

R Notebook

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 3.4.2
```

```
## -- Attaching packages -----
```

```
## v ggplot2 3.1.0      v purrr  0.2.5
```

```
## v tibble  2.0.1      v dplyr  0.7.8
```

```
## v tidyr   0.8.0      v stringr 1.3.1
```

```
## v readr   1.1.1      v forcats 0.3.0
```

```
## Warning: package 'ggplot2' was built under R version 3.4.4
```

```
## Warning: package 'tibble' was built under R version 3.4.4
```

```
## Warning: package 'tidyr' was built under R version 3.4.3
```

```
## Warning: package 'purrr' was built under R version 3.4.4
```

```
## Warning: package 'dplyr' was built under R version 3.4.4
```

```
## Warning: package 'forcats' was built under R version 3.4.3
```

```
## -- Conflicts -----
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()     masks stats::lag()
```

```
library(knitr)
```

```
## Warning: package 'knitr' was built under R version 3.4.3
```

Document clustering, or text clustering, is a very popular application of clustering algorithms. A web search engine, like Google, often returns thousands of results for a simple query. For example, if you type the search term “jaguar” into Google, around 200 million results are returned. This makes it very difficult to browse or find relevant information, especially if the search term has multiple meanings. If we search for “jaguar”, we might be looking for information about the animal, the car, or the Jacksonville Jaguars football team.

Clustering methods can be used to **automatically group search results into categories**, making it easier to find relevant results. This method is used in the search engines PolyMeta and HelioId, as well as on FirstGov.gov, the official Web portal for the U.S. government. The two most common algorithms used for document clustering are Hierarchical and k-means.

In this problem, we'll be clustering articles published on Daily Kos, an American political blog that publishes news and opinion articles written from a progressive point of view. Daily Kos was founded by Markos Moulitsas in 2002, and as of September 2014, the site had an average weekday traffic of hundreds of thousands of visits.

The file **dailykos.csv** contains data on **3,430 news articles or blogs that have been posted on Daily Kos**. These articles were posted in 2004, leading up to the United States Presidential Election. The leading candidates were incumbent President George W. Bush (republican) and John Kerry (democratic). Foreign policy was a dominant topic of the election, specifically, the 2003 invasion of Iraq.

Each of the variables in the dataset is a word that has appeared in at least 50 different articles (1,545 words in total). The set of words has been trimmed according to some of the techniques covered in the previous week on text analytics (punctuation has been removed, and stop words have been removed). For each document, **the variable values are the number of times that word appeared in the document.**

1.1) Hierarchical Clustering

Let's start by building a hierarchical clustering model. First, read the data set into R. Then, compute the distances (using method="euclidean"), and use hclust to build the model (using method="ward.D"). You should cluster on all of the variables.

Running the dist function will probably take you a while. Why? Select all that apply.

