

George_Smith_HW4_IST772

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Introduction

The Federalist Papers were a series of eighty-five essays urging the citizens of New York to ratify the new United States Constitution. The essays originally appeared anonymously in New York newspapers in 1787 and 1788 under the pen names “Publius”. It was not until 1818 that the authors Alexander Hamilton, James Madison, and John Jay were identified by name. Using clustering algorithms, k-Means, EM, and HAC I am going to solve the mystery of who wrote each of the Federalist Papers.

installs

```
# install.packages('wordcloud')  
# install.packages('tm')  
# install.packages('slam')  
# install.packages('quanteda')  
# install.packages('SnowballC')  
# install.packages('arules')  
# install.packages('proxy')  
# install.packages('cluster')  
# install.packages('stringi')  
# install.packages('Matrix')  
# install.packages('tidytext')  
# install.packages('plyr')  
# install.packages('ggplot2')  
# install.packages('factoextra')  
# install.packages('mclust')  
# install.packages('dplyr')  
# install.packages('rdduplus')  
# install.packages('corpus')  
# install.packages('quanteda')  
# install.packages('tm')  
# install.packages('Rcpp')
```

```
library(wordcloud)
```

```
## Loading required package: RColorBrewer
```

```
library(tm)
```

```
## Loading required package: NLP
```

```
library(slam)  
library(quanteda)
```

```
## Package version: 3.0.0  
## Unicode version: 10.0  
## ICU version: 61.1
```

```
## Parallel computing: 12 of 12 threads used.
```

```
## See https://quanteda.io for tutorials and examples.
```

```
##  
## Attaching package: 'quanteda'
```

```
## The following object is masked from 'package:tm':  
##  
##     stopwords
```

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

```
library(SnowballC)  
library(arules)
```

```
## Loading required package: Matrix
```

```
##  
## Attaching package: 'arules'
```

```
## The following object is masked from 'package:tm':  
##  
##     inspect
```

```
## The following objects are masked from 'package:base':  
##  
##     abbreviate, write
```

```
library(proxy)
```

```
##  
## Attaching package: 'proxy'
```

```
## The following object is masked from 'package:Matrix':  
##  
##     as.matrix
```

```
## The following objects are masked from 'package:stats':  
##  
##   as.dist, dist
```

```
## The following object is masked from 'package:base':  
##  
##   as.matrix
```

```
library(cluster)  
library(stringi)  
library(Matrix)  
library(tidytext)  
library(plyr)  
library(ggplot2)
```

```
##  
## Attaching package: 'ggplot2'
```

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

```
library(factoextra)
```

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

```
library(mclust)
```

```
## Package 'mclust' version 5.4.7  
## Type 'citation("mclust")' for citing this R package in publications.
```

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:plyr':  
##  
##   arrange, count, desc, failwith, id, mutate, rename, summarise,  
##   summarize
```

```
## The following objects are masked from 'package:arules':  
##  
##   intersect, recode, setdiff, setequal, union
```

```
## The following objects are masked from 'package:stats':  
##  
##   filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
##   intersect, setdiff, setequal, union
```

```
library(rdwplus)
```

```
## Loading required package: raster
```

```
## Loading required package: sp
```

```
##
```

```
## Attaching package: 'raster'
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      select
```

```
## Loading required package: rgrass7
```

```
## Loading required package: XML
```

```
## GRASS GIS interface loaded with GRASS version: (GRASS not running)
```

```
library(corpus)
```

```
library(tm)
```

```
library(Rcpp)
```

read in file

```
FederalistPapers <- read.csv("C:/Users/GeorgeSmith/Documents/fedPapers85.csv", row.names = 2, na.string = "NA")
```

Create backup of FederalistPapers in case it's needed

```
FederalistPapers_Orig <- FederalistPapers
```

Check for NAs and missing values

```
sum(is.na(FederalistPapers))
```

```
## [1] 0
```

```
FederalistPapers <- FederalistPapers[,-1]
```

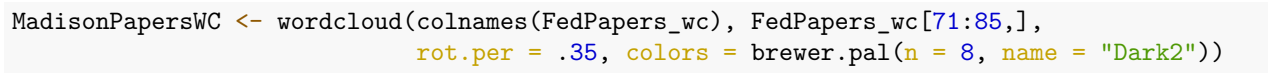
There are no NAs in this data set

first, remove the file and author names for a word cloud gallery

```
FedPapers_wc <- as.matrix(as.dfm(FederalistPapers)) #FederalistPapers[,3:72]
hamPapers = FedPapers_wc[12:62,]
DisputedPapersWC <- wordcloud(colnames(FedPapers_wc), FedPapers_wc[11,],
                               rot.per = .35, colors = brewer.pal(n = 8, name = "Dark2"))
```



```
HamiltonPapersWC <- wordcloud(colnames(FedPapers_wc), FedPapers_wc[12:62,],
                               rot.per = .35, colors = brewer.pal(n = 8, name = "Dark2"))
```





Make the file names the row names. Need a dataframe of numerical values for k-means

```
rownames(FedPapers_km) <- FedPapers_km[,1]
FedPapers_km[,1] <- NULL
```

Set seed for fixed random seed

```
set.seed(20)
```

run k-means

```
Clusters <- kmeans(FedPapers_km, 6)
FedPapers_km$Clusters <- as.factor(Clusters$cluster)
str(Clusters)
```

```
## List of 9
## $ cluster      : Named int [1:85] 1 6 1 6 6 1 6 5 1 6 ...
## ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:6, 1:70] 0.297 0.216 0.16 0.299 0.363 ...
## ..- attr(*, "dimnames")=List of 2
## .. ..$ : chr [1:6] "1" "2" "3" "4" ...
## .. ..$ : chr [1:70] "a" "all" "also" "an" ...
## $ totss       : num 12.6
## $ withinss    : num [1:6] 2.039 0.423 0.599 1.791 1.12 ...
## $ tot.withinss: num 7.29
## $ betweenss   : num 5.28
## $ size        : int [1:6] 20 5 5 24 13 18
## $ iter        : int 3
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

```
Clusters$centers[]
```

```
##           a           all           also           an           and           any           are
## 1 0.2971500 0.05520000 0.008900000 0.06480000 0.3287000 0.04300000 0.07375000
## 2 0.2156000 0.05760000 0.013000000 0.05200000 0.4990000 0.01960000 0.08540000
## 3 0.1598000 0.03600000 0.019800000 0.02520000 0.7152000 0.03760000 0.08520000
## 4 0.2992083 0.05291667 0.003333333 0.08833333 0.3406667 0.05016667 0.07466667
## 5 0.3633846 0.05938462 0.005923077 0.07476923 0.3697692 0.03261538 0.07800000
## 6 0.2888889 0.04872222 0.008444444 0.05772222 0.3925000 0.04238889 0.07872222
##           as           at           be           been           but           by           can
## 1 0.1350000 0.03375000 0.3415000 0.05160000 0.03130000 0.1271000 0.03205000
## 2 0.0700000 0.04640000 0.1196000 0.03280000 0.02400000 0.1648000 0.01620000
## 3 0.1568000 0.03600000 0.2754000 0.02680000 0.04920000 0.1362000 0.03300000
```

```

## 4 0.1300417 0.04579167 0.3185000 0.06358333 0.03237500 0.0992500 0.03883333
## 5 0.1078462 0.05784615 0.2678462 0.06276923 0.03176923 0.1120769 0.03284615
## 6 0.1222222 0.04583333 0.3148333 0.07777778 0.03138889 0.1623333 0.04322222
##      do      down      even      every      for.      from
## 1 0.005200000 0.001600000 0.00685000 0.02845000 0.0929500 0.06675000
## 2 0.002400000 0.002000000 0.00560000 0.01060000 0.0784000 0.08560000
## 3 0.008200000 0.000000000 0.00760000 0.00600000 0.0960000 0.09100000
## 4 0.006916667 0.0032916667 0.01600000 0.02170833 0.0907500 0.08166667
## 5 0.009230769 0.0003846154 0.01307692 0.02253846 0.0750000 0.08923077
## 6 0.004944444 0.0002222222 0.01177778 0.03144444 0.1158889 0.08016667
##      had      has      have      her      his      if.      in.
## 1 0.01495000 0.03380000 0.08525000 0.001900000 0.02135000 0.02500000 0.3358500
## 2 0.05560000 0.05240000 0.06180000 0.012200000 0.07520000 0.01140000 0.2538000
## 3 0.01640000 0.02880000 0.08680000 0.014800000 0.00900000 0.05260000 0.2714000
## 4 0.01895833 0.04387500 0.10241667 0.002333333 0.04329167 0.02833333 0.3377083
## 5 0.01953846 0.05815385 0.10607692 0.022384615 0.01684615 0.02730769 0.3194615
## 6 0.02394444 0.04916667 0.09822222 0.009333333 0.01816667 0.02600000 0.2985556
##      into      is      it      its      may      more      must
## 1 0.02330000 0.1675500 0.1681500 0.04660000 0.06170000 0.04545000 0.02920000
## 2 0.02420000 0.1258000 0.1008000 0.05360000 0.02600000 0.05080000 0.01100000
## 3 0.04460000 0.0936000 0.2048000 0.03340000 0.05680000 0.08680000 0.02120000
## 4 0.01712500 0.1724167 0.1709167 0.05666667 0.06895833 0.03566667 0.03720833
## 5 0.02984615 0.1249231 0.1183846 0.04084615 0.05069231 0.04392308 0.03238462
## 6 0.02438889 0.1707222 0.1550556 0.04738889 0.07177778 0.04738889 0.04166667
##      my      no      not      now      of      on      one
## 1 0.002150000 0.03830000 0.09590000 0.006000000 0.9746500 0.08920000 0.03815000
## 2 0.005000000 0.02900000 0.04040000 0.007600000 0.8950000 0.07960000 0.04460000
## 3 0.001800000 0.01500000 0.10800000 0.006600000 0.6390000 0.07460000 0.08140000
## 4 0.003208333 0.03362500 0.09483333 0.005333333 0.9127917 0.04333333 0.03512500
## 5 0.002538462 0.02423077 0.08361538 0.008307692 1.0096154 0.05407692 0.04130769
## 6 0.005000000 0.03572222 0.10211111 0.004777778 0.8388889 0.08827778 0.03855556
##      only      or      our      shall      should      so      some
## 1 0.02600000 0.09765000 0.00715000 0.02180000 0.02600000 0.02510000 0.01570000
## 2 0.01100000 0.07320000 0.00720000 0.01180000 0.00700000 0.02180000 0.01780000
## 3 0.04340000 0.16080000 0.06600000 0.01740000 0.04140000 0.04460000 0.02140000
## 4 0.02125000 0.10008333 0.01804167 0.02175000 0.03425000 0.03000000 0.01683333
## 5 0.01792308 0.08992308 0.04330769 0.01476923 0.02300000 0.03038462 0.01984615
## 6 0.02277778 0.08494444 0.02500000 0.01655556 0.02083333 0.03255556 0.02883333
##      such      than      that      the      their      then      there
## 1 0.02825000 0.04755000 0.2320000 1.476150 0.07090000 0.005800000 0.02600000
## 2 0.02060000 0.03680000 0.1330000 1.337800 0.09840000 0.007800000 0.00820000
## 3 0.05120000 0.06280000 0.2434000 0.854400 0.14160000 0.008000000 0.01400000
## 4 0.03233333 0.03804167 0.2109583 1.332833 0.07741667 0.006708333 0.03520833
## 5 0.02192308 0.04346154 0.1880000 1.123000 0.07807692 0.002461538 0.03800000
## 6 0.02772222 0.04500000 0.2218889 1.210833 0.09883333 0.007166667 0.01511111
##      things      this      to      up      upon      was      were
## 1 0.003200000 0.09230000 0.5008500 0.000500000 0.02400000 0.02060000 0.01340000
## 2 0.001800000 0.06880000 0.4004000 0.005400000 0.01220000 0.08340000 0.03980000
## 3 0.001400000 0.05320000 0.4834000 0.000000000 0.00180000 0.02480000 0.02880000
## 4 0.002708333 0.08795833 0.6470000 0.005875000 0.04745833 0.02062500 0.01879167
## 5 0.004615385 0.09692308 0.5103077 0.006769231 0.03961538 0.02084615 0.02076923
## 6 0.001166667 0.08716667 0.4968889 0.001666667 0.01555556 0.02650000 0.02150000
##      what      when      which      who      will      with      would
## 1 0.01340000 0.013850000 0.1522000 0.02625000 0.12565000 0.07495000 0.11200000

```

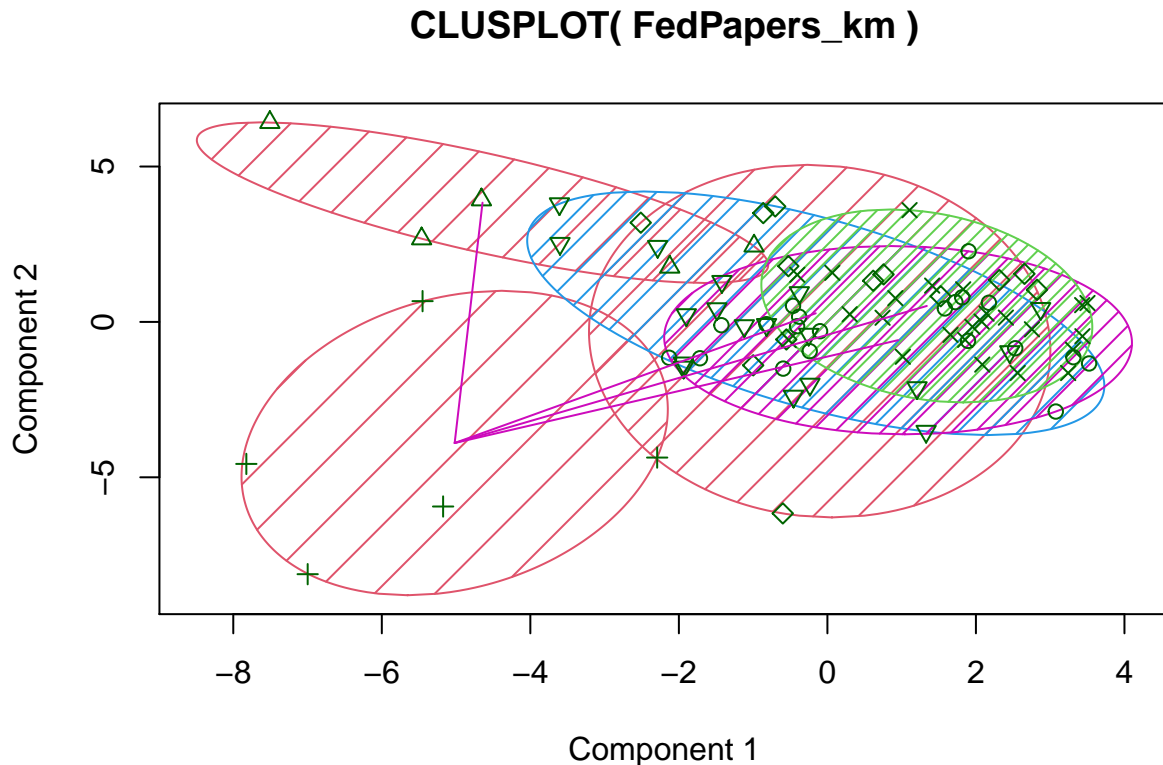
```
## 2 0.00520000 0.011400000 0.1484000 0.03820000 0.03660000 0.09600000 0.04120000
## 3 0.01840000 0.021000000 0.0986000 0.05160000 0.12600000 0.09500000 0.12520000
## 4 0.01475000 0.011916667 0.1616667 0.03833333 0.09187500 0.07570833 0.10587500
## 5 0.01284615 0.009769231 0.1603846 0.02130769 0.08169231 0.09515385 0.14823077
## 6 0.01033333 0.008111111 0.1758333 0.03300000 0.09955556 0.07027778 0.06144444
##      your
## 1 0.0000000000
## 2 0.0000000000
## 3 0.0064000000
## 4 0.0009166667
## 5 0.0007692308
## 6 0.0060000000
```

Add clusters to dataframe original dataframe with author name

```
FedPapers_km2 <- FederalistPapers
FedPapers_km2$Clusters <- as.factor(Clusters$cluster)
```

Plot results

```
clusplot(FedPapers_km, FedPapers_km$Clusters, color=TRUE, shade=TRUE, labels=0, lines=0)
clusplot(FedPapers_km, FedPapers_km$Clusters, color=TRUE, shade=TRUE, labels=0, lines=T)
```



These two components explain 16.39 % of the point variability.

word clouds based on authorship

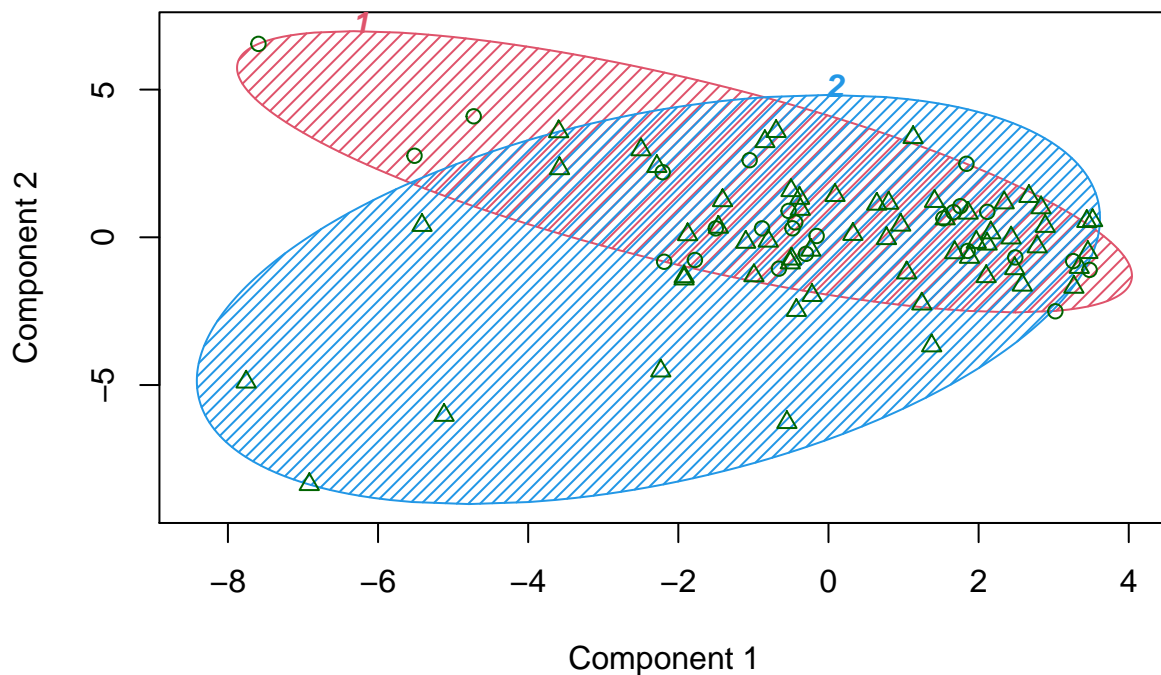
#Loop

```
cluster_loop <- c(2,3,4,5,6,7,8,9)
set.seed(20)
for (x in cluster_loop){
  print(x)
  # run k-means
  Clusters <- kmeans(FedPapers_km, x)
  FedPapers_km$Clusters <- as.factor(Clusters$cluster)
  str(Clusters)
  #print(Clusters$centers)
  # Plot results
  clusplot(FedPapers_km, FedPapers_km$Clusters, color=T, shade=T, labels=4, lines=T)
}
```

```
## [1] 2
## List of 9
## $ cluster      : Named int [1:85] 1 2 1 2 2 1 2 2 1 2 ...
##   .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:2, 1:71] 0.28084 0.2984 0.05568 0.05165 0.00972 ...
##   .. attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:2] "1" "2"
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 295
## $ withinss     : num [1:2] 7.2 65.8
```

```
## $ tot.withinss: num 73
## $ betweenss   : num 222
## $ size        : int [1:2] 25 60
## $ iter        : int 1
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

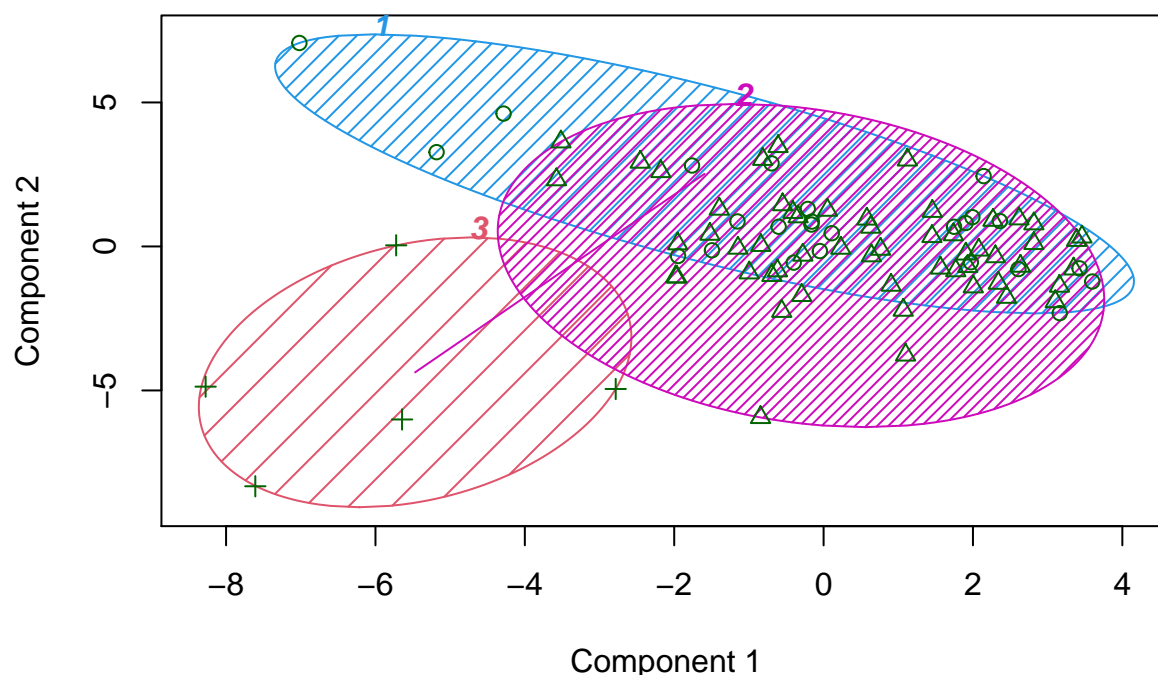
CLUSPLOT(FedPapers_km)



These two components explain 16.43 % of the point variability.

