

Sharat_Sripada_HW4

```
# install.packages('tm')
# install.packages('tmap')
# install.packages('quanteda')
# install.packages('philentropy')
# install.packages('factoextra')

library(tm)

## Loading required package: NLP

library(tmap)
library(quanteda)

## Package version: 2.1.1
## Parallel computing: 2 of 4 threads used.
## See https://quanteda.io for tutorials and examples.
##
## Attaching package: 'quanteda'

## The following objects are masked from 'package:tm':
##
##   as.DocumentTermMatrix, stopwords

## The following objects are masked from 'package:NLP':
##
##   meta, meta<-

## The following object is masked from 'package:utils':
##
##   View

library(RColorBrewer)
library(wordcloud)
library(philentropy)
library(factoextra)

## Loading required package: ggplot2

##
## Attaching package: 'ggplot2'

## The following object is masked from 'package:NLP':
##
##   annotate
```

```
## Welcome! Want to learn more? See two factoextra-related books at  
https://goo.gl/ve3WBa
```

Introduction

This week delves on concepts of clustering viz. k-means, HAC and various distance measurements that aid in the process namely, Euclidean and cosine methods. In particular, the home-work will attempt to solve the problem of classifying disputed papers between authors Hamilton and Madison.

We begin our analysis by ingesting a corpus of documents and running through the following pipelines:

- loading the documents using the R Corpus function
- build a document term matrix (DTM)
- visualize wordclouds
- dive into the core concepts of clustering
- classify disputed documents from results of clustering

```
#Load the data/corpus  
FedPapersCorpus <-  
Corpus(DirSource("/Users/venkatasharatsripada/Downloads/IST707repo-  
master/FedPapersCorpus"))  
numFedPapers <- length(FedPapersCorpus)
```

```
summary(FedPapersCorpus)
```

```
##           Length Class           Mode  
## dispt_fed_49.txt      2 PlainTextDocument list  
## dispt_fed_50.txt      2 PlainTextDocument list  
## dispt_fed_51.txt      2 PlainTextDocument list  
## dispt_fed_52.txt      2 PlainTextDocument list  
## dispt_fed_53.txt      2 PlainTextDocument list  
## dispt_fed_54.txt      2 PlainTextDocument list  
## dispt_fed_55.txt      2 PlainTextDocument list  
## dispt_fed_56.txt      2 PlainTextDocument list  
## dispt_fed_57.txt      2 PlainTextDocument list  
## dispt_fed_62.txt      2 PlainTextDocument list  
## dispt_fed_63.txt      2 PlainTextDocument list  
## Hamilton_fed_1.txt    2 PlainTextDocument list  
## Hamilton_fed_11.txt   2 PlainTextDocument list  
## Hamilton_fed_12.txt   2 PlainTextDocument list  
## Hamilton_fed_13.txt   2 PlainTextDocument list  
## Hamilton_fed_15.txt   2 PlainTextDocument list  
## Hamilton_fed_16.txt   2 PlainTextDocument list  
## Hamilton_fed_17.txt   2 PlainTextDocument list  
## Hamilton_fed_21.txt   2 PlainTextDocument list
```

[illegible]

```

## Jay_fed_64.txt      2      PlainTextDocument list
## Madison_fed_10.txt  2      PlainTextDocument list
## Madison_fed_14.txt  2      PlainTextDocument list
## Madison_fed_37.txt  2      PlainTextDocument list
## Madison_fed_38.txt  2      PlainTextDocument list
## Madison_fed_39.txt  2      PlainTextDocument list
## Madison_fed_40.txt  2      PlainTextDocument list
## Madison_fed_41.txt  2      PlainTextDocument list
## Madison_fed_42.txt  2      PlainTextDocument list
## Madison_fed_43.txt  2      PlainTextDocument list
## Madison_fed_44.txt  2      PlainTextDocument list
## Madison_fed_45.txt  2      PlainTextDocument list
## Madison_fed_46.txt  2      PlainTextDocument list
## Madison_fed_47.txt  2      PlainTextDocument list
## Madison_fed_48.txt  2      PlainTextDocument list
## Madison_fed_58.txt  2      PlainTextDocument list

# meta(FedPapersCorpus[[1]])

#Ignore extremely rare words - <2% of documents
(minTermFreq <- 0.02 * numFedPapers)

## [1] 1.7

#Also, ignore common words - >75%-95% of documents
(maxTermFreq <- 0.95 * numFedPapers)

## [1] 80.75

#
Papers_DTM <- DocumentTermMatrix(FedPapersCorpus,
                                control=list(
                                  stopwords=TRUE,
                                  wordLengths=c(3,15),
                                  removePunctuation=T,
                                  removeNumbers=T,
                                  tolower=T,
                                  stemming=T,
                                  remove_separators=T,
                                  bounds=list(global=c(minTermFreq,
maxTermFreq)))
                                ))

DTM <- as.matrix(Papers_DTM)
(DTM[1:11,1:10])

##              Terms
## Docs      abandon abat abb abet abil abl ablest abolish abolit
abort
##  dispt_fed_49.txt      0      0      0      0      0      2      0      0      0
0
##  dispt_fed_50.txt      0      0      0      0      0      0      0      0      0

```

```

0
## dispt_fed_51.txt      0    0    0    0    0    1    0    0    0
0
## dispt_fed_52.txt      0    0    0    0    1    1    0    0    0
0
## dispt_fed_53.txt      0    1    0    0    0    0    0    0    0
0
## dispt_fed_54.txt      0    0    0    0    0    0    0    0    0
0
## dispt_fed_55.txt      0    0    0    0    0    0    0    0    0
0
## dispt_fed_56.txt      0    0    0    0    0    0    0    0    0
0
## dispt_fed_57.txt      0    0    0    0    0    0    0    0    0
0
## dispt_fed_62.txt      0    0    0    0    0    1    0    0    0
0
## dispt_fed_63.txt      0    0    0    0    0    4    0    0    0
0

```

```

col_WordFreq <- colSums(as.matrix(Papers_DTM))
(head(col_WordFreq))

```

```

## abandon    abat    abb    abet    abil    abl
##          9      2      5      2     15     74

```

```

#Length of all words
(length(col_WordFreq))

```

```

## [1] 3370

```

```

(row_WordFreq <- rowSums(as.matrix(Papers_DTM)))

```

```

## dispt_fed_49.txt dispt_fed_50.txt dispt_fed_51.txt
dispt_fed_52.txt
##          677          480          783
743
## dispt_fed_53.txt dispt_fed_54.txt dispt_fed_55.txt
dispt_fed_56.txt
##          903          766          865
649
## dispt_fed_57.txt dispt_fed_62.txt dispt_fed_63.txt
Hamilton_fed_1.txt
##          889          983          1244
659
## Hamilton_fed_11.txt Hamilton_fed_12.txt Hamilton_fed_13.txt
Hamilton_fed_15.txt
##          1020          901          400
1256
## Hamilton_fed_16.txt Hamilton_fed_17.txt Hamilton_fed_21.txt
Hamilton_fed_22.txt