```
daily_kos = read.csv('dailykos.csv')
```

```
head(daily_kos)
```

```
##  abandon abc ability abortion absolute abstain abu abuse accept access
## 1      0  0      0      0      0      0      0  0      0      0      0
## 2      0  0      0      0      0      0      0  0      0      0      0
## 3      0  0      0      0      0      0      1  0      0      0      0
## 4      0  0      0      0      0      0      0  0      0      0      0
## 5      0  0      0      0      0      0      0  0      0      0      0
## 6      0  0      0      0      0      0      0  0      0      0      0
##  accomplish account accurate accusations achieve acknowledge act action
## 1      0      0      0      0      0      0      0      0  0  0      2
## 2      0      0      0      0      0      0      0      0  0  0      0
## 3      0      2      0      0      0      0      0      0  0  0      0
## 4      0      0      0      0      2      0      0      0  0  0      0
## 5      0      0      0      0      0      0      0      0  0  0      0
## 6      0      0      0      0      0      0      0      0  0  0      0
##  active activist actual add added addition address admin administration
## 1      0      0      0  0      1      0      0      0      0      1
## 2      0      0      0  0      0      0      0      0      0      0
## 3      0      0      0  0      0      0      0      1      0      0
## 4      0      0      0  0      0      0      0      0      0      0
## 5      0      0      0  0      1      0      0      0      0      0
## 6      0      0      0  0      0      0      0      0      0      0
##  admit advance advantage advertise advised affair affect affiliate
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      1      0      0      0      0      0
## 4      0      0      1      0      0      0      0      0      0
## 5      1      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##  afghanistan afraid afternoon age agencies agenda agree ahead aid aim air
## 1      0      0      0  0      0      0      0      0      0  0  0  0
## 2      0      0      0  0      0      0      0      0      0  0  0  0
## 3      0      0      0  0      0      0      0      0      0  0  0  0
## 4      0      0      0  0      0      0      0      0      0  1  0  0
## 5      0      0      0  0      0      0      0      0      0  1  0  0
## 6      0      0      0  0      0      0      0      0      0  0  0  0
##  alaska allegation allegory allied allowed alternative altsite amazing
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      1      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##  amendment america american amount amp analysis analyst anecdotal anger
## 1      0      0      0      0  0      0      0      0      0  0
```

## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	1	0
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	1	0	0	1	0	0	0
## 6	0	0	0	0	0	0	0	0	0
##	angry	announce	annual	answer	apologies	apparent	appeal	appearance	
## 1	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	
## 4	0	0	0	1	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	
##	applied	appointed	approach	approval	apr	april	arab	area	arent
## 1	0	0	0	1	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0
##	argument	arizona	arm	armstrong	army	arrest	arrive	article	asap
## 1	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	1
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0
##	assess	assist	associate	assume	atlanta	atrios	attack	attempt	attend
## 1	0	0	0	0	0	0	1	0	0
## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	1	0	0	1	0
## 4	0	0	0	0	0	0	2	0	0
## 5	0	0	0	0	0	0	4	0	0
## 6	0	0	0	0	0	0	0	0	0
##	attention	attorney	audience	aug	august	author	average	avoid	backed
## 1	0	0	0	1	0	0	0	0	0
## 2	0	0	0	1	0	0	0	0	0
## 3	0	0	0	1	0	0	0	0	0
## 4	0	0	0	1	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0
## 6	0	0	0	1	0	0	0	0	0
##	background	bad	baghdad	balance	bald	ballot	ban	bank	bar
## 1	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	1	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0
## 6	0	1	0	0	0	0	0	0	0
##	basic	battle	battleground	bear	beat	began	begin	behalf	behavior
## 1	0	1	0	0	0	0	0	0	1
## 2	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0

##	ben	benefit	bet	bid	bigger	biggest	bill	billion	bin	bit	black	blades
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	blame	blast	block	blog	blogger	blogosphere	bloomfield	blow	blue	board		
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	boat	bob	bodies	bomb	book	boost	boston	bottom	bounce	boxblogroll		
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	1	0	0	0	1	0	0	0	0	0	0	0
## 3	0	1	0	0	0	0	0	0	0	0	1	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	boxfeed_listing	boxrdf_feeds	boy	brad	bradnickel	braun	break.	bring				
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	1	0	1	0	0	1	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	broadcast	broke	brought	btw	budget	build	built	bunch	burn	bush		
## 1	0	0	0	0	0	0	0	0	0	2	0	0
## 2	0	0	0	0	0	0	0	0	0	1	0	0
## 3	0	0	0	0	0	0	0	0	0	2	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	12	0	0
## 6	0	0	0	0	0	0	0	0	0	4	0	0
##	bushcheney	bushsux	business	buy	calculate	california	calistan	call	camp			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	1	0	0	0
## 3	0	1	0	0	1	0	0	1	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	campaign	candidacy	candidate	capabilities	capture	car	card	care	career			
## 1	0	0	0	0	0	0	0	0	0	1	0	0
## 2	1	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	5	0	0	0	0	0	0	0	0	0	0	0
## 5	2	0	1	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	carolina	carried	carson	cases	cash	cast	casualties	catch	caucus	caught		
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	1	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	1	0

## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	caused	cbs	cell	center	central	cfr	chair	chairman	challenge	chance	
## 1	0	0	0	0	0	0	0	1	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	4	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	chandler	change	charge	check	chedrcheez	chemical	cheney	chicago	chief		
## 1	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	1	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	children	choice	choose	chose	chris	christopher	cia	cited	cities	citizen	
## 1	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	1	0	0	1	1
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	2	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	civil	civilian	claim	clark	class	clean	clear	click	clinton	clock	close
## 1	0	0	0	0	0	0	0	1	0	0	0
## 2	0	0	0	0	0	0	0	1	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	1	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	closer	club	cnn	coalition	coburn	cole	collect	college	colorado	column	
## 1	0	0	0	1	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	1	0	
##	combat	combination	coming	command	commercial	commission	commit	committee			
## 1	0	0	0	0	0	0	0	0	0	2	
## 2	0	0	0	0	1	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	common	communicate	communities	companies	compare	comparison	compete				
## 1	0	0	0	0	0	2	0	0			
## 2	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	1	0	0			
## 6	0	0	0	0	0	0	0	0			
##	competition	complain	complaint	complete	computer	con	concern	conclude			
## 1	0	0	0	0	0	0	0	0	1		
## 2	0	0	0	0	0	0	0	0	0		

## 3	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0		
##	conclusion	condition	conduct	conference	confidence	confirm	conflict			
## 1	2	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0			
##	congress	congressional	congressman	connect	consequence	conservation				
## 1	0	2	2	0	0	0				
## 2	0	0	0	0	0	0				
## 3	0	0	0	0	0	0				
## 4	1	0	1	0	0	0				
## 5	1	0	1	0	0	0				
## 6	0	0	0	0	0	0				
##	considerable	considered	consistent	constitution	consultant	contact				
## 1	0	1	0	0	0	0				
## 2	0	0	0	0	0	0				
## 3	0	0	0	0	0	0	1			
## 4	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0			
##	content	contest	continue	contrast	contribute	control	controversial			
## 1	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	1			
## 5	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0			
##	convention	conversation	converted	convince	cool	coordinated	corner	corp		
## 1	0	0	0	0	0	0	0	0		
## 2	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0		
## 5	1	0	0	1	0	0	0	0		
## 6	0	0	0	0	0	0	0	0		
##	corporate	correct	corrupt	cost	couldnt	council	count	countdown	counties	
## 1	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	1	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	
##	countries	couple	court	courtesy	cover	coverage	crazy	create	credibility	
## 1	1	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	1	0	
## 4	0	0	0	0	1	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	1	
## 6	0	0	0	0	0	0	0	0	0	
##	credit	crime	criminal	crisis	critic	crooks	cross	crowd	cultural	current

## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	1	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	1	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	1	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	cut	cycle	dakota	damage	damn	dan	danger	danielua	daschle	data	date	david
## 1	0	0	0	1	1	0	0	0	0	0	0	0
## 2	0	0	0	0	1	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	1	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	1	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	dccc	dday	dead	deadline	deal	dean	death	debate	dec	decade	december	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	1	0	0	0	
## 3	0	0	0	0	0	0	0	1	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	decide	decision	declaration	decline	deep	defeat	defend	defense	deficit			
## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	1	0			
## 6	0	0	0	0	0	0	0	0	0			
##	define	delay	delegate	deliver	dem	demand	democracy	democrat	demonstrate			
## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	1	0	0	0	2			
## 4	0	1	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0	0			
##	denied	department	depend	describe	deserve	design	desperate	destroy				
## 1	0	0	0	0	0	0	0	0				
## 2	0	0	0	0	0	0	0	0				
## 3	0	0	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0	0				
## 5	0	0	0	0	0	0	0	0				
## 6	0	0	0	0	0	0	0	0				
##	destruction	detail	determination	develop	dhinmi	dick	die	differ				
## 1	1	0	0	0	0	0	0	0				
## 2	0	0	0	0	0	0	0	0				
## 3	0	0	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0	0				
## 5	0	0	0	0	0	0	0	0				
## 6	0	0	0	0	0	0	0	0				
##	difficult	diplomatic	direct	director	direwolf	disaster	discussed	dismiss				
## 1	0	0	0	0	0	0	0	0				
## 2	0	0	0	0	0	0	0	0				
## 3	0	1	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0	1				
## 5	0	0	1	0	0	0	0	1				

## 6	0	0	0	0	0	0	0	0	0	0
##	dispute	district	divide	division	dkos	dkosopedia	dnc	document	dollar	
## 1	0	2	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	1	0	0	0	
## 4	0	1	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	
##	domestic	dominance	donald	donate	donor	door	doubt	dozen	draft	draw
## 1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0
##	drfranklives	drive	drop	drug	dryfly	dsc	duderino	due	duties	earlier
## 1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	1	0	0	0	0	0	0	0
## 3	0	0	0	0	1	0	0	0	0	0
## 4	0	0	1	0	0	0	0	0	0	0
## 5	0	1	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0
##	early	earn	easier	easily	east	easy	economic	economy	edge	edit
## 1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	1	0	0
## 4	0	0	0	1	0	0	0	0	0	0
## 5	0	0	0	0	0	0	1	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0
##	educated	edward	effect	effort	egon	elect	electoral	email	emerge	employ
## 1	0	0	0	0	0	0	1	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	1	2	0	0	0
## 4	0	0	0	0	0	2	0	0	0	0
## 5	0	0	0	0	0	3	1	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0
##	employee	encourage	end	endorse	endspan	enemies	energy	engage	enjoy	
## 1	0	0	0	0	0	0	0	1	0	
## 2	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	1	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	
## 5	0	0	1	0	0	0	1	0	0	
## 6	0	0	0	0	0	0	0	0	0	
##	ensure	enter	entire	entries	environment	equal	error	essential	establish	
## 1	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	
##	estimate	ethic	evening	event	eventual	evidence	excellent	excited		
## 1	0	0	0	0	0	0	0	0		
## 2	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	0	0	0		

## 4	0	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0	0			
##	executed	exist	exit	expect	expense	experience	expert	explain	explosion			
## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	0	0			