```
## [1] 3
## List of 9
## $ cluster      : Named int [1:85] 1 2 1 2 2 1 2 2 1 2 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:3, 1:71] 0.2808 0.311 0.1598 0.0557 0.0531 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:3] "1" "2" "3"
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 30.2
## $ withinss     : num [1:3] 3.197 5.543 0.599
## $ tot.withinss : num 9.34
## $ betweenss    : num 20.9
## $ size         : int [1:3] 25 55 5
## $ iter         : int 3
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

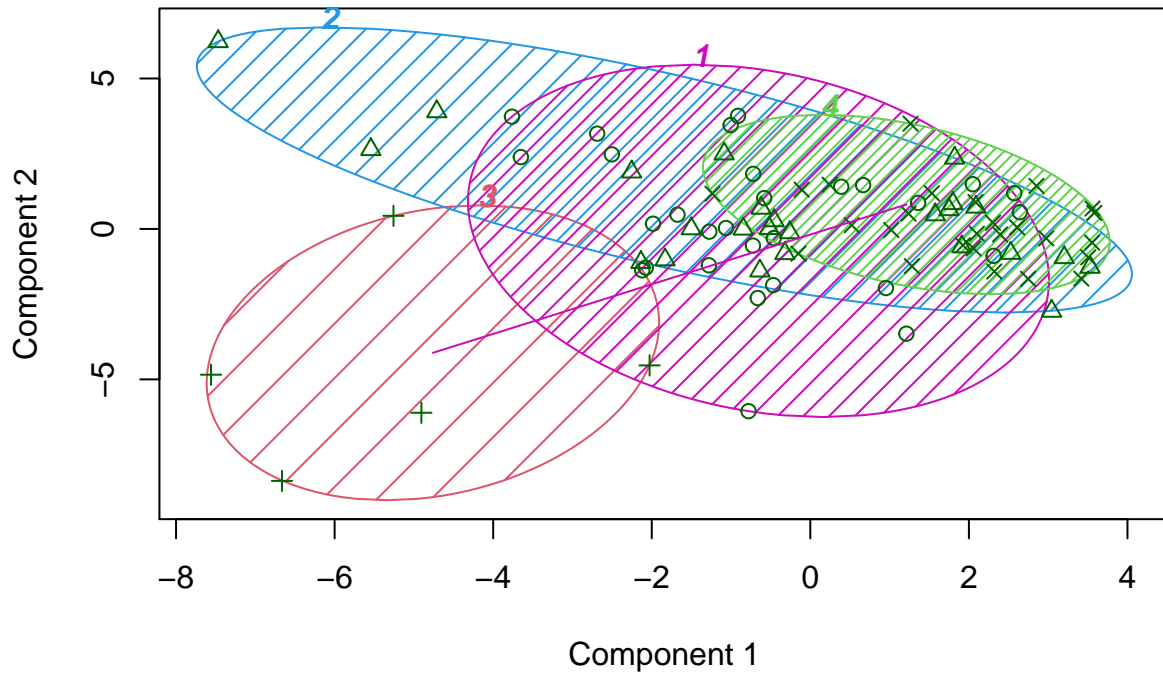
CLUSPLOT(FedPapers_km)



These two components explain 16.67 % of the point variability.

```
## [1] 4
## List of 9
## $ cluster      : Named int [1:85] 2 1 2 4 1 2 1 1 2 1 ...
##   .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:4, 1:71] 0.3168 0.2808 0.1598 0.3045 0.0537 ...
##   .. attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:4] "1" "2" "3" "4"
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss       : num 37.9
## $ withinss    : num [1:4] 2.713 3.197 0.599 2.017
## $ tot.withinss: num 8.53
## $ betweenss   : num 29.3
## $ size        : int [1:4] 29 25 5 26
## $ iter        : int 2
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

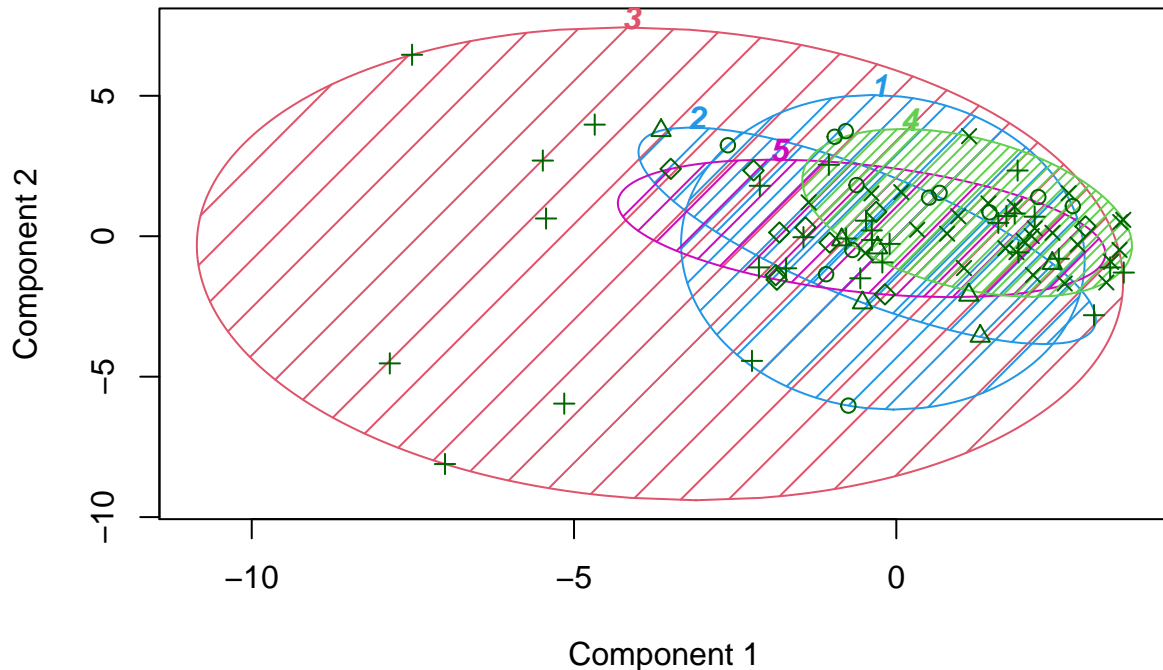
CLUSPLOT(FedPapers_km)



These two components explain 16.52 % of the point variability.

```
## [1] 5
## List of 9
## $ cluster      : Named int [1:85] 3 2 3 4 5 3 2 1 3 5 ...
##   .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:5, 1:71] 0.355 0.256 0.261 0.305 0.314 ...
##   .. attr(*, "dimnames")=List of 2
##     .. $ : chr [1:5] "1" "2" "3" "4" ...
##     .. $ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 141
## $ withinss      : num [1:5] 1.008 0.511 10.619 2.017 0.562
## $ tot.withinss : num 14.7
## $ betweenss     : num 127
## $ size          : int [1:5] 12 7 30 26 10
## $ iter          : int 3
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

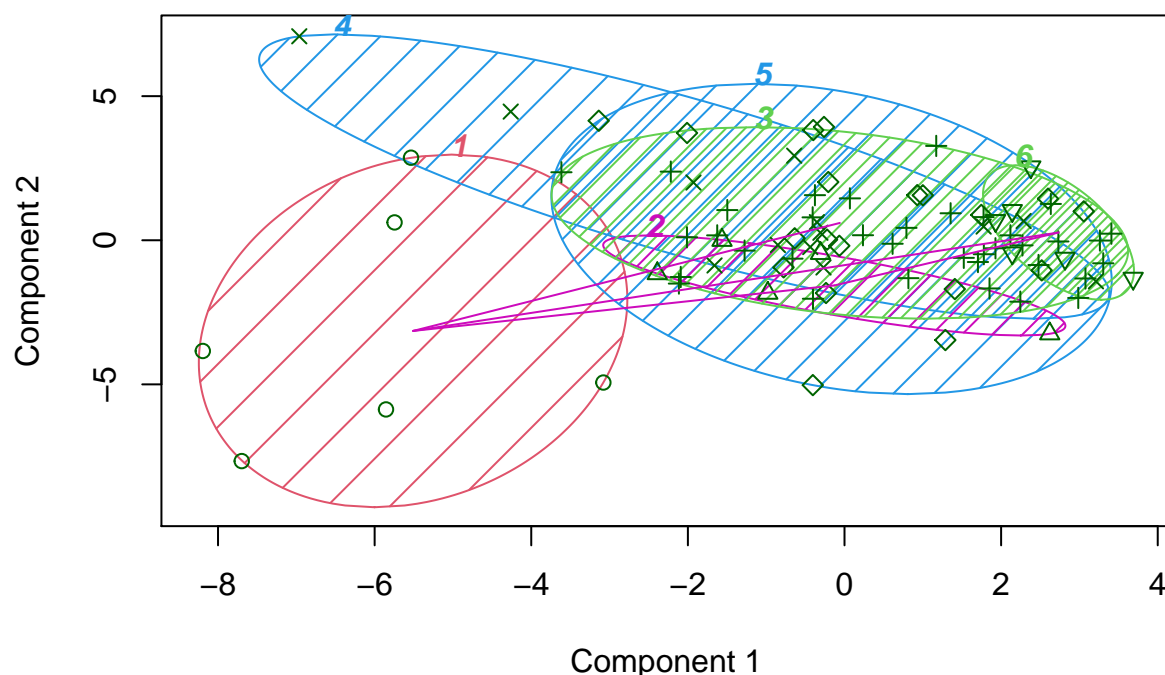
CLUSPLOT(FedPapers_km)



These two components explain 16.4 % of the point variability.

```
## [1] 6
## List of 9
## $ cluster      : Named int [1:85] 4 5 2 3 3 4 5 5 4 3 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:6, 1:71] 0.171 0.261 0.307 0.264 0.319 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:6] "1" "2" "3" "4" ...
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 131
## $ withinss      : num [1:6] 0.858 0.416 10.386 1.234 6.293 ...
## $ tot.withinss  : num 19.6
## $ betweenss     : num 111
## $ size          : int [1:6] 6 5 36 13 19 6
## $ iter          : int 2
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

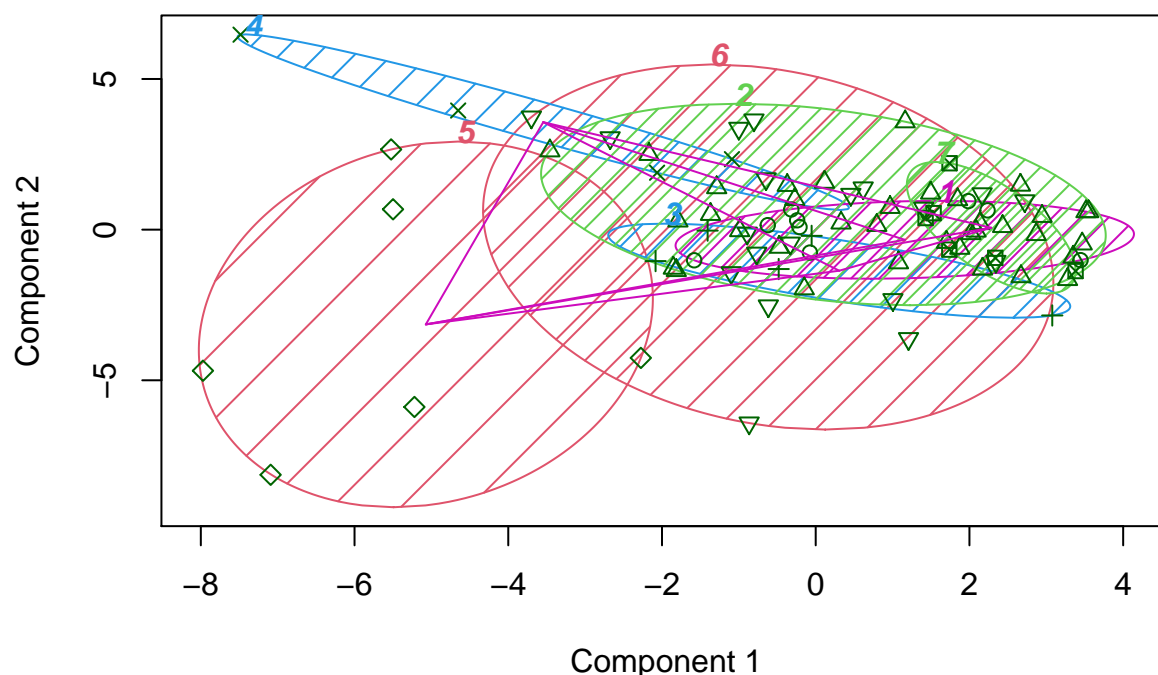

CLUSPLOT(FedPapers_km)



These two components explain 16.74 % of the point variability.

```
## [1] 7
## List of 9
## $ cluster      : Named int [1:85] 1 6 3 2 2 1 6 6 1 2 ...
##   .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:7, 1:71] 0.287 0.307 0.261 0.213 0.171 ...
##   .. attr(*, "dimnames")=List of 2
##     .. $ : chr [1:7] "1" "2" "3" "4" ...
##     .. $ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 153
## $ withinss      : num [1:7] 0.651 3.164 0.416 0.264 0.858 ...
## $ tot.withinss : num 7.61
## $ betweenss     : num 145
## $ size          : int [1:7] 9 36 5 4 6 19 6
## $ iter          : int 2
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

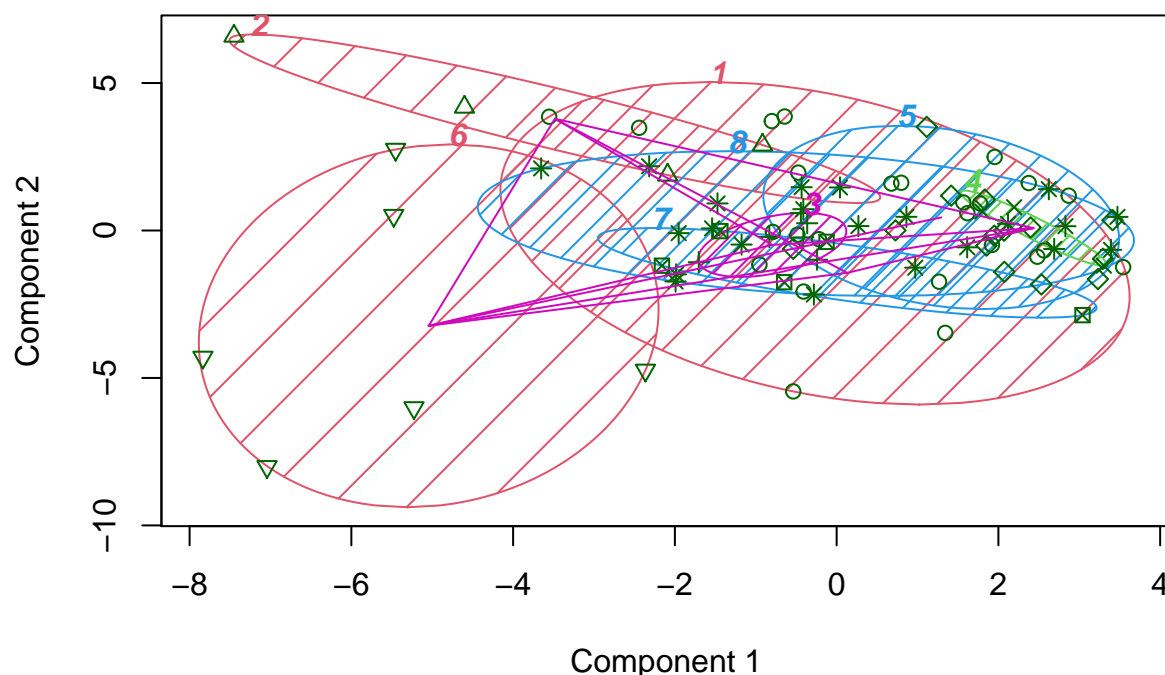
CLUSPLOT(FedPapers_km)



These two components explain 16.43 % of the point variability.

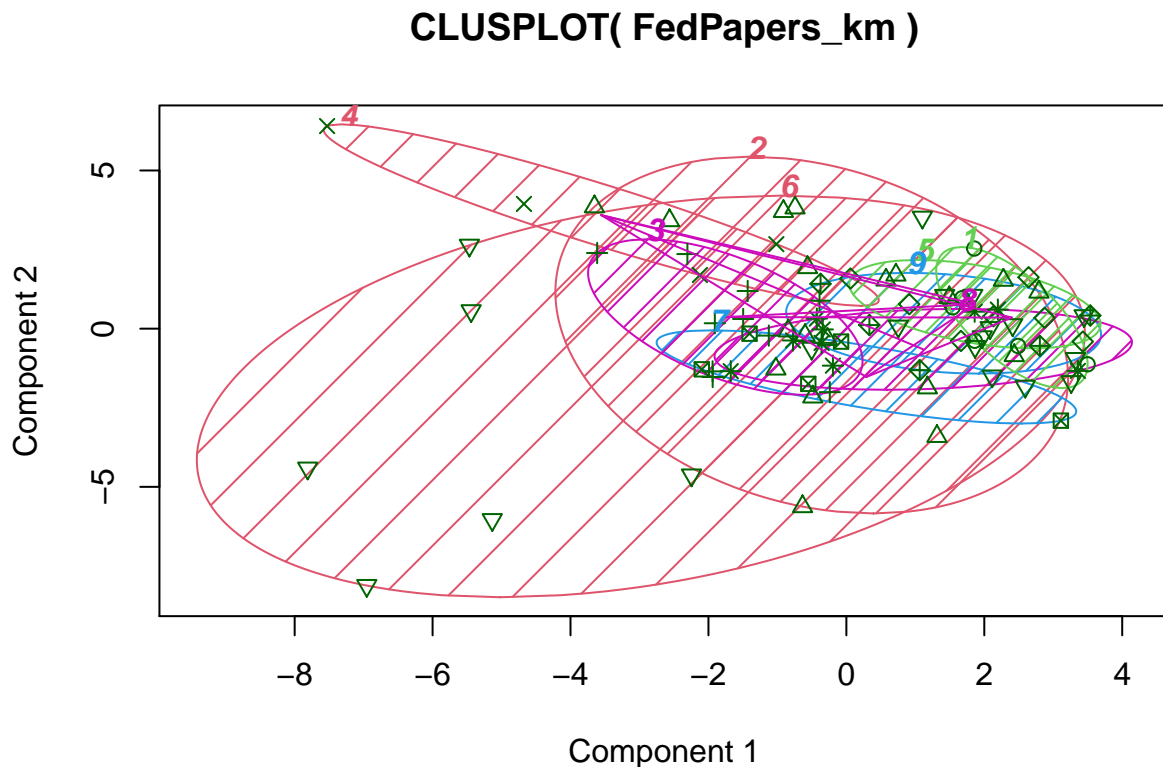
```
## [1] 8
## List of 9
## $ cluster      : Named int [1:85] 3 1 7 8 8 3 1 1 3 8 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:8, 1:71] 0.324 0.213 0.277 0.305 0.263 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. .$ : chr [1:8] "1" "2" "3" "4" ...
##     .. .$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 358
## $ withinss      : num [1:8] 7.64 0.264 0.325 0.18 0.925 ...
## $ tot.withinss : num 12.3
## $ betweenss     : num 346
## $ size          : int [1:8] 25 4 6 3 14 6 5 22
## $ iter          : int 2
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

CLUSPLOT(FedPapers_km)



These two components explain 16.43 % of the point variability.

```
## [1] 9
## List of 9
## $ cluster      : Named int [1:85] 8 2 7 3 3 8 2 2 8 3 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:9, 1:71] 0.343 0.319 0.305 0.213 0.37 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. .$ : chr [1:9] "1" "2" "3" "4" ...
##     .. .$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 676
## $ withinss     : num [1:9] 0.388 1.872 0.537 0.264 0.386 ...
## $ tot.withinss : num 15
## $ betweenss    : num 661
## $ size         : int [1:9] 6 19 10 4 7 20 5 9 5
## $ iter         : int 2
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```



These two components explain 16.43 % of the point variability.

Hierarchical Clustering Algorithms (HAC)

Remove author names from dataset

```
FedPapers_HAC <- FederalistPapers[,c(2:72)]
```

Make the file names the row names. Need a dataframe of numerical values for HAC

```
rownames(FedPapers_HAC) <- FedPapers_HAC[,1]
FedPapers_HAC[,1] <- NULL
View(FedPapers_HAC)
```

Calculate distance in a variety of ways

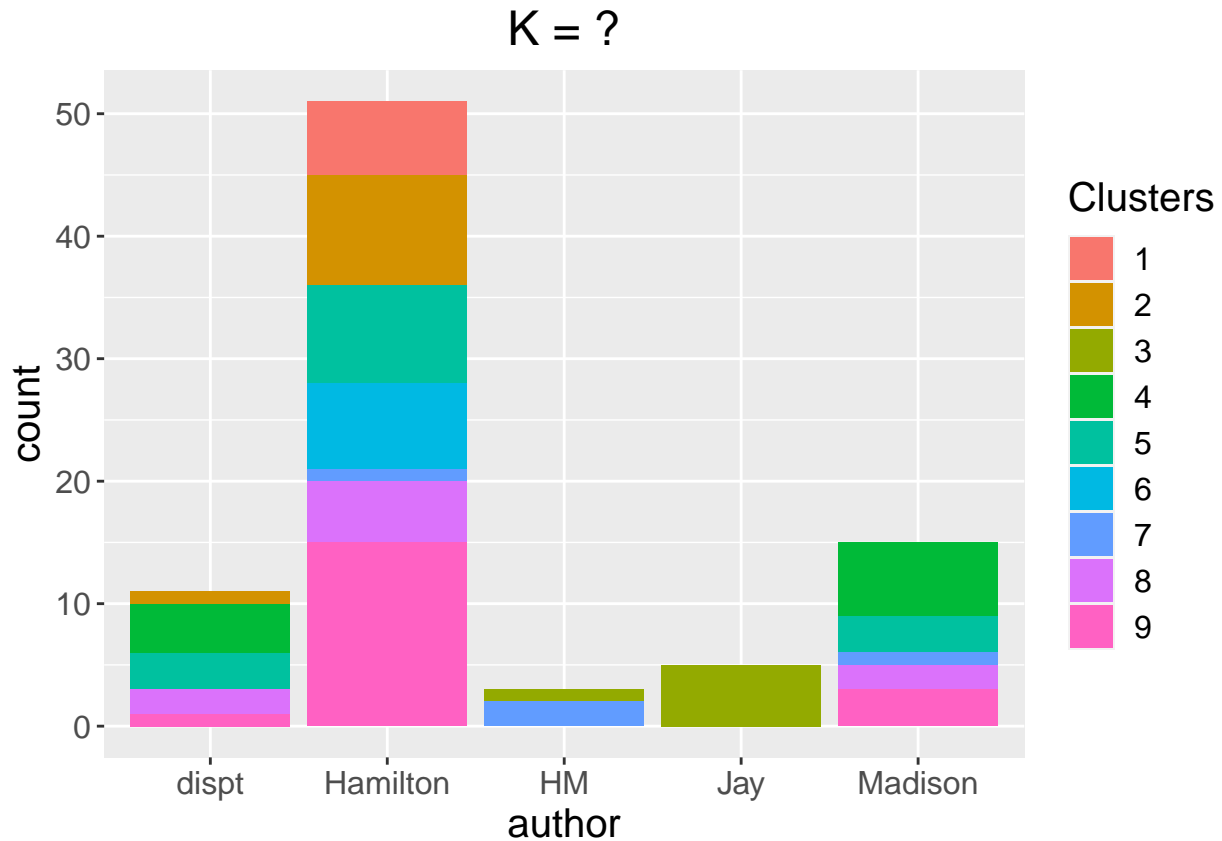
```
distance <- dist(FedPapers_HAC, method = "euclidean")
distance2 <- dist(FedPapers_HAC, method = "maximum")
distance3 <- dist(FedPapers_HAC, method = "manhattan")
```

```
distance4 <- dist(FedPapers_HAC, method = "canberra")
distance5 <- dist(FedPapers_HAC, method = "binary")
distance6 <- dist(FedPapers_HAC, method = "minkowski", p = 3)
```

```
Clusters1 <- kmeans(FedPapers_km, 9)
FedPapers_km2$Clusters <- as.factor(Clusters1$cluster)
str(Clusters)
```

```
## List of 9
## $ cluster      : Named int [1:85] 8 2 7 3 3 8 2 2 8 3 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:9, 1:71] 0.343 0.319 0.305 0.213 0.37 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:9] "1" "2" "3" "4" ...
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss       : num 676
## $ withinss    : num [1:9] 0.388 1.872 0.537 0.264 0.386 ...
## $ tot.withinss: num 15
## $ betweenss   : num 661
## $ size        : int [1:9] 6 19 10 4 7 20 5 9 5
## $ iter        : int 2
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

```
ggplot(data=FedPapers_km2, aes(x=author, fill=Clusters))+
  geom_bar(stat="count") +
  labs(title = "K = ?") +
  theme(plot.title = element_text(hjust=0.5), text=element_text(size=15))
```



Madison essays

```
Madison_Leaning <- FederalistPapers[which(FedPapers_km2$Clusters[c(1:11)]== 8 | FedPapers_km$Clusters[c(1:11)]== 8), ]
Madison_Leaning
```

```
## [1] "dispt_fed_50.txt" "dispt_fed_52.txt" "dispt_fed_53.txt" "dispt_fed_55.txt"
## [5] "dispt_fed_62.txt" "dispt_fed_63.txt"
```

A loop to plot multiple HACs

```
hac_loop <- c(2,3,4,5,6,7,8,9)
for (y in hac_loop) {
  HAC <- hclust(distance, method="complete")
  plot(HAC, cex=0.6, hang=-1, main = c("HAC Cluster Euclidean Complete", y, "Clusters"))
  rect.hclust(HAC, k = y, border=2:5)

  HACSingle <- hclust(distance, method="single")
  plot(HACSingle, cex=0.6, hang=-1, main = c("HAC Cluster Euclidean Single", y, "Clusters"))
  rect.hclust(HACSingle, k = y, border=2:5)

  HAC2 <- hclust(distance2, method="complete")
  plot(HAC2, cex=.1, hang=-1, main = c("HAC Cluster Maximum Complete", y, "Clusters"))
  rect.hclust(HAC2, k =y, border=2:5)
```

```

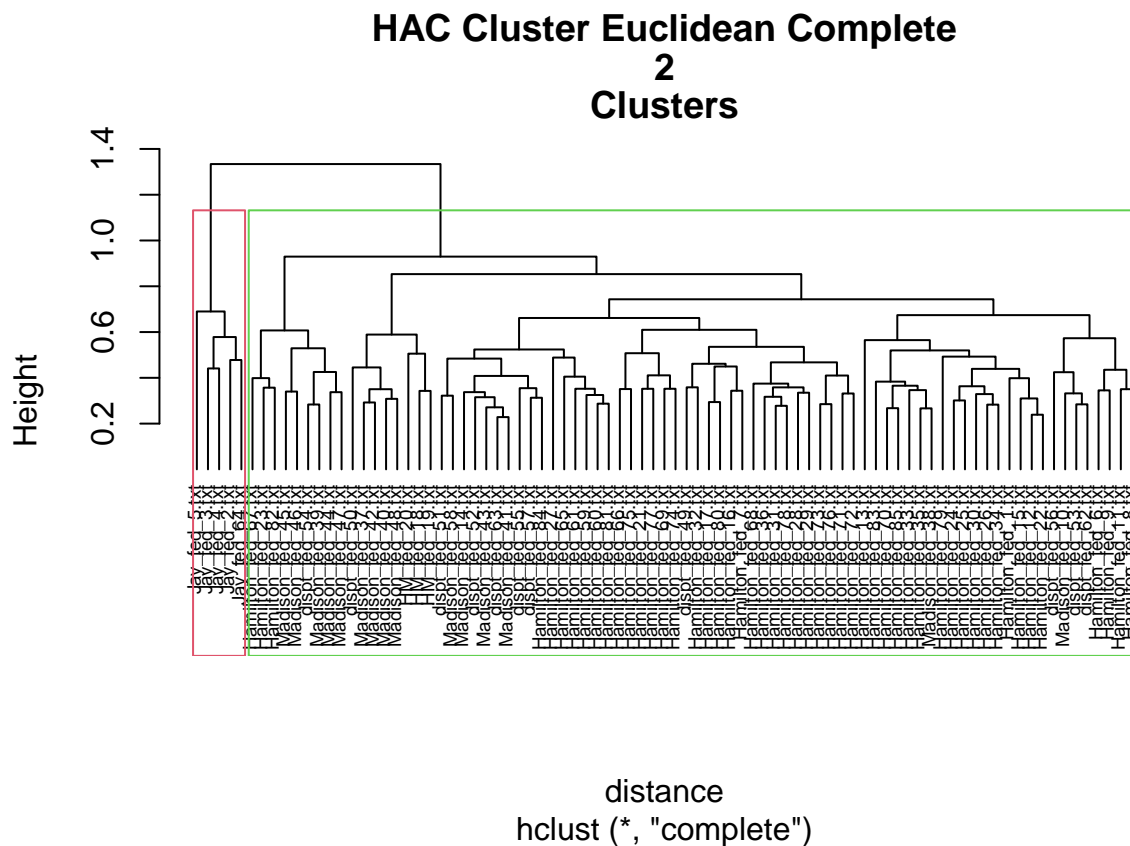
HAC3 <- hclust(distance3, method="complete")
plot(HAC3, cex=0.6, hang=-1, main = c("HAC Cluster Manhattan Complete", y, "Clusters"))
rect.hclust(HAC3, k =y, border=2:5)