```

##	814	663	823
1494			
## Hamilton_fed_23.txt	Hamilton_fed_24.txt	Hamilton_fed_25.txt	
Hamilton_fed_26.txt			
##	717	826	825
983			
## Hamilton_fed_27.txt	Hamilton_fed_28.txt	Hamilton_fed_29.txt	
Hamilton_fed_30.txt			
##	573	639	876
819			
## Hamilton_fed_31.txt	Hamilton_fed_32.txt	Hamilton_fed_33.txt	
Hamilton_fed_34.txt			
##	673	589	640
883			
## Hamilton_fed_35.txt	Hamilton_fed_36.txt	Hamilton_fed_59.txt	
Hamilton_fed_6.txt			
##	942	1095	720
868			
## Hamilton_fed_60.txt	Hamilton_fed_61.txt	Hamilton_fed_65.txt	
Hamilton_fed_66.txt			
##	892	591	816
899			
## Hamilton_fed_67.txt	Hamilton_fed_68.txt	Hamilton_fed_69.txt	
Hamilton_fed_7.txt			
##	688	604	1174
952			
## Hamilton_fed_70.txt	Hamilton_fed_71.txt	Hamilton_fed_72.txt	
Hamilton_fed_73.txt			
##	1295	677	842
941			
## Hamilton_fed_74.txt	Hamilton_fed_75.txt	Hamilton_fed_76.txt	
Hamilton_fed_77.txt			
##	422	822	796
798			
## Hamilton_fed_78.txt	Hamilton_fed_79.txt	Hamilton_fed_8.txt	
Hamilton_fed_80.txt			
##	1245	421	892
974			
## Hamilton_fed_81.txt	Hamilton_fed_82.txt	Hamilton_fed_83.txt	
Hamilton_fed_84.txt			
##	1581	642	2374
1656			
## Hamilton_fed_85.txt	Hamilton_fed_9.txt	HM_fed_18.txt	
HM_fed_19.txt			
##	1114	808	926
907			
## HM_fed_20.txt	Jay_fed_2.txt	Jay_fed_3.txt	
Jay_fed_4.txt			
##	692	709	622
663			

```
##      Jay_fed_5.txt      Jay_fed_64.txt  Madison_fed_10.txt
Madison_fed_14.txt
##              605              966              1316
882
##  Madison_fed_37.txt  Madison_fed_38.txt  Madison_fed_39.txt
Madison_fed_40.txt
##              1122              1348              981
1132
##  Madison_fed_41.txt  Madison_fed_42.txt  Madison_fed_43.txt
Madison_fed_44.txt
##              1479              1140              1344
1178
##  Madison_fed_45.txt  Madison_fed_46.txt  Madison_fed_47.txt
Madison_fed_48.txt
##              810              980              1167
738
##  Madison_fed_58.txt
##              847
```

Normalization

#create a normalized version of Papers_DTM

```
Papers_M <- as.matrix(Papers_DTM)
Papers_M_N1 <- apply(Papers_M, 1, function(i) round(i/sum(i),3))
Papers_Matrix_Norm <- t(Papers_M_N1)
```

#compare the original and normalized version

```
(Papers_M[c(1:11),c(1000:1010)])
```

```
##              Terms
## Docs      edit effect effectu efficaci effici effort eight eighth
## dispt_fed_49.txt    0      1      1      0      0      0      0      0
## dispt_fed_50.txt    0      3      0      0      0      0      0      0
## dispt_fed_51.txt    0      0      0      0      0      0      0      0
## dispt_fed_52.txt    0      1      1      0      0      0      0      0
## dispt_fed_53.txt    0      2      1      0      0      0      0      0
## dispt_fed_54.txt    0      3      0      2      0      0      0      0
## dispt_fed_55.txt    0      0      0      0      0      0      1      0
## dispt_fed_56.txt    0      2      0      0      0      0      3      0
## dispt_fed_57.txt    0      0      2      0      0      0      0      0
## dispt_fed_62.txt    0      4      0      0      0      0      0      0
## dispt_fed_63.txt    0      2      2      0      0      0      0      0
##              Terms
## Docs      either elaps elect
## dispt_fed_49.txt    1      0      1
## dispt_fed_50.txt    3      0      2
## dispt_fed_51.txt    0      0      1
## dispt_fed_52.txt    0      0     21
## dispt_fed_53.txt    2      1     20
## dispt_fed_54.txt    0      0      1
## dispt_fed_55.txt    2      0      3
```

```
## dispt_fed_56.txt      2      0      3
## dispt_fed_57.txt      0      0     10
## dispt_fed_62.txt      0      0      2
## dispt_fed_63.txt      0      0     14
```

```
(Papers_Matrix_Norm[c(1:11),c(1000:1010)])
```

```
##                      Terms
## Docs                edit effect effectu efficaci effici effort eight eighth
## dispt_fed_49.txt      0 0.001 0.001 0.000      0      0 0.000      0
## dispt_fed_50.txt      0 0.006 0.000 0.000      0      0 0.000      0
## dispt_fed_51.txt      0 0.000 0.000 0.000      0      0 0.000      0
## dispt_fed_52.txt      0 0.001 0.001 0.000      0      0 0.000      0
## dispt_fed_53.txt      0 0.002 0.001 0.000      0      0 0.000      0
## dispt_fed_54.txt      0 0.004 0.000 0.003      0      0 0.000      0
## dispt_fed_55.txt      0 0.000 0.000 0.000      0      0 0.001      0
## dispt_fed_56.txt      0 0.003 0.000 0.000      0      0 0.005      0
## dispt_fed_57.txt      0 0.000 0.002 0.000      0      0 0.000      0
## dispt_fed_62.txt      0 0.004 0.000 0.000      0      0 0.000      0
## dispt_fed_63.txt      0 0.002 0.002 0.000      0      0 0.000      0
```

```
##                      Terms
## Docs                either elaps elect
## dispt_fed_49.txt      0.001 0.000 0.001
## dispt_fed_50.txt      0.006 0.000 0.004
## dispt_fed_51.txt      0.000 0.000 0.001
## dispt_fed_52.txt      0.000 0.000 0.028
## dispt_fed_53.txt      0.002 0.001 0.022
## dispt_fed_54.txt      0.000 0.000 0.001
## dispt_fed_55.txt      0.002 0.000 0.003
## dispt_fed_56.txt      0.003 0.000 0.005
## dispt_fed_57.txt      0.000 0.000 0.011
## dispt_fed_62.txt      0.000 0.000 0.002
## dispt_fed_63.txt      0.000 0.000 0.011
```