## 3	0	0	1	0	0	1	0	0	0			
## 4	0	0	0	1	1	0	0	0	0			
## 5	0	0	0	2	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0	0			
##	expose	express	extend	extra	extreme	eye	face	fact	factor	fail	failure	
## 1	0	0	0	0	0	0	0	1	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	1	0	0	0	
## 5	0	0	0	0	0	0	2	1	0	2	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	fair	faith	fall	fallujah	familiar	families	faq	farmer	fast	father	favor	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	1	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	1	0	0	0	
##	favorite	fbi	fear	feature	feb	february	federal	feel	fellow	felt	fewer	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	field	fight	figure	file	fill	final	finance	financial	find	finish	fire	
## 1	0	0	0	0	0	1	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	1	0	0	0	0	0	0	0	0	0	
## 6	0	0	1	0	0	0	0	0	0	0	0	
##	firm	fit	fix	flip	floor	florida	fly	focus	folks	follow	force	foreign
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	1	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	1	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	forget	form	forward	fought	found	foundation	fox	francisco	frank	free		
## 1	0	0	0	0	0	0	0	0	0	0		
## 2	0	0	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	1	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	freedom	fresh	friday	friend	front	full	fully	fun	function.	fund		
## 1	0	0	0	0	0	0	0	0	0	0		

## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	1	0	0	0	0	1	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	fundamental	fundraise	funny	future	gain	gallup	game	garner	gatana	gather	
## 1		0	0	0	0	0	0	0	0	0	
## 2		0	0	0	0	0	0	0	0	0	
## 3		0	0	0	0	0	0	0	0	0	
## 4		0	0	0	0	0	0	0	0	0	
## 5		0	0	0	0	0	0	0	0	0	
## 6		0	0	0	0	0	0	0	0	0	
##	gave	gay	gen	general	generate	george	georgia	gephardt	ghraib	giving	
## 1	0	0	0	2	0	0	0	0	0	0	
## 2	0	0	0	1	0	0	0	0	0	0	
## 3	0	0	0	2	0	0	0	0	0	0	
## 4	1	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	1	1	0	0	0	0	0	
## 6	0	0	0	1	0	0	0	0	0	0	
##	global	goal	god	gooper	gop	gore	gotv	gov	govern	governor	grand
## 1	0	0	0	1	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	2	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	1
## 6	0	0	0	0	0	0	0	0	1	0	0
##	greater	greatest	green	ground	group	grow	growth	guard	guess	guest	gun
## 1	1		0	0	0	0	0	0	0	0	0
## 2	0		0	0	0	2	0	0	0	0	0
## 3	0		0	0	0	0	0	0	0	0	0
## 4	0		0	0	0	0	0	0	0	0	0
## 5	0		0	0	0	0	0	0	0	0	0
## 6	0		0	0	0	0	0	0	0	0	0
##	guy	hadnt	half	hall	hampshire	hand	handle	hang	happen	happy	hard
## 1	0	0	0	0		0	0	0	0	0	0
## 2	0	0	0	0		0	0	0	0	0	0
## 3	0	0	0	0		0	0	0	0	0	0
## 4	0	0	0	0		0	0	0	0	0	0
## 5	0	0	0	0		0	0	0	0	0	0
## 6	2	0	0	0		0	0	0	0	0	0
##	hasnt	hate	havent	head	headline	health	hear	heard	heart	heat	heavily
## 1	0	0		0		0	0	0	0	0	0
## 2	0	0		0		0	0	0	0	0	0
## 3	0	0		0		0	0	0	0	0	0
## 4	0	0		0		0	0	0	0	0	0
## 5	0	0		0		0	0	0	0	0	0
## 6	0	0		0		0	0	0	0	0	0
##	heavy	heck	held	hell	helped	heres	herseth	hey	hide	high	higher
## 1	0	0	0	0	0	0		0	0	0	0
## 2	0	0	0	0	0	0		0	0	0	0
## 3	0	0	0	0	0	0		0	0	0	0
## 4	0	0	0	0	0	0		0	0	0	0
## 5	0	0	0	0	0	0		0	0	0	0
## 6	0	0	0	0	0	0		0	0	0	0

##	highlight	hire	historic	history	hit	hold	homepage	honest	honor	hoodies	
## 1	0	0	0	0	0	1	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	1	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	1	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	hope	host	hostile	hot	hotshotxi	hour	house	howard	hstewart	huge	human
## 1	0	0	0	0	0	0	2	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	2	0	0	0	0
## 4	0	0	0	0	0	0	2	0	0	0	0
## 5	1	0	0	0	0	0	1	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	hundred	hurt	husband	hussein	idea	identified	ideological	idetestthispres			
## 1	0	0	0	0	0	0	0	0		0	
## 2	0	0	0	0	0	0	0	0		0	
## 3	0	0	0	0	0	0	0	0		0	
## 4	0	0	0	0	0	0	0	0		0	
## 5	0	0	0	0	0	0	0	0		0	
## 6	0	0	0	0	0	0	0	0		0	
##	ignorance	ill	illinois	image	imagination	immediately	impact	importance			
## 1	0	0	0	0	0	0	0	0		0	
## 2	0	1	0	0	0	0	0	0		0	
## 3	0	1	0	0	0	0	0	0		0	
## 4	0	0	0	0	0	0	0	0		0	
## 5	0	0	0	0	0	0	0	0		0	
## 6	0	0	0	0	0	0	0	0		0	
##	impossible	impressed	improve	incident	incidentally	include	income				
## 1	0	0	0	0	0	1	0	0			
## 2	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0			
##	increase	incumbency	independence	indicating	indicted	individual					
## 1	0	0	1	0	0	0					
## 2	0	0	0	0	0	0					
## 3	0	0	0	0	0	0					
## 4	0	1	0	0	0	0					
## 5	0	4	0	0	0	0					
## 6	0	0	0	0	0	0					
##	industrial	inevitable	influence	info	inform	initial	inside	insist			
## 1	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	2			
## 3	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0			
##	instance	institute	insurgency	intelligence	intent	interest	internal				
## 1	0	0	0	3	0	0	2				
## 2	0	0	0	0	0	0	0				
## 3	0	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0				

## 5	0	0	0	0	0	0	0	0					
## 6	0	0	0	0	0	0	0	0					
##	internet	interview	invasion	investigate	involve	iowa	iraq	iraqi	isnt				
## 1	0	0	0	0	0	0	2	0	0				
## 2	0	0	0	0	0	0	0	0	0				
## 3	0	0	0	0	0	0	0	0	0				
## 4	0	2	0	0	0	0	0	0	0				
## 5	0	0	0	0	0	0	1	0	0				
## 6	0	0	0	0	0	0	0	0	0				
##	issue	itll	ive	james	jan	january	jeff	jerome	jersey	jiacinto	jim	job	joe
## 1	0	0	1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	1	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	2	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0	0
##	john	join	josh	journal	journalist	juan	judge	jul	july	jump	jun	june	
## 1	1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	1	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0	0
##	juppon	justice	karl	katerina	keeping	ken	kennedy	kentucky	kerry				
## 1	0	0	0	0	0	0	0	0	3				
## 2	0	0	0	0	0	0	0	0	7				
## 3	0	0	0	0	0	0	0	0	2				
## 4	0	0	0	0	1	0	0	0	0				
## 5	0	0	0	0	0	0	0	0	5				
## 6	0	0	0	0	0	0	0	0	4				
##	kerryedwards	key	kick	kid	kill	kind	king	kingelection	kitty	knew	knowles		
## 1		0	0	0	0	0	0		0	0	0	0	0
## 2		0	0	0	0	0	0		0	0	0	0	0
## 3		0	0	0	1	0	0		0	1	0	0	0
## 4		0	0	0	0	0	0		0	0	0	0	0
## 5		0	0	0	0	0	0		0	0	0	0	0
## 6		0	0	0	0	0	0		0	0	0	0	0
##	kossacks	kucinich	labor	lack	laden	land	language	large	larger	largest			
## 1	0	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0	0
## 3	1	0	1	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0	0
##	larry	late	latest	launch	law	lawnorder	lawyer	lay	lead	leader	leadership		
## 1	0	0	0	1	0	0	0	0	1	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	1	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0	0
##	leak	lean	learn	leave	led	left	legal	legislation	legislature	legitimate			
## 1	0	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0	0

## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	lets	letter	level	liar	liberal	liberalrakkasan	lie	lieberman	life	light	
## 1	0	1	0	0	0		0	0	0	0	0
## 2	0	0	0	0	0		0	0	0	0	0
## 3	0	0	0	0	0		1	0	0	0	0
## 4	0	0	1	0	0		0	0	0	0	0
## 5	0	0	0	0	0		0	0	0	0	0
## 6	0	0	0	0	0		0	0	0	0	0
##	limit	line	linked	listed	listen	literally	live	local	located	login	long
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	1	1	0
## 4	0	0	0	0	0	0	0	1	0	0	0
## 5	0	0	0	0	0	0	0	1	0	0	0
## 6	0	1	0	0	0	0	0	0	0	0	0
##	longer	looked	lose	loss	lost	lot	louisiana	love	low	lower	lud lzmd
## 1	0	0	0	0	0	0	0	0	0	1	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	1	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	machinations	magazine	main	mainstream	maintain	major	makes	man	manage		
## 1		0	0	0	0	1	0	0	0	0	
## 2		0	0	0	0	0	0	0	0	0	
## 3		0	0	0	0	0	0	0	0	0	
## 4		0	0	0	0	0	0	0	0	1	
## 5		0	0	0	0	0	0	0	0	0	
## 6		0	0	0	0	0	0	0	0	0	
##	mar	march	margin	marine	mark	market	marriage	married	mass	massachusetts	
## 1	0	0	0	0	0	0	0	0	1	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	1	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	massive	match	material	materiel	matter	mccain	meaning	meant	measure		
## 1	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	1	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	media	medical	meet	melanie	member	memo	memorial	men	mention	menu	mess
## 1	0	0	0	0	1	0	0	0	0	0	0
## 2	0	0	0	0	1	0	0	0	0	0	0
## 3	1	0	0	0	0	0	0	0	0	1	0
## 4	1	0	0	0	1	0	0	0	0	0	0
## 5	3	0	0	0	1	0	0	0	1	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	message	met	meteor	meter	mexico	mich	michael	michigan	midday	middle	mike

## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	1	0	0	0	0
## 3	0	0	0	1	0	0	0	0	1	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	miles	military	milller	million	mind	mine	minister	minnesota	minor	minute	
## 1	0	5	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	miss	mission	missouri	mistake	misterajc	mix	model	moderate	mom	moment	
## 1	0	0	0	1	0	0	0	0	0	0	0
## 2	0	2	0	0	0	0	0	0	0	1	
## 3	0	0	0	0	0	0	0	0	0	1	
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	1	0	0	0	0	0	0	0	0
##	momentum	monday	money	montclair	month	moore	morning	morrison	mosh	mother	
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	1	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	1	0	0	0	0
## 5	0	0	0	0	1	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	motivated	move	movement	mrliberal	mydd	nader	nag	named	nancy	narrow	
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	1	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	nation	nationwide	natural	nbc	needed	negative	neighbor	net	network		
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	1	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	4	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	nevada	news	newspaper	newsweek	newwindow	nice	night	nominate	nominee		
