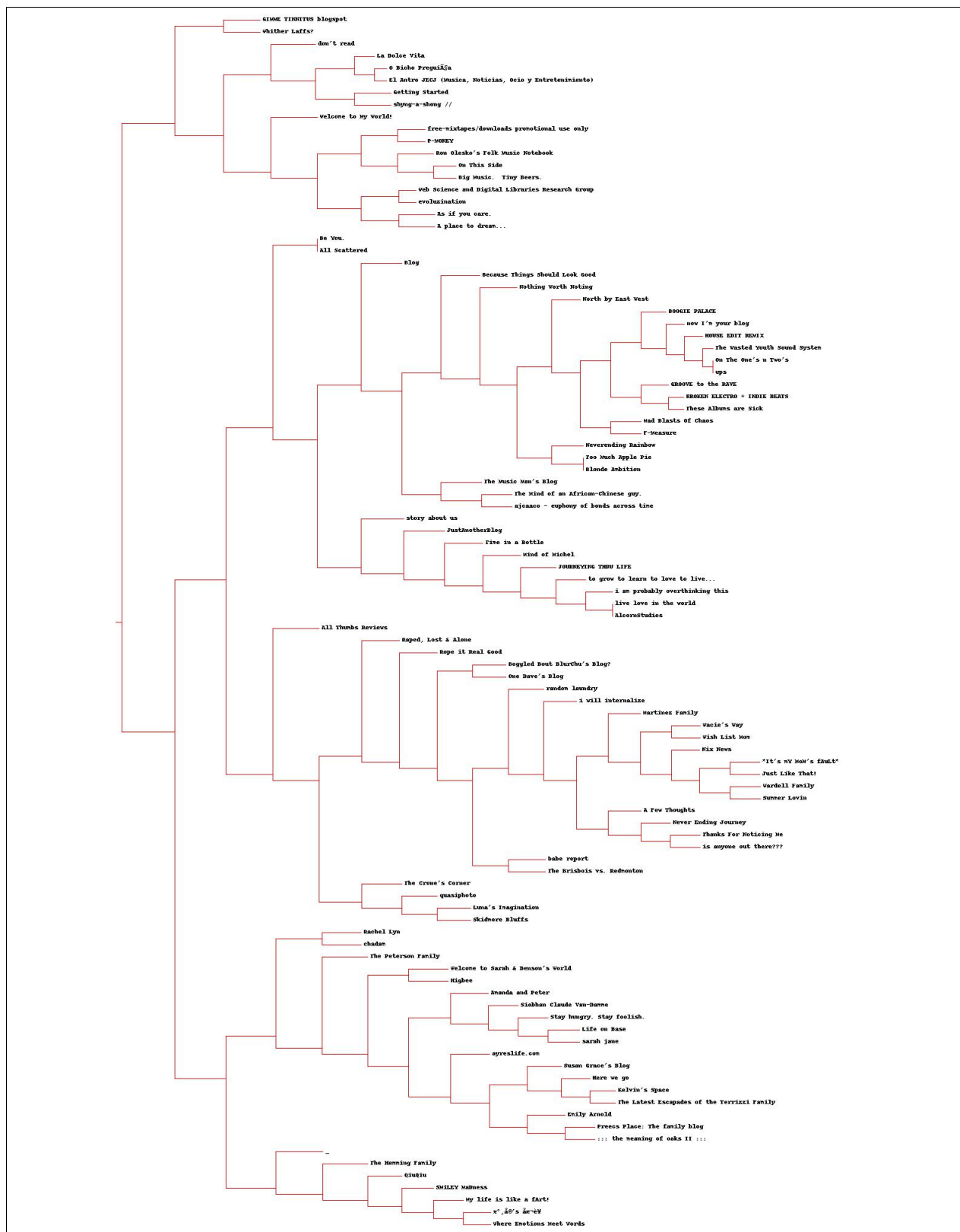
sys.path.insert(0, '/Users/vneblitt/Documents/cs595-f13/assignment09/library ')

import clusters

# Create clusters
blognames, words, data=clusters.readfile('/Users/vneblitt/Documents/cs595-f13/assignment09/q01
/blogdata1.txt ')
clust=clusters.hcluster(data)

# Create ASCII dendrogram
clusters.printclust(clust, labels=blognames)

# Create Nicer dendrogram with PIL
clusters.drawdendrogram(clust, blognames, jpeg='blogclust.jpg')
```



## Question 3

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

## Answer to Question 3

Listing 2: getKMeans.py