*#verify for word 'embarrass' in document 'dispt_fed_62.txt' if the
#normalization math is correct*

```
(row_WordFreq)
```

```
## dispt_fed_49.txt      dispt_fed_50.txt      dispt_fed_51.txt
dispt_fed_52.txt
##                677                480                783
743
## dispt_fed_53.txt      dispt_fed_54.txt      dispt_fed_55.txt
dispt_fed_56.txt
##                903                766                865
649
## dispt_fed_57.txt      dispt_fed_62.txt      dispt_fed_63.txt
Hamilton_fed_1.txt
##                889                983                1244
659
```



```

## Hamilton_fed_11.txt Hamilton_fed_12.txt Hamilton_fed_13.txt
Hamilton_fed_15.txt
##                1020                901                400
1256
## Hamilton_fed_16.txt Hamilton_fed_17.txt Hamilton_fed_21.txt
Hamilton_fed_22.txt
##                814                663                823
1494
## Hamilton_fed_23.txt Hamilton_fed_24.txt Hamilton_fed_25.txt
Hamilton_fed_26.txt
##                717                826                825
983
## Hamilton_fed_27.txt Hamilton_fed_28.txt Hamilton_fed_29.txt
Hamilton_fed_30.txt
##                573                639                876
819
## Hamilton_fed_31.txt Hamilton_fed_32.txt Hamilton_fed_33.txt
Hamilton_fed_34.txt
##                673                589                640
883
## Hamilton_fed_35.txt Hamilton_fed_36.txt Hamilton_fed_59.txt
Hamilton_fed_6.txt
##                942                1095                720
868
## Hamilton_fed_60.txt Hamilton_fed_61.txt Hamilton_fed_65.txt
Hamilton_fed_66.txt
##                892                591                816
899
## Hamilton_fed_67.txt Hamilton_fed_68.txt Hamilton_fed_69.txt
Hamilton_fed_7.txt
##                688                604                1174
952
## Hamilton_fed_70.txt Hamilton_fed_71.txt Hamilton_fed_72.txt
Hamilton_fed_73.txt
##                1295                677                842
941
## Hamilton_fed_74.txt Hamilton_fed_75.txt Hamilton_fed_76.txt
Hamilton_fed_77.txt
##                422                822                796
798
## Hamilton_fed_78.txt Hamilton_fed_79.txt Hamilton_fed_8.txt
Hamilton_fed_80.txt
##                1245                421                892
974
## Hamilton_fed_81.txt Hamilton_fed_82.txt Hamilton_fed_83.txt
Hamilton_fed_84.txt
##                1581                642                2374
1656
## Hamilton_fed_85.txt Hamilton_fed_9.txt HM_fed_18.txt
HM_fed_19.txt

```

```
##          1114          808          926
907
##      HM_fed_20.txt      Jay_fed_2.txt      Jay_fed_3.txt
Jay_fed_4.txt
##          692          709          622
663
##      Jay_fed_5.txt      Jay_fed_64.txt  Madison_fed_10.txt
Madison_fed_14.txt
##          605          966          1316
882
##  Madison_fed_37.txt  Madison_fed_38.txt  Madison_fed_39.txt
Madison_fed_40.txt
##          1122          1348          981
1132
##  Madison_fed_41.txt  Madison_fed_42.txt  Madison_fed_43.txt
Madison_fed_44.txt
##          1479          1140          1344
1178
##  Madison_fed_45.txt  Madison_fed_46.txt  Madison_fed_47.txt
Madison_fed_48.txt
##          810          980          1167
738
##  Madison_fed_58.txt
##          847
```

#dispt_fed_62 has 798 words in total

#there are 2x words of 'embarrass' so, $2/798 = 0.0025 \sim 0.003$ (3 places after decimal)

Data-structures

```
Papers_dtm_matrix <- as.matrix(Papers_DTM)
str(Papers_dtm_matrix)
```

```
##  num [1:85, 1:3370] 0 0 0 0 0 0 0 0 0 0 ...
##  - attr(*, "dimnames")=List of 2
##    ..$ Docs : chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt"
"dispt_fed_51.txt" "dispt_fed_52.txt" ...
##    ..$ Terms: chr [1:3370] "abandon" "abat" "abb" "abet" ...
```

```
Papers_dtm_matrix[c(1:11),c(2:10)]
```

```
##          Terms
## Docs      abat abb abet abil abl ablest abolish abolit abort
##  dispt_fed_49.txt      0  0  0  0  2      0      0      0  0
##  dispt_fed_50.txt      0  0  0  0  0      0      0      0  0
##  dispt_fed_51.txt      0  0  0  0  1      0      0      0  0
##  dispt_fed_52.txt      0  0  0  1  1      0      0      0  0
##  dispt_fed_53.txt      1  0  0  0  0      0      0      0  0
##  dispt_fed_54.txt      0  0  0  0  0      0      0      0  0
##  dispt_fed_55.txt      0  0  0  0  0      0      0      0  0
##  dispt_fed_56.txt      0  0  0  0  0      0      0      0  0
```

