

muller

please exoneration
flynn fact
release just hoax piece democrats
like bill now
seen trump went sally
general dont barr see one
yates peoplewont know
report lied
based

smollet

dragged
pill chicago hate cosby community
knows get
amp faked r one completely
getting smollets charges white
kelly now black just crime
made case chris rock

jussie

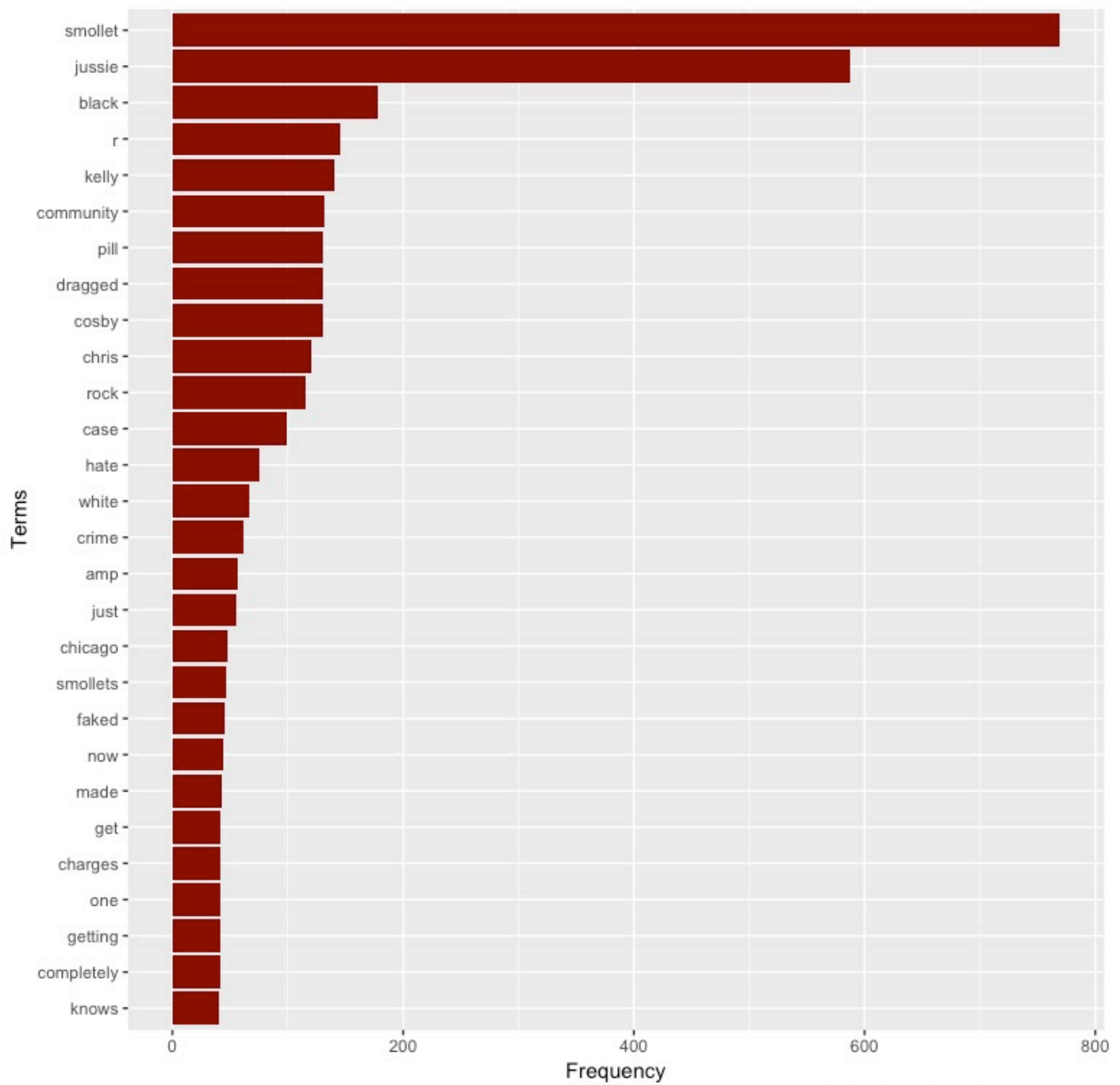
> sort1.termdf_Muller

	term	freq
muller	<u>muller</u>	<u>690</u>
report	<u>report</u>	<u>411</u>
know	<u>know</u>	<u>128</u>
seen	<u>seen</u>	<u>124</u>
barr	<u>barr</u>	<u>122</u>
trump	<u>trump</u>	<u>111</u>
fact	<u>fact</u>	<u>94</u>
dont	<u>dont</u>	<u>72</u>
release	<u>release</u>	<u>72</u>
just	<u>just</u>	<u>69</u>
one	<u>one</u>	<u>65</u>
went	<u>went</u>	<u>58</u>
hoax	<u>hoax</u>	<u>56</u>
exoneration	<u>exoneration</u>	<u>56</u>
democrats	<u>democrats</u>	<u>53</u>
see	<u>see</u>	<u>49</u>
lied	<u>lied</u>	<u>48</u>
like	<u>like</u>	<u>48</u>
based	<u>based</u>	<u>47</u>
wont	<u>wont</u>	<u>47</u>
please	<u>please</u>	<u>46</u>