```
import sys

sys.path.insert(0, '/Users/vneblitt/Documents/cs595-f13/assignment09/library ')

import clusters

blognames, words, data=clusters.readfile('/Users/vneblitt/Documents/cs595-f13/assignment09/q01
/blogdata1.txt ')

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

print 'k=10'
kclust=clusters.kcluster(data,k=10)

print 'k=20'
kclust=clusters.kcluster(data,k=20)
```

| Value of k | No. of Iterations |
|------------|-------------------|
| 5          | 6                 |
| 10         | 7                 |
| 20         | 100               |

Table 1: K-Means Iterations

## Question 4

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

### Answer to Question 4

I had to reduce the MDS image down to 20% in order to include it in this report, but that makes it too hard to read. The full one is located in GitHub at <https://github.com/vneblitt/cs595-f13/blob/master/assignment09/q04/blogs2d.jpg>

When I first ran the code, I encounter a divide-by-zero error that would not allow the code to fully execute. I had to remove the following four blogs from the blog-term matrix since their entire row in the matrix contained zeros.

- Blonde Ambition
- ups
- Alcorn Studios
- All Scattered

I ran into this problem earlier when I had only managed to get 60 terms from Q1 and I thought I would not encounter this problem since I included the summaries and descriptions had achieved 3200+ terms and then reduced the terms down to the 500 most frequent.

Listing 3: getMDS.py

```
import sys

sys.path.insert(0, '/Users/vneblitt/Documents/cs595-f13/assignment09/library')

import clusters

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

coords = clusters.scaledown(data)
clusters.draw2d(coords, blognames, jpeg='blogs2d.jpg')
```

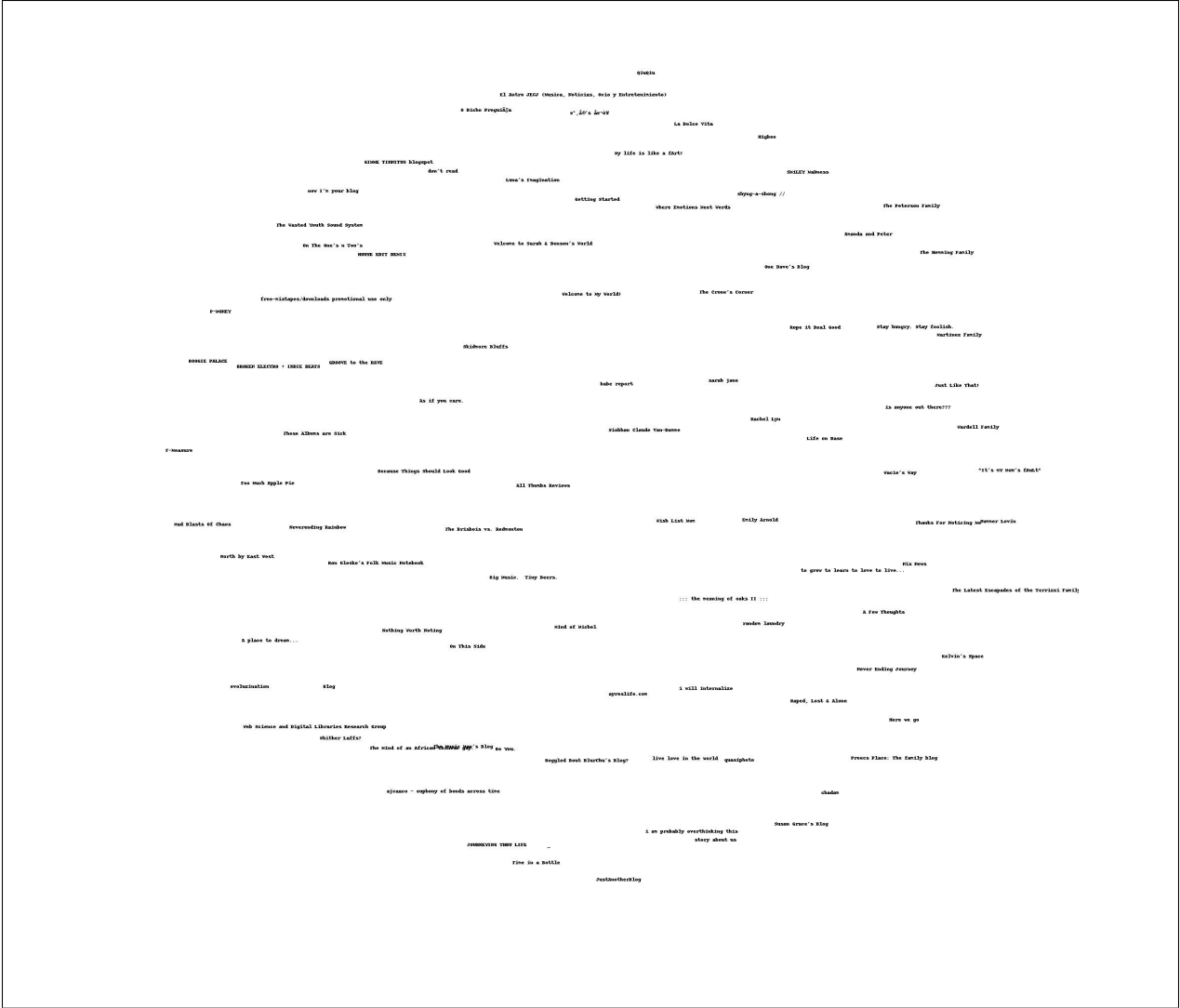


Figure 3: Multi-Dimenional Scaling



## Question 5 - Extra Credit (5 points)

Re-run Q2, 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 file to GitHub.

Compare and contrast the resulting dendrogram with the dendrogram from Q2.

Note: Ideally you would not reuse the same 500 terms and instead come up with TFIDF scores for all the terms and choose the top 500 from that list, but I am trying to limit the amount of work necessary.

### Answer to Question 5

Not attempted.

## A Full Version of getFeedList.py

Listing 4: getFeedList.py