##	dispt_fed_57.txt	0	0	0	0	0	0	0	0	0
##	dispt_fed_62.txt	0	0	0	0	1	0	0	0	0
##	dispt_fed_63.txt	0	0	0	0	4	0	0	0	0

Convert to a data-frame

```
Papers_DF <- as.data.frame(as.matrix(Papers_DTM))
str(Papers_DF)
```

```
## 'data.frame':    85 obs. of  3370 variables:
## $ abandon      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abat         : num  0 0 0 0 1 0 0 0 0 0 ...
## $ abb          : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abet         : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abil         : num  0 0 0 1 0 0 0 0 0 0 ...
## $ abl          : num  2 0 1 1 0 0 0 0 0 1 ...
## $ ablest       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abolish      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abolit       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abort        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abound       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abridg       : num  0 0 0 1 0 0 0 0 0 0 ...
## $ abroad       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ absolut      : num  0 2 2 1 0 0 0 0 0 0 ...
## $ absorb       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abstain      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abstract     : num  0 0 0 0 0 0 0 0 0 0 ...
## $ absurd       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abund        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abus         : num  1 1 2 1 1 0 0 0 0 0 ...
## $ abyss        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ acced        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accept       : num  0 0 0 0 0 0 0 0 0 1 ...
## $ access       : num  0 0 0 2 0 0 0 0 0 0 ...
## $ accid        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accident     : num  0 0 0 1 0 0 0 0 0 0 ...
## $ accommod     : num  0 0 0 0 1 0 0 0 0 0 ...
## $ accompani    : num  0 0 0 0 0 0 0 1 0 0 ...
## $ accomplic    : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accomplish   : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accord       : num  0 0 0 0 1 2 2 1 1 0 ...
## $ account      : num  0 0 0 0 0 0 1 0 0 0 ...
## $ accumul      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accur        : num  1 0 0 0 1 0 0 0 0 1 ...
## $ accuraci     : num  0 0 0 0 0 1 0 0 0 0 ...
## $ accus        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ accustom     : num  0 0 0 0 0 0 0 0 0 0 ...
## $ achaeaan     : num  0 0 0 0 0 0 0 0 0 0 ...
## $ acknowledg    : num  0 1 0 0 0 0 0 0 0 1 ...
## $ acquaint     : num  1 0 0 0 2 0 0 2 0 1 ...
## $ acquiesc     : num  0 0 0 0 0 0 0 0 0 0 ...
```

```

## $ acquire      : num 1 0 0 0 5 0 0 2 0 0 ...
## $ acquisit     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ act          : num 0 0 0 1 2 1 0 1 0 1 ...
## $ action       : num 0 0 1 0 0 0 0 0 0 1 ...
## $ activ        : num 0 4 0 0 0 0 0 0 0 0 ...
## $ actor        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ actual       : num 1 2 0 0 4 0 0 0 1 0 ...
## $ actualat     : num 0 0 0 0 0 0 1 0 1 0 ...
## $ adapt        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ add          : num 0 0 0 0 1 0 0 1 1 0 ...
## $ addict       : num 0 0 0 0 0 0 0 0 0 0 ...
## $ addit        : num 0 0 1 1 0 0 0 0 1 1 ...
## $ address      : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adduc        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adept        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adequ        : num 1 1 0 0 0 0 0 0 0 0 ...
## $ adher        : num 0 0 1 0 0 1 0 0 0 0 ...
## $ adjourn      : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adjud        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adjust       : num 0 0 0 0 0 1 0 0 0 0 ...
## $ administ     : num 0 0 2 0 0 0 0 0 0 1 ...
## $ administr    : num 1 2 1 0 0 0 0 0 1 0 ...
## $ admir        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ admiralti    : num 0 0 0 0 0 0 0 0 0 0 ...
## $ admiss       : num 0 0 0 0 0 1 0 0 1 1 ...
## $ admit        : num 1 0 3 0 1 5 2 0 1 0 ...
## $ admitt       : num 0 0 0 0 0 0 0 0 0 0 ...
## $ admonish     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ admonit      : num 0 0 0 0 0 0 0 0 0 1 ...
## $ adopt        : num 0 0 0 1 0 1 0 0 0 1 ...
## $ advanc       : num 0 0 0 0 1 0 0 1 1 2 ...
## $ advantag     : num 4 1 0 2 2 4 0 1 0 7 ...
## $ adventiti    : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adventur     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ advers       : num 2 0 0 0 0 0 0 0 0 0 ...
## $ adversari    : num 0 0 0 0 0 0 0 0 0 0 ...
## $ advert       : num 0 0 0 0 0 0 0 0 0 0 ...
## $ advertis     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ advic        : num 0 0 0 0 0 0 0 0 0 1 ...
## $ advis        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ advoc        : num 0 0 0 0 0 1 0 1 0 0 ...
## $ affair       : num 0 0 1 0 9 0 1 5 0 4 ...
## $ affect       : num 0 0 0 1 0 0 0 0 1 1 ...
## $ affin        : num 0 0 0 0 0 0 0 0 0 0 ...
## $ affirm       : num 0 0 0 0 2 0 0 0 0 1 ...
## $ afford       : num 0 0 0 0 1 0 0 0 0 0 ...
## $ affront      : num 0 0 0 0 0 0 0 0 0 0 ...
## $ afraid       : num 0 0 0 0 0 0 1 0 0 0 ...
## $ afterward    : num 0 0 0 0 0 0 0 0 0 0 ...
## $ age          : num 0 0 0 1 0 0 0 0 0 2 ...

```

```
## $ agenc      : num  0 0 1 0 0 0 0 0 0 1 ...
## $ agent      : num  1 1 0 0 0 0 0 0 0 0 ...
## $ aggrand    : num  0 0 0 0 0 0 0 0 0 0 ...
## $ aggrandiz  : num  1 0 0 0 0 0 0 0 1 0 ...
## $ aggreg     : num  0 0 0 0 0 2 0 0 0 0 ...
## $ aggress    : num  0 0 0 0 0 0 0 0 0 0 ...
## $ aggressor  : num  0 0 0 0 0 0 0 0 0 0 ...
## $ agit       : num  0 0 0 0 0 0 0 0 0 0 ...
## [list output truncated]
```

Example word cloud

Breaking the word clouds based on the document list: - 1:11 -> disputed papers - 12:62 -> Hamilton papers - 63:70 -> Ignoring HM_fed, Jay_fed papers - 71:85 -> Madison papers

```
disputedpaperswc <- wordcloud(colnames(Papers_dtm_matrix),
Papers_dtm_matrix[11,])

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
repres
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
branch
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
bodi
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
exampl
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
small
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
## american could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
member
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
senat
## could not be fit on page. It will not be plotted.

## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :
## maryland could not be fit on page. It will not be plotted.
```

```
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## charact could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## passion could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## advantag could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## independ could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
hous  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
former  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## without could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
might  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
mani  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
known  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
defect  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
measur  
## could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
## legislatur could not be fit on page. It will not be plotted.  
  
## Warning in wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[11, :  
two  
## could not be fit on page. It will not be plotted.
```



```
(head(sort(as.matrix(Papers_DTM)[11,], decreasing = TRUE), n=50))
```