HAC4 <- hclust(distance4, method="complete")
plot(HAC4, cex=0.6, hang=-1, main = c("HAC Cluster Canberra Complete", y, "Clusters"))
rect.hclust(HAC4, k =y, border=2:5)

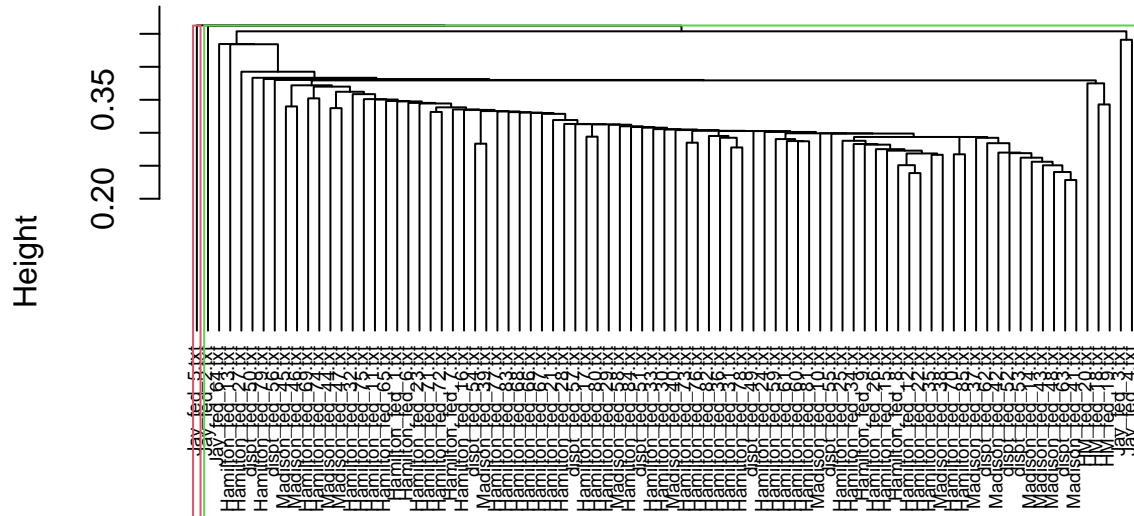
HAC5 <- hclust(distance5, method="complete")
plot(HAC5, cex=0.6, hang=-1, main = c("HAC Cluster Minkowski Complete", y, "Clusters"))
rect.hclust(HAC5, k =y, border=2:5)

HAC6 <- hclust(distance6, method="complete")
plot(HAC6, cex=0.6, hang=-1, main = c("HAC Cluster Maximum Complete", y, "Clusters"))
rect.hclust(HAC6, k =y, border=2:5)
}

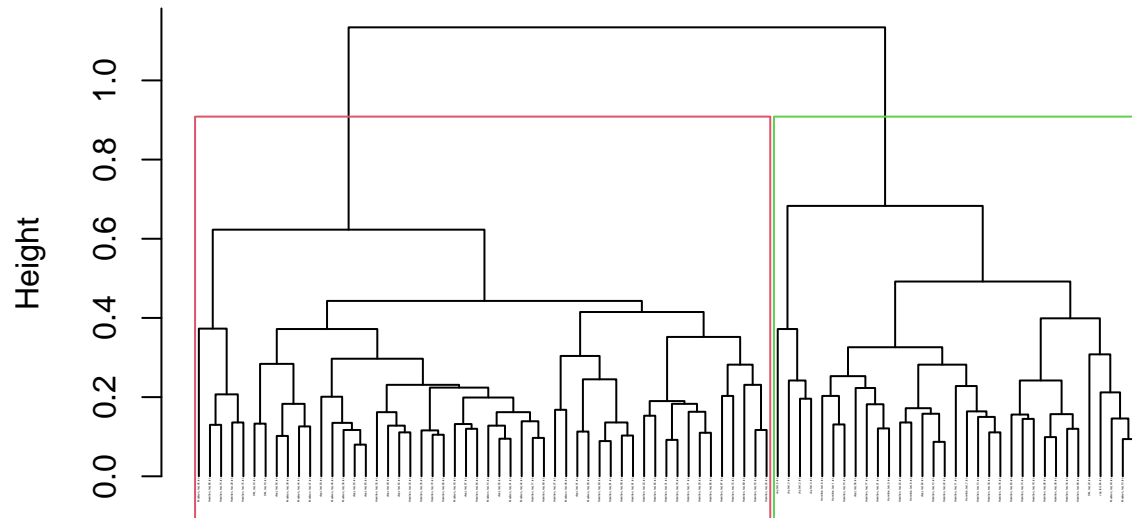
```



HAC Cluster Euclidean Single 2 Clusters

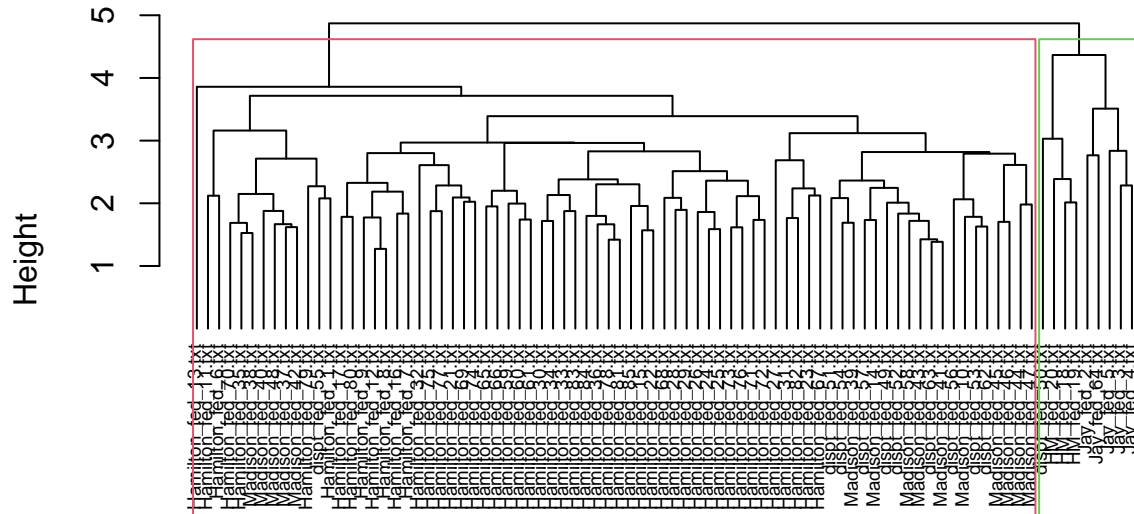


HAC Cluster Maximum Complete 2 Clusters



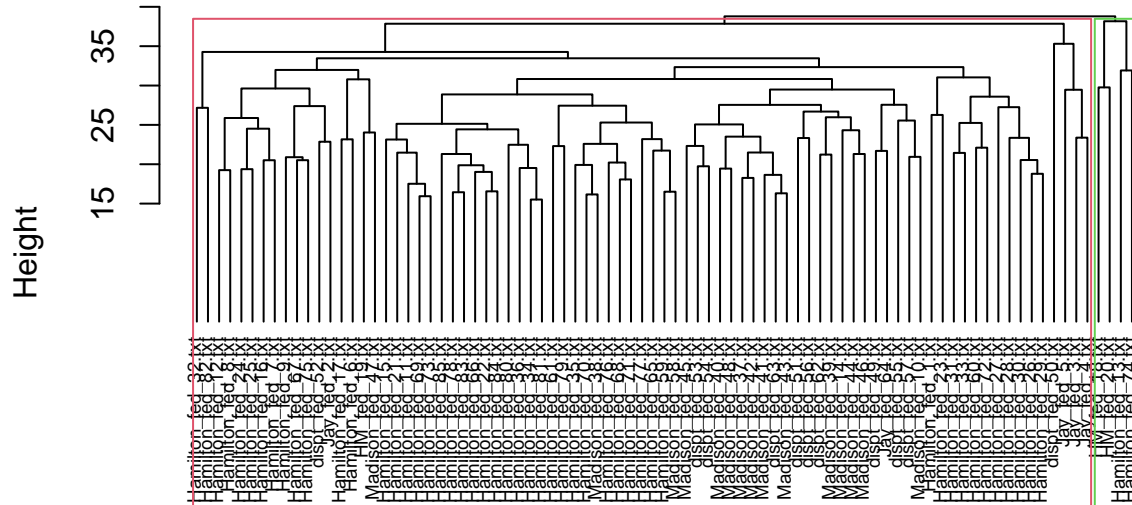
distance2
hclust (*, "complete")

HAC Cluster Manhattan Complete 2 Clusters



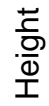
distance3
hclust (*, "complete")

HAC Cluster Canberra Complete 2 Clusters



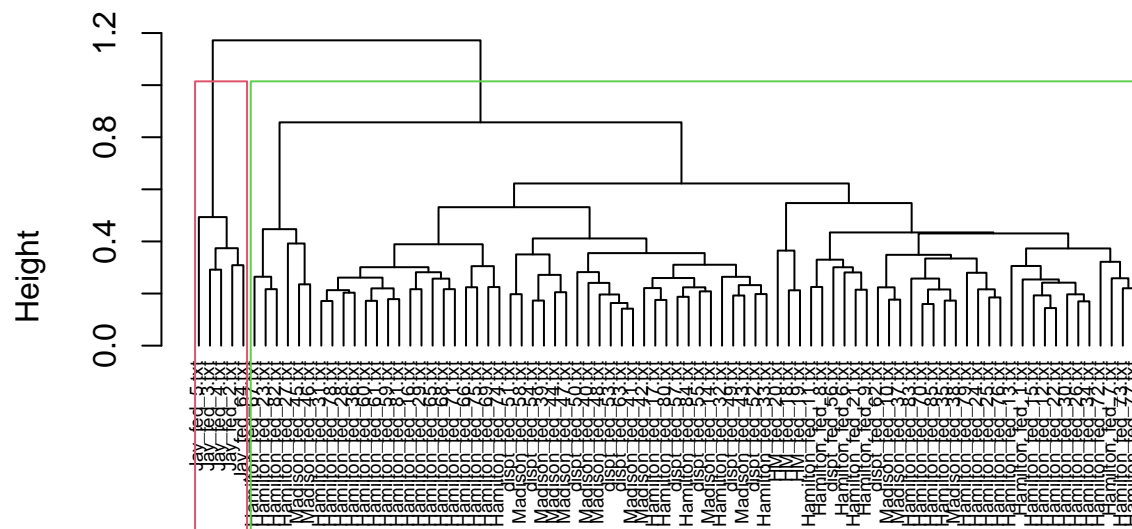
distance4
hclust (*, "complete")

2 Clusters



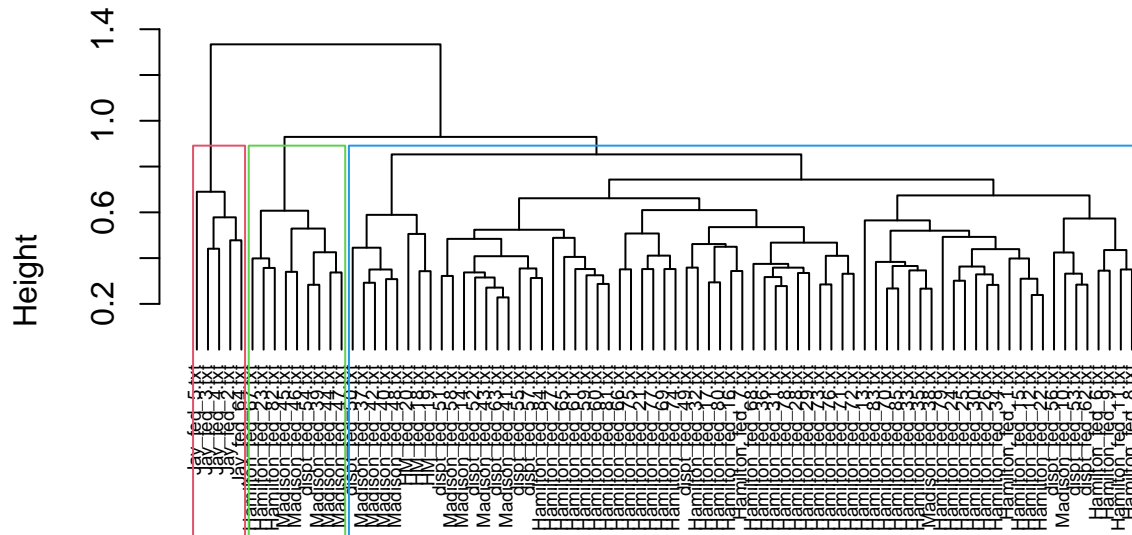
```
distance5
hclust (*, "complete")
```

HAC Cluster Maximum Complete 2 Clusters

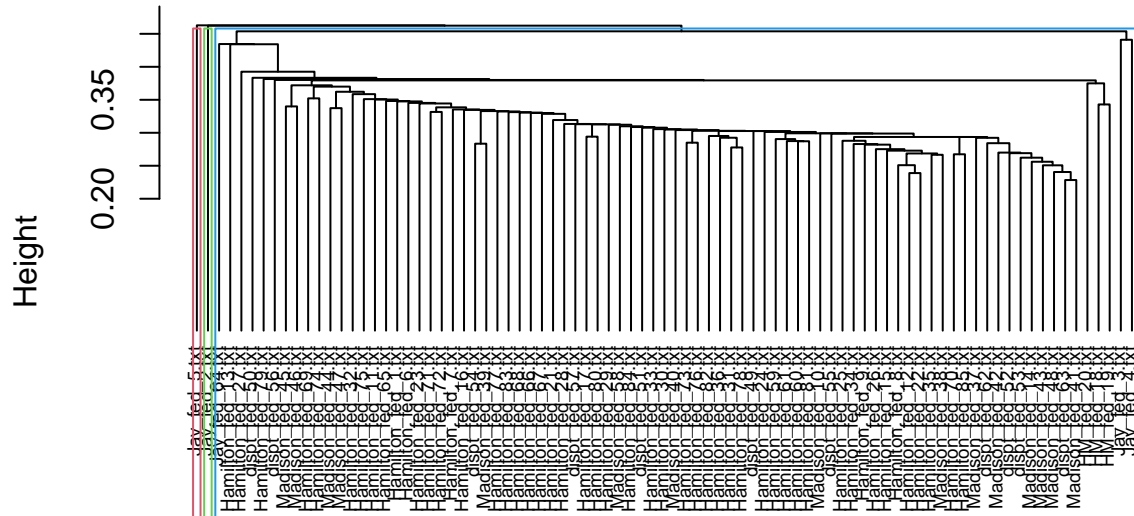


distance6
hclust (*, "complete")

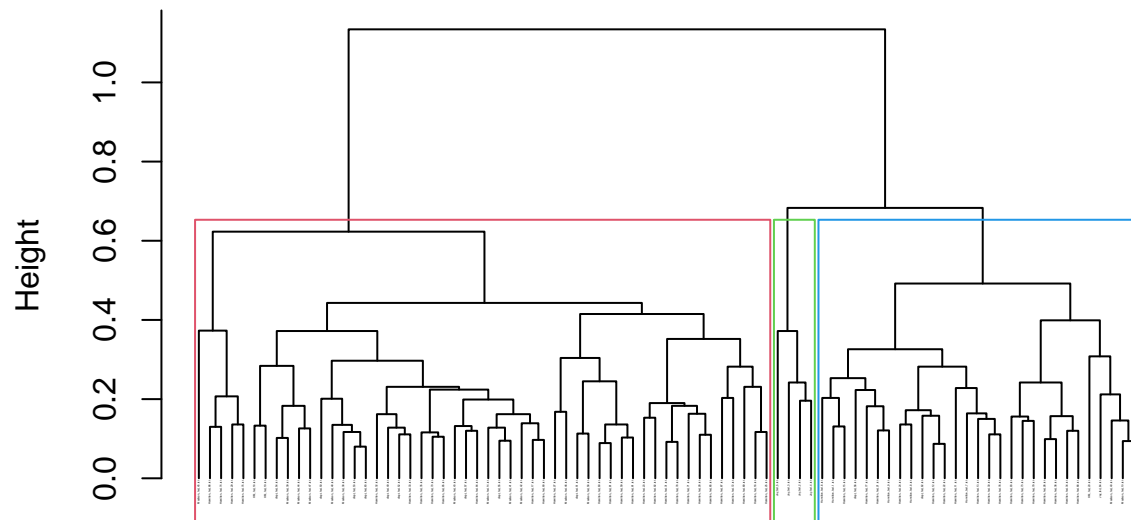
HAC Cluster Euclidean Complete 3 Clusters



HAC Cluster Euclidean Single 3 Clusters

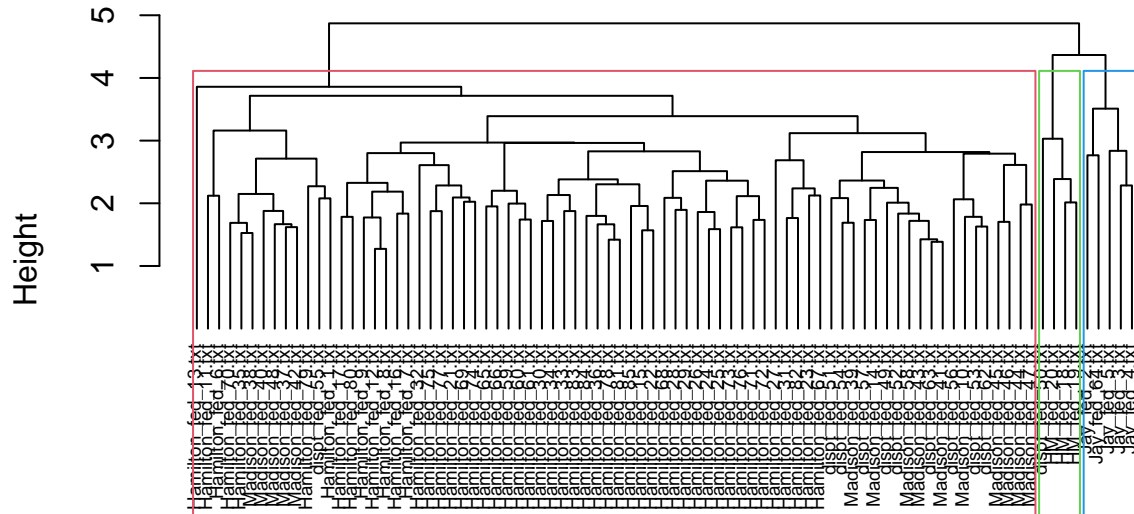


HAC Cluster Maximum Complete 3 Clusters



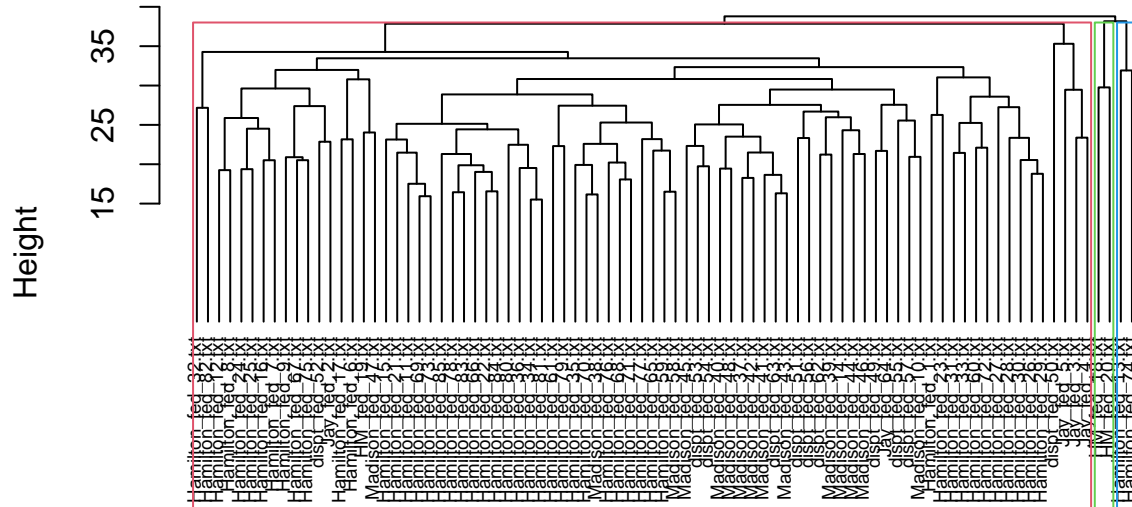
distance2
hclust (*, "complete")

HAC Cluster Manhattan Complete 3 Clusters



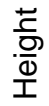
distance3
hclust (*, "complete")

HAC Cluster Canberra Complete 3 Clusters



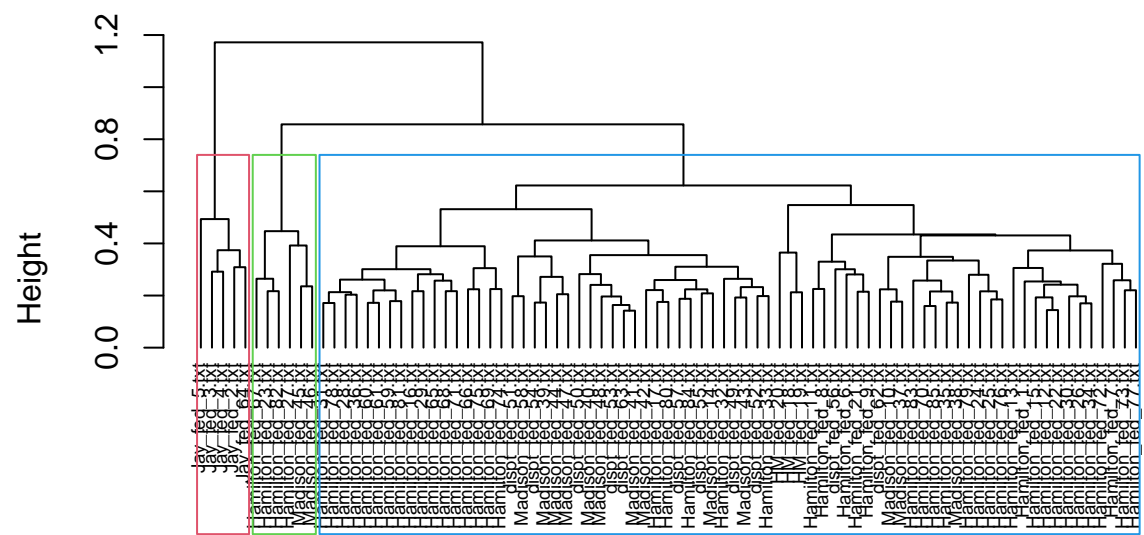
distance4
hclust (*, "complete")

3 Clusters



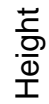
```
distance5
hclust (*, "complete")
```

HAC Cluster Maximum Complete
3
Clusters



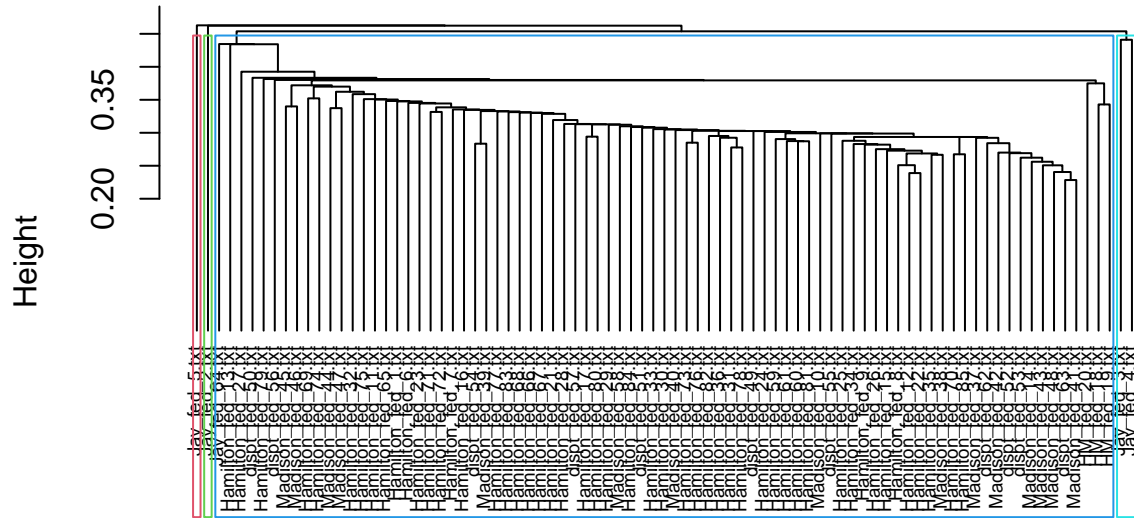
distance6
hclust (*, "complete")

4 Clusters



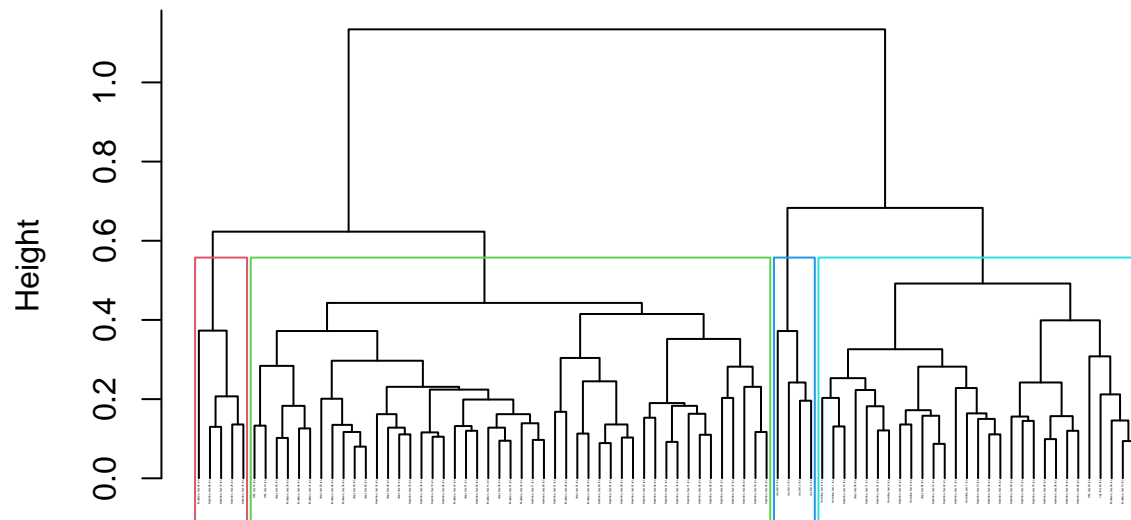
```
distance
hclust (*, "complete")
```

HAC Cluster Euclidean Single 4 Clusters



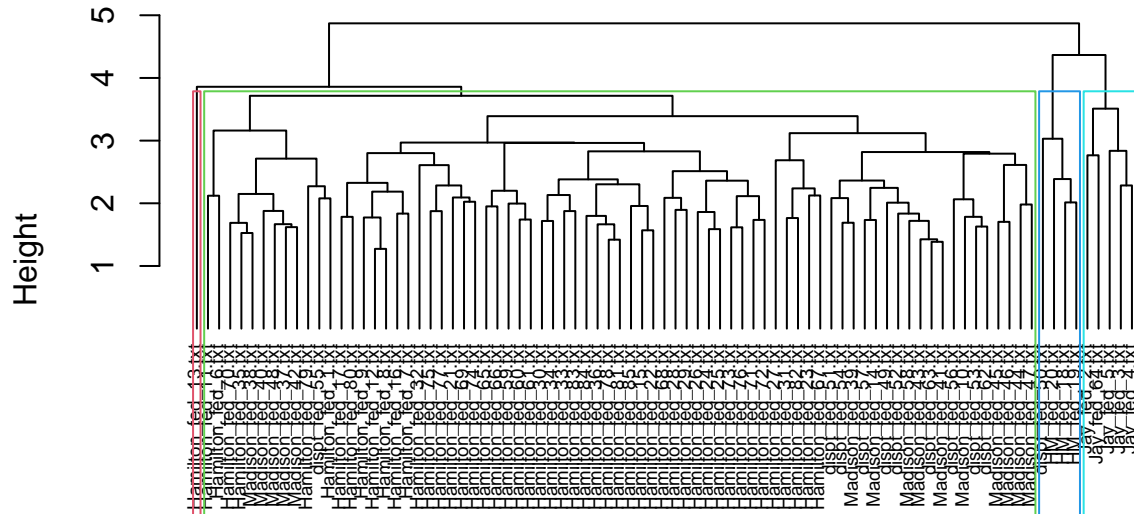
distance
hclust (*, "single")

HAC Cluster Maximum Complete 4 Clusters



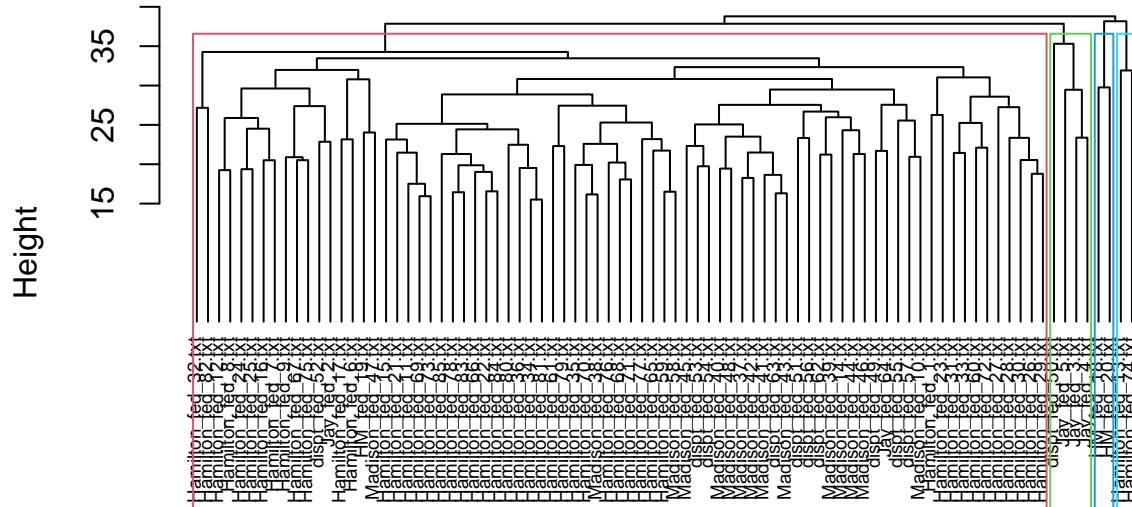
distance2
hclust (*, "complete")

HAC Cluster Manhattan Complete 4 Clusters



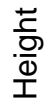
distance3
hclust (*, "complete")

HAC Cluster Canberra Complete 4 Clusters



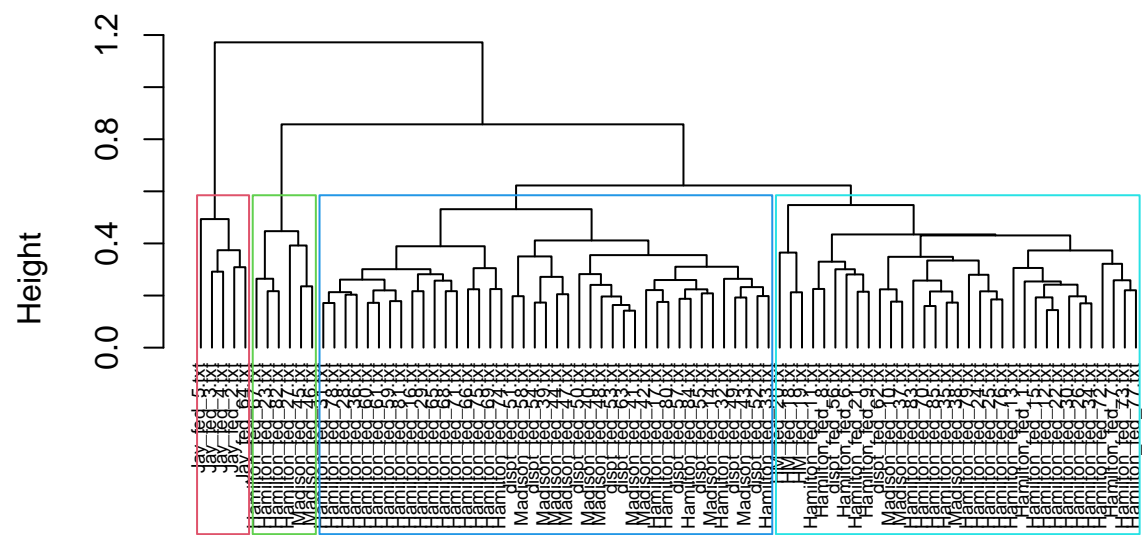
distance4
hclust (*, "complete")

4 Clusters



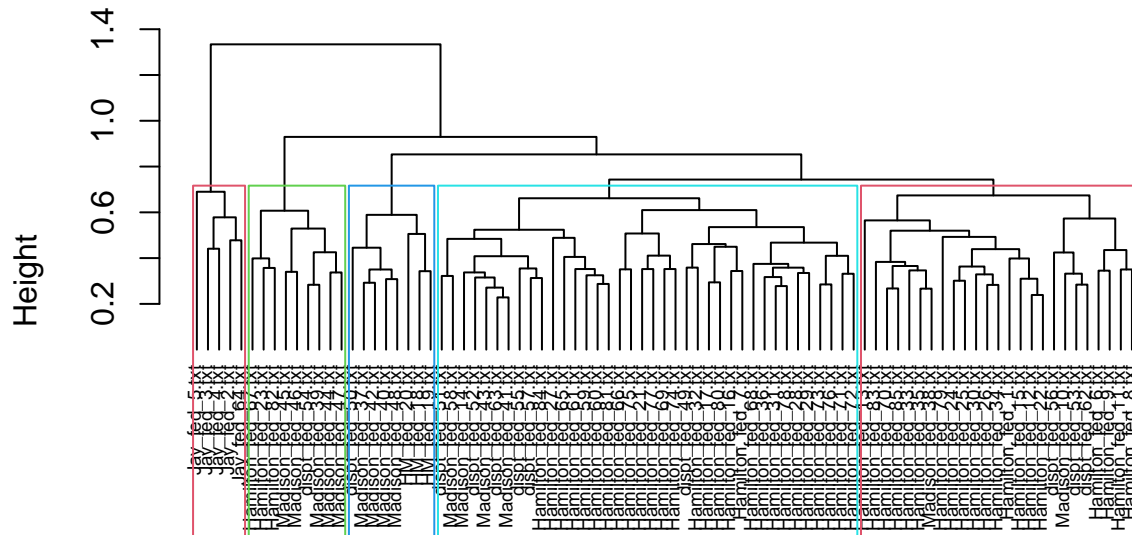
```
distance5
hclust (*, "complete")
```

HAC Cluster Maximum Complete
4
Clusters



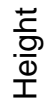
distance6
hclust (*, "complete")

HAC Cluster Euclidean Complete 5 Clusters



distance
hclust (*, "complete")

5 Clusters



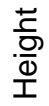
```
distance
hclust (*, "single")
```

5 Clusters



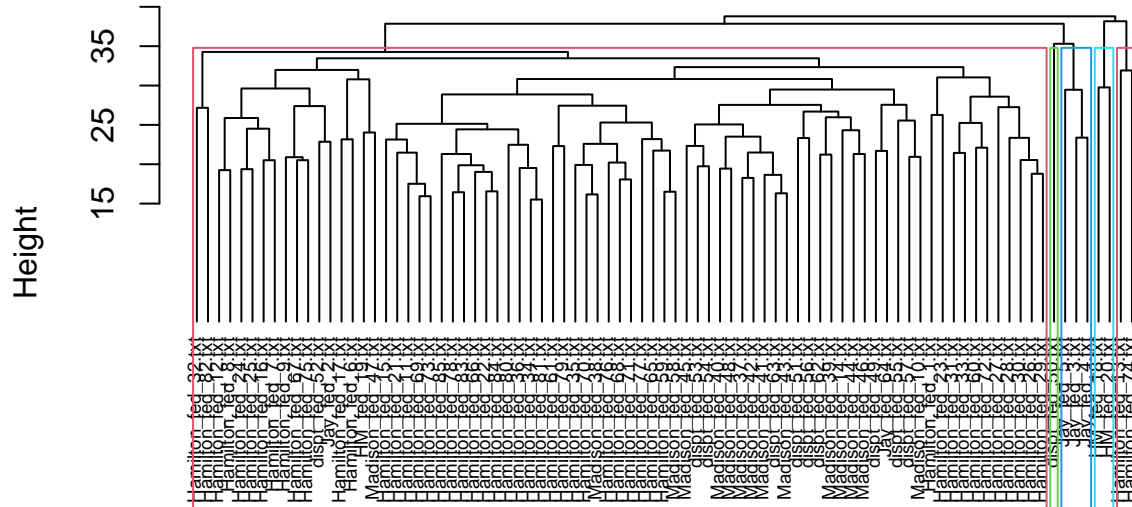
```
distance2
hclust (*, "complete")
```

5 Clusters

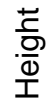