general	<u>general</u>	<u>46</u>
flynn	<u>flynn</u>	<u>45</u>
sally	<u>sally</u>	<u>45</u>
yates	<u>yates</u>	<u>45</u>
now	<u>now</u>	<u>45</u>
people	<u>people</u>	<u>45</u>
bill	<u>bill</u>	<u>42</u>
piece	<u>piece</u>	<u>40</u>

> sort1.termdf_Smollet

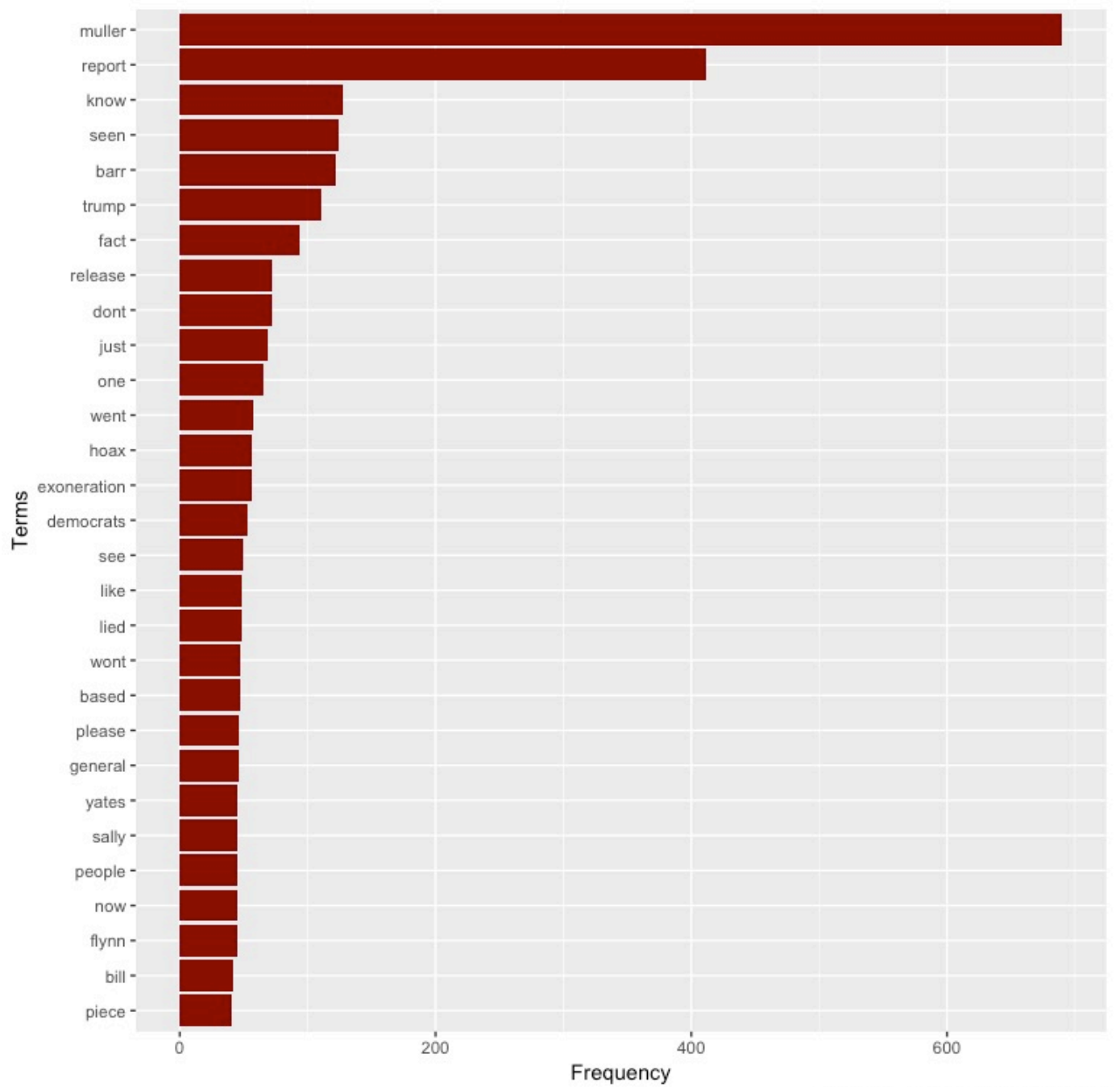
	term	freq
smollet	<u>smollet</u>	<u>769</u>
jussie	<u>jussie</u>	<u>587</u>
black	<u>black</u>	<u>178</u>
r	<u>r</u>	<u>145</u>
kelly	<u>kelly</u>	<u>140</u>
community	<u>community</u>	<u>132</u>
cosby	<u>cosby</u>	<u>131</u>
dragged	<u>dragged</u>	<u>131</u>
pill	<u>pill</u>	<u>131</u>
chris	<u>chris</u>	<u>121</u>
rock	<u>rock</u>	<u>116</u>
case	<u>case</u>	<u>99</u>
hate	<u>hate</u>	<u>75</u>
white	<u>white</u>	<u>67</u>
crime	<u>crime</u>	<u>62</u>
amp	<u>amp</u>	<u>57</u>
just	<u>just</u>	<u>55</u>
chicago	<u>chicago</u>	<u>48</u>
smollets	<u>smollets</u>	<u>47</u>
faked	<u>faked</u>	<u>45</u>
now	<u>now</u>	<u>44</u>
made	<u>made</u>	<u>43</u>
charges	<u>charges</u>	<u>42</u>
get	<u>get</u>	<u>42</u>
getting	<u>getting</u>	<u>41</u>