```
from bs4 import BeautifulSoup
from urllib2 import HTTPError
import urllib2
import time

f = open('feedlist.txt', 'w')

# These two are automatically in the list
firstfeed = 'http://f-measure.blogspot.com/feeds/posts/default'
secondfeed = 'http://ws-dl.blogspot.com/feeds/posts/default'

feedlist = []
feedlist.append(firstfeed)
feedlist.append(secondfeed)

while len(feedlist) <= 100:
    try:
        nextblog = 'http://www.blogger.com/next-blog?navBar=true&blogID=3471633091411211117'

        html = urllib2.urlopen(nextblog).read()

        soup = BeautifulSoup(html)

        atomfeedurl = soup.find_all('link', attrs = {'type' : 'application/atom+xml'})[0].
            attrs['href']

    except HTTPError as h:
        pass

    except IndexError as i:
        pass

    if atomfeedurl:

        if atomfeedurl not in feedlist:

            print atomfeedurl
            feedlist.append(atomfeedurl)

        time.sleep(1)

for item in feedlist:
    f.write(item + '?max-results=200' + '\n')

f.close()
```

## B Full Version of generatefeedvector.py

Listing 5: generatefeedvector.py

```
import feedparser
import re

# Returns title and dictionary of word counts for an RSS feed
def getwordcounts(url):
    # Parse the feed
    d=feedparser.parse(url)
    wc={}

    return d.title, wc
```

```

# Loop over all the entries
for e in d.entries:
    if 'summary' in e: summary=e.summary
    else: summary=e.description

    # Extract a list of words
    words=getwords(e.title+' '+summary)
    for word in words:
        wc.setdefault(word,0)
        wc[word]+=1
return d.feed.title,wc

def getwords(html):
    # 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!='']

apcount={}
wordcounts={}
feedlist=[line for line in file('feedlist.txt')]
for feedurl in feedlist:
    try:
        title,wc=getwordcounts(feedurl)
        wordcounts[title]=wc
        for word,count in wc.items():
            apcount.setdefault(word,0)
            if count>1:
                apcount[word]+=1
    except:
        print 'Failed to parse feed %s' % feedurl

wordlist=[]
for w,bc in apcount.items():
    frac=float(bc)/len(feedlist)
    if frac>0.1 and frac<0.5:
        wordlist.append(w)

out=file('blogdata.txt','w')
out.write('Blog')
for word in wordlist: out.write('\t%s' % word)
out.write('\n')
for blog,wc in wordcounts.items():
    print blog
    blog = blog.encode('UTF-8')
    out.write(blog)
    for word in wordlist:
        if word in wc: out.write('\t%d' % wc[word])
        else: out.write('\t0')
    out.write('\n')

```

## C Full Version of reduceTerms.py

Listing 6: reduceTerms.py

```

import sys

sys.path.insert(0, '/Users/vneblitt/Documents/cs595-f13/assignment09/library')

```