```
distance3
hclust (*, "complete")
```

HAC Cluster Canberra Complete 5 Clusters

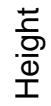


5 Clusters



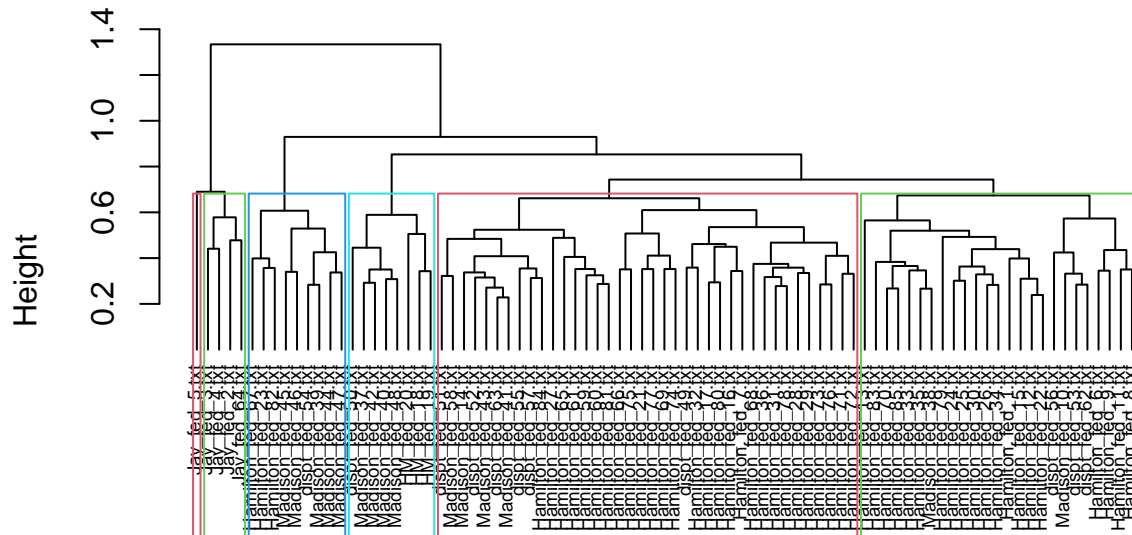
```
distance5
hclust (*, "complete")
```

5 Clusters



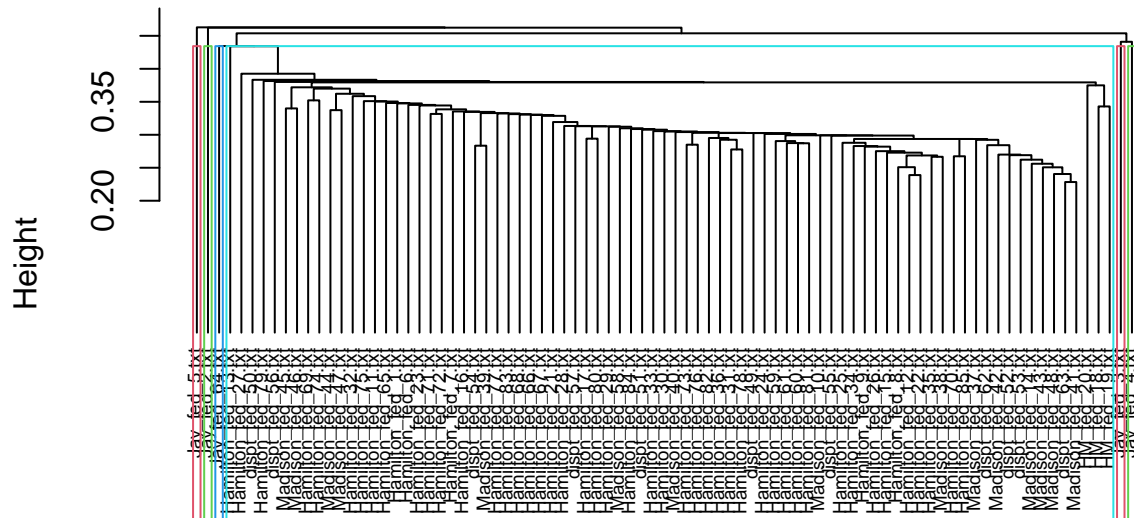
```
distance6
hclust (*, "complete")
```

HAC Cluster Euclidean Complete 6 Clusters

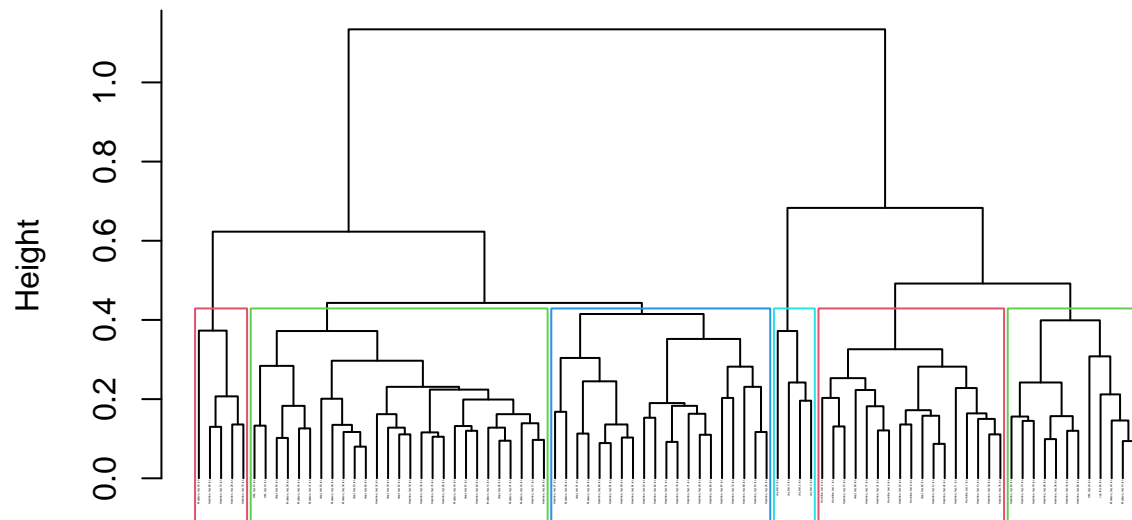


distance
hclust (*, "complete")

HAC Cluster Euclidean Single 6 Clusters

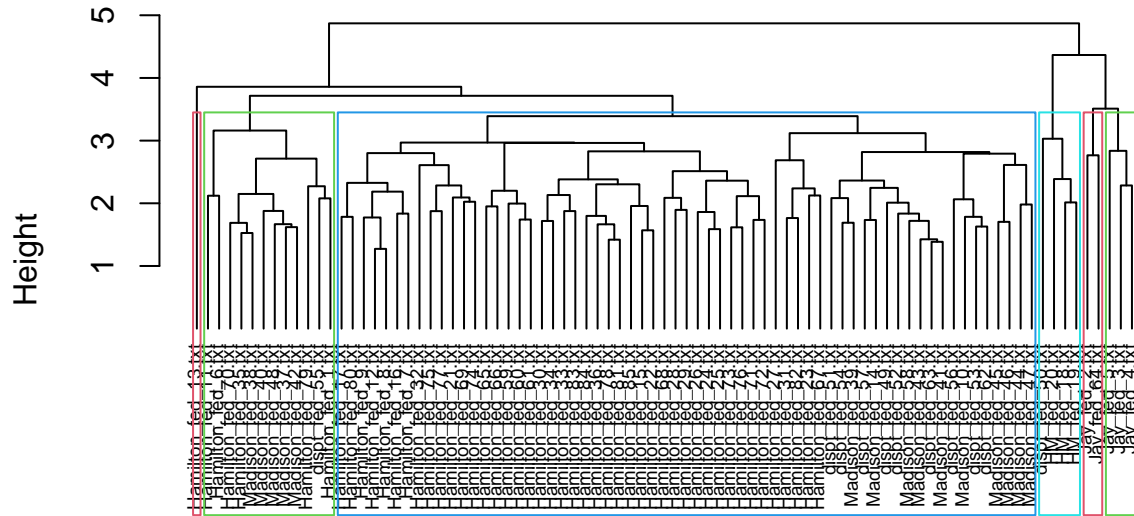


HAC Cluster Maximum Complete 6 Clusters



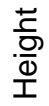
distance2
hclust (*, "complete")

HAC Cluster Manhattan Complete 6 Clusters



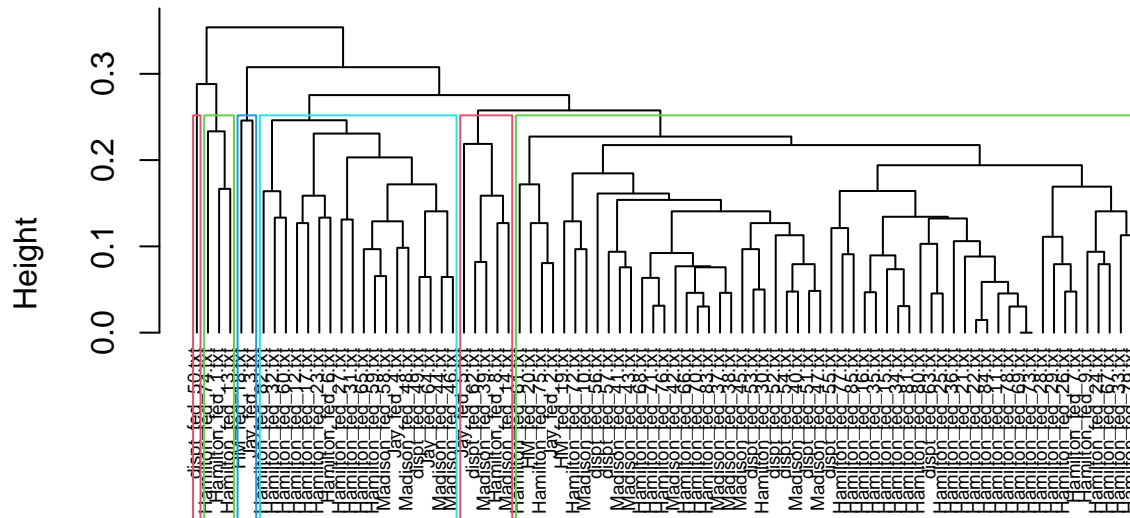
distance3
hclust (*, "complete")

6 Clusters