Bar Plot of #Smollet Tweets



(Based on data retrieved from Twitter)

Bar Plot of #Muller Tweets



(Based on data retrieved from Twitter)

Correlations with Word Associations (Muller):

\$exoneration

smoky	traded	room	nam	player	rahm	maybe	jussie	barr
0.98	0.98	0.96	0.96	0.94	0.94	0.83	0.79	0.41

\$hoax

flynn	know	sally	yates	fact	general
0.89	0.89	0.89	0.89	0.88	0.88

based	lied	went	disgrace	killary	ruined
0.87	0.86	0.78	0.41		0.41 0.41

victim	russiagate	national	life	another	called
0.39	0.37	0.36	0.36	0.33	0.29

Correlations with Word Associations (Smollet):

\$dropped

charges	annalise	keating	wiped	clean	record	getting
0.86	0.83	0.83	0.83	0.80	0.80	0.64

\$hoax

diagram	droppe	overlaps	russia	scientific	showing
0.37	0.37	0.37	0.37	0.37	0.37

hateful	pulled	responsible	likely	main
0.33	0.33	0.30		0.29 0.29

andy	effecthoax	lgbt	ngo
0.29	0.29	0.29	0.29

Clustering with K-MEANS algorithm (Muller):

cluster 1: report muller seen dont trump

cluster 2: gun muller rodger went years

cluster 3: report release democrats barr bill

cluster 4: muller know fact trump barr

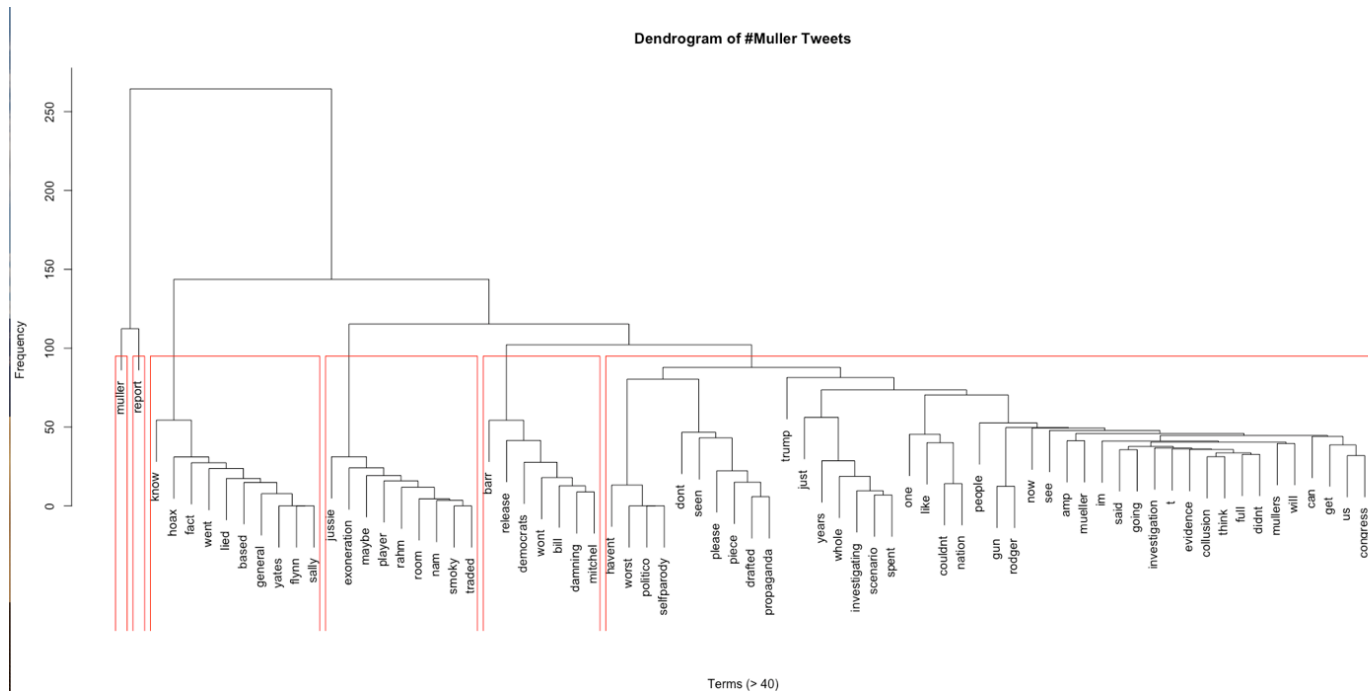
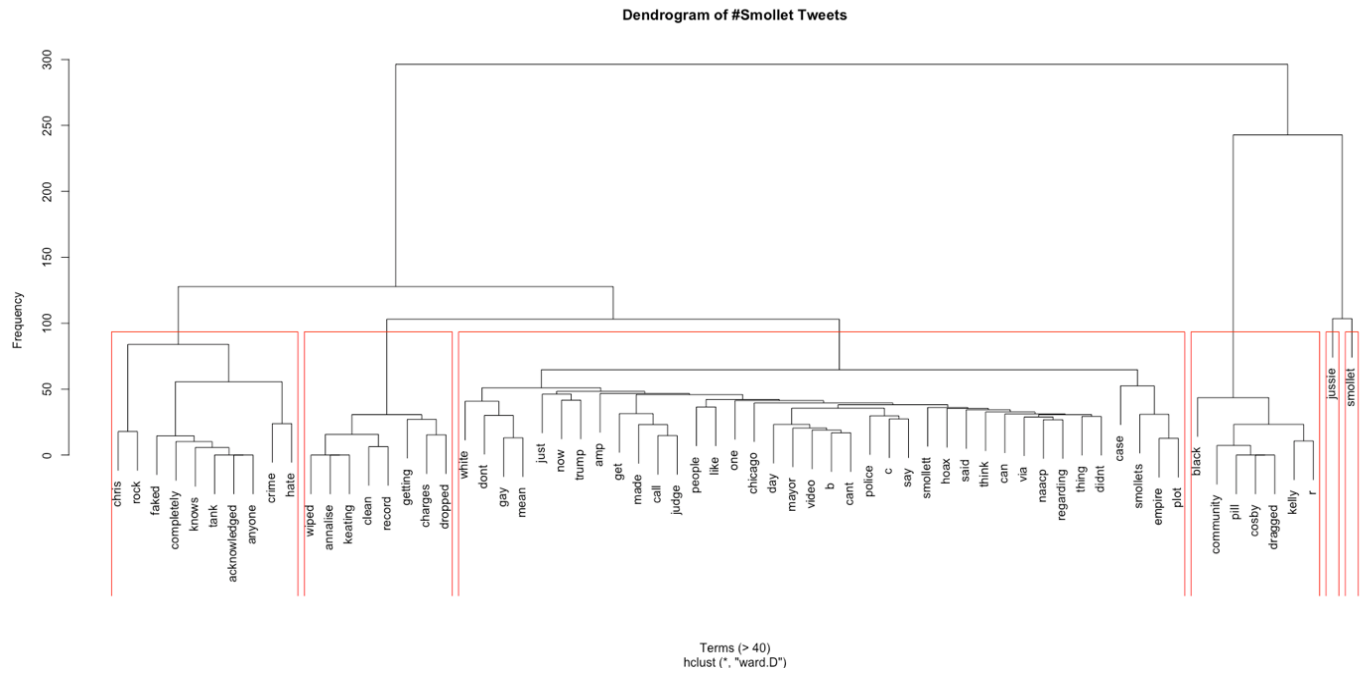
Clustering with K-MEANS algorithm (Smollet):