## 1	0	1	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	1	0	0	0	0	0	0
## 4	0	1	0	0	0	0	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0	0	0
## 6	1	1	0	0	0	0	0	0	0	0	0
##	north	note	notice	nov	november	nprigo	nro	nuclear	number	obama	object
## 1	0	0	0	0	1	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	1	10	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0

## 6	0	0	0	0	0	0	0	0	0	0	0
##	observation	obtain	obvious	occupation	occur	oct	october	offense	offer		
## 1		0	0	0	0	0	0	0	0	0	
## 2		0	0	0	0	0	0	0	0	0	
## 3		0	0	0	0	0	0	0	0	0	
## 4		0	0	0	0	0	0	0	0	0	
## 5		0	0	0	0	0	0	0	0	0	
## 6		0	0	0	0	0	0	0	0	0	
##	office	official	ohio	oil	oklahoma	online	opened				
## 1	0		0	0	0	0	0				
## 2	0		0	0	0	1	0				
## 3	0		0	1	0	1	0				
## 4	0		0	0	0	0	0				
## 5	1		0	1	0	0	0				
## 6	0		0	0	0	0	0				
##	openhttpwwedwardsforprezcomdailykoshtml	operate	opinion	opponent							
## 1					0	0	0	0			
## 2					0	2	0	0			
## 3					1	0	0	0			
## 4					0	0	0	2			
## 5					0	0	0	2			
## 6					0	0	0	0			
##	opportunities	oppose	opposite	optimist	option	order	oregon	organization			
## 1		2	0	0	0	0	0	0	0		
## 2		0	0	0	0	0	0	0	0		
## 3		0	0	0	0	0	0	0	0		
## 4		0	0	0	0	0	0	0	0		
## 5		0	0	0	0	0	0	0	0		
## 6		0	0	0	0	0	0	0	0		
##	original	osama	ourcongressorg	outrage	overwhelming	page	paid	panel	paper		
## 1	0	0		0	0	0	0	0	0	0	
## 2	0	0		0	0	0	0	0	0	0	
## 3	0	0		1	0	0	0	0	0	0	
## 4	0	0		0	0	0	2	0	0	0	
## 5	0	0		0	0	0	0	0	0	0	
## 6	0	0		0	0	0	0	0	0	0	
##	parecommend	parenthesis	participants	parties	partisan	partly	partner				
## 1	0		0		0	0	0	0	0		
## 2	0		0		0	0	0	0	0		
## 3	0		0		0	0	0	0	0		
## 4	0		0		0	0	0	0	0		
## 5	0		0		0	0	0	0	0		
## 6	0		0		0	0	0	0	0		
##	pass	password	past	pat	patriot	paul	pay	pdf	peace	peanut	pennsylvania
## 1	0		0	0	0	0	0	0	0		1
## 2	0		0	0	0	0	0	0	0		0
## 3	0		1	0	0	0	0	0	0	1	0
## 4	0		0	1	0	0	0	0	0		0
## 5	0		0	0	0	0	0	0	0		0
## 6	0		0	0	0	0	0	4	0	0	0
##	pentagon	people	percent	percentage	perfect	perform	period	person			
## 1	0		0	4	0	0	0	0	0		
## 2	0		0	0	0	0	0	0	0		
## 3	0		0	0	0	0	0	0	0		

## 4	0	0	0	0	0	0	0	0	0	
## 5	0	1	4	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	
##	personnel	perspective	philly	phone	photo	pick	pickup	picture	piece	place
## 1	0		0	0	0	1	1	0	0	0
## 2	0		0	0	0	0	0	0	0	1
## 3	0		0	1	0	0	0	0	0	0
## 4	0		0	0	0	0	0	0	0	0
## 5	0		0	0	0	0	0	0	0	0
## 6	0		0	0	0	0	0	0	0	0
##	plan	plane	play	plenty	pointed	police	policies	political	politician	poll
## 1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	2	0	0	0	0	0	0	0
## 3	1	0	0	0	0	0	0	0	0	4
## 4	0	0	0	0	0	3	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	2
## 6	0	0	0	0	0	0	0	0	0	0
##	pollster	poor	popular	population	position	possibilities	posted	potential		
## 1	0	0		0	0	0	0	0	0	0
## 2	0	0		0	0	0	0	0	0	0
## 3	0	0		0	0	1	0	0	0	0
## 4	0	0		0	0	0	0	0	0	0
## 5	0	0		0	0	2	0	0	0	0
## 6	0	0		0	0	0	0	0	0	0
##	powell	power	practical	predict	prefer	preparation	presence	present		
## 1	0	0		0	0	0	0	0	0	0
## 2	0	0		0	0	0	0	0	0	0
## 3	0	1		0	0	0	0	0	0	0
## 4	0	1		0	0	0	0	0	0	0
## 5	0	0		0	0	0	0	0	0	0
## 6	0	0		0	0	0	0	0	0	0
##	presided	presidential	press	pressure	pretend	pretty	prevent	previous		
## 1	1		0	0	0	0	0	0	0	0
## 2	0		0	0	0	0	0	0	0	0
## 3	0		0	0	0	0	0	0	0	0
## 4	0		0	1	1	0	1	0	0	0
## 5	7		1	0	0	0	1	0	0	0
## 6	0		0	0	0	0	0	0	0	0
##	price	pride	primaries	prime	prior	priorities	prison	private	problem	
## 1	0	0		0	0	0	0	0	0	0
## 2	0	0		0	0	0	0	0	0	0
## 3	0	0		1	0	0	0	0	0	0
## 4	0	0		0	0	0	0	0	2	0
## 5	0	0		0	0	0	0	0	0	0
## 6	0	0		0	0	0	0	0	0	0
##	process	produce	product	profile	program	progress	project	promise	promote	
## 1	0	0		0	0	0	0	0	0	0
## 2	0	0		0	0	0	0	0	0	0
## 3	0	0		0	0	0	0	0	0	0
## 4	0	0		0	0	0	0	0	0	0
## 5	0	0		0	0	0	0	0	0	0
## 6	0	0		0	0	0	0	0	0	0
##	proof	proper	proposal	prospect	protect	protest	proud	prove	provide	
## 1	0	0		0	0	0	0	1	0	0

## 2	0	0	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	1	1	0	0	0		
## 4	0	0	0	0	0	0	0	1	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	public	publish	pull	pundit	punkmonk	purpose	push	put	qaeda	qaqaa	quarter	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	3	0	3	0	0	0	0	0	0	0	0	
## 5	0	0	1	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	question	quick	quiet	quote	race	racine	rad	radio	raise	rallies	ralph	ran
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	2	1	1	0	0	0	0	0
## 4	1	0	0	0	2	0	0	0	0	0	0	0
## 5	0	0	0	0	1	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	range	rank	rapid	rare	rasmussen	rate	reach	reaction	read	reader	ready	
## 1	0	0	0	0	0	0	1	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	1	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	reagan	real	realities	realization	reason	receive	recently	recognize				
## 1	0	1	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	recommend	record	recruit	red	reduce	reelect	refer	reflect	reform	refusal		
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	1	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	1	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	regime	region	register	registrants	regular	reject	related	release				
## 1	0	0	0	0	0	0	1	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	1	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	relevance	religious	remain	remark	remember	remind	remove	rep	repeat.			
## 1	0	0	0	0	0	0	0	1	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	1	0	0	0	1	0	0	0	0	0
## 5	0	0	1	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0

##	replace	report	represent	repub	republican	republicans	forkerry	request				
## 1	0	0	0	0	2		0	0				
## 2	0	0	0	0	0		0	1				
## 3	0	1	0	1	2		1	0				
## 4	0	3	0	0	0		0	0				
## 5	0	0	0	0	1		0	0				
## 6	0	0	0	0	0		0	0				
##	require	research	reserve	resign	resist	resource	respect	respond	response			
## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	1	0			
## 3	0	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0	0			
##	rest	result	retire	return	reuters	reveal	reverse	review	rhetoric	rice		
## 1	0	1	0	0	0	1	0	0	0	0		
## 2	0	0	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0	3	0	0	0	0	0	0	0	0		
##	richard	ridiculous	rightly	rightwing	rise	risk	rival	rnc	road	robert		
## 1	1	0	0	0	0	0	0	0	0	0		
## 2	0	0	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	role	roll	room	root	rough	round	rove	rule	rumor	rumsfeld	running	saddam
## 1	0	0	0	0	0	0	0	0	0	0	0	1
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	safe	salazar	sample	san	sappy	saturday	save	scandal	scare	scene	schaller	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	1	0	0	0	0	0	0	0	0	0	
##	schedule	school	science	scientist	scoop	score	scott	screen	seamus	search		
## 1	0	0	0	0	0	0	0	0	0	0		
## 2	0	0	0	0	0	0	0	0	0	0		
## 3	0	0	0	0	1	0	0	0	1	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	season	seat	secret	secretaries	section	secure	seek	select	sen	senate		
## 1	0	1	0	0	0	0	0	0	0	0		
## 2	0	0	0	2	0	0	0	0	0	0		
## 3	0	0	0	0	1	0	0	1	0	2		
## 4	0	0	0	0	0	0	0	0	0	0		

## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	3
##	senategovernors	send	senior	sense	sep	sept	september	series	serve	server	
## 1		0	0	1	0	0	0	0	0	0	0
## 2		0	0	0	0	0	0	0	0	0	0
## 3		1	0	0	0	0	0	0	0	0	0
## 4		0	0	0	0	0	0	0	0	0	0
## 5		0	0	0	0	0	0	0	0	0	0
## 6		0	0	0	0	0	0	0	0	0	0
##	service	session	set	shape	share	sharpton	shes	shift	shit	shock	short
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	1	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	15	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	1	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	shot	shouldnt	showed	shown	side	sign	signature	significance	similar		
## 1	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	1	0	0	16	0	0	0	0	
## 5	0	0	0	0	0	2	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	
##	simple	simply	single	sit	sites	situation	slate	slight	slip	slow	small
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	1	0	0	0	0	0	0	0
## 5	0	0	0	1	0	1	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	smear	social	soldier	solid	son	sort	sound	source	south	southern	space
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	1	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	1	0	0	0	0
## 5	0	0	0	0	0	0	1	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	speak	speaker	special	specific	specter	speculate	speech	spend	spent	spin	
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	split	spoke	spokesman	spot	spread	staff	staffer	stage	stand	standard	
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	1	0	0	0	0	0	0	0	0
## 3	1	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0
##	star	stark	start	startspan	state	statement	statewide	station	statistical		
## 1	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	2	0	0	0	0	0

## 3	0	0	0	1	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	1	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	status	stay	steal	step	steve	stick	stock	stolen	stop	store	stories	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	1	1	0	0	0	0	1	0	1	0	
## 4	0	0	2	0	0	1	0	1	0	0	0	
## 5	0	1	0	1	0	1	0	1	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	straight	strategic	strategies	strategist	street	strength	strike	strong				
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	1	0	0	0	0	0	0	0	0	0	0	
## 4	1	0	0	0	0	1	0	0	0	0	0	
## 5	0	0	1	0	0	1	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	stronger	struck	student	studies	stuff	stupid	subject	success	sudden			
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	1	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	suffer	suggest	suit	summer	sun	sunday	sunzoo	support	suppose	suppress		
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	2	0	0	1	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	supreme	surely	surge	surprise	survey	surveyusa	survival	susa	suspect			
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	4	0	1	0	0	0	
##	swift	swing	switch	system	table	tactic	takes	talked	tank	target	task	tax
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	1	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	team	televised	telling	ten	term	territory	terror	terrorist	test			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	1	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	testimony	texas	theme	theories	theyll	theyre	theyve	thinking	thought			

## 1	0	0	0	0	0	0	0	0	0			
## 2	0	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	0	0	0	0			
## 4	0	0	0	0	0	0	0	0	0			
## 5	0	0	1	0	0	0	0	0	0			
## 6	0	0	0	0	0	0	0	0	0			
##	thousand	threat	threaten	throw	thursday	ticket	tie	tight	time	tip	today's	
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	1	0	0	
## 3	0	0	0	0	0	1	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	3	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	tom	tomorrow	tonight	tony	tool	top	topic	total	tough	town	track	trade
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	1	0	2	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	tradesports	tradition	traffic	trail	train	transfer	trapper	travel	treat			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	treatment	trend	trial	trip	troop	trouble	TRUE.	trust	truth	tuesday		
## 1	0	0	0	0	0	0	1	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	1	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	1	0	0	0	0	0	0
## 5	0	0	0	0	0	1	0	0	0	0	0	0
## 6	0	1	0	0	0	0	0	0	0	0	0	0
##	tunesmith	turn	turnout	type	typical	ultimate	undecided	understand	union			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	1	0	0	0	0	0	0	0	0
## 5	0	0	0	1	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	unit	universal	unlike	updated	upenn	urge	usa	username	usual			
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	1	1	0	0	1	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	vaantirepublican	values	var	vast	verification	version	veteran	vice				
## 1	0	0	0	0	0	0	0	1	0	0	0	0
## 2	0	0	0	0	0	0	0	2	0	0	0	0
## 3	0	0	1	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0

## 6		0	0	0	0		0	0	0	0		
##	victim	victories	video	vietnam	view	violate	violence	virginia				
## 1	0	0	0	0	1	0	0	0				
## 2	0	0	0	2	0	0	0	0				
## 3	0	1	0	0	0	0	0	0				
## 4	0	0	0	0	0	0	0	0				
## 5	0	0	0	0	0	0	0	0				
## 6	0	0	0	0	0	0	0	0				
##	virginiadem	virtual	visit	voice	volunteer	vote	voter	vulnerable	wage			
## 1	0	0	0	0	0	1	2	0	0			
## 2	0	0	0	0	0	0	0	0	0			
## 3	0	0	0	0	0	8	2	0	0			
## 4	0	0	0	0	0	0	0	0	0			
## 5	0	0	0	0	0	0	1	0	1			
## 6	0	0	0	0	0	0	0	0	0			
##	wait	walk	wall	wanted	wapo	war	warn	washington	wasnt	waste	watch	
## 1	0	0	0	0	0	4	0	0	0	0	0	
## 2	0	0	0	0	0	2	0	0	1	0	2	
## 3	1	0	0	0	0	1	0	0	0	0	2	
## 4	0	0	0	0	0	0	0	0	1	0	0	
## 5	0	0	0	1	0	1	0	1	0	0	1	
## 6	0	0	0	0	0	0	0	1	0	0	0	
##	watchers	water	ways	wclathe	weak	weapon	web	website	wednesday	weekend		
## 1	0	0	0	0	0	1	0	0	0	0		
## 2	0	0	0	0	0	0	0	0	0	0		
## 3	1	0	0	0	0	0	0	0	0	0		
## 4	0	0	0	0	0	0	0	0	0	0		
## 5	0	0	0	0	0	0	0	0	0	0		
## 6	0	0	0	0	0	0	0	0	0	0		
##	weekly	werent	west	weve	whats	white	whos	wide	wife	william	win	wingnut
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	1	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	1	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	winner	wisconsin	witness	woes	woman	women	won	wonderful	wont	word	worked	
## 1	0	0	0	0	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0
##	worker	worried	worse	worst	worth	wouldnt	wouldve	wound	write	writer		
## 1	0	0	0	0	0	0	0	0	0	0	0	
## 2	0	0	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	
## 4	0	0	0	0	0	0	0	0	0	0	0	
## 5	0	0	0	0	0	0	0	0	0	0	0	
## 6	0	0	0	0	0	0	0	0	0	0	0	
##	written	wrong	wrote	yahoo	yeah	year	yesterday	york	youll	young	youre	
## 1	0	2	1	0	0	0	0	0	0	0	0	
## 2	0	1	0	0	0	0	0	0	0	0	0	
## 3	0	0	0	0	0	0	0	0	0	0	0	