```
distance4
hclust (*, "complete")
```

HAC Cluster Minkowski Complete 6 Clusters



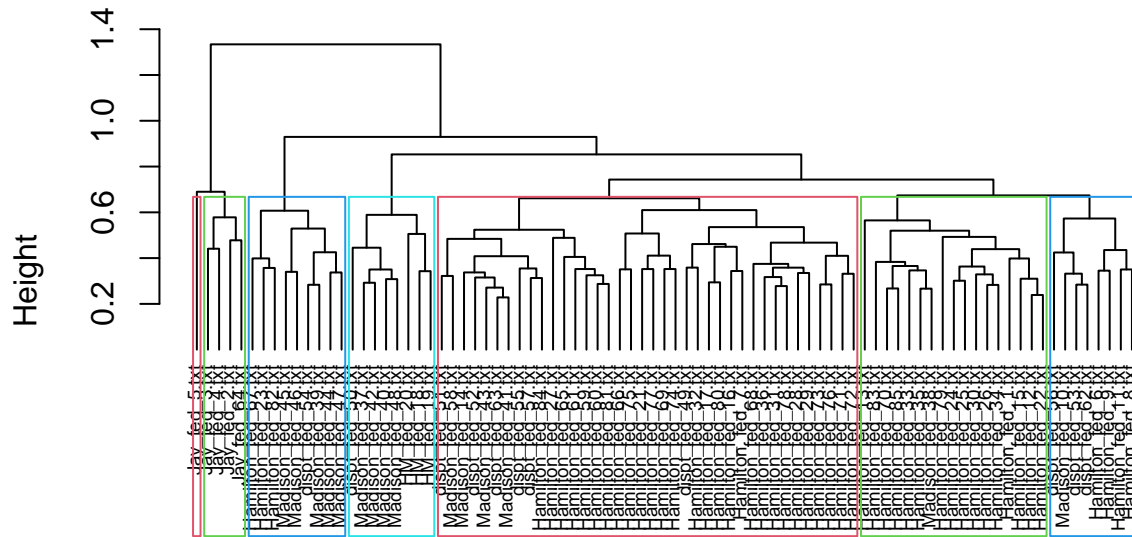
distance5
hclust (*, "complete")

6 Clusters

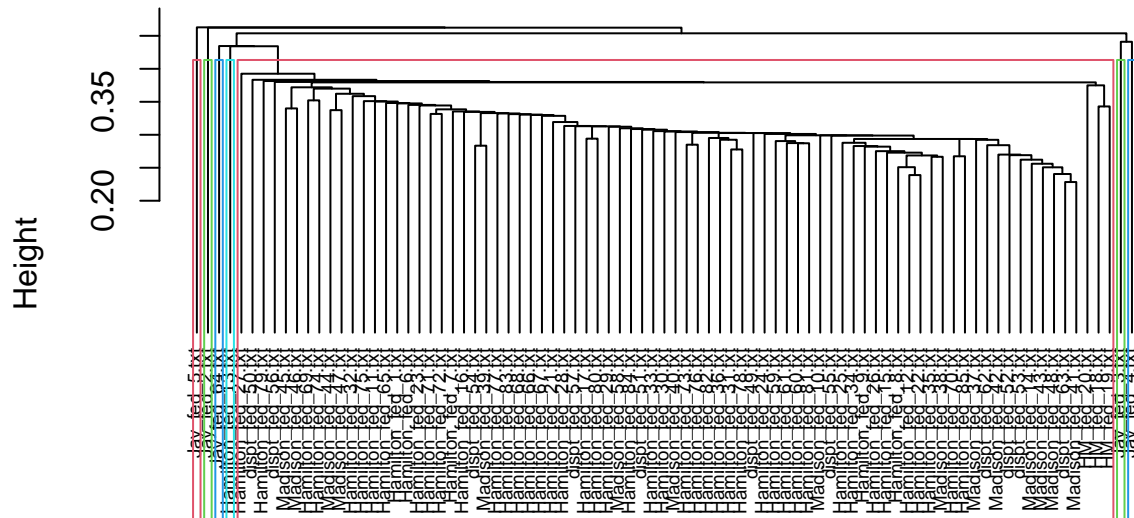


```
distance6
hclust (*, "complete")
```

HAC Cluster Euclidean Complete 7 Clusters

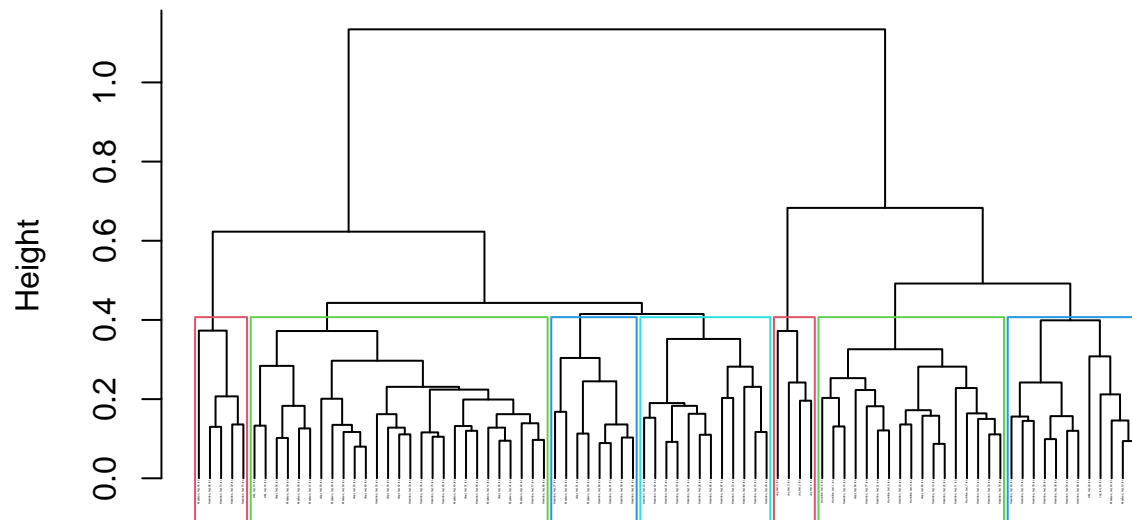


HAC Cluster Euclidean Single 7 Clusters



distance
hclust (*, "single")

HAC Cluster Maximum Complete 7 Clusters



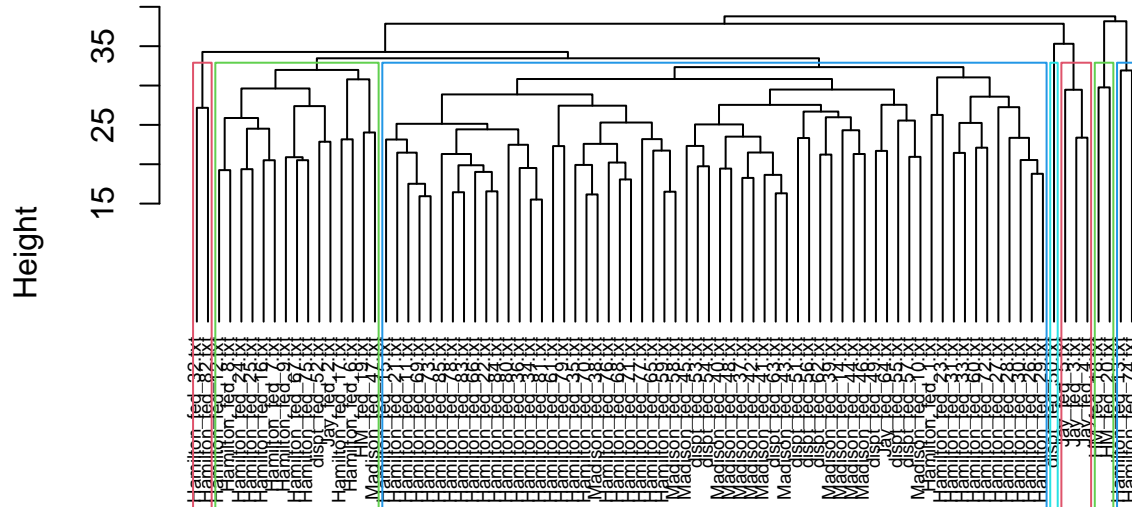
distance2
hclust (*, "complete")

7 Clusters



```
distance3
hclust (*, "complete")
```

HAC Cluster Canberra Complete 7 Clusters



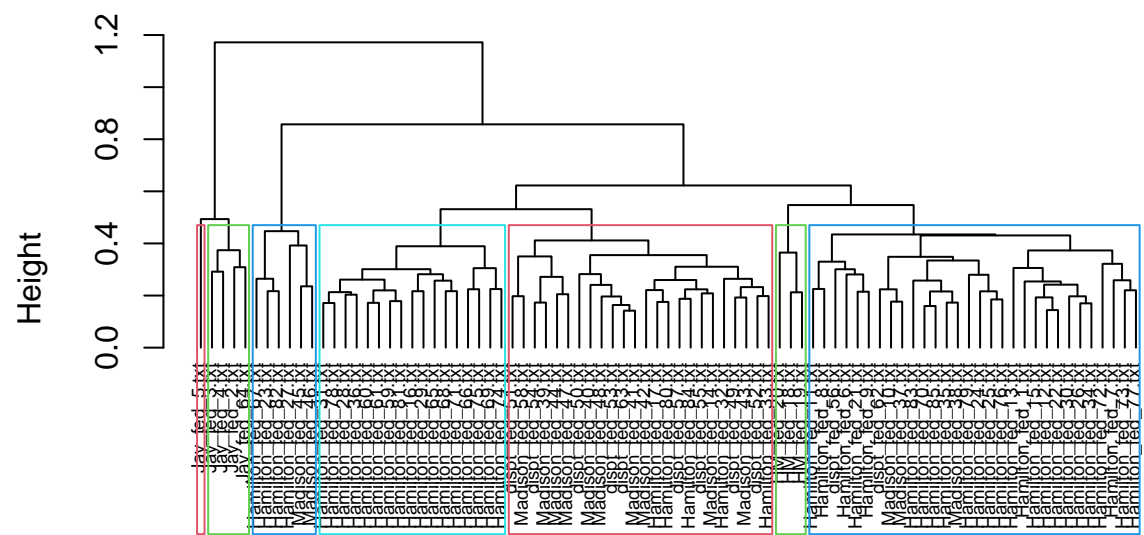
distance4
hclust (*, "complete")

7 Clusters

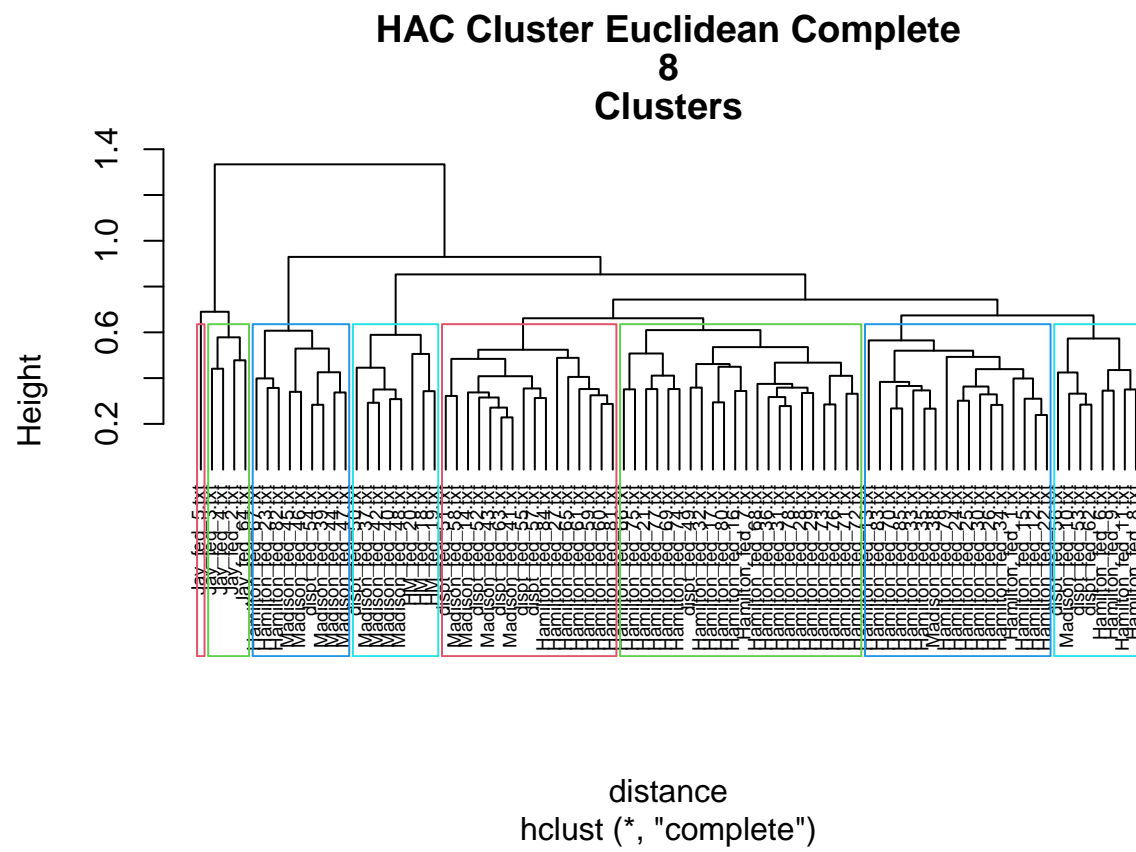


```
distance5
hclust (*, "complete")
```

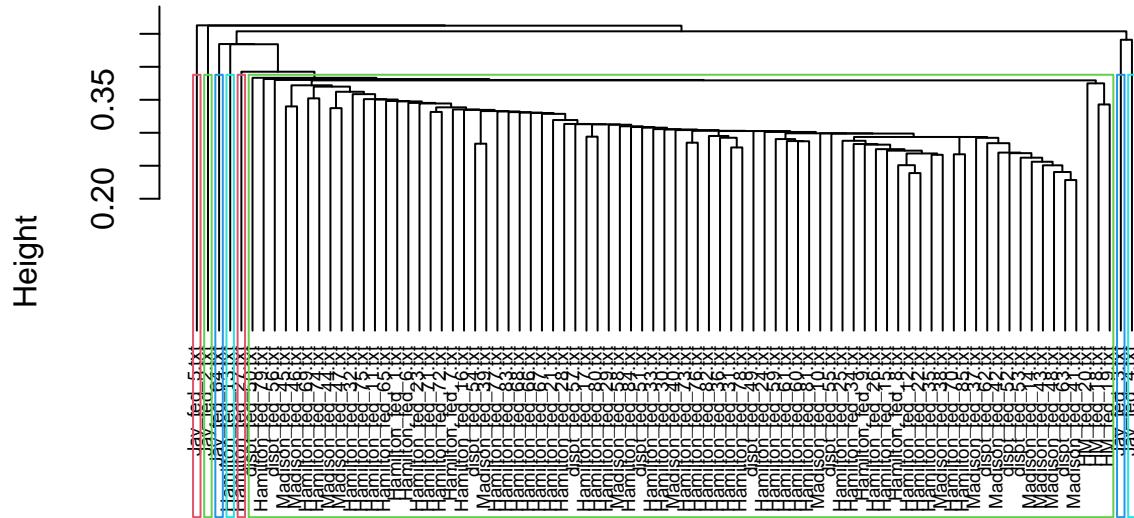
HAC Cluster Maximum Complete
7
Clusters



distance6
hclust (*, "complete")

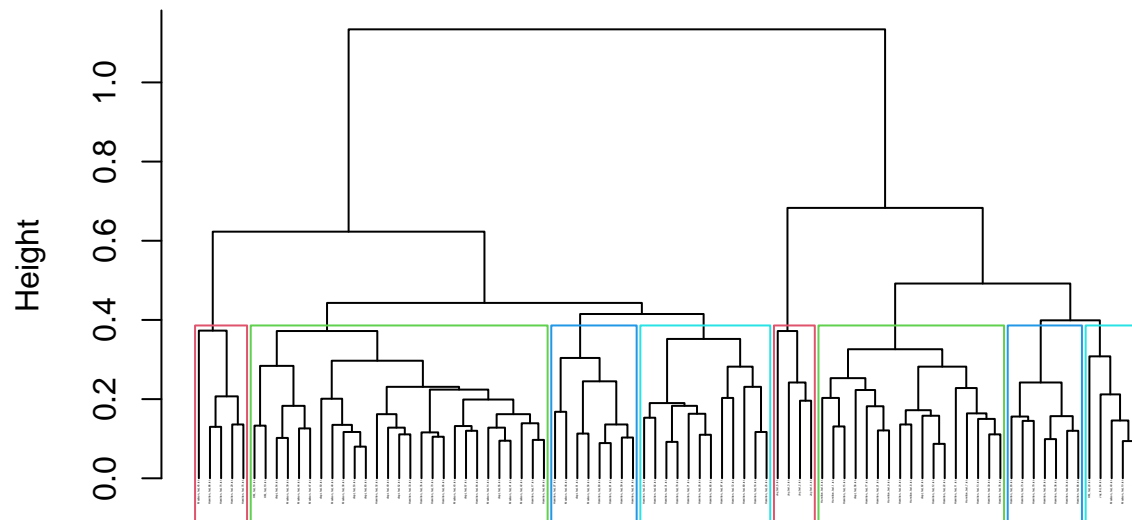


HAC Cluster Euclidean Single 8 Clusters



distance
hclust (*, "single")

HAC Cluster Maximum Complete 8 Clusters



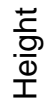
distance2
hclust (*, "complete")

8 Clusters



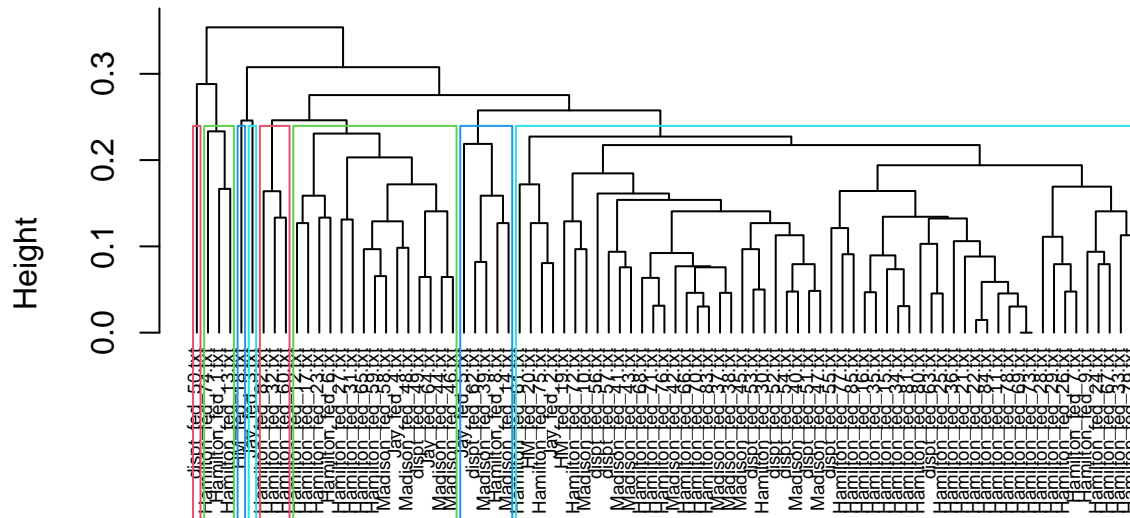
```
distance3
hclust (*, "complete")
```

8 Clusters



```
distance4
hclust (*, "complete")
```

HAC Cluster Minkowski Complete 8 Clusters



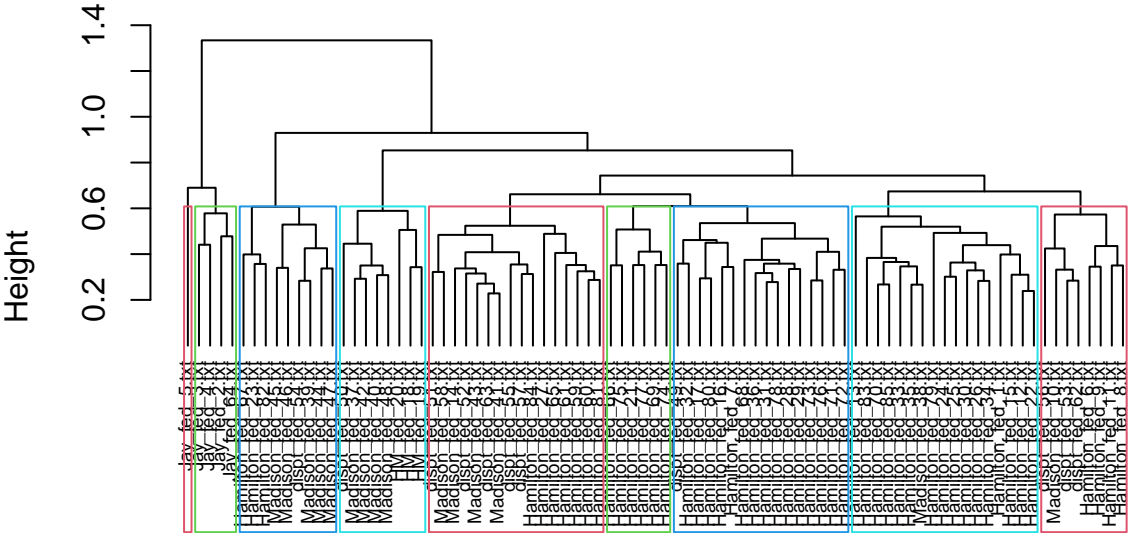
distance5
hclust (*, "complete")

8 Clusters

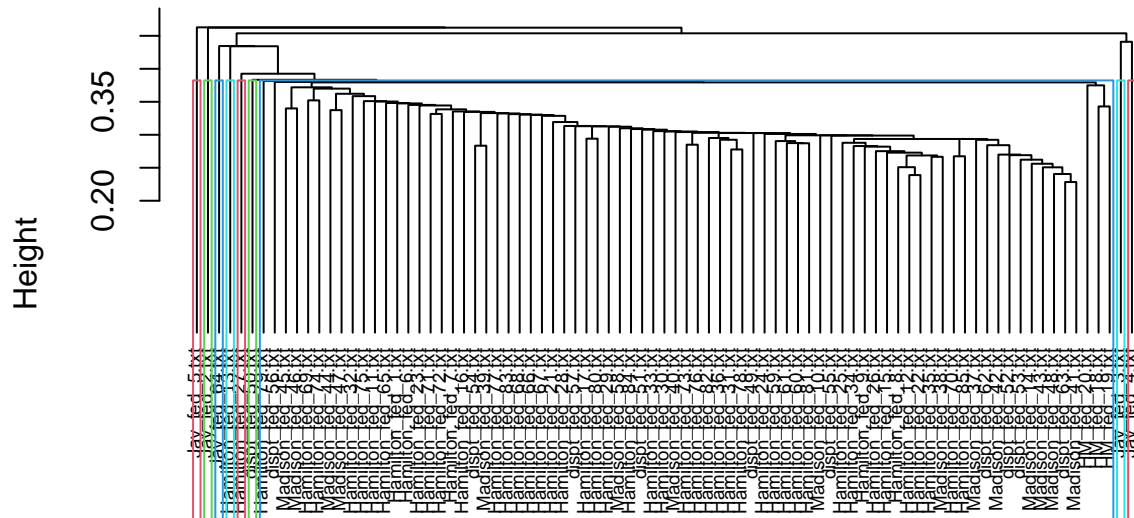