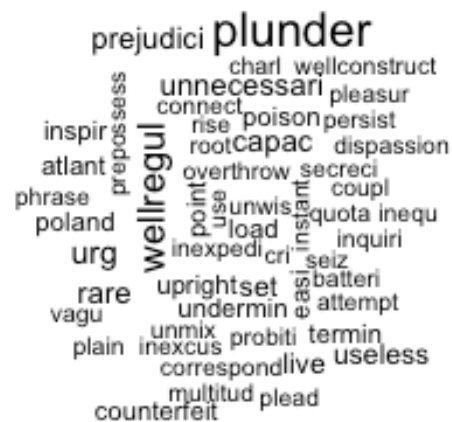
##	senat	repres	bodi	can	elect	measur
corrupt						
##	24	18	15	14	14	11
9						
##	nation	constitut	former	reason	year	assembl
exempl						
##	9	8	8	8	8	7
7						
##	two	annual	danger	everi	evid	feder
import						
##	7	6	6	6	6	6
6						
##	latter	object	particular	public	advantag	ancient
answer						
##	6	6	6	6	5	5
5						
##	appear	charact	fact	first	hous	institut
less						
##	5	5	5	5	5	5
5						
##	mani	member	might	oper	order	popular

```

probal
##      5      5      5      5      5      5
5
##      republ      respons      small      term      time      whole
without
##      5      5      5      5      5      5
5
##      abl
##      4

HamiltonPapersWC <- wordcloud(colnames(Papers_dtm_matrix),
Papers_dtm_matrix[12:62, ])