```
## 4      0      0      0      0      0      0      2      0      0      1      0
## 5      0      0      0      0      0      1      1      0      0      1      0
## 6      0      0      0      0      0      0      0      0      0      0      0
##   youve zogby zone
## 1      0      0      1
## 2      0      0      0
## 3      0      0      0
## 4      0      0      0
## 5      0      0      0
## 6      0      0      0
```

```
include_graphics('1.1.png')
```

☒ We have a lot of observations, so it takes a long time to compute the distances for all observations. ✓

☒ We have a lot of variables, so the distance computation is long. ✓

☐ Our variables have a wide range of values, so the distances are more sensitive to outliers.

☐ The euclidean distance is known to take a long time to compute, especially for large datasets.



Explanation

You can read in the data set, compute the distances, and build the hierarchical clustering using the following commands:

```
dailykos = read.csv("dailykos.csv")
```

```
kosDist = dist(dailykos, method="euclidean")
```

```
kosHierClust = hclust(kosDist, method="ward.D")
```

The distance computation can take a long time if you have a lot of observations.

As we saw in recitation, it might not even work if you have too many observations.

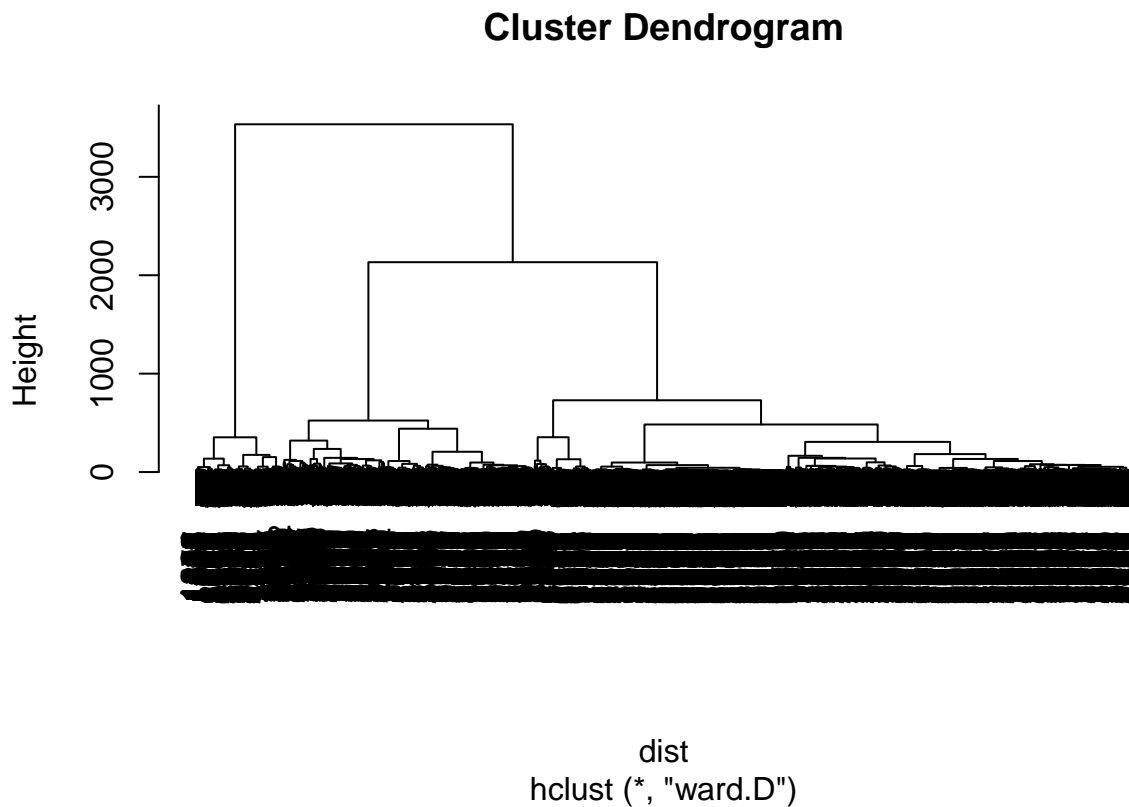
```
dist = dist(daily_kos, method = 'euclidian')
```

```
cluster = hclust(dist, method = 'ward.D')
```

1.2) Hierarchical Clustering

Plot the dendrogram of your hierarchical clustering model. Just looking at the dendrogram, which of the following seem like good choices for the number of clusters? Select all that apply.

```
plot(cluster)
```



```
include_graphics('1.2.png')
```


☒ 2 ✓

☒ 3 ✓

☐ 5

☐ 6



Explanation

You can plot the dendrogram with the command:

```
plot(kosHierClust)
```

where "kosHierClust" is the name of your clustering model.

The choices 2 and 3 are good cluster choices according to the dendrogram. The horizontal lines in the dendrogram in those cut off spots (draw a line across 2 or 3 vertical lines). The choices of 5 and 6 do not seem good because there is very little space.

1.3) Hierarchical Clustering

In this problem, we are trying to cluster news articles or blog posts into groups. This can be used to show readers categories to choose from when trying to decide what to read. Just thinking about this application, what are good choices for the number of clusters? Select all that apply.

```
include_graphics('1.3.png')
```

☐ 2

☒ 3

☐ 7 ✓

☒ 8 ✓

Explanation

Thinking about the application, it is probably better to show the reader categories would probably be too broad to be useful. Seven or eight ca

1.4) Hierarchical Clustering

Let's pick 7 clusters. This number is reasonable according to the dendrogram, and also seems reasonable for the application. Use the `cutree` function to split your data into 7 clusters.

Now, we don't really want to run `tapply` on every single variable when we have over 1,000 different variables. Let's instead use the `subset` function to subset our data by cluster. Create 7 new datasets, each containing the observations from one of the clusters.

How many observations are in cluster 3?