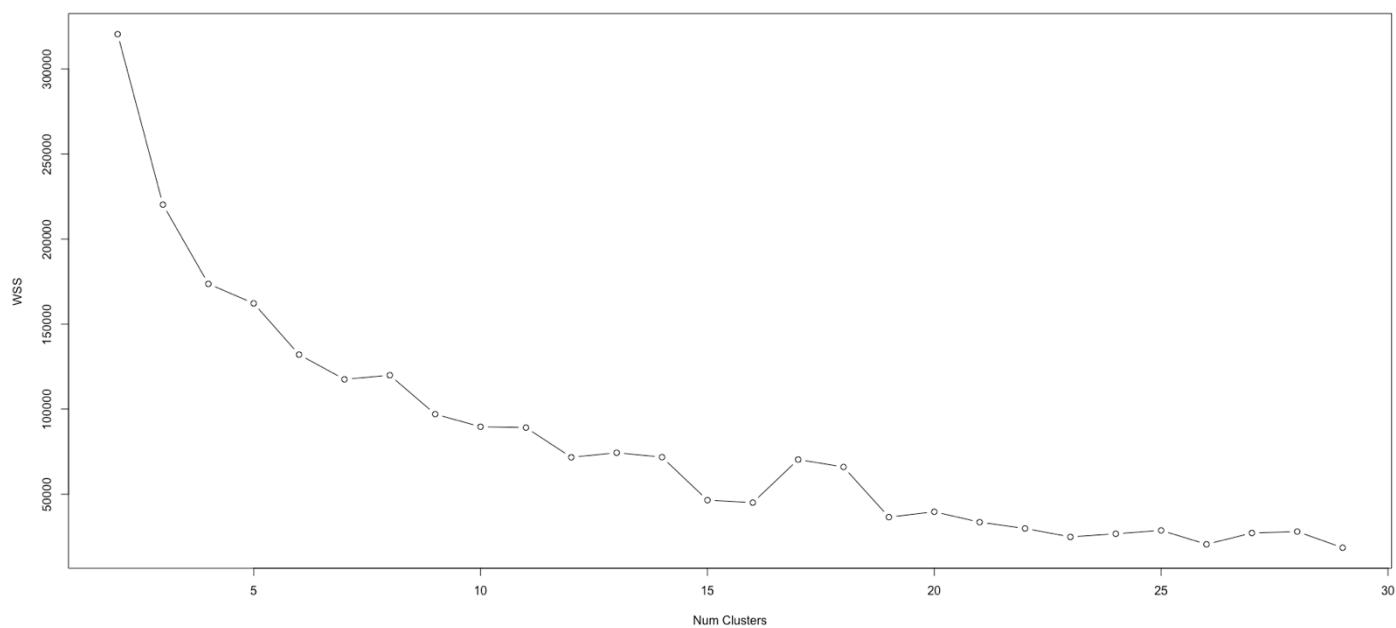
cluster 1: smollet jussie case white just

cluster 2: amp chicago jussie smollet police

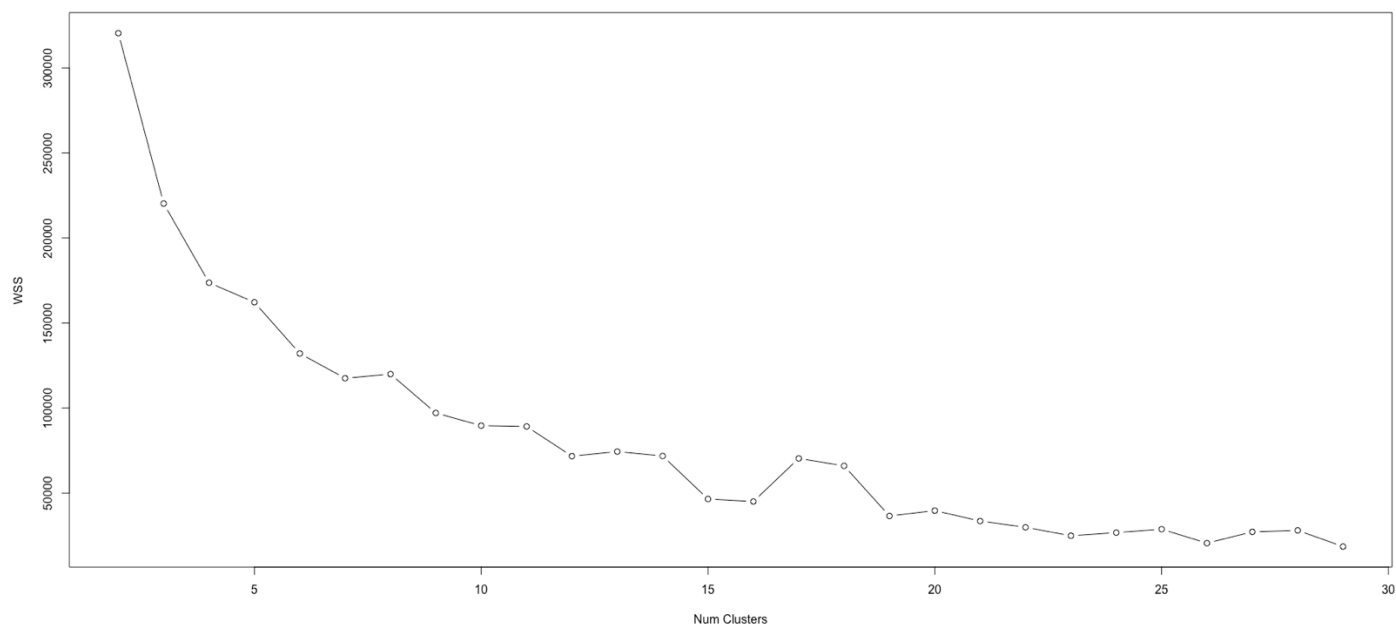
cluster 3: jussie smollet black community cosby