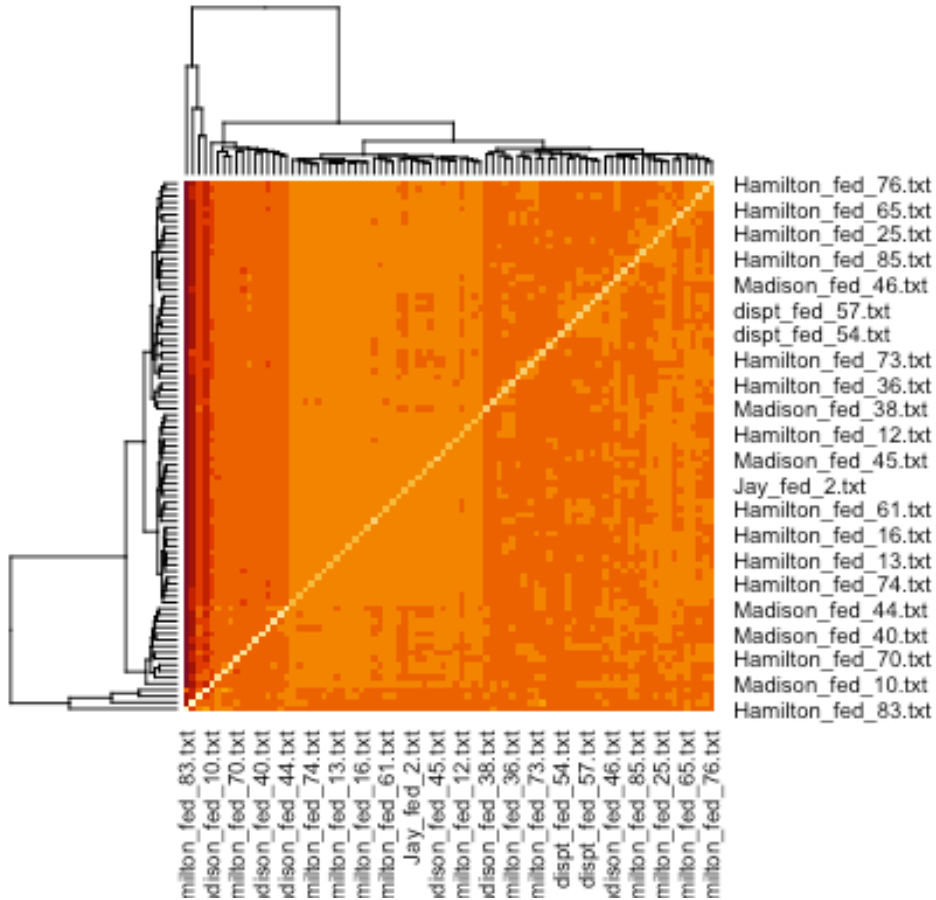
```



```

MadisonPapersWC <- wordcloud(colnames(Papers_dtm_matrix),
Papers_dtm_matrix[71:85, ])

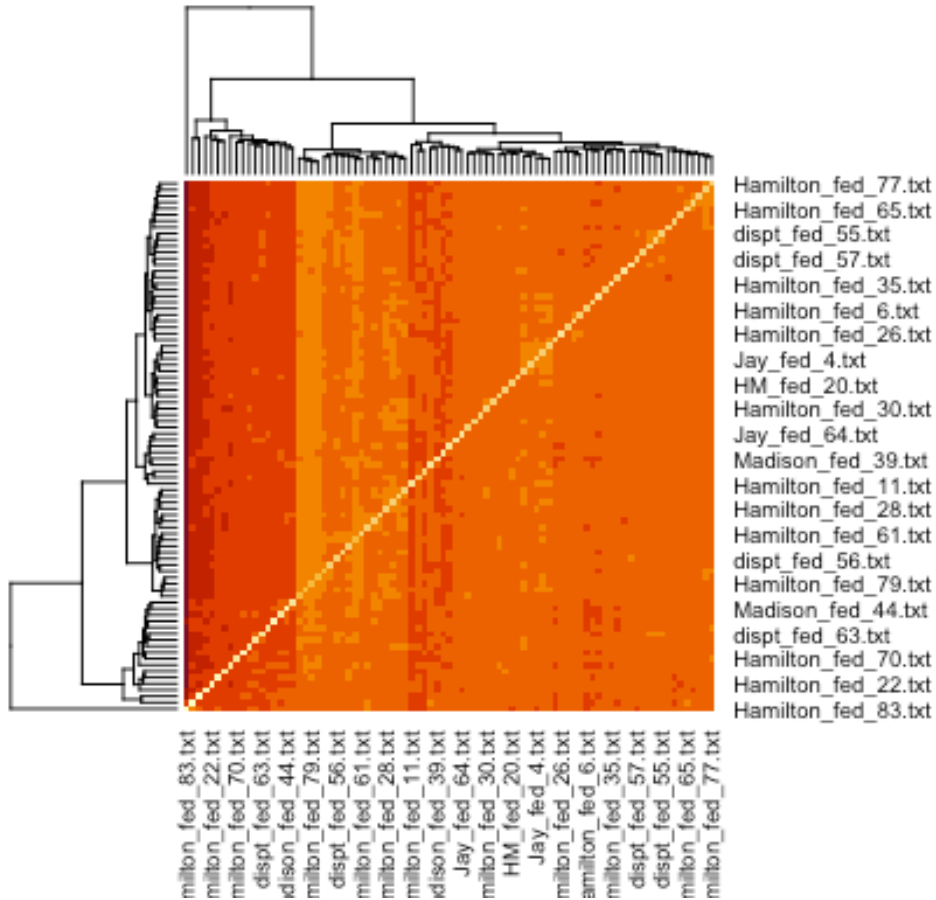
```

```
distMatrix_M <- distance(m, method='manhattan', use.row.names = TRUE)

## Metric: 'manhattan'; comparing: 85 vectors.

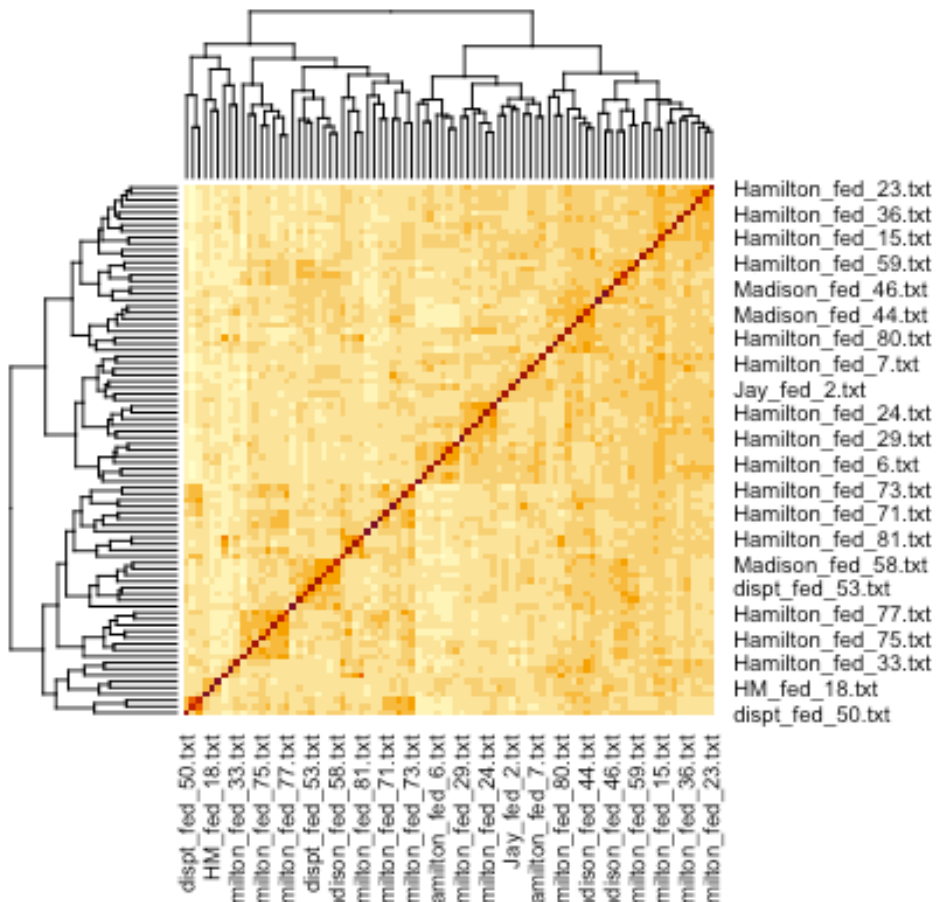
# print(distMatrix_M)
heatmap(distMatrix_M)
```



```
distMatrix_C <- distance(m, method = 'cosine', use.row.names = TRUE)

## Metric: 'cosine'; comparing: 85 vectors.

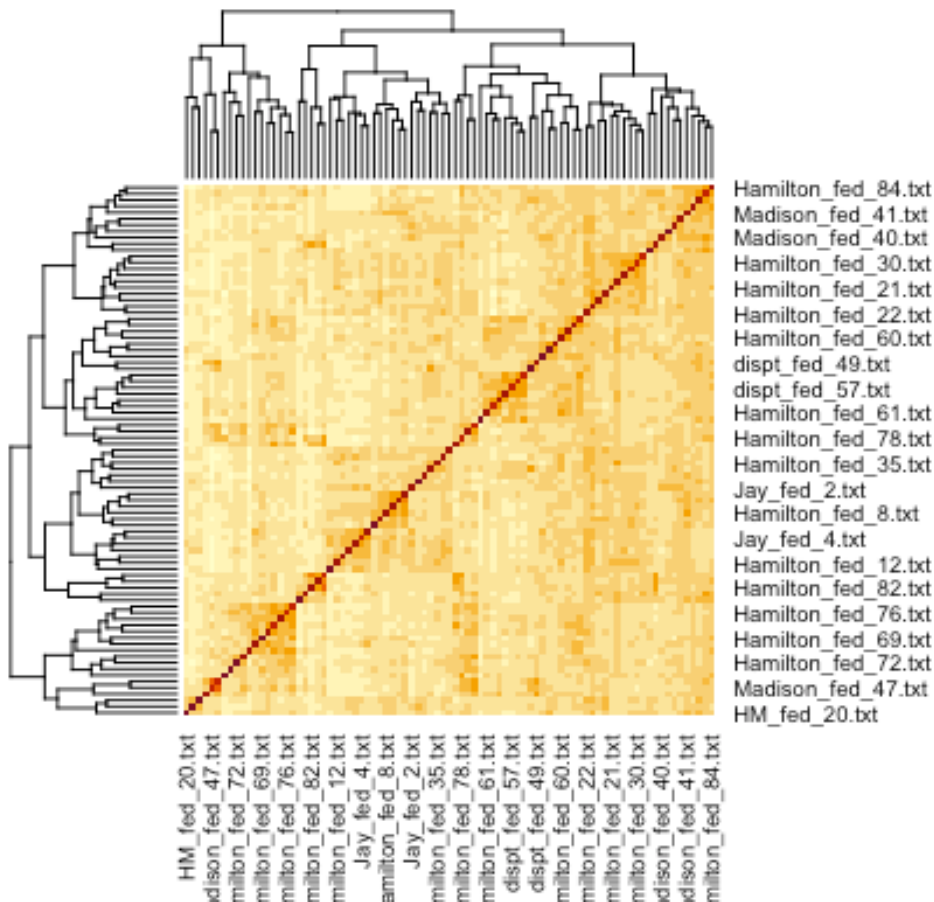
# print(distMatrix_C)
heatmap(distMatrix_C)
```



```
distMatrix_C_norm <- distance(m_norm, method='cosine', use.row.names = TRUE)

## Metric: 'cosine'; comparing: 85 vectors.

# print(distMatrix_C_norm)
heatmap(distMatrix_C_norm)
```