```
clusters_no = cutree(cluster, 7)
```

```
table(clusters_no)
```

```
## clusters_no
##      1      2      3      4      5      6      7
## 1266   321   374   139   407   714   209
```

```
clusters_no
```

```
##      [1] 1 1 2 3 4 1 5 6 1 6 6 2 2 6 1 3 1 1 1 1 6 6 7 2 6 5 5 1 6 1 1 3 6 6
##      [35] 2 1 6 5 3 1 5 1 1 6 2 1 6 1 1 4 4 6 6 3 1 1 1 2 3 1 1 5 6 6 1 5 6 5
##      [69] 2 1 1 1 1 6 1 4 4 3 1 2 1 1 1 1 5 3 1 5 6 2 6 1 4 1 1 6 1 6 1 2 1 6
##     [103] 1 4 4 5 1 1 6 6 6 6 2 1 1 1 1 1 1 1 1 1 6 6 2 1 1 2 1 1 3 1 1 1 1
##     [137] 5 1 1 6 2 5 6 1 4 1 1 2 6 6 2 1 1 1 1 5 5 6 1 4 1 1 2 1 1 5 1 6 6 4
##     [171] 3 1 5 1 2 1 6 6 1 1 1 1 3 1 4 5 2 6 6 3 6 5 1 2 3 6 3 3 6 5 3 3 2 3
##     [205] 6 5 4 6 6 1 6 2 6 1 6 6 6 6 6 2 1 3 6 6 5 5 5 1 6 3 2 3 2 1 1 1 1 1
##     [239] 4 1 1 2 6 6 3 5 1 5 6 6 3 2 3 6 6 5 6 1 6 1 4 2 1 3 4 6 4 6 3 5 6 2
```

```

## [273] 3 1 1 1 1 6 6 1 2 1 3 1 1 4 1 3 6 6 5 2 4 1 3 1 6 6 6 1 1 2 6 3 6 1
## [307] 6 6 6 1 2 6 6 6 1 5 1 6 1 1 2 6 6 1 3 6 1 1 6 4 2 5 1 2 6 3 1 1 1 1
## [341] 1 6 2 1 1 3 1 6 6 6 5 1 2 1 1 1 6 1 1 6 4 3 1 2 3 1 6 1 1 1 6 1 3
## [375] 7 1 2 3 1 4 1 1 1 1 1 1 6 2 1 3 1 1 3 6 3 1 1 2 5 1 3 1 6 1 4 5 5 1
## [409] 2 1 3 3 1 4 1 6 1 5 6 6 6 2 1 5 1 6 1 1 1 1 1 1 1 3 1 2 6 5 5 6 6 6
## [443] 3 4 6 2 6 6 2 1 1 6 7 6 7 1 5 2 6 6 1 7 1 6 1 5 1 1 2 1 1 1 6 1 1 6
## [477] 1 1 1 4 2 4 6 1 1 1 5 4 1 1 6 1 2 3 3 1 5 1 4 3 5 6 1 1 2 1 4 3 1 5
## [511] 1 1 1 6 6 2 1 6 1 1 4 1 6 1 6 1 6 5 2 4 4 1 1 5 5 6 6 1 1 2 6 3 6 1
## [545] 5 1 6 1 4 1 6 1 2 4 1 3 3 1 1 4 5 1 1 1 1 2 1 5 2 6 1 1 1 1 3 6 3 1
## [579] 6 2 7 1 1 6 3 1 6 6 6 1 1 1 2 6 3 1 3 3 6 1 3 1 1 4 1 2 1 5 1 1 1 4
## [613] 1 1 1 6 1 5 2 3 1 5 5 6 1 1 1 1 2 5 5 4 6 1 1 1 1 5 1 1 6 2 1 1 5 1
## [647] 1 6 6 1 2 6 7 6 3 1 3 1 6 1 1 2 6 6 3 5 5 1 5 3 1 1 4 2 5 1 6 6 3 1
## [681] 1 6 3 6 2 3 2 1 6 1 1 6 1 1 6 5 3 1 2 3 6 6 1 1 1 3 1 1 1 2 6 6 6 4
## [715] 6 1 3 1 4 3 2 3 5 3 1 6 6 6 6 6 6 6 2 1 3 6 6 5 1 5 1 1 6 2 1 6 1 1
## [749] 1 4 1 1 5 6 3 2 1 1 4 6 1 1 4 1 1 1 2 1 3 6 3 1 1 6 3 1 1 2 1 1 1 5
## [783] 1 3 1 6 1 1 1 1 4 2 6 6 6 1 1 1 1 1 2 6 3 2 3 1 6 6 1 3 1 3 1 2 6 1
## [817] 1 6 6 6 5 1 1 1 2 1 1 1 1 6 1 5 6 1 2 1 3 5 6 6 6 1 6 6 5 4 2 1 1 5
## [851] 1 6 5 1 3 1 6 2 6 6 4 3 1 4 1 1 1 2 1 6 1 6 6 6 3 3 1 1 5 2 3 6 1 1
## [885] 6 1 1 1 6 2 6 3 3 5 6 5 3 1 1 5 1 2 1 1 6 6 6 3 5 1 1 1 2 5 1 2 1 6
## [919] 1 6 6 1 1 1 1 3 2 1 6 3 5 1 3 1 1 1 2 1 3 3 6 5 6 6 1 2 1 6 6 1 6 5
## [953] 6 6 1 1 2 1 1 1 1 5 1 1 1 4 6 1 5 2 1 1 3 1 1 6 5 6 5 2 5 1 6 5 5 1
## [987] 1 5 6 1 5 1 6 5 1 4 1 5 5 1 1 2 1 1 1 1 3 5 5 3 1 2 1 1 4 6 1 1 1 6
## [1021] 1 4 1 2 1 5 2 5 6 1 1 1 6 5 6 6 3 6 3 5 6 6 5 1 1 5 3 1 1 2 5 1 1 1
## [1055] 5 3 1 1 1 1 2 1 1 6 1 1 5 5 6 1 1 5 2 6 6 1 1 3 1 5 1 4 6 6 2 3 3 1
## [1089] 1 1 1 1 1 1 6 2 6 1 1 1 1 5 1 1 4 1 3 1 2 6 5 6 6 6 5 5 1 1 1 3 2 5
## [1123] 5 5 5 5 4 5 5 1 1 1 2 5 5 1 1 1 4 6 5 1 2 6 1 2 6 4 4 3 6 1 1 1 6 6
## [1157] 2 6 5 6 1 5 5 4 1 4 1 5 1 2 1 6 1 5 5 6 3 4 5 1 3 2 1 5 1 3 1 5 1 1
## [1191] 1 1 1 5 2 6 3 1 6 1 5 4 5 6 1 5 2 5 5 1 5 3 1 1 4 1 6 1 2 1 5 3 5 1
## [1225] 5 1 1 5 2 3 1 6 1 3 5 5 1 1 6 4 2 5 1 5 1 5 6 1 1 1 5 2 1 1 1 5 3 1
## [1259] 3 1 3 3 1 2 6 3 2 5 1 6 5 3 5 1 1 6 5 1 2 6 6 5 5 5 6 1 1 4 2 1 5 2
## [1293] 1 5 5 1 1 6 5 1 1 5 5 2 1 5 4 1 5 1 5 1 6 5 6 3 6 1 5 5 1 6 5 6 1 3
## [1327] 2 5 6 1 3 4 5 6 3 3 1 6 2 1 3 3 1 6 1 1 1 1 1 1 6 2 1 6 6 1 6 5 5 5
## [1361] 1 6 6 2 1 1 1 6 3 1 1 5 6 6 1 2 5 3 3 1 6 5 1 3 3 1 1 2 1 2 1 3 1 5
## [1395] 1 3 1 6 3 1 2 1 1 5 5 3 1 3 1 1 5 2 5 3 3 1 3 6 5 1 5 2 6 6 5 5 3 5
## [1429] 1 1 1 2 1 4 6 1 6 3 1 5 3 6 1 1 2 1 1 1 5 4 1 6 1 1 6 6 6 2 1 1 5 1
## [1463] 1 1 6 6 4 6 5 4 2 1 1 4 4 1 1 4 1 1 1 5 2 1 1 6 5 5 3 5 3 1 1 1 2 3
## [1497] 5 6 5 1 5 4 3 1 1 2 5 1 2 1 1 6 6 6 4 1 3 3 4 1 2 1 1 4 7 4 5 6 3 1
## [1531] 1 3 2 3 3 5 6 5 1 1 3 3 5 1 6 2 6 5 3 1 6 1 1 1 6 1 2 1 1 1 1 1 1 7
## [1565] 3 2 1 7 1 5 1 3 1 1 3 6 7 1 1 1 3 4 6 6 2 3 3 3 3 3 1 1 2 3 1 3 6 1
## [1599] 1 5 6 3 3 2 5 6 4 7 1 6 3 1 1 2 2 1 1 6 4 1 3 6 7 4 1 6 2 3 6 3 3 3
## [1633] 6 3 6 6 2 5 3 7 5 3 3 7 1 2 7 7 6 6 6 7 7 6 7 3 2 6 6 3 7 3 6 6 5 3
## [1667] 7 6 2 1 6 4 5 5 6 3 1 5 3 2 6 3 3 6 1 1 6 6 6 6 6 6 2 7 1 1 1 7 1 6
## [1701] 4 6 2 1 3 3 6 3 3 3 1 1 1 6 3 3 2 7 1 6 6 1 1 6 3 1 1 6 2 1 2 7 1 6
## [1735] 1 5 5 1 3 2 6 6 5 1 1 5 6 3 5 6 1 2 7 6 7 6 5 1 1 6 6 5 2 1 7 1 1 3
## [1769] 6 3 7 2 5 3 1 1 3 5 7 6 2 7 3 7 7 6 6 1 6 7 6 1 7 3 6 3 7 3 7 5 6 6
## [1803] 2 4 3 6 7 7 7 5 5 3 2 3 1 1 3 7 6 7 5 1 6 7 2 1 7 7 7 7 6 3 7 1 6 7
## [1837] 2 1 7 2 6 6 6 3 7 1 7 1 6 2 6 6 6 6 6 1 1 1 3 6 7 1 2 6 7 1 6 7 6 6
## [1871] 6 5 2 6 1 1 7 7 7 7 7 1 7 1 2 5 5 6 6 6 6 1 6 7 2 4 6 7 1 6 6 7 3 6
## [1905] 7 1 2 1 1 6 5 1 6 7 6 7 7 2 1 6 7 3 5 5 5 3 7 6 2 1 7 6 7 1 7 1 1 7
## [1939] 1 1 3 7 2 6 6 7 6 3 7 6 1 1 7 7 3 6 6 2 7 1 1 7 1 1 1 1 6 7 2 1 4 7
## [1973] 7 7 6 6 1 1 6 6 2 7 3 6 1 7 7 1 7 7 1 1 3 2 1 3 6 7 7 7 3 6 1 3 7 2
## [2007] 1 7 5 6 3 7 7 7 2 3 1 7 1 6 7 7 1 7 1 2 7 7 6 6 1 7 1 6 7 7 2 3 3 1
## [2041] 1 7 7 7 6 1 2 5 5 5 5 7 6 7 7 1 7 7 2 1 3 1 1 7 1 1 1 3 5 2 1 3 2 7
## [2075] 1 5 3 1 1 3 6 5 7 1 2 7 7 6 6 3 6 6 1 1 5 1 2 6 1 7 1 1 1 6 1 1 2 7

```