cluster 4: chris rock smollet jussie faked

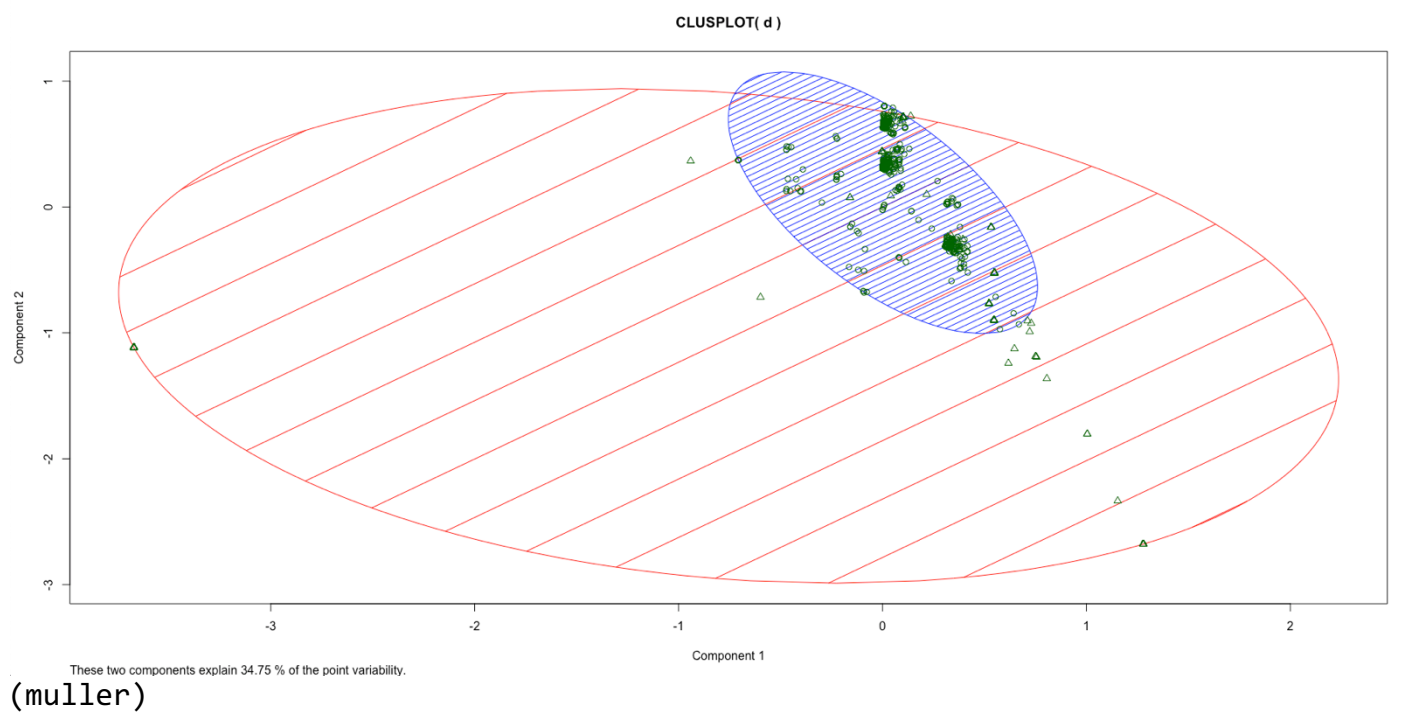
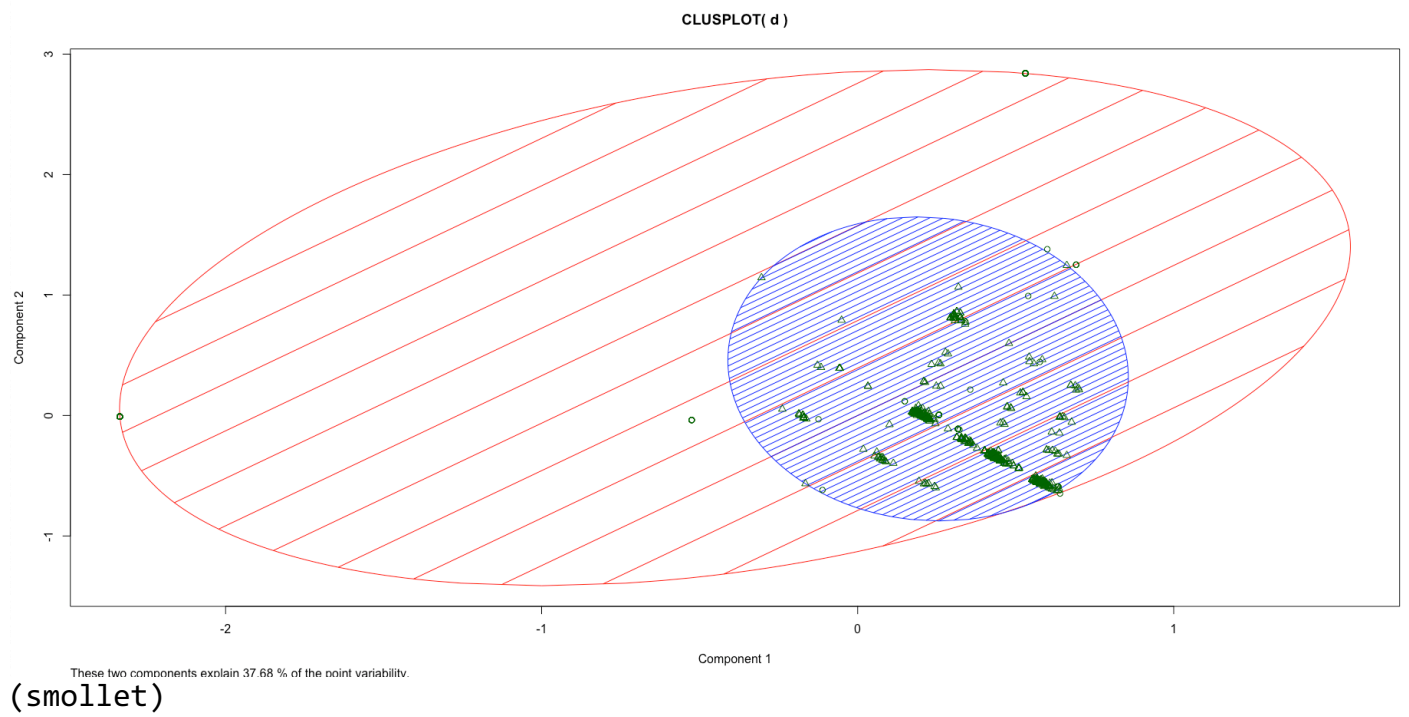