The `dist()` function has issues with 'cosine' methods. Instead, used `distance()` function and obtain cosine similarity visualization. Heat-maps prove cosine similarity measurements are likely more suitable for document analysis.

Data

We will explore the following two methods to cluster the data and determine an author to the disputed papers:

- K-means algorithm
- HAC algorithm

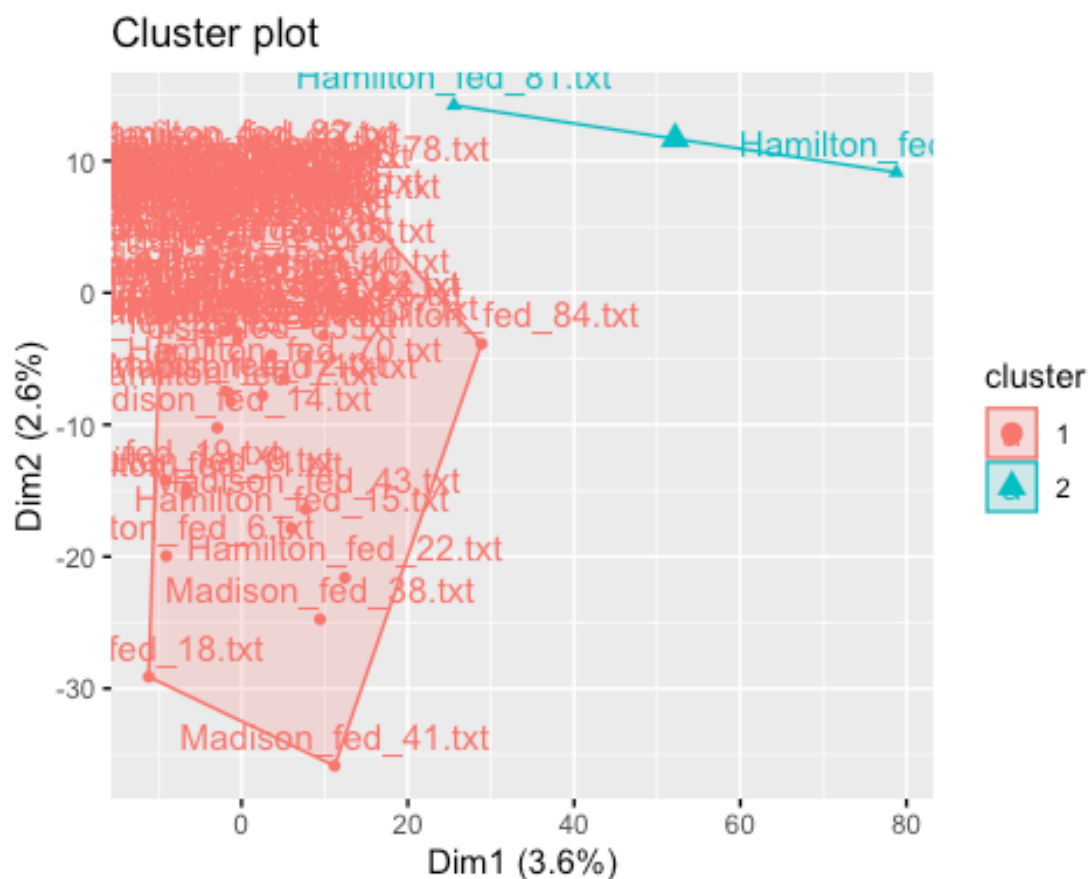
Given that the number of authors here are namely Hamilton and Madison, we will start with choosing number of clusters = 2.

First, is the k-means algorithm:

```
k <- 2
set.seed(5)
km.res <- kmeans(Papers_dtm_matrix, k, nstart=100, iter.max=50)
str(km.res)
```

```
## List of 9
## $ cluster      : Named int [1:85] 1 1 1 1 1 1 1 1 1 1 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt"
"dispt_fed_51.txt" "dispt_fed_52.txt" ...
## $ centers      : num [1:2, 1:3370] 0.1084 0 0.0241 0 0.0602 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:2] "1" "2"
##     .. ..$ : chr [1:3370] "abandon" "abat" "abb" "abet" ...
## $ totss       : num 202176
## $ withinss    : num [1:2] 174195 6448
## $ tot.withinss: num 180642
## $ betweenss   : num 21533
## $ size        : int [1:2] 83 2
## $ iter        : int 1
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"

#plot a visualization
fviz_cluster(km.res, Papers_dtm_matrix)
```



```
k <- 7
km.res <- kmeans(Papers_Matrix_Norm, k, nstart=50, iter.max=50)
str(km.res)
```

```
## List of 9
## $ cluster      : Named int [1:85] 1 1 1 5 5 5 5 5 7 ...
## ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt"
"dispt_fed_51.txt" "dispt_fed_52.txt" ...
## $ centers      : num [1:7, 1:3370] 7.69e-05 0.00 7.14e-05 0.00 0.00 ...
## ..- attr(*, "dimnames")=List of 2
## .. ..$ : chr [1:7] "1" "2" "3" "4" ...
## .. ..$ : chr [1:3370] "abandon" "abat" "abb" "abet" ...
## $ totss       : num 0.226
## $ withinss    : num [1:7] 0.03396 0.00231 0.02952 0.00754 0.02239 ...
## $ tot.withinss: num 0.174
## $ betweenss   : num 0.0514
## $ size        : int [1:7] 13 2 14 4 10 5 37
## $ iter        : int 4
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

#plot a visualization

```
fviz_cluster(km.res, Papers_Matrix_Norm)
```



Now, we explore the HAC algorithms

#Euclidean distance measure

```
dist.eul <- as.dist(distMatrix_E)
```

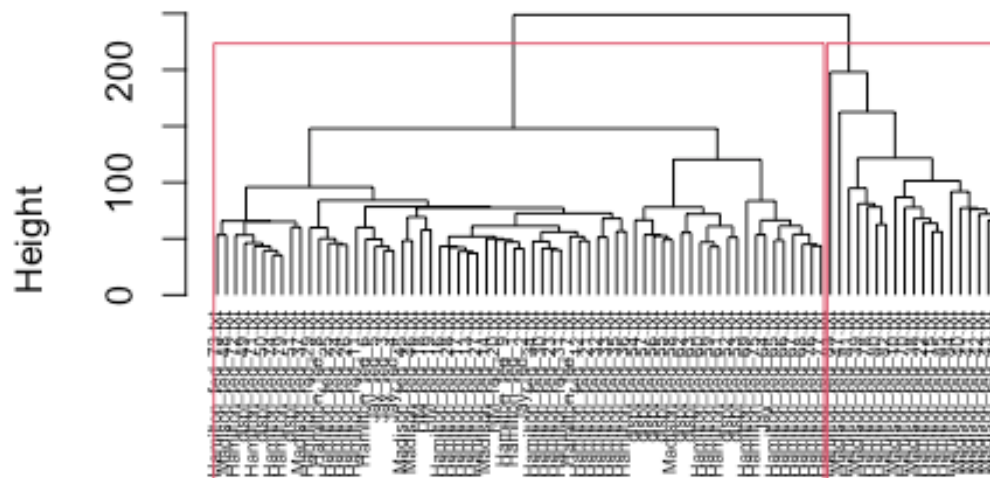
```
groups_E <- hclust(dist.eul, method='ward.D')
```

#Visualizations