```

import clusters

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

# row
# print 'blog name: ' + blognames[0]

# column
# print 'term: ' + words[3]

# junction
# print 'value: ' + str(data[0][3])

# print 'number of blogs: ' + str(len(blognames))
# print 'number of words: ' + str(len(words))

wordsums={}

for j in range(len(words)):
    sum = 0
    for i in range(len(blognames)):
        sum = sum + data[i][j]
    # print words[j] + ' ' + str(sum)
    wordsums[j] = int(sum)

# print wordsums

a = sorted(wordsums, key=wordsums.get, reverse=True)[0:500]

# print (len(a))

# print a

print 'Blog',

for m in range(len(words)):
    if m in a:
        print '\t' + words[m],
print

for i in range(len(blognames)):
    print blognames[i],

    for j in range(len(words)):
        if j in a:
            print '\t' + str(int(data[i][j])),

    print

```

## D Full Version of clusters.py

This code is used for Q2, Q3, and Q4.

Listing 7: clusters.py

```

from PIL import Image, ImageDraw

def readfile(filename):
    lines=[line for line in file(filename)]

    # First line is the column titles
    colnames=lines[0].strip().split('\t')[1:]
    rownames=[]
    data=[]
    for line in lines[1:]:
        p=line.strip().split('\t')

```

```

    # First column in each row is the rowname
    rownames.append(p[0])
    # The data for this row is the remainder of the row
    data.append([float(x) for x in p[1:]])
    return rownames, colnames, data

from math import sqrt

def pearson(v1,v2):
    # Simple sums
    sum1=sum(v1)
    sum2=sum(v2)

    # Sums of the squares
    sum1Sq=sum([pow(v,2) for v in v1])
    sum2Sq=sum([pow(v,2) for v in v2])

    # Sum of the products
    pSum=sum([v1[i]*v2[i] for i in range(len(v1))])

    # Calculate r (Pearson score)
    num=pSum-(sum1*sum2/len(v1))
    den=sqrt((sum1Sq-pow(sum1,2)/len(v1))*(sum2Sq-pow(sum2,2)/len(v1)))
    if den==0: return 0

    return 1.0-num/den

class bicluster:
    def __init__(self, vec, left=None, right=None, distance=0.0, id=None):
        self.left=left
        self.right=right
        self.vec=vec
        self.id=id
        self.distance=distance

def hcluster(rows, distance=pearson):
    distances={}
    currentclustid=-1

    # Clusters are initially just the rows
    clust=[bicluster(rows[i], id=i) for i in range(len(rows))]

    while len(clust)>1:
        lowestpair=(0,1)
        closest=distance(clust[0].vec, clust[1].vec)

        # loop through every pair looking for the smallest distance
        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:
                    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 ' ',
    if clust.id<0:
        # negative id means that this is branch
        print '-'
    else:
        # positive id means that this is an endpoint
        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)

def getheight(clust):
    # Is this an endpoint? Then the height is just 1
    if clust.left==None and clust.right==None: return 1

    # Otherwise the height is the same of the heights of
    # each branch
    return getheight(clust.left)+getheight(clust.right)

def getdepth(clust):
    # The distance of an endpoint is 0.0
    if clust.left==None and clust.right==None: return 0

    # The distance of a branch is the greater of its two sides
    # plus its own distance
    return max(getdepth(clust.left), getdepth(clust.right))+clust.distance

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, labels)
    img.save(jpeg, 'JPEG')

def drawnode(draw, clust, x, y, scaling, labels):
    if clust.id<0:
        h1=getheight(clust.left)*20
        h2=getheight(clust.right)*20
        top=y-(h1+h2)/2
        bottom=y+(h1+h2)/2
        # Line length

```

```

    ll=clust.distance*scaling
    # Vertical line from this cluster to children
    draw.line((x,top+h1/2,x,bottom-h2/2),fill=(255,0,0))

    # Horizontal line to left item
    draw.line((x,top+h1/2,x+ll,top+h1/2),fill=(255,0,0))

    # Horizontal line to right item
    draw.line((x,bottom-h2/2,x+ll,bottom-h2/2),fill=(255,0,0))

    # Call the function to draw the left and right nodes
    drawnode(draw,clust.left,x+ll,top+h1/2,scaling,labels)
    drawnode(draw,clust.right,x+ll,bottom-h2/2,scaling,labels)
else:
    # If this is an endpoint, draw the item label
    draw.text((x+5,y-7),labels[clust.id],(0,0,0))

def rotatematrix(data):
    newdata=[]
    for i in range(len(data[0])):
        newrow=[data[j][i] for j in range(len(data))]
        newdata.append(newrow)
    return newdata

import random

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]
    for i in range(len(rows[0]))] for j in range(k)]

    lastmatches=None
    for t in range(100):
        print 'Iteration %d' % t
        bestmatches=[] for i in range(k)