```
distance6
hclust (*, "complete")
```

HAC Cluster Euclidean Complete
9
Clusters

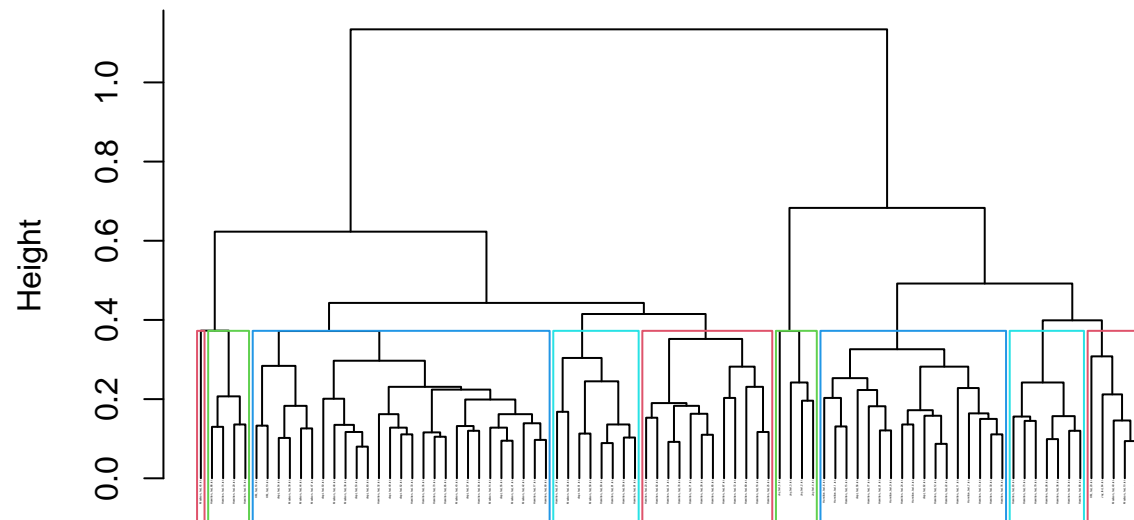


HAC Cluster Euclidean Single 9 Clusters



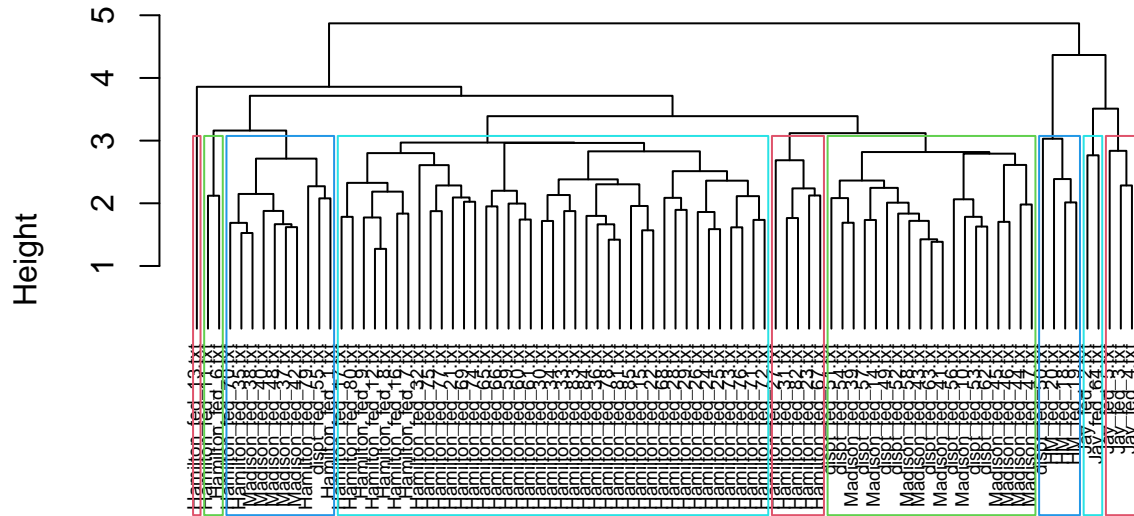
distance
hclust (*, "single")

HAC Cluster Maximum Complete 9 Clusters



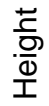
distance2
hclust (*, "complete")

HAC Cluster Manhattan Complete 9 Clusters



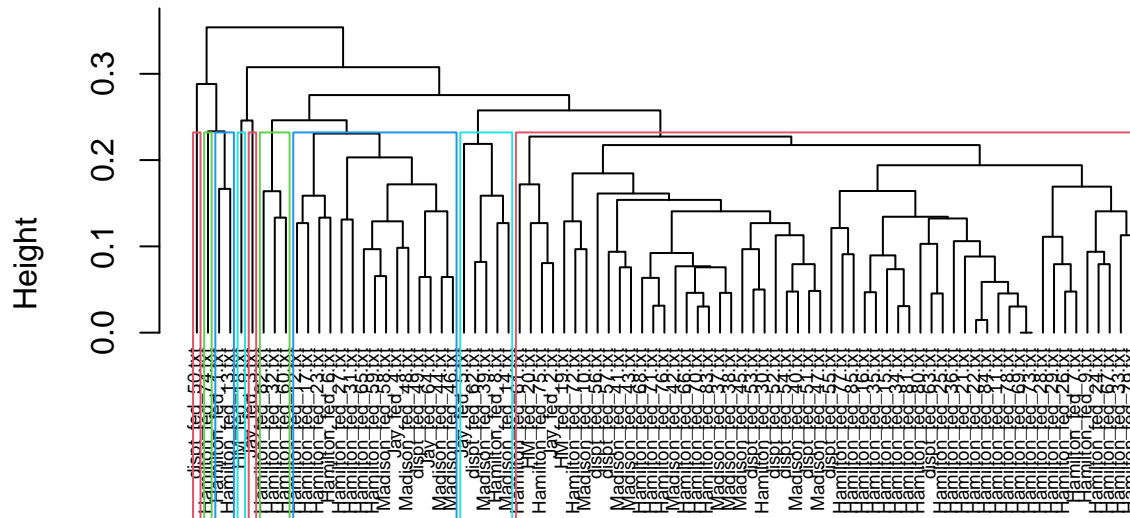
distance3
hclust (*, "complete")

9 Clusters

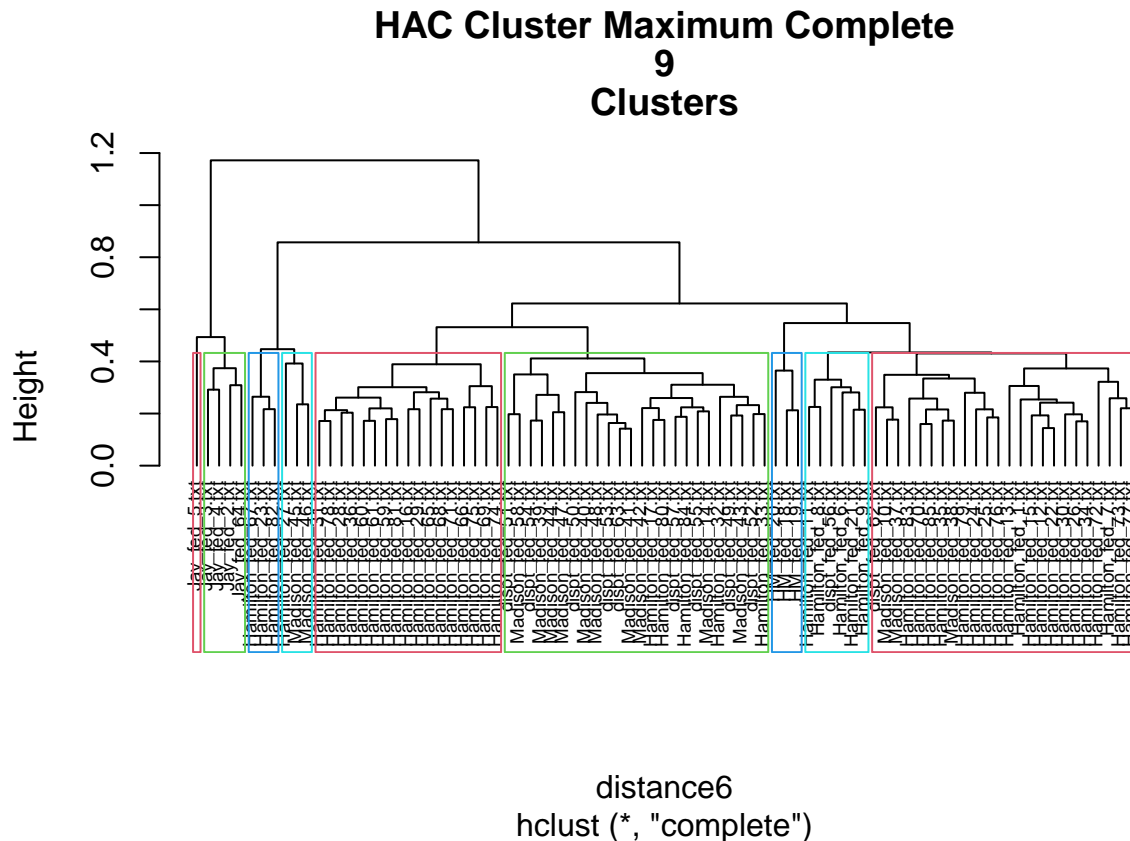


```
distance4
hclust (*, "complete")
```

HAC Cluster Minkowski Complete 9 Clusters



distance5
hclust (*, "complete")



Other analysis

Load Data (as Corpus).

In this example, we will load the data in corpus form. We will need to do much of the data cleaning, text processing, ourselves.

```
###Load Fed Papers Corpus
```

```
FedPapersCorpus <- Corpus(DirSource("C:/Users/GeorgeSmith/Desktop/FedPapersCorpus"))
(numberFedPapers<-length(FedPapersCorpus))
```

```
## [1] 85
```

```
##The following will show you that you read in all the documents
```

```
(summary(FedPapersCorpus))
```

```
##          Length Class          Mode
## dispt_fed_49.txt      2 PlainTextDocument list
## dispt_fed_50.txt      2 PlainTextDocument list
```

[illegible]

```

## Hamilton_fed_81.txt 2      PlainTextDocument list
## Hamilton_fed_82.txt 2      PlainTextDocument list
## Hamilton_fed_83.txt 2      PlainTextDocument list
## Hamilton_fed_84.txt 2      PlainTextDocument list
## Hamilton_fed_85.txt 2      PlainTextDocument list
## Hamilton_fed_9.txt 2       PlainTextDocument list
## HM_fed_18.txt 2           PlainTextDocument list
## HM_fed_19.txt 2           PlainTextDocument list
## HM_fed_20.txt 2           PlainTextDocument list
## Jay_fed_2.txt 2           PlainTextDocument list
## Jay_fed_3.txt 2           PlainTextDocument list
## Jay_fed_4.txt 2           PlainTextDocument list
## Jay_fed_5.txt 2           PlainTextDocument list
## 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

```

Data Cleaning

Here we investigate the data and vectorize it using DocumentTermMatrix.

We will ignore very infrequent words and very frequent words during the vectorization process.

Note: The DocumentTermMatrix method will perform much data cleaning for us.

Data Preparation and Transformation on Fed Papers

Remove punctuation,numbers, and space


```
(getTransformations())
```

```
## [1] "removeNumbers"      "removePunctuation" "removeWords"  
## [4] "stemDocument"        "stripWhitespace"
```

```
(nFedPapersCorpus<-length(FedPapersCorpus))
```

```
## [1] 85
```

ignore extremely rare words i.e. terms that appear in less than 1% of the documents

```
(minTermFreq <- nFedPapersCorpus * 0.0001)
```

```
## [1] 0.0085
```

```
(minTermFreqNum <- 30) # min terms as a number
```

```
## [1] 30
```

###Ignore overly common words i.e. terms that appear in more than 50% of the documents

```
(maxTermFreq <- nFedPapersCorpus * 1)
```

```
## [1] 85
```

```
(maxTermFreqNum <- 1000) # max terms as a number
```

```
## [1] 1000
```

```
MyStopwords <- c("will","one","two", "may","less", "well","might","withou","small", "single", "several"  
                "but", "very", "can", "must", "also", "very", "can", "any", "and", "are", "however",  
                "into", "almost", "can","for","add")
```

```
(STOPS <-stopwords('english'))
```

```
## [1] "i"      "me"      "my"      "myself"  "we"  
## [6] "our"    "ours"    "ourselves" "you"     "your"  
## [11] "yours"  "yourself" "yourselves" "he"      "him"  
## [16] "his"    "himself" "she"       "her"     "hers"  
## [21] "herself" "it"      "its"       "itself"  "they"  
## [26] "them"   "their"   "theirs"    "themselves" "what"  
## [31] "which"  "who"     "whom"      "this"     "that"  
## [36] "these"  "those"   "am"        "is"       "are"  
## [41] "was"    "were"    "be"        "been"     "being"  
## [46] "have"   "has"     "had"       "having"   "do"  
## [51] "does"   "did"     "doing"     "would"    "should"
```

```
## [56] "could"      "ought"      "i'm"        "you're"     "he's"
## [61] "she's"      "it's"       "we're"      "they're"    "i've"
## [66] "you've"     "we've"      "they've"    "i'd"        "you'd"
## [71] "he'd"       "she'd"      "we'd"       "they'd"     "i'll"
## [76] "you'll"     "he'll"      "she'll"     "we'll"      "they'll"
## [81] "isn't"      "aren't"     "wasn't"     "weren't"    "hasn't"
## [86] "haven't"    "hadn't"     "doesn't"    "don't"      "didn't"
## [91] "won't"      "wouldn't"   "shan't"     "shouldn't"  "can't"
## [96] "cannot"     "couldn't"   "mustn't"    "let's"      "that's"
## [101] "who's"      "what's"     "here's"     "there's"    "when's"
## [106] "where's"    "why's"      "how's"      "a"          "an"
## [111] "the"        "and"        "but"        "if"         "or"
## [116] "because"    "as"         "until"      "while"      "of"
## [121] "at"         "by"         "for"        "with"       "about"
## [126] "against"    "between"    "into"       "through"    "during"
## [131] "before"     "after"      "above"      "below"      "to"
## [136] "from"       "up"         "down"       "in"         "out"
## [141] "on"         "off"        "over"       "under"      "again"
## [146] "further"    "then"       "once"       "here"       "there"
## [151] "when"       "where"      "why"        "how"        "all"
## [156] "any"        "both"       "each"       "few"        "more"
## [161] "most"       "other"      "some"       "such"       "no"
## [166] "nor"        "not"        "only"       "own"        "same"
## [171] "so"         "than"       "too"        "very"       "will"
```

```
Papers_DTM <- DocumentTermMatrix(FedPapersCorpus, control = list( stopwords = TRUE, wordLengths=c(3, 15),
                                                                    removePunctuation = T, removeNumbers = F),
                                  stopwords = MyStopwords, bounds = list(1, 10000))
```

use the “built-in” STOP words

#inspect FedPapers Document Term Matrix (DTM)

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

```
##              Terms
## Docs      abandon abat abb abet abhorr abil abject abl ablest abolish
## dispt_fed_49.txt      0  0  0  0      0  0      0  2      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  0      0  1      0      0
## dispt_fed_52.txt      0  0  0  0      0  1      0  1      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      1  0      0  0      0      0
## dispt_fed_62.txt      0  0  0  0      0  0      0  1      0      0
## dispt_fed_63.txt      0  0  0  0      0  0      0  4      0      0
```

Inspect Initial Cleaning Results

Look at word frequencies

```
WordFreq <- colSums(as.matrix(Papers_DTM))
(head(WordFreq))
```

```
## abandon    abat    abb    abet  abhorr    abil
##          9      2      5      2      1      15
```

```
(length(WordFreq))
```

```
## [1] 4900
```

```
ord <- order(WordFreq)
(WordFreq[head(ord)])
```

```
## abhorr  abject  abraham  abreg  absenc  absolv
##        1        1        1        1        1        1
```

```
(WordFreq[tail(ord)])
```

```
## constitut    may    power    govern    will    state
##          686    811    937    1040    1263    1662
```

```
(Row_Sum_Per_doc <- rowSums((as.matrix(Papers_DTM))))
```

```
## dispt_fed_49.txt  dispt_fed_50.txt  dispt_fed_51.txt  dispt_fed_52.txt
##              758              530              923              853
## dispt_fed_53.txt  dispt_fed_54.txt  dispt_fed_55.txt  dispt_fed_56.txt
##             1035             882             968             765
## dispt_fed_57.txt  dispt_fed_62.txt  dispt_fed_63.txt  Hamilton_fed_1.txt
##             1023             1124             1432             767
## Hamilton_fed_11.txt Hamilton_fed_12.txt Hamilton_fed_13.txt Hamilton_fed_15.txt
##             1164             1044             479             1411
## Hamilton_fed_16.txt Hamilton_fed_17.txt Hamilton_fed_21.txt Hamilton_fed_22.txt
##              918              767              937             1692
## Hamilton_fed_23.txt Hamilton_fed_24.txt Hamilton_fed_25.txt Hamilton_fed_26.txt
##              828              925              927             1093
## Hamilton_fed_27.txt Hamilton_fed_28.txt Hamilton_fed_29.txt Hamilton_fed_30.txt
##              690              755             1010             948
## Hamilton_fed_31.txt Hamilton_fed_32.txt Hamilton_fed_33.txt Hamilton_fed_34.txt
##              797              686              773             1020
## Hamilton_fed_35.txt Hamilton_fed_36.txt Hamilton_fed_59.txt Hamilton_fed_6.txt
##             1052             1272             860             984
## Hamilton_fed_60.txt Hamilton_fed_61.txt Hamilton_fed_65.txt Hamilton_fed_66.txt
##             1006             681             912             997
## Hamilton_fed_67.txt Hamilton_fed_68.txt Hamilton_fed_69.txt Hamilton_fed_7.txt
```

```
##           781           683           1359           1073
## Hamilton_fed_70.txt Hamilton_fed_71.txt Hamilton_fed_72.txt Hamilton_fed_73.txt
##           1436           766           925           1061
## Hamilton_fed_74.txt Hamilton_fed_75.txt Hamilton_fed_76.txt Hamilton_fed_77.txt
##           478           905           883           887
## Hamilton_fed_78.txt Hamilton_fed_79.txt Hamilton_fed_80.txt Hamilton_fed_81.txt
##           1376           478           998           1132
## Hamilton_fed_82.txt Hamilton_fed_83.txt Hamilton_fed_84.txt Hamilton_fed_85.txt
##           1798           749           2620           1907
## Hamilton_fed_86.txt Hamilton_fed_87.txt HM_fed_18.txt HM_fed_19.txt
##           1264           931           1029           1023
## HM_fed_20.txt Jay_fed_2.txt Jay_fed_3.txt Jay_fed_4.txt
##           776           804           736           780
## Jay_fed_5.txt Jay_fed_6.txt Madison_fed_10.txt Madison_fed_11.txt
##           657           1072           1437           1016
## Madison_fed_12.txt Madison_fed_13.txt Madison_fed_14.txt Madison_fed_15.txt
##           1268           1529           1169           1340
## Madison_fed_16.txt Madison_fed_17.txt Madison_fed_18.txt Madison_fed_19.txt
##           1701           1330           1601           1382
## Madison_fed_20.txt Madison_fed_21.txt Madison_fed_22.txt Madison_fed_23.txt
##           1018           1233           1306           846
## Madison_fed_24.txt
##           978
```

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

```
##           Terms
## Docs      crude cruel crush culpabl cultiv columnni cun cupid cure
## dispt_fed_49.txt      0      0      0      0      0      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      0      0      0      0      0
## dispt_fed_52.txt      0      0      0      0      0      0      0      0      0
## dispt_fed_53.txt      0      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      1      0      0      0      0      0      0
##           Terms
## Docs      curios curious
## dispt_fed_49.txt      0      0
## dispt_fed_50.txt      0      0
```

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

Terms

```
## function (x)
## UseMethod("Terms")
## <bytecode: 0x0000000018e869c0>
## <environment: namespace:tm>
```

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

```
##               Terms
## Docs      crude cruel crush culpabl cultiv columni cun cupid cure
## dispt_fed_49.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_50.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_51.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_52.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_53.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_54.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_55.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_56.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_57.txt      0      0 0.000      0 0.000      0 0      0 0
## dispt_fed_62.txt      0      0 0.000      0 0.001      0 0      0 0
## dispt_fed_63.txt      0      0 0.001      0 0.000      0 0      0 0
##               Terms
## Docs      curios curious
## dispt_fed_49.txt 0.000      0
## dispt_fed_50.txt 0.000      0
## dispt_fed_51.txt 0.000      0
## dispt_fed_52.txt 0.000      0
## dispt_fed_53.txt 0.001      0
## dispt_fed_54.txt 0.000      0
## dispt_fed_55.txt 0.000      0
## dispt_fed_56.txt 0.000      0
## dispt_fed_57.txt 0.000      0
## dispt_fed_62.txt 0.000      0
## dispt_fed_63.txt 0.000      0
```

Data Structures

Convert to matrix and view

```
Papers_dtm_matrix = as.matrix(Papers_DTM)
str(Papers_dtm_matrix)
```

```
## num [1:85, 1:4900] 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 [1:4900] "abandon" "abat" "abb" "abet" ...
```

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

```
##               Terms
## Docs          abat abb abet abhorr abil abject abl ablest abolish
## dispt_fed_49.txt  0  0  0      0      0  0      0  2      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  0      0  1      0      0
## dispt_fed_52.txt  0  0  0      0      1  0      0  1      0      0
## dispt_fed_53.txt  1  0  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      1      0  0      0  0      0      0
## dispt_fed_62.txt  0  0  0      0      0  0      0  1      0      0
## dispt_fed_63.txt  0  0  0      0      0  0      0  4      0      0
```

#Also convert to DF

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

```
## 'data.frame':   85 obs. of  4900 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 ...
## $ abhorr       : num  0 0 0 0 0 0 0 0 1 0 ...
## $ abil         : num  0 0 0 1 0 0 0 0 0 0 ...
## $ abject       : num  0 0 0 0 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 ...
## $ abraham      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ abreg        : 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 ...
## $ abrog        : num  0 0 0 0 0 0 0 0 0 0 ...
## $ absenc       : num  0 0 0 0 0 0 0 0 0 0 ...
## $ absolut      : num  0 2 2 1 0 0 0 0 0 0 ...
## $ absolv       : num  0 0 0 0 0 0 0 0 0 0 ...
```

## \$ absorb	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ abstain	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ abstract	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ abstrus	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ absurd	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ abund	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ abus	: num	1 1 2 1 1 0 0 0 0 0 0 ...
## \$ abyss	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ acced	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ acceler	: num	0 0 0 0 1 0 0 0 0 0 0 ...
## \$ accept	: num	0 0 0 0 0 0 0 0 0 0 1 ...
## \$ access	: num	0 0 0 2 0 0 0 0 0 0 0 ...
## \$ accid	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ accident	: num	0 0 0 1 0 0 0 0 0 0 0 ...
## \$ accommod	: num	0 0 0 0 1 0 0 0 0 0 0 ...
## \$ accomod	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ accompani	: num	0 0 0 0 0 0 0 0 1 0 0 ...
## \$ accomplic	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ accomplish	: num	0 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 ...
## \$ accret	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ accru	: num	0 0 0 0 0 0 0 0 0 0 1 ...
## \$ accumul	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ accur	: num	1 0 0 0 1 0 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 0 ...
## \$ accustom	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ achaeae	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ achaeus	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ achaia	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ achiev	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ acknowledg	: num	0 1 0 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 0 ...
## \$ acquir	: num	1 0 0 0 5 0 0 2 0 0 ...
## \$ acquisit	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ acquit	: num	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ acr	: num	0 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 ...
## \$ actuat	: num	0 0 0 0 0 0 1 0 1 0 ...
## \$ acut	: num	0 0 0 0 0 0 0 0 0 0 ...
## \$ adag	: num	0 0 0 0 0 0 0 0 0 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 ...
## $ adjac      : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adjoin     : num 0 0 0 0 0 0 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 ...
## $ adjudg     : 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 ...
## $ admiralgener : num 0 0 0 0 0 0 0 0 0 0 ...
## $ admiralalti : 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 ...
## $ adroit     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ adul       : num 0 0 0 0 0 0 0 0 0 0 ...
## $ 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 ...
## [list output truncated]
```

```
(Papers_DF$abolit)
```

```
## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [39] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 3 0 0 1 1 0 0 0 0 0 0 0 1 0 0 1 0 0
## [77] 0 0 0 0 0 0 0 0 0
```

```
(nrow(Papers_DF)) ## Each row is Paper
```

```
## [1] 85
```

Add row names

```
Papers_DF1<- Papers_DF%>%add_rownames()
```

```
## Warning: 'add_rownames()' was deprecated in dplyr 1.0.0.
## Please use 'tibble::rownames_to_column()' instead.
```

```
names(Papers_DF1)[1]<-"Author"
Papers_DF1[1:11,1]="dispt"
Papers_DF1[12:65,1]="hamil"
Papers_DF1[66:70,1]="jay"
Papers_DF1[71:85,1]="madis"
head(Papers_DF1[,1:2],20)
```



```
## # A tibble: 20 x 2
##   Author abandon
##   <chr>      <dbl>
## 1 dispt      0
## 2 dispt      0
## 3 dispt      0
## 4 dispt      0
## 5 dispt      0
## 6 dispt      0
## 7 dispt      0
## 8 dispt      0
## 9 dispt      0
## 10 dispt     0
## 11 dispt     0
## 12 hamil     0
## 13 hamil     0
## 14 hamil     0
## 15 hamil     0
## 16 hamil     2
## 17 hamil     0
## 18 hamil     0
## 19 hamil     0
## 20 hamil     0
```

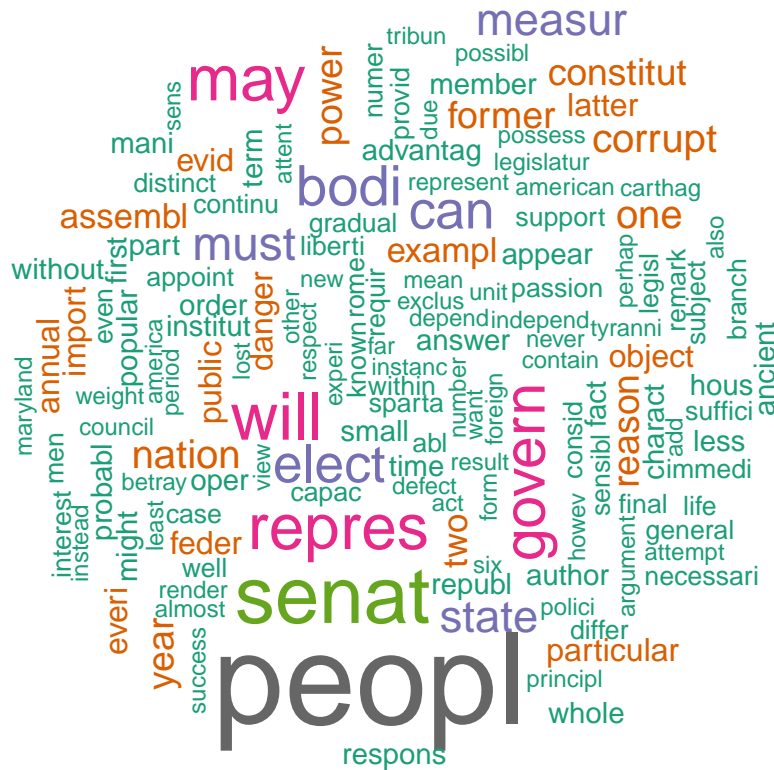
```
tail(Papers_DF1[,1:2],20)
```

```
## # A tibble: 20 x 2
##   Author abandon
##   <chr>      <dbl>
## 1 jay        0
## 2 jay        0
## 3 jay        0
## 4 jay        0
## 5 jay        0
## 6 madis      0
## 7 madis      0
## 8 madis      0
## 9 madis      1
## 10 madis     1
## 11 madis     0
## 12 madis     0
## 13 madis     0
## 14 madis     0
## 15 madis     0
## 16 madis     0
## 17 madis     0
## 18 madis     0
## 19 madis     0
## 20 madis     0
```

Example Word Cloud

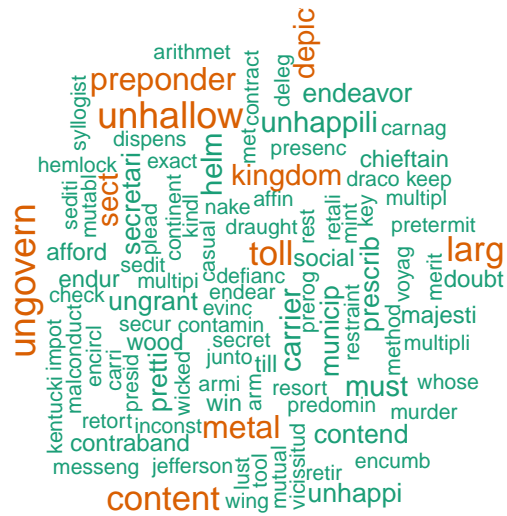
Wordcloud Visualization Hamilton, Madison and Disputed Papers

```
DisputedPapersWC<- wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[[11, ],
                                     rot.per = .35, colors = brewer.pal(n = 8, name = "Dark2"))
```