```
## [2109] 3 5 5 1 5 1 1 6 5 5 2 7 5 3 7 7 7 7 3 1 1 1 2 1 3 1 5 6 1 1 1 2 6 1
## [2143] 3 7 7 5 7 1 3 2 1 1 1 1 3 5 6 7 5 2 1 1 6 3 1 1 7 1 7 4 1 3 6 2 1 1
## [2177] 6 6 1 6 4 3 1 5 1 2 1 2 1 7 1 1 7 3 5 1 7 7 2 1 7 6 7 1 3 7 4 1 3 6
## [2211] 2 7 1 7 6 6 1 1 4 1 2 1 3 7 1 6 5 1 1 7 6 1 6 2 5 1 5 7 1 1 7 7 5 3
## [2245] 1 2 7 3 7 1 6 5 5 4 6 1 2 5 7 3 7 7 6 1 3 7 2 7 3 7 1 3 5 7 7 1 2 1
## [2279] 3 6 1 1 3 5 6 1 6 3 2 7 6 5 3 7 5 3 7 6 5 5 6 2 3 6 2 1 1 1 1 1 6 6
## [2313] 1 1 2 1 1 5 6 1 3 7 1 3 6 3 2 1 3 1 7 5 6 1 1 6 3 1 2 7 5 3 1 6 1 7
## [2347] 3 6 6 2 5 1 7 3 3 3 6 1 5 3 6 2 7 3 3 6 1 7 1 6 5 1 1 1 2 1 3 5 5 6
## [2381] 7 1 1 1 7 2 1 1 1 7 7 7 1 1 3 1 6 2 1 5 3 7 1 1 3 1 1 1 6 2 4 5 1 1
## [2415] 7 1 1 3 1 1 6 2 6 7 2 5 1 1 6 6 3 3 1 6 7 2 1 1 6 6 1 1 6 1 1 3 1 5
## [2449] 1 2 7 1 6 5 7 3 3 1 1 6 1 2 1 1 6 6 1 1 5 5 5 3 7 1 2 6 1 1 7 7 7 3
## [2483] 7 1 3 2 5 5 3 6 1 6 1 5 1 5 6 2 5 1 6 5 6 5 6 6 1 5 2 1 3 5 3 6 5 1
## [2517] 6 1 1 2 6 1 5 3 5 6 6 6 6 1 1 2 7 1 1 1 1 6 3 7 7 6 3 6 2 6 2 1 7 7
## [2551] 1 5 6 6 1 7 5 2 5 6 3 1 7 5 1 5 3 2 3 1 5 1 7 5 1 6 1 5 2 1 6 1 1 1
## [2585] 1 7 1 3 1 2 6 6 1 1 1 6 5 1 3 7 7 2 7 7 6 6 3 7 1 1 3 3 2 1 5 6 3 1
## [2619] 6 5 6 1 1 1 2 1 3 7 6 6 7 7 7 5 7 2 3 5 3 3 5 1 7 1 1 2 1 4 6 3 1 5
## [2653] 1 7 5 3 7 6 2 6 1 2 6 5 1 1 3 7 6 3 6 1 2 6 5 7 1 6 7 1 6 6 6 2 1 6
## [2687] 6 1 5 6 3 5 6 2 6 1 6 5 1 6 6 6 3 1 1 2 1 6 1 3 3 6 6 1 1 5 6 1 2 1
## [2721] 6 5 3 4 1 1 6 1 2 1 4 1 5 6 1 5 3 4 1 1 2 6 1 1 1 4 4 6 5 4 1 6 2 6
## [2755] 4 1 6 1 1 1 1 6 2 6 6 1 1 4 1 1 1 6 2 1 6 2 3 6 1 6 6 1 1 1 6 1 2 6
## [2789] 6 1 6 1 1 1 1 4 4 5 6 2 1 3 6 6 6 6 4 1 6 2 6 1 1 2 6 6 1 6 5 2 1 6
## [2823] 6 6 5 6 1 1 2 1 3 6 1 1 1 1 1 6 5 1 6 2 1 1 1 1 1 1 1 5 3 3 6 1 5 1
## [2857] 4 1 6 3 1 2 5 5 5 1 1 4 4 4 4 5 2 1 6 6 6 1 6 1 1 1 2 5 2 4 1 1 1 5
## [2891] 5 5 3 5 2 1 6 6 1 1 1 1 2 6 4 1 1 5 5 1 4 1 3 2 1 1 5 3 1 1 4 1 1 6
## [2925] 6 2 1 6 1 5 6 6 1 1 6 1 2 6 6 1 1 5 1 3 1 3 6 2 1 3 1 1 6 6 1 6 6 1
## [2959] 6 1 2 1 1 3 3 5 6 5 1 1 4 2 1 3 1 1 1 1 1 1 5 2 4 1 5 1 3 1 6 3 1 1
## [2993] 5 2 3 2 1 1 1 6 3 1 1 5 6 2 3 6 6 1 1 1 1 5 1 1 2 6 6 3 6 5 5 1 1 5
## [3027] 6 1 2 3 5 5 1 3 6 6 3 6 1 1 2 4 1 1 1 1 6 5 1 3 6 2 6 3 1 5 1 6 6 6
## [3061] 1 1 1 1 2 1 6 1 5 1 1 1 1 2 1 1 5 5 1 1 1 1 3 6 2 6 3 1 1 1 1 6 4 1
## [3095] 4 2 1 1 6 5 1 1 6 5 5 6 1 2 6 1 2 1 4 4 1 1 6 1 6 3 1 2 6 1 4 6 1 1
## [3129] 3 5 1 5 2 5 5 5 3 6 6 3 5 3 1 2 1 1 6 1 1 3 5 1 4 4 2 4 6 1 5 5 1 6
## [3163] 1 6 6 1 2 6 3 6 1 5 6 1 1 1 2 1 1 3 1 1 1 6 6 1 1 1 3 3 1 6 6 1 6 4
## [3197] 1 1 6 1 3 3 6 1 1 1 6 1 5 1 2 1 1 1 1 5 4 1 6 3 6 3 3 2 1 4 2 3 6 6
## [3231] 1 6 1 6 4 1 1 5 2 1 4 1 1 3 1 1 1 5 1 2 5 1 1 1 1 1 1 5 1 5 2 1 6 6
## [3265] 5 6 6 5 2 6 1 1 1 6 1 1 1 6 2 6 6 5 5 1 1 5 1 2 1 1 6 6 6 1 1 6 2 1
## [3299] 1 1 3 5 1 1 6 2 6 6 6 5 1 6 1 1 6 2 6 1 6 1 5 1 3 6 1 1 2 5 1 2 4 1
## [3333] 1 5 6 1 5 1 4 1 1 3 1 4 1 1 1 4 6 1 6 6 2 6 1 5 5 6 1 5 4 1 6 5 2 3
## [3367] 1 1 1 1 5 4 5 5 1 1 2 6 1 5 1 6 1 1 6 6 1 3 6 6 1 1 1 5 6 3 5 5 2 1
## [3401] 6 3 6 5 1 5 1 1 6 5 3 6 1 1 1 1 5 1 1 1 1 2 4 1 1 1 6 1 1 5
```

Now, we don't really want to run `tapply` on every single variable when we have over 1,000 different variables. Let's instead use the `subset` function to subset our data by cluster. Create 7 new datasets, each containing the observations from one of the clusters.