(smollet)



(muller)



R CODE:

```
#### TWITTER SCRAPER #####
```

```
install.packages("twitter")
install.packages("ROAuth")
install.packages("RCurl")
install.packages("RJSONIO")
install.packages("stringr")
install.packages("httr")
install.packages("tm")
install.packages("wordcloud")
```

```
library(twitter)
library(ROAuth)
library(RCurl)
library(RJSONIO)
library(stringr)
library(httr)
library(tm)
library(ggplot2)
library(wordcloud)
library(cluster)
library(stats)
```

```
#Consumer API keys
```

```
api_key <- "VYzwdUqxvZuxjmjW6Ulo4bS6N" #(API key)
api_secret <- "eUhotijA79Xt0G7vgdG60ojQzHeHVCiu81MtZVbPnVLPG#####" #(API secret key,
##### for security)
```

```
#Access token & access token secret
```

```
access_token <- "1109130826619473921-uvCCKi1AjTzCbGLzAlwXZasQKxmMED" #(Access token)
access_token_secret <- "G7gCQ3fNEQ7YVwuwYs6Wp6kukVvKpLfrJz4rFOrX#####" #(Access token
secret, ##### for security)
#connect to Twitter
setup_twitter_oauth(api_key, api_secret, access_token, access_token_secret)
```

```
tweets_Smollet <- searchTwitter("Smollet", n=1000, lang = "en", resultType =
"recent")
tweetsSmolletCopy <- tweets_Smollet
tweets <- tweets_Smollet
tweetsDF1 <- twListToDF(tweets)
```

```
#check info
```

```
dim(tweetsDF1)
str(tweetsDF1)
class(tweetsDF1)
length(tweetsDF1)
attributes(tweetsDF1)
writeLines(tweetsDF1$text[1])
```

```
# check content of several tweets
```

```

for (i in 25:30) {
  cat(paste("[", i, "]", sep = ""))
  writeLines(strwrap(tweets[[i]]$getText(), width = 73))
}

rm(myCorpus)

##### BEGIN CORPUS HERE #####

# create corpus called 'myCorpus'
myCorpus <- Corpus(VectorSource(tweetsDF1$text))

##### DOCUMENT INSPECTION FUNCTION #####
quickCheck <- function(x) {
  for (i in c(25:30)){
    cat(paste0("[", i, "]))
    writeLines(strwrap(as.character(myCorpus[[i]]), width = 60))
  }
}

##### BEGIN CLEANING HERE #####

#### 1. convert myCorpus into lowercase
myCorpus <- tm_map(myCorpus, content_transformer(tolower))

#### 2. remove URL
removeURL <- function(x) gsub("http[^\s:]*", "", x)
myCorpus <- tm_map(myCorpus, content_transformer(removeURL))
quickCheck(myCorpus)

### 3. remove Alias
removeAlias <- function(y) gsub("@[^\s:]*", "", y)
myCorpus <- tm_map(myCorpus, content_transformer(removeAlias))
quickCheck(myCorpus)

### 4. remove anything other than English letters or space
removeNumPunct <- function(x) gsub("[^\s\p{L}]", "", x)
myCorpus <- tm_map(myCorpus, content_transformer(removeNumPunct))
quickCheck(myCorpus)

### 5. remove stopwords
#add custom stopwords
mystopwords <- c(stopwords("english"), "rt", "íí", "o", "d", "got")
#remove stopwords
myCorpus <- tm_map(myCorpus, removeWords, mystopwords)

### 6. remove extra whitespace
myCorpus <- tm_map(myCorpus, stripWhitespace)
quickCheck(myCorpus)

rm(myTDM)