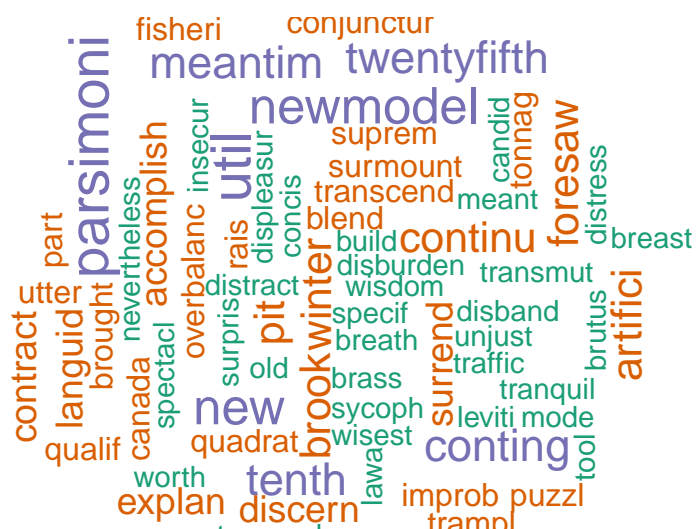


```
(head(sort(as.matrix(Papers_DTM)[11,], decreasing = TRUE), n=50))
```

##	peopl	senat	will	may	repres	govern	bodi
##	42	24	19	18	18	16	15
##	can	elect	must	measur	state	corrupt	nation
##	14	14	12	11	11	9	9
##	one	constitut	former	power	reason	year	assembl
##	9	8	8	8	8	8	7
##	exampl	two	annual	danger	everi	evid	feder
##	7	7	6	6	6	6	6
##	import	latter	object	particular	public	advantag	ancient
##	6	6	6	6	6	5	5
##	answer	appear	author	charact	fact	first	hous
##	5	5	5	5	5	5	5
##	institut	less	mani	member	might	oper	order
##	5	5	5	5	5	5	5
##	part						
##	5						



```
JayPapersHW <- wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[66:70, ],
  rot.per = .35, colors = brewer.pal(n = 8, name = "Dark2"))
```



```
DisputedWC <- wordcloud(colnames(Papers_dtm_matrix), Papers_dtm_matrix[1:11, ],
                        rot.per = .35, colors = brewer.pal(n = 8, name = "Dark2"))
```



```
##### # Analysis ##### # Distance Metrics
##### #Computing different distance matrices to determine which seems to
work the best! ###Distance Measure
```

```
m <- Papers_dtm_matrix
m_norm <- Papers_Matrix_Norm
```

```
#m <- [1:2, 1:3]
```

```
distMatrix_E <- dist(m, method="euclidean")
```

```
#print(distMatrix_E)
```

```
distMatrix_M <- dist(m, method="manhattan")
```

```
#print(distMatrix_M)
```

```
distMatrix_C <- dist(m, method="cosine")
```

```
print(distMatrix_C)
```

```
distMatrix_C_norm <- dist(m_norm, method="cosine")
```

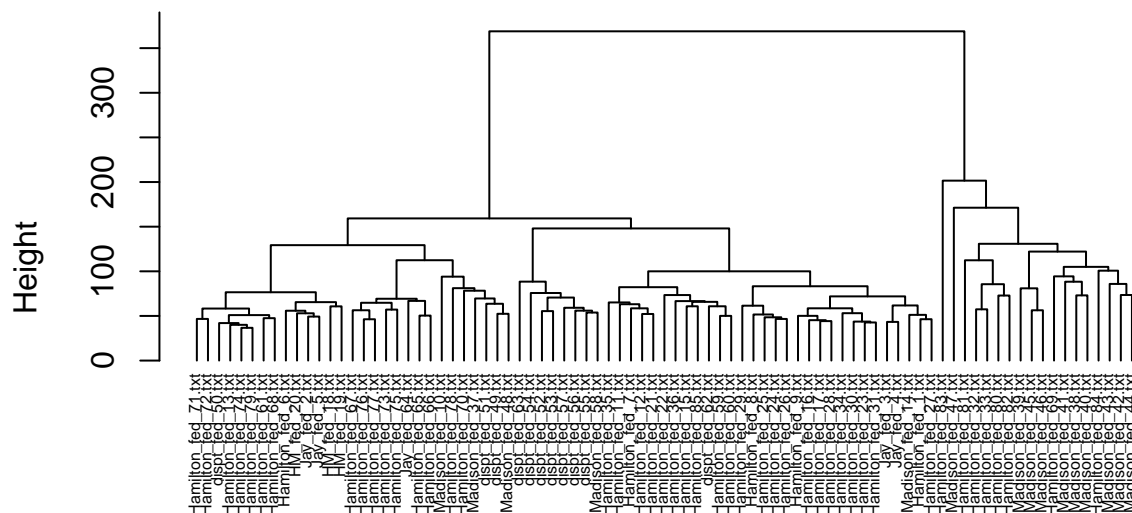
```
#print(distMatrix_C_norm)
```

Clustering

```
###Clustering Methods: ## HAC: Hierarchical Algorithm Clustering Method ## Euclidean
```

```
groups_E <- hclust(distMatrix_E,method="ward.D")
plot(groups_E, cex=0.5, font=22, hang=-1, main = "HAC Cluster Dendrogram with Euclidean Similarity")
```

HAC Cluster Dendrogram with Euclidean Similarity



```
distMatrix_E
hclust (*, "ward.D")
```

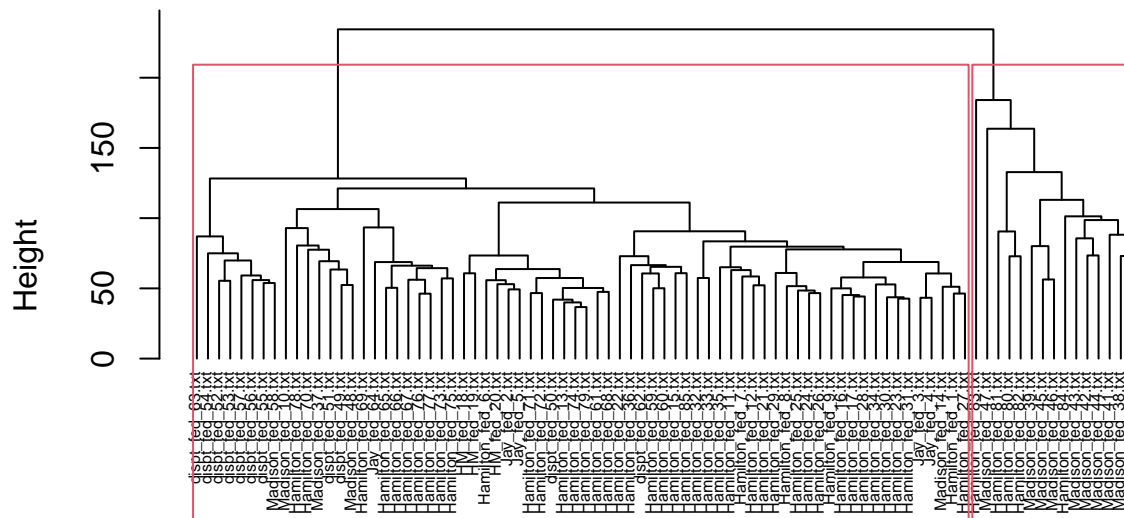
```
# Plots the separations
```

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

```
#HAC Cluster Dendrogram with Euclidean Similarity
```

```
distMatrix_E1 <- hclust(distMatrix_E, "ward.D2")
plot(distMatrix_E1, cex=0.5, font=22, hang=-1, main = "HAC Cluster Dendrogram with Euclidean Similarity")
rect.hclust(distMatrix_E1, k=2)
```

HAC Cluster Dendrogram with Euclidean Similarity #2

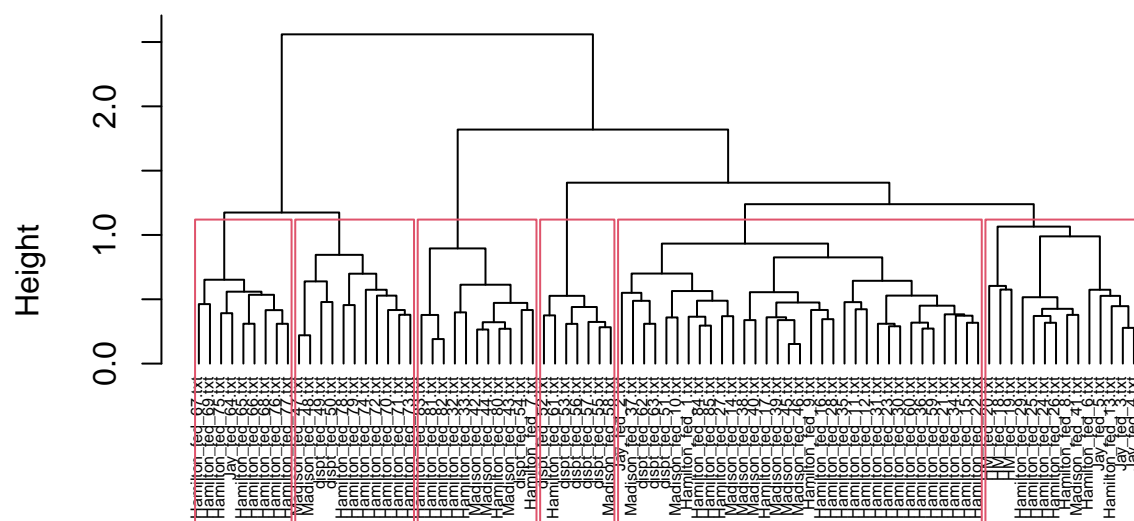


```
distMatrix_E
hclust (*, "ward.D2")
```

```
# HAC Cluster Dendrogram with Cosine Similarity
```

```
groups_C <- hclust(distMatrix_C,method="ward.D")
plot(groups_C, cex=0.5,font=22, hang=-1,main = "HAC Cluster Dendrogram with Cosine Similarity")
rect.hclust(groups_C, k=6)
```


HAC Cluster Dendrogram with Cosine Similarity

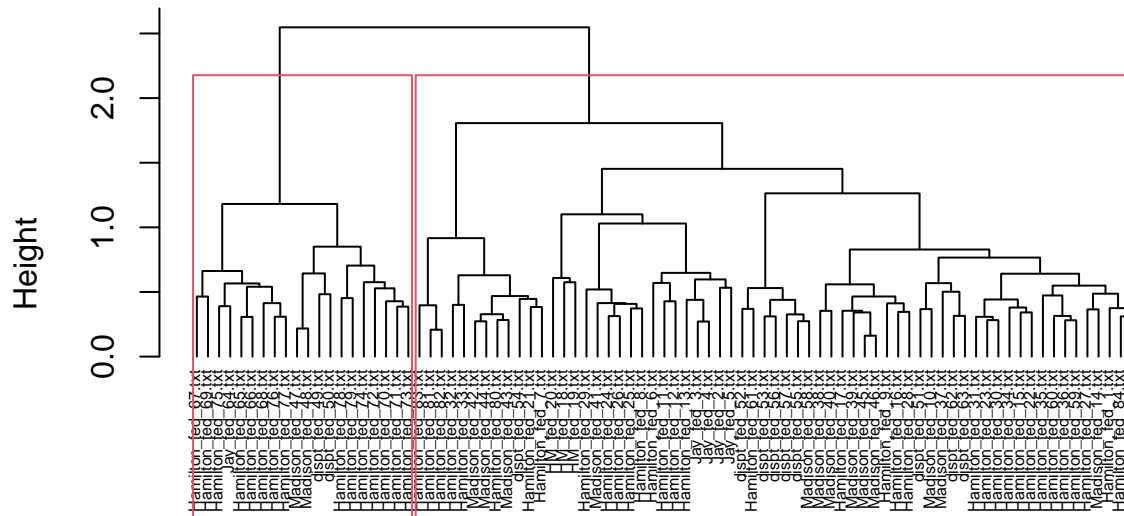


```
distMatrix_C
hclust (*, "ward.D")
```

```
## Cosine Similarity for Normalized Matrix
```

```
groups_C_n <- hclust(distMatrix_C_norm,method="ward.D")
plot(groups_C_n, cex=0.5, font=22, hang=-1, main = "HAC Cluster Dendrogram with Cosine Similarity Normalized")
rect.hclust(groups_C_n, k=2)
```

HAC Cluster Dendrogram with Cosine Similarity Normalized Matrix



distMatrix_C_norm
hclust (*, "ward.D")

k means clustering Methods

```
X <- m_norm
k2 <- kmeans(X, centers = 2, nstart = 100, iter.max = 50)
str(k2)
```

```
## List of 9
## $ cluster      : Named int [1:85] 1 1 1 2 2 2 1 2 2 1 ...
## .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:2, 1:4900] 1.09e-04 6.67e-05 1.82e-05 3.33e-05 9.09e-05 ...
## .. attr(*, "dimnames")=List of 2
## .. ..$ : chr [1:2] "1" "2"
## .. ..$ : chr [1:4900] "abandon" "abat" "abb" "abet" ...
## $ totss       : num 0.216
## $ withinss    : num [1:2] 0.1231 0.0794
## $ tot.withinss: num 0.203
## $ betweenss   : num 0.0137
## $ size        : int [1:2] 55 30
## $ iter        : int 1
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

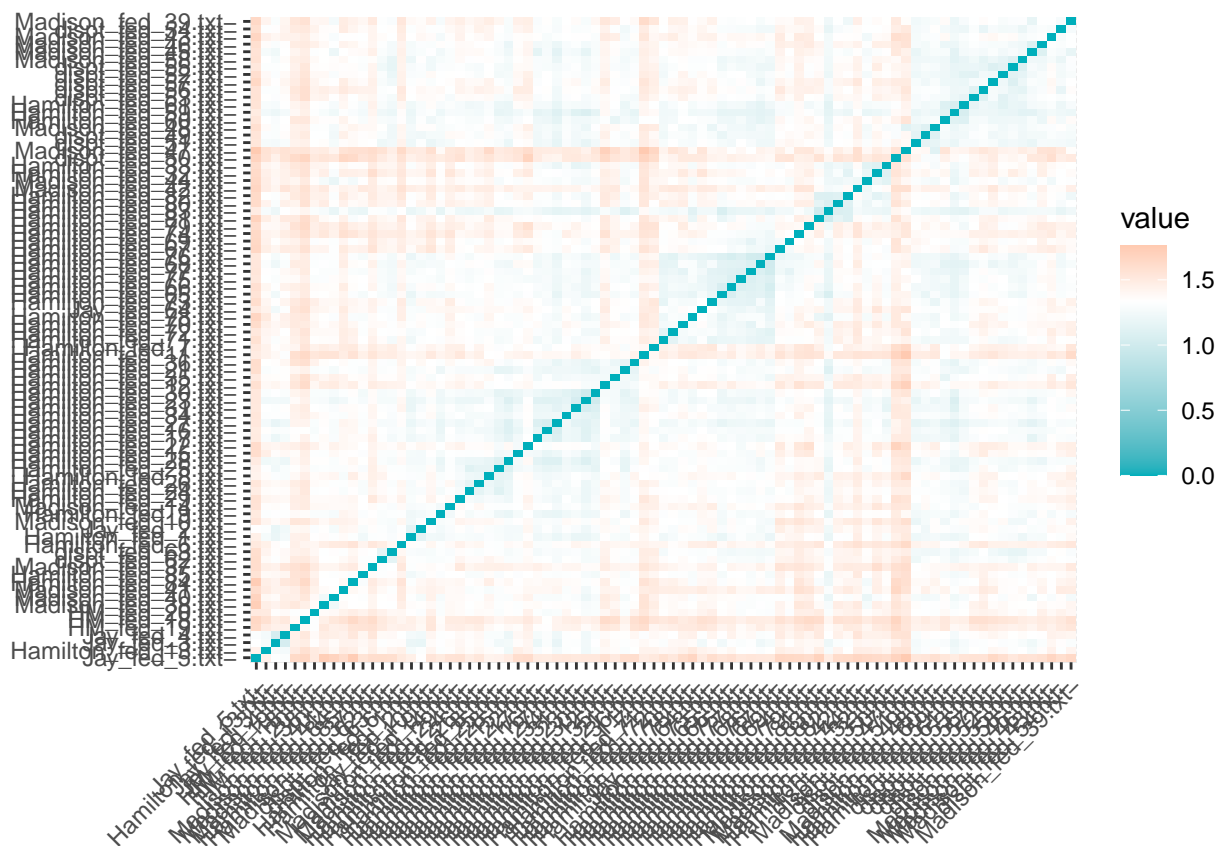
```
k3 <- kmeans(X, centers = 7, nstart = 50, iter.max= 50)
str(k3)
```

```
## List of 9
```

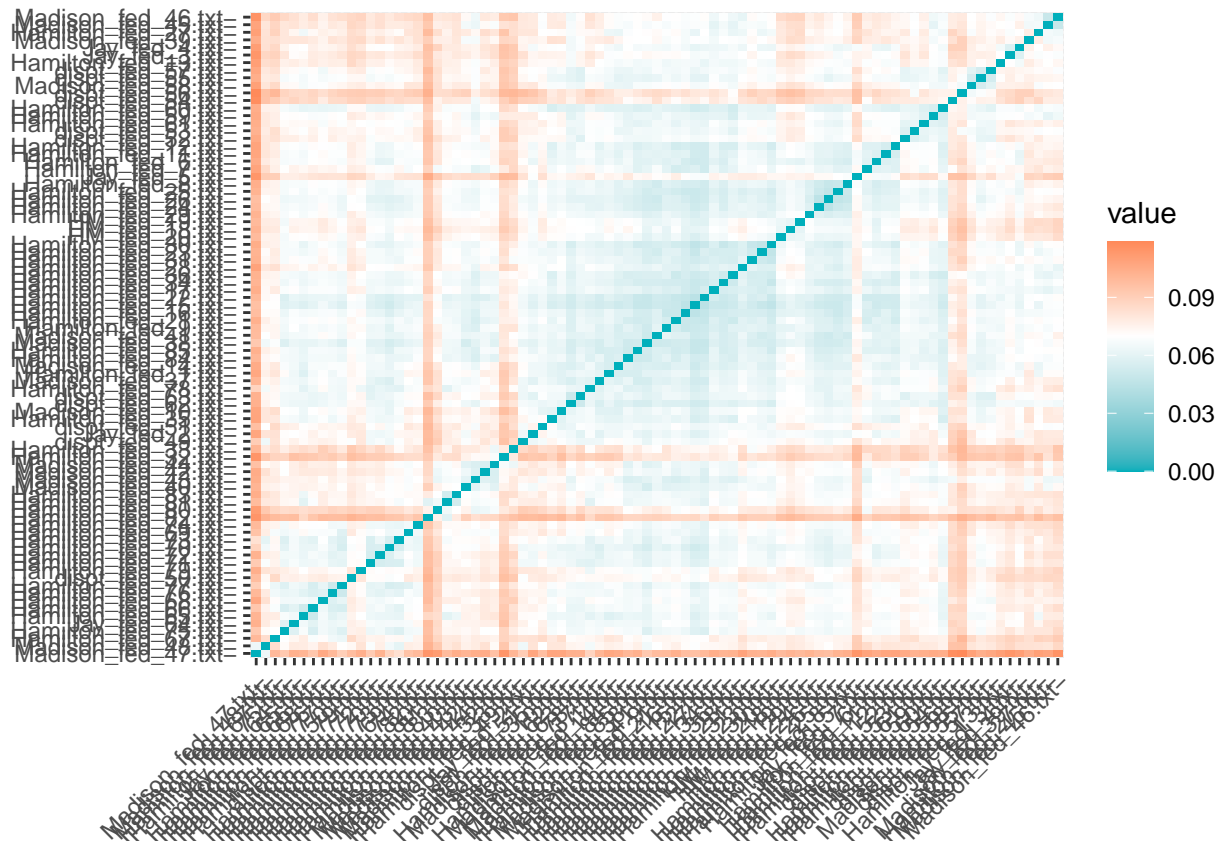
```
## $ cluster      : Named int [1:85] 7 7 1 4 4 4 4 4 1 ...
##   .. attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:7, 1:4900] 0.000214 0 0.000125 0 0 ...
##   .. attr(*, "dimnames")=List of 2
##     .. $ : chr [1:7] "1" "2" "3" "4" ...
##     .. $ : chr [1:4900] "abandon" "abat" "abb" "abet" ...
## $ totss       : num 0.216
## $ withinss    : num [1:7] 0.02827 0.01163 0.05749 0.01622 0.00201 ...
## $ tot.withinss: num 0.163
## $ betweenss   : num 0.0531
## $ size        : int [1:7] 14 6 32 8 2 4 19
## $ iter        : int 3
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"
```

k means visualization results

```
distance1 <- get_dist(X,method = "manhattan")
fviz_dist(distance1, gradient = list(low = "#00AFBB", mid = "white", high = "#FC4E07"))
```



```
distance2 <- get_dist(X,method = "euclidean")
fviz_dist(distance2, gradient = list(low = "#00AFBB", mid = "white", high = "#FC4E07"))
```



```
distance3 <- get_dist(X,method = "spearman")
fviz_dist(distance3, gradient = list(low = "#00AFBB", mid = "white", high = "#FC4E07", title= "Distance
```



```

x <- c(2,3,4,5,6,7,8,9)
set.seed(20)
for (val in x){
  print(val)
  # run k-means
  Clusters <- kmeans(FedPapers_km, val)
  FedPapers_km$Clusters <- as.factor(Clusters$cluster)
  str(Clusters)
  Clusters$centers

  # Add clusters to dataframe original dataframe with author name
  FedPapers_km2 <- FederalistPapers
  FedPapers_km2$Clusters <- as.factor(Clusters$cluster)
  # Plot results
  #clusplot(FedPapers_km, FedPapers_km$Clusters, color=TRUE, shade=TRUE, labels=0, lines=0)

  clusplot(FedPapers_km, FedPapers_km$Clusters, color=T, shade=T, labels=4, lines=T)
}

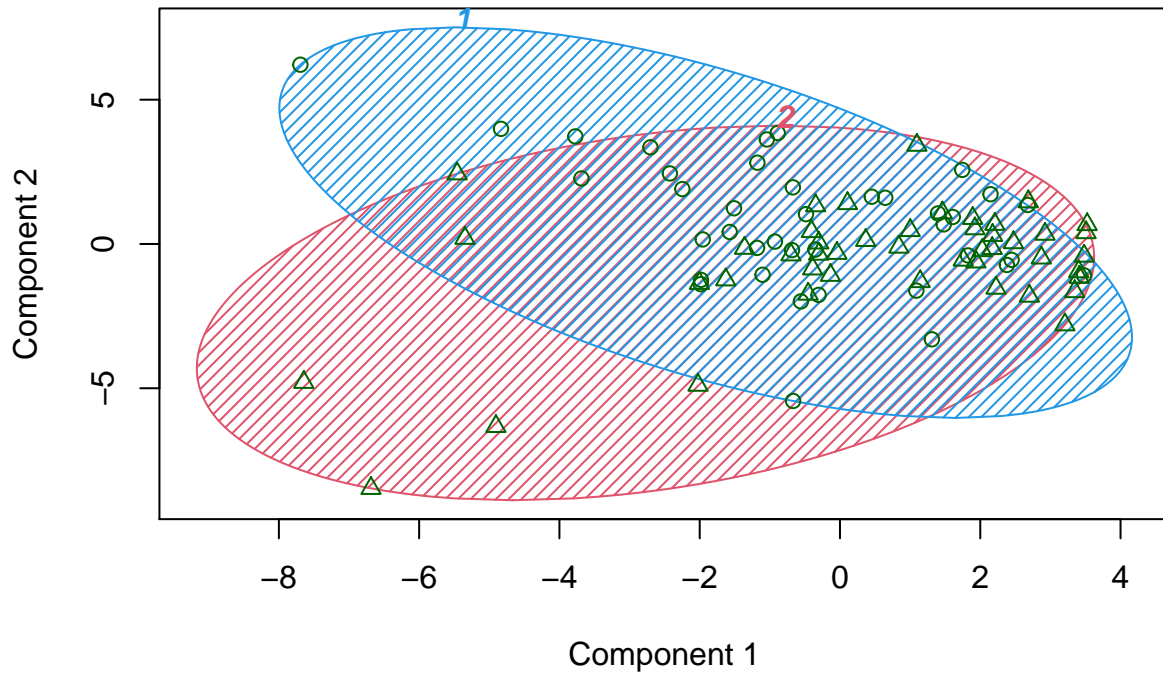
```

```

## [1] 2
## List of 9
## $ cluster      : Named int [1:85] 2 1 2 1 1 2 1 1 2 1 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers      : num [1:2, 1:71] 0.3081 0.28063 0.05051 0.0548 0.00828 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:2] "1" "2"
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss       : num 515
## $ withinss    : num [1:2] 33.1 79.5
## $ tot.withinss: num 113
## $ betweenss   : num 403
## $ size        : int [1:2] 39 46
## $ iter        : int 1
## $ ifault      : int 0
## - attr(*, "class")= chr "kmeans"

```

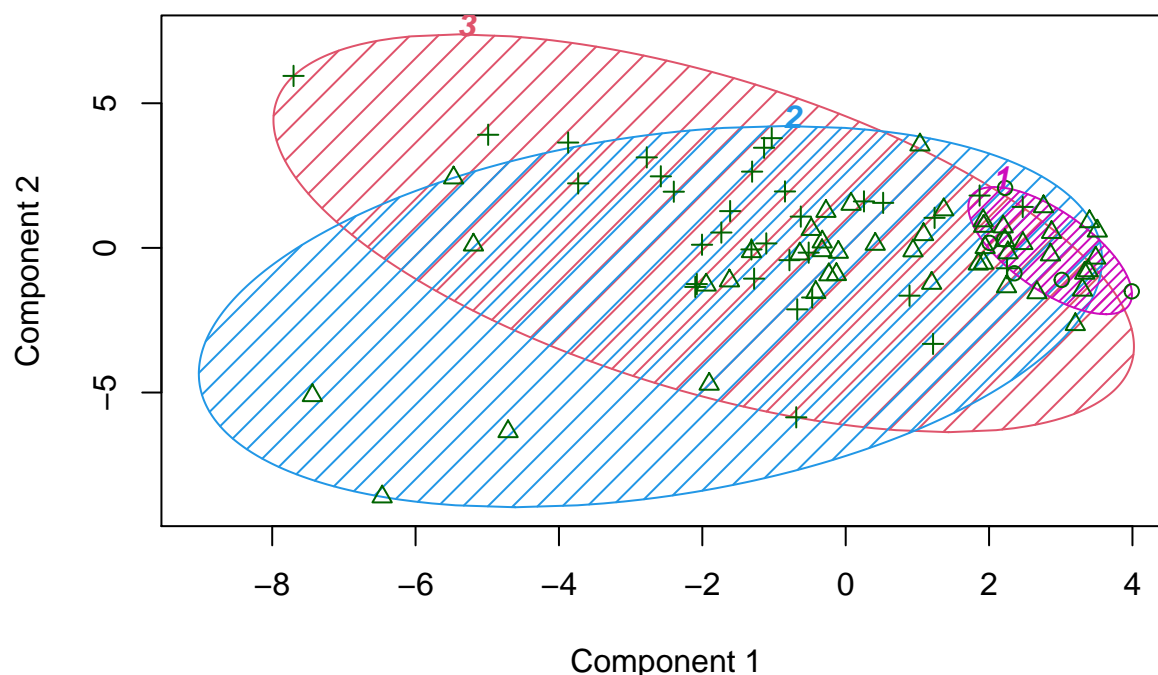
CLUSPLOT(FedPapers_km)



These two components explain 16.53 % of the point variability.