        # Find which centroid is the closest for each row
        for j in range(len(rows)):
            row=rows[j]
            bestmatch=0
            for i in range(k):
                d=distance(clusters[i],row)
                if d<distance(clusters[bestmatch],row): bestmatch=i
            bestmatches[bestmatch].append(j)

        # If the results are the same as last time, this is complete
        if bestmatches==lastmatches: break
        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

def tanamoto(v1,v2):
    c1,c2,shr=0,0,0

```

```

for i in range(len(v1)):
    if v1[i]!=0: c1+=1 # in v1
    if v2[i]!=0: c2+=1 # in v2
    if v1[i]!=0 and v2[i]!=0: shr+=1 # in both

return 1.0-(float(shr)/(c1+c2-shr))

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)]

    # Randomly initialize the starting points of the locations in 2D
    loc=[[random.random(),random.random()] for i in range(n)]
    fakedist=[[0.0 for j in range(n)] for i in range(n)]

    lasterror=None
    for m in range(0,1000):
        # Find projected distances
        for i in range(n):
            for j in range(n):
                fakedist[i][j]=sqrt(sum([pow(loc[i][x]-loc[j][x],2)
                                           for x in range(len(loc[i]))]))

        # Move points
        grad=[[0.0,0.0] for i in range(n)]

        totalerror=0
        for k in range(n):
            for j in range(n):
                if j==k: continue
                # The error is percent difference between the distances
                errorterm=(fakedist[j][k]-realdist[j][k])/realdist[j][k]

                # Each point needs to be moved away from or towards the other
                # point in proportion to how much error it has
                grad[k][0]+=((loc[k][0]-loc[j][0])/fakedist[j][k])*errorterm
                grad[k][1]+=((loc[k][1]-loc[j][1])/fakedist[j][k])*errorterm

                # Keep track of the total error
                totalerror+=abs(errorterm)
        print totalerror

        # If the answer got worse by moving the points, we are done
        if lasterror and lasterror<totalerror: break
        lasterror=totalerror

        # Move each of the points by the learning rate times the gradient
        for k in range(n):
            loc[k][0]-=rate*grad[k][0]
            loc[k][1]-=rate*grad[k][1]

    return loc

def draw2d(data,labels,jpeg='mds2d.jpg'):
    img=Image.new('RGB',(2000,2000),(255,255,255))
    draw=ImageDraw.Draw(img)
    for i in range(len(data)):
        x=(data[i][0]+0.5)*1000
        y=(data[i][1]+0.5)*1000
        draw.text((x,y),labels[i],[0,0,0])
    img.save(jpeg,'JPEG')

```



## E Output from getKMeans.py

Listing 8: kmeansoutput.txt

```
k=5
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
k=10
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
k=20
Iteration 0
Iteration 1
Iteration 2
Iteration 3
Iteration 4
Iteration 5
Iteration 6
Iteration 7
Iteration 8
Iteration 9
Iteration 10
Iteration 11
Iteration 12
Iteration 13
Iteration 14
Iteration 15
Iteration 16
Iteration 17
Iteration 18
Iteration 19
Iteration 20
Iteration 21
Iteration 22
Iteration 23
Iteration 24
Iteration 25
Iteration 26
Iteration 27
Iteration 28
Iteration 29
Iteration 30
Iteration 31
Iteration 32
Iteration 33
Iteration 34
Iteration 35
Iteration 36
Iteration 37
Iteration 38
Iteration 39
Iteration 40
Iteration 41
Iteration 42
Iteration 43
Iteration 44
Iteration 45
Iteration 46
Iteration 47
```

|              |
|--------------|
| Iteration 48 |
| Iteration 49 |
| Iteration 50 |
| Iteration 51 |
| Iteration 52 |
| Iteration 53 |
| Iteration 54 |
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| Iteration 89 |
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| Iteration 91 |
| Iteration 92 |
| Iteration 93 |
| Iteration 94 |
| Iteration 95 |
| Iteration 96 |
| Iteration 97 |
| Iteration 98 |
| Iteration 99 |

## References

- [1] SEGARAN, T. *Programming Collective Intelligence: Building Smart Web 2.0 Applications*. O'Reilly, 2007.