```

```
##### BUILD TERM DOCUMENT MATRIX #####
```

```
myTDM <- TermDocumentMatrix(myCorpus, control = list(wordLengths=c(1,Inf)))
myTDM
```

```
### Frequent Terms
findFreqTerms(myTDM, lowfreq = 40)
termFrequency <- rowSums(as.matrix(myTDM))
termFrequency <- subset(termFrequency, termFrequency>=40)
```

```
# create data frame from termFrequency
termdf <- data.frame(term= names(termFrequency), freq = termFrequency)
termdf_Smollet <- termdf
```

```
## sort frequent terms by order
attach(termdf_Smollet)
sort1.termdf_Smollet <- termdf_Smollet[order(freq, decreasing = TRUE), ]
sort1.termdf_Smollet
```

```
### plot term frequency dataframe w/ ggplot2
ggplot(termdf_Smollet, aes(x = reorder(term, freq), y = freq)) +
  geom_bar(stat = "identity", fill="darkred") + coord_flip() +
  labs(title = "Bar Plot of #Smollet Tweets", caption = "(Based on data retrieved
from Twitter)") + ylab("Frequency") + xlab("Terms")
```

```
### create a Word Cloud
set.seed(142)
wordcloud(words = names(termFrequency), freq = termFrequency, max.words = 40, scale =
c(6, .5), colors = brewer.pal(6, "Dark2"))
```

```
### create a cluster DENDROGRAM
# simplify by removing sparse terms
myTDMsparse <- removeSparseTerms(myTDM, 0.98)
myTDMsparse
```

```
# transform to Matrix without sparse terms
m <- as.matrix(myTDMsparse) # from LIST to MATRIX
is.matrix(m)
```

```
set.seed(122)
# cluster centers
distMatrix <- dist(scale(m))
```

```
fit <- hclust(distMatrix, method = "ward.D")
# plot DENDROGRAM
plot(fit, main = "Dendrogram of #Smollet Tweets", xlab = "Terms (> 40)", ylab =
"Frequency")
rect.hclust(fit, k = 6)
```

```
### find correlations with Word Associations
```

```
findAssocs(myTDM, 'dropped', 0.25)
findAssocs(myTDM, 'hoax', 0.25)
```

```
### Clustering with K-MEANS algorithm
```

```
m2 <- t(m)      #transpose matrix 'm' to 'm2'
set.seed(169)
k <- 4  #choose number of K
kmeansResult <- kmeans(m2, k) #use KMEANS function w/ 'k' number of centroids
round(kmeansResult$centers, digits = 2) #print out data
```

```
## sort top 5 terms by centroid
for (i in 1:k) {
  cat(paste("cluster ", i, ": ", sep = ""))
  s <- sort(kmeansResult$centers[i,], decreasing = T)
  cat(names(s)[1:5], "\n")
}
```

```
#### K-MEANS Cluster Plot
```

```
d <- dist(m2) # create distance vector with transposed matrix: m2
kfit <- kmeans(d, 2, nstart = 100)
clusplot(d, kfit$cluster, diss = T, color = T, shade = T, labels = 0, lines = 0)
```

```
## WSS elbow method (w/ info, re-do K Means if necessary)
```

```
wss <- 2:29
for (i in 2:29) wss[i] <- sum(kmeans(d, centers = i, nstart = 25)$withinss)
plot(2:29, wss[2:29], type = "b", xlab="Num Clusters", ylab= "WSS")
```