```
plot(groups_E, cex=0.5, font=22, hang=-1, main="HAC cluster dendrogram with  
Euclidean Similarity")
```

```
rect.hclust(groups_E, k=2)
```

HAC cluster dendrogram with Euclidean Similarity



dist.eul
hclust (*, "ward.D")

#Cosine distance measure

```
dist.cos <- as.dist(distMatrix_C)
```

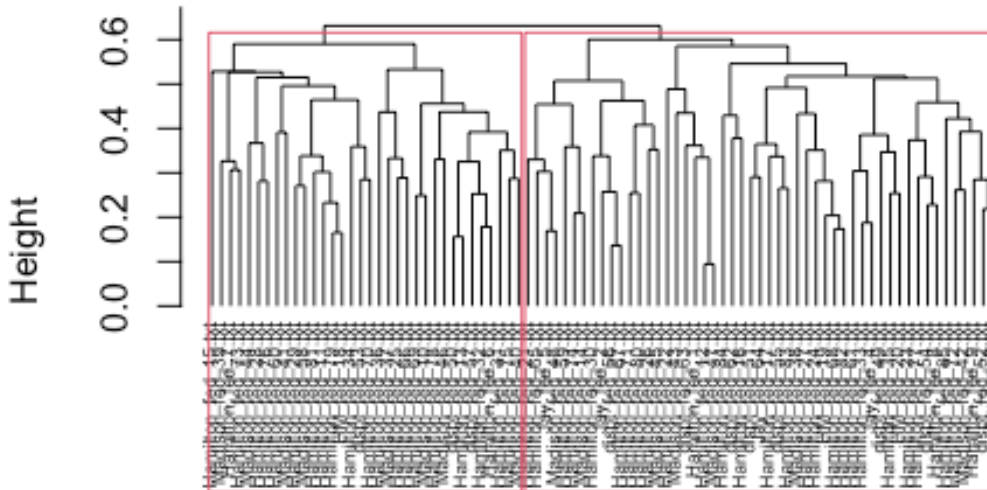
```
groups_C <- hclust(dist.cos, method='ward.D')
```

#Visualizations

```
plot(groups_C, cex=0.5, font=22, hang=-1, main="HAC cluster dendrogram with  
Cosine Similarity")
```

```
rect.hclust(groups_C, k=2)
```


HAC cluster dendrogram with Cosine Similarity

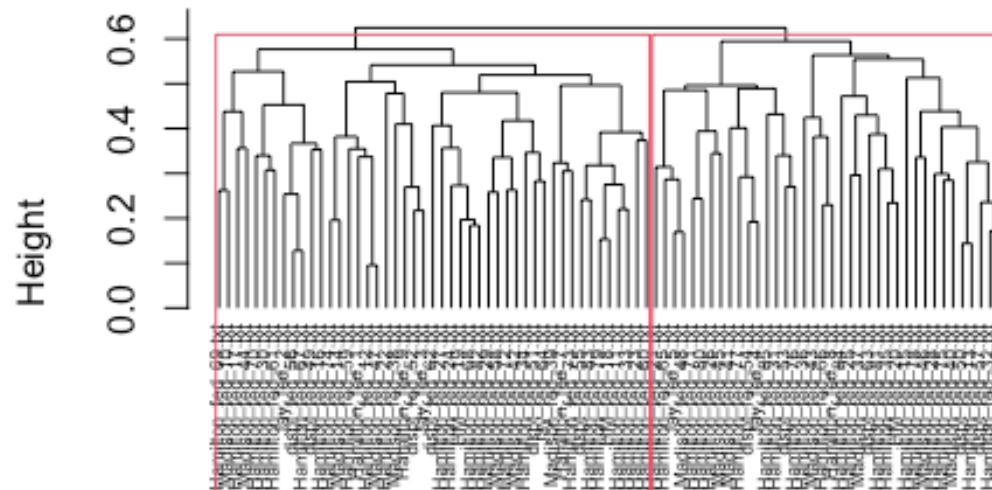


```
dist.cos
hclust (*, "ward.D")
```

```
#Cosine distance measure (Normalized)
dist.cosnorm <- as.dist(distMatrix_C_norm)
groups_C_norm <- hclust(dist.cosnorm, method='ward.D')

#Visualizations
plot(groups_C_norm, cex=0.5, font=22, hang=-1, main="HAC cluster dendrogram
with Cosine Similarity (Normalized)")
rect.hclust(groups_C_norm, k=2)
```

AC cluster dendrogram with Cosine Similarity (Normalized)



dist.cosnorm
hclust (*, "ward.D")

Analysis and Results

K-Means

Here are some results/observations with experiments around different cluster sizes:

- cluster-size=2

SSEs

Within cluster sum of squares by cluster is high:

[1] 174194.7 6447.5

This is an indication of high deviation between data-points and the centroid which we would ideally like to be lower. To explore k-means further, we could consider using the k-medoids/expectation-max or PAM algorithms.

Data

Most of the data-points were grouped into cluster-1 and this did not help to clearly determine the author for the disputed papers.

- cluster-size=7

SSE

SSEs look a lot better with increased cluster-size

Within cluster sum of squares by cluster:

```
[1] 0.00754175 0.03396400 0.06862076 0.00231200 0.02952307 0.00990520  
0.02239410
```

Data

Disputed papers were placed in clusters - 2, 7, 3:

- Number of disputed papers in cluster-2 = 3
- Number of disputed papers in cluster-7 = 7
- Number of disputed papers in cluster-3 = 1

Cluster-7 that has the highest papers does not have sufficient majority of Hamilton/Madison papers to make a decision.

Overall, k-means does not seem like a good algorithm for document analysis use-cases.

HAC algorithm

In comparison, seems like plotting and analyzing dendograms, seems a plausible means to realize the exercise. To a very large extent we can classify the disputed documents to the corresponding authors.

Conclusions

With Hierarchical Agglomerative Clustering (HAC) techniques (and dendograms to analyze the results) we conclude by analyzing one disputed document `dispt_fed_49.txt` across:

- Euclidian

In plot 'HAC cluster dendogram with Euclidean Similarity', see document 'dispt_fed_49.txt' present in the first-cluster on the left and is associated by nodes/leafs that belong to Hamilton so, we can conclude it was written by author Hamilton with moderate confidence.

- Cosine

In plot 'HAC cluster dendogram with Cosine Similarity', see document 'dispt_fed_49.txt' belonging to a cluster towards the end. Again, the nodes/leafs around it are documents by author Hamilton.

- Cosine-Normalized

Likewise, in plot 'HAC cluster dendogram with Cosine Similarity (Normalized)' the surrounding nodes/leafs are related to author Hamilton.

In similar lines, we could extend the study to all disputed documents and hence classify them between the two authors.