```
## [1] 3
## List of 9
## $ cluster      : Named int [1:85] 2 3 2 3 3 2 3 3 2 3 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:3, 1:71] 0.3432 0.2806 0.3017 0.0372 0.0548 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:3] "1" "2" "3"
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 33.7
## $ withinss     : num [1:3] 0.388 7.415 3.492
## $ tot.withinss : num 11.3
## $ betweenss    : num 22.4
## $ size         : int [1:3] 6 46 33
## $ iter         : int 2
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

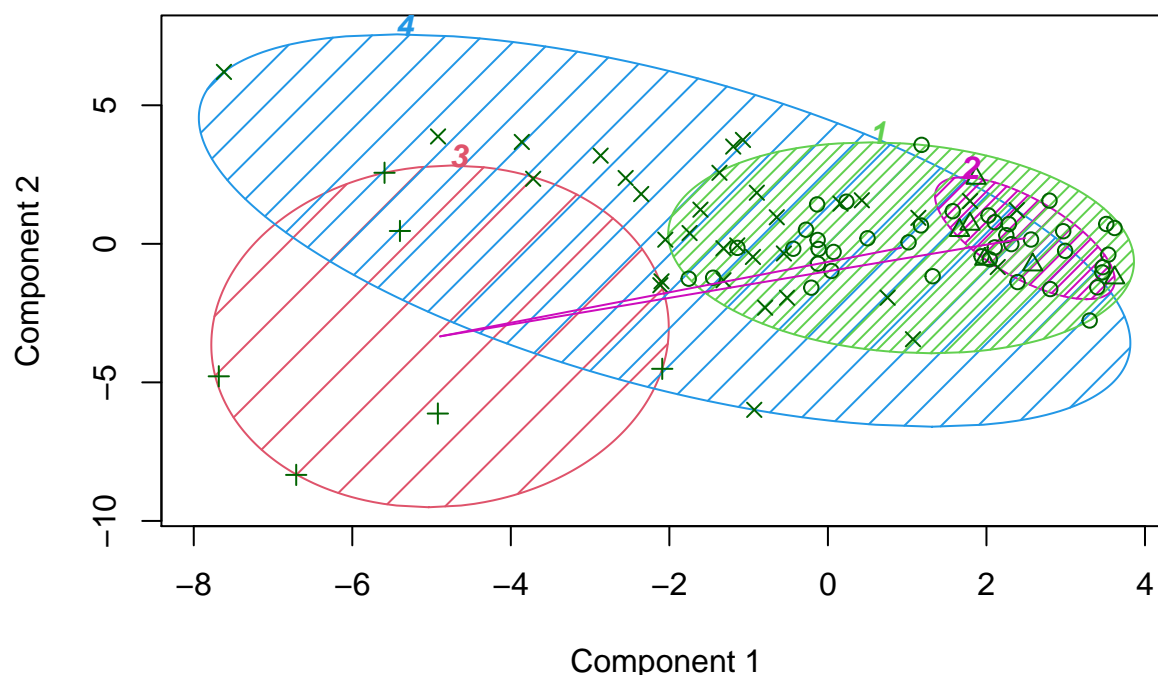
CLUSPLOT(FedPapers_km)



These two components explain 16.64 % of the point variability.

```
## [1] 4
## List of 9
## $ cluster      : Named int [1:85] 1 4 1 4 4 1 4 4 1 4 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:4, 1:71] 0.2971 0.3432 0.1707 0.3017 0.0572 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. .$ : chr [1:4] "1" "2" "3" "4"
##     .. .$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 43
## $ withinss     : num [1:4] 3.996 0.388 0.858 3.492
## $ tot.withinss : num 8.73
## $ betweenss    : num 34.3
## $ size         : int [1:4] 40 6 6 33
## $ iter         : int 3
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

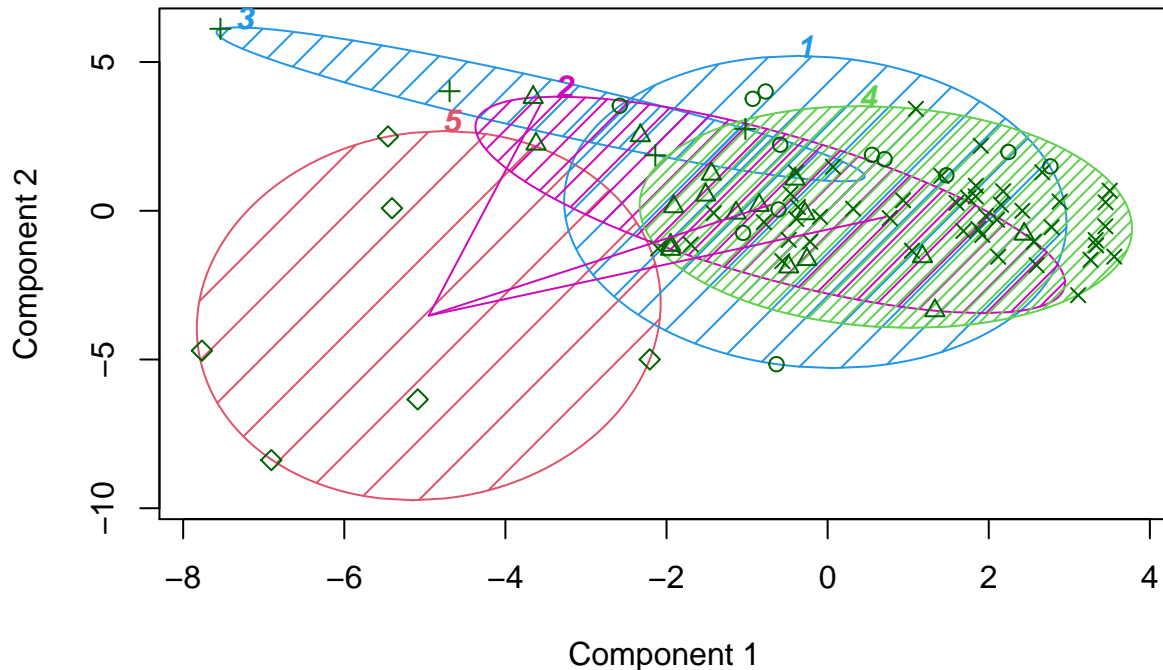

CLUSPLOT(FedPapers_km)



These two components explain 16.77 % of the point variability.

```
## [1] 5
## List of 9
## $ cluster      : Named int [1:85] 4 2 4 2 2 4 2 1 4 2 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:5, 1:71] 0.355 0.285 0.213 0.303 0.171 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:5] "1" "2" "3" "4" ...
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 179
## $ withinss      : num [1:5] 1.008 1.195 0.264 9.921 0.858
## $ tot.withinss  : num 13.2
## $ betweenss     : num 165
## $ size          : int [1:5] 12 17 4 46 6
## $ iter          : int 3
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

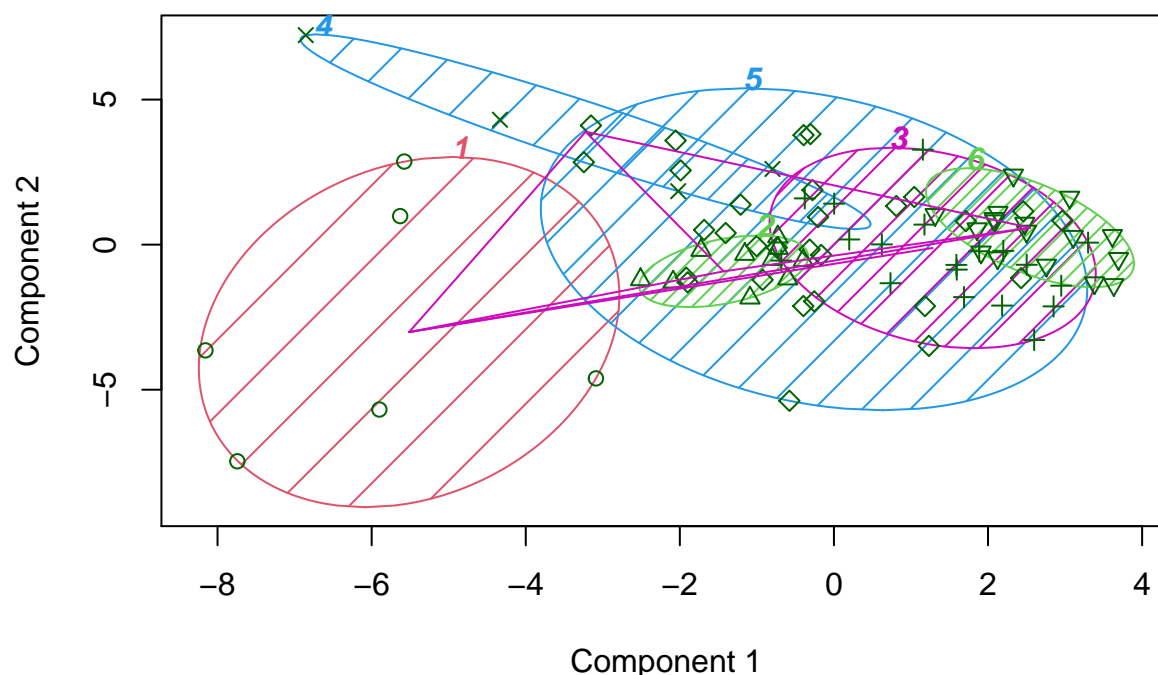
CLUSPLOT(FedPapers_km)



These two components explain 16.51 % of the point variability.

```
## [1] 6
## List of 9
## $ cluster      : Named int [1:85] 2 5 2 5 5 2 5 5 2 5 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:6, 1:71] 0.171 0.27 0.286 0.213 0.314 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. .$ : chr [1:6] "1" "2" "3" "4" ...
##     .. .$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss         : num 144
## $ withinss      : num [1:6] 0.858 0.755 1.416 0.264 9.734 ...
## $ tot.withinss : num 14.5
## $ betweenss     : num 130
## $ size          : int [1:6] 6 10 19 4 29 17
## $ iter          : int 3
## $ ifault        : int 0
## - attr(*, "class")= chr "kmeans"
```

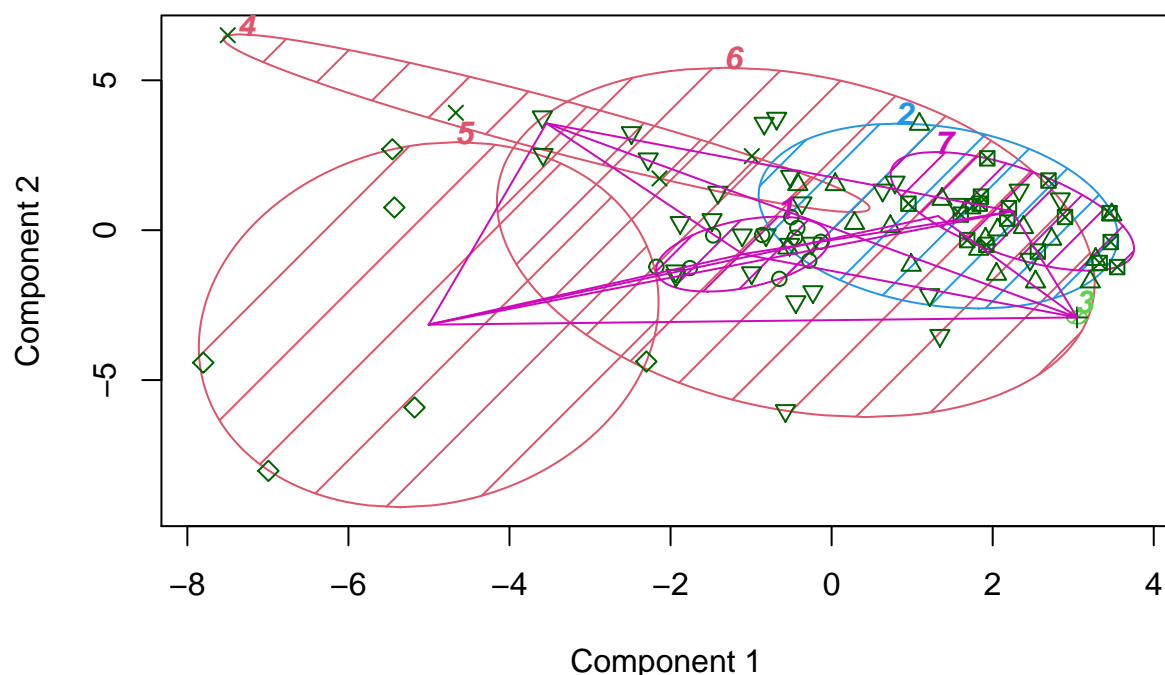
CLUSPLOT(FedPapers_km)



These two components explain 16.83 % of the point variability.

```
## [1] 7
## List of 9
## $ cluster      : Named int [1:85] 1 6 1 6 6 1 6 6 1 6 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:7, 1:71] 0.27 0.286 0.27 0.213 0.171 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:7] "1" "2" "3" "4" ...
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 222
## $ withinss     : num [1:7] 0.755 1.26 0 0.264 0.858 ...
## $ tot.withinss : num 7.3
## $ betweenss    : num 215
## $ size         : int [1:7] 10 18 1 4 6 29 17
## $ iter         : int 2
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

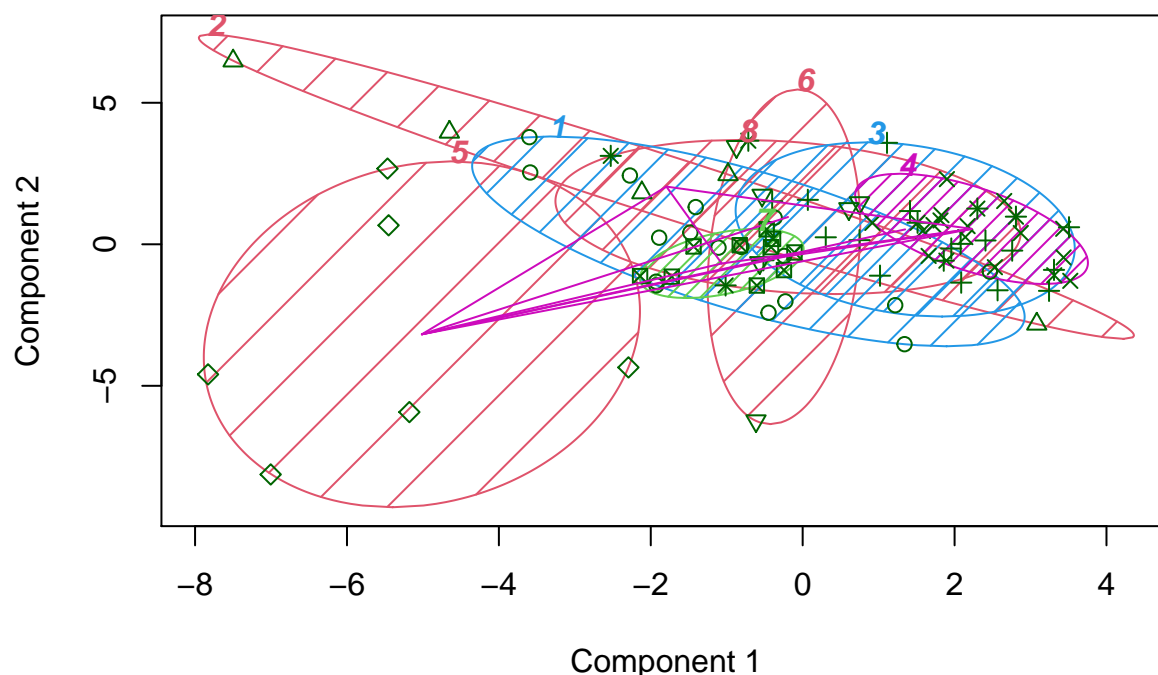
CLUSPLOT(FedPapers_km)



These two components explain 16.4 % of the point variability.

```
## [1] 8
## List of 9
## $ cluster      : Named int [1:85] 7 1 7 1 1 7 1 8 7 1 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:8, 1:71] 0.285 0.225 0.286 0.342 0.171 ...
##   ..- attr(*, "dimnames")=List of 2
##     .. ..$ : chr [1:8] "1" "2" "3" "4" ...
##     .. ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 423
## $ withinss     : num [1:8] 1.195 1.34 1.26 1.462 0.858 ...
## $ tot.withinss : num 7.66
## $ betweenss    : num 416
## $ size         : int [1:8] 17 5 18 17 6 6 10 6
## $ iter         : int 3
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

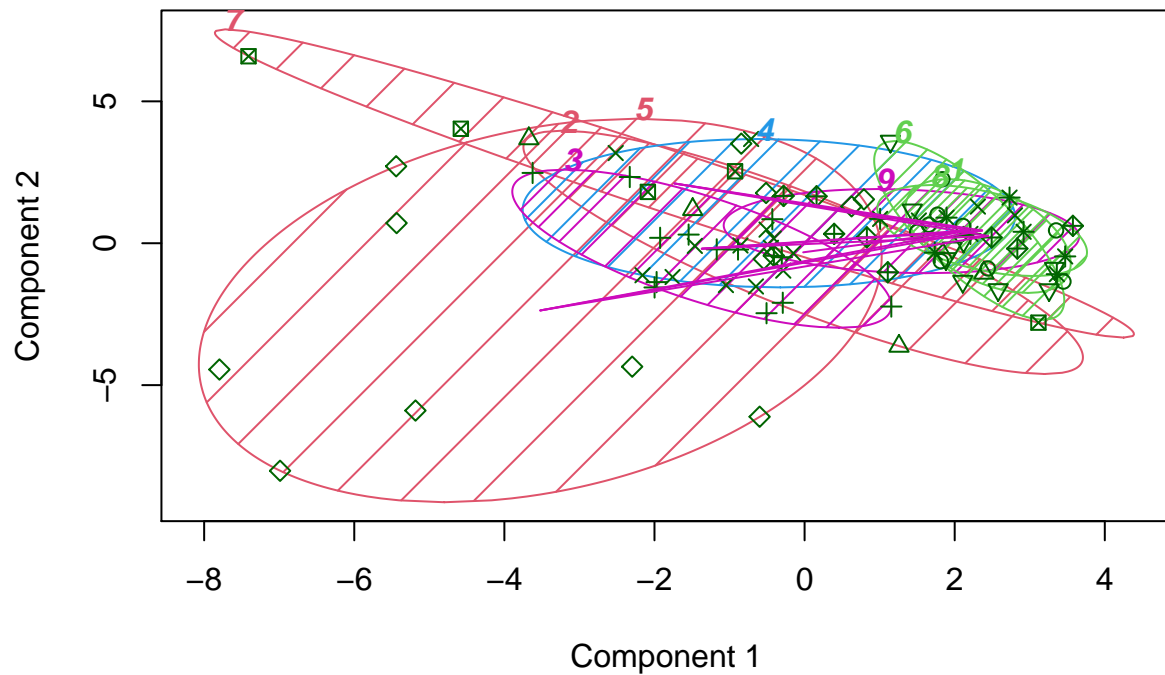
CLUSPLOT(FedPapers_km)



These two components explain 16.39 % of the point variability.

```
## [1] 9
## List of 9
## $ cluster      : Named int [1:85] 4 2 4 2 3 4 3 4 4 3 ...
##   ..- attr(*, "names")= chr [1:85] "dispt_fed_49.txt" "dispt_fed_50.txt" "dispt_fed_51.txt" "dispt_f
## $ centers       : num [1:9, 1:71] 0.329 0.245 0.297 0.308 0.254 ...
##   ..- attr(*, "dimnames")=List of 2
##     ..$ : chr [1:9] "1" "2" "3" "4" ...
##     ..$ : chr [1:71] "a" "all" "also" "an" ...
## $ totss        : num 419
## $ withinss     : num [1:9] 0.675 0.289 0.744 5.623 5.194 ...
## $ tot.withinss : num 14.7
## $ betweenss    : num 404
## $ size         : int [1:9] 9 4 13 16 12 9 5 8 9
## $ iter         : int 3
## $ ifault       : int 0
## - attr(*, "class")= chr "kmeans"
```

CLUSPLOT(FedPapers_km)

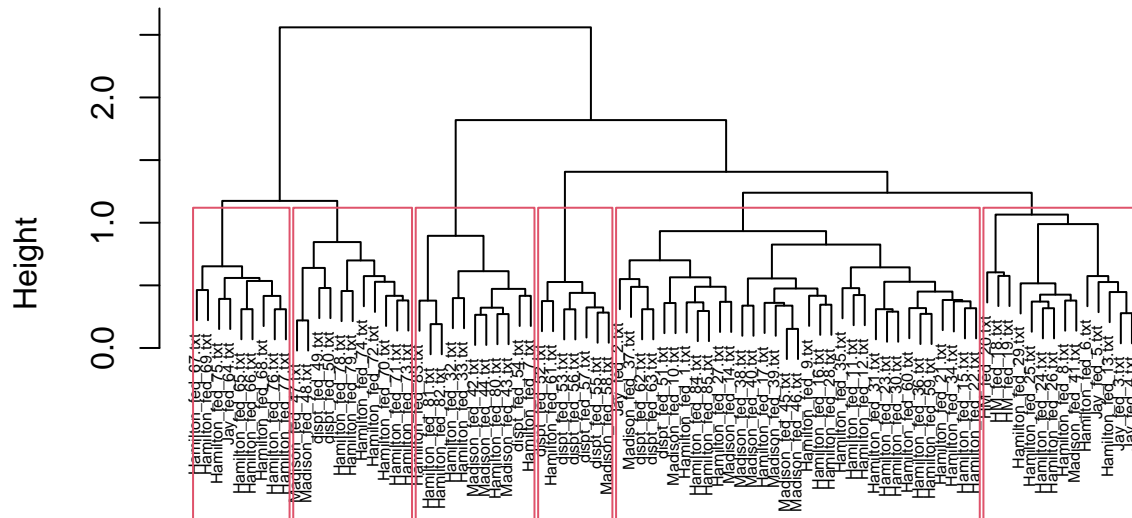


These two components explain 16.41 % of the point variability.

#Cosine Assignment of essays

```
plot(groups_C, main = "Fed Paper Cosine Clustering", cex = 0.5)
rect.hclust(groups_C, k=6)
```

Fed Paper Cosine Clustering



distMatrix_C
hclust (*, "ward.D")

```
authorCut <- cutree(groups_C, k = 6)

(Madison_cos <- FedPapers_km2[which((authorCut == "1") & FedPapers_km2$author == "dispt"),c(1,2)])

##   author      filename
## 1  dispt dispt_fed_49.txt
## 2  dispt dispt_fed_50.txt

(Hamilton_cos <- FedPapers_km2[which((authorCut == "2") & FedPapers_km2$author == "dispt"),c(1,2)])

##   author      filename
## 3  dispt dispt_fed_51.txt
## 10 dispt dispt_fed_62.txt
## 11 dispt dispt_fed_63.txt
```

conclusion

using clustering algorithms k-Means, EM, and HAC techniques the authors of the federalist papers are no longer a mystery. I was able to generate multiple images that give clarity as to who wrote the disputed essays.