```
cluster1 = subset(daily_kos, clusters_no == 1)
cluster2 = subset(daily_kos, clusters_no == 2)
cluster3 = subset(daily_kos, clusters_no == 3)
cluster4 = subset(daily_kos, clusters_no == 4)
cluster5 = subset(daily_kos, clusters_no == 5)
cluster6 = subset(daily_kos, clusters_no == 6)
cluster7 = subset(daily_kos, clusters_no == 7)
```

How many observations are in cluster 3?

374

Which cluster has the most observations?

Cluster 1

Which cluster has the fewest observations?

Cluster 4

1.5) Hierarchical Clustering

Instead of looking at the average value in each variable individually, we'll just look at the top 6 words in each cluster. To do this for cluster 1, type the following in your R console (where "HierCluster1" should be replaced with the name of your first cluster subset):

```
tail(sort(colMeans(cluster1)))
```

```
##      state republican      poll democrat      kerry      bush
## 0.7575039 0.7590837 0.9036335 0.9194313 1.0624013 1.7053712
```

This computes the mean frequency values of each of the words in cluster 1, and then outputs the 6 words that occur the most frequently. The colMeans function computes the column (word) means, the sort function orders the words in increasing order of the mean values, and the tail function outputs the last 6 words listed, which are the ones with the largest column means.

What is the most frequent word in this cluster, in terms of average value? Enter the word exactly how you see it in the output:

Explanation

After running the R command given above, we can see that the most frequent word on average is "bush". This corresponds to President George W. Bush.

1.6) Hierarchical Clustering

Now repeat the command given in the previous problem for each of the other clusters, and answer the following questions.

Which words best describe cluster 2?

```
tail(sort(colMeans(cluster2)))
```

```
##      bush democrat challenge      vote      poll november
## 2.847352 2.850467 4.096573 4.398754 4.847352 10.339564
```

Which cluster could best be described as the cluster related to the Iraq war?

In 2004, one of the candidates for the Democratic nomination for the President of the United States was Howard Dean, John Kerry was the candidate who won the democratic nomination, and John Edwards with the running mate of John Kerry (the Vice President nominee). Given this information, which cluster best corresponds to the democratic party?

```
tail(sort(colMeans(cluster3)))
```

```
##      elect      parties      state republican democrat      bush
## 1.647059 1.665775 2.320856 2.524064 3.823529 4.406417
```

```
tail(sort(colMeans(cluster4)))
```

```
## campaign voter presided      poll      bush      kerry
## 1.431655 1.539568 1.625899 3.589928 7.834532 8.438849
```

```
tail(sort(colMeans(cluster5)))
```

```
##      american      presided administration      war      iraq
##      1.090909      1.120393      1.230958      1.776413      2.427518
##      bush
##      3.941032
```

```
tail(sort(colMeans(cluster6)))
```

```
##      race      bush      kerry      elect democrat      poll
## 0.4579832 0.4887955 0.5168067 0.5350140 0.5644258 0.5812325
```

```
tail(sort(colMeans(cluster7)))
```

```
## democrat      clark      edward      poll      kerry      dean
## 2.148325 2.497608 2.607656 2.765550 3.952153 5.803828
```

Explanation

You can repeat the command on each of the clusters by typing the following:

```
tail(sort(colMeans(HierCluster2)))
```

```
tail(sort(colMeans(HierCluster3)))
```

```
tail(sort(colMeans(HierCluster4)))
```

```
tail(sort(colMeans(HierCluster5)))
```

```
tail(sort(colMeans(HierCluster6)))
```

```
tail(sort(colMeans(HierCluster7)))
```

You can see that the words that best describe Cluster 2 are november, poll, vote, and challenge. The most common words in Cluster 5 are bush, iraq, war, and administration, so it is the cluster that can best be described as corresponding to the Iraq war. And the most common words in Cluster 7 are dean, kerry, poll, and edward, so it looks like the democratic cluster.

2.2) K-Means Clustering

Now, run k-means clustering, setting the seed to 1000 right before you run the kmeans function. Again, pick the number of clusters equal to 7. You don't need to add the `iters.max` argument.

```
set.seed(1000)
```

```
k_cluster = kmeans(daily_kos, centers = 7)
```

Subset your data into the 7 clusters (7 new datasets) by using the "cluster" variable of your kmeans output.

```
cluster1_k = subset(daily_kos, k_cluster$cluster == 1)
cluster2_k = subset(daily_kos, k_cluster$cluster == 2)
cluster3_k = subset(daily_kos, k_cluster$cluster == 3)
cluster4_k = subset(daily_kos, k_cluster$cluster == 4)
cluster5_k = subset(daily_kos, k_cluster$cluster == 5)
cluster6_k = subset(daily_kos, k_cluster$cluster == 6)
cluster7_k = subset(daily_kos, k_cluster$cluster == 7)
```

How many observations are in Cluster 3?

```
nrow(cluster3_k)
```

```
## [1] 277
```

Which cluster has the most observations?

Which cluster has the fewest number of observations?

```
table(k_cluster$cluster)
```

```
##
##      1      2      3      4      5      6      7
## 146  144  277 2063  163  329  308
```

Explanation

You can run k-means clustering by using the following commands:

```
set.seed(1000)
```

```
KmeansCluster = kmeans(dailykos, centers=7)
```

Then, you can subset your data into the 7 clusters by using the following commands:

```
KmeansCluster1 = subset(dailykos, KmeansCluster$cluster == 1)
```

```
KmeansCluster2 = subset(dailykos, KmeansCluster$cluster == 2)
```

```
KmeansCluster3 = subset(dailykos, KmeansCluster$cluster == 3)
```

```
KmeansCluster4 = subset(dailykos, KmeansCluster$cluster == 4)
```

```
KmeansCluster5 = subset(dailykos, KmeansCluster$cluster == 5)
```

```
KmeansCluster6 = subset(dailykos, KmeansCluster$cluster == 6)
```

```
KmeansCluster7 = subset(dailykos, KmeansCluster$cluster == 7)
```

Alternatively, you could answer these questions by looking at the output of `table(KmeansCluster$cluster)`.

More Advanced Approach:

There is a very useful function in R called the “split” function. Given a vector assigning groups like `KmeansCluster$cluster`, you could split `dailykos` into the clusters by typing:

```
KmeansCluster = split(dailykos, KmeansCluster$cluster)
```

Then cluster 1 can be accessed by typing `KmeansCluster[[1]]`, cluster 2 can be accessed by typing `KmeansCluster[[2]]`, etc. If you have a variable in your current R session called “split”, you will need to remove it with `rm(split)` before using the split function.

2.2) K-Means Clustering

Now, output the six most frequent words in each cluster, like we did in the previous problem, for each of the k-means clusters.

Which k-means cluster best corresponds to the Iraq War?

Which k-means cluster best corresponds to the democratic party? (Remember that we are looking for the names of the key democratic party leaders.)

```
tail(sort(colMeans(cluster1_k)))
```

```
##      state      iraq      kerry administration      presided
## 1.609589  1.616438  1.636986      2.664384      2.767123
##      bush
## 11.431507
```

```
tail(sort(colMeans(cluster2_k)))
```

```
## primaries democrat edward clark kerry dean
## 2.319444 2.694444 2.798611 3.090278 4.979167 8.277778
tail(sort(colMeans(cluster3_k)))

## administration iraqi american bush war
## 1.389892 1.610108 1.685921 2.610108 3.025271
## iraq
## 4.093863
tail(sort(colMeans(cluster4_k)))

## elect republican kerry poll democrat bush
## 0.6010664 0.6175473 0.6495395 0.7474552 0.7891420 1.1473582
tail(sort(colMeans(cluster5_k)))

## race senate state parties republican democrat
## 2.484663 2.650307 3.521472 3.619632 4.638037 6.993865
tail(sort(colMeans(cluster6_k)))

## democrat bush challenge vote poll november
## 2.899696 2.960486 4.121581 4.446809 4.872340 10.370821
tail(sort(colMeans(cluster7_k)))

## presided voter campaign poll bush kerry
## 1.324675 1.334416 1.383117 2.788961 5.970779 6.480519
```

Explanation

You can output the most frequent words in each of the k-means clusters by using the following commands:

```
tail(sort(colMeans(KmeansCluster1)))
tail(sort(colMeans(KmeansCluster2)))
tail(sort(colMeans(KmeansCluster3)))
tail(sort(colMeans(KmeansCluster4)))
tail(sort(colMeans(KmeansCluster5)))
tail(sort(colMeans(KmeansCluster6)))
tail(sort(colMeans(KmeansCluster7)))
```

By looking at the output, you can see that the cluster best corresponding to the Iraq War is cluster 3 (top words are iraq, war, and bush) and the cluster best corresponding to the democratic party is cluster 2 (top words dean, kerry, clark, and edward).

2.3 - K-Means Clustering

For the rest of this problem, we'll ask you to compare how observations were assigned to clusters in the two different methods. Use the table function to compare the cluster assignment of hierarchical clustering to the cluster assignment of k-means clustering.

Which Hierarchical Cluster best corresponds to K-Means Cluster 2?

```
table(clusters_no)

## clusters_no
## 1 2 3 4 5 6 7
```



```
## 1266 321 374 139 407 714 209
tail(sort(colMeans(cluster1)))

##      state republican      poll democrat      kerry      bush
## 0.7575039 0.7590837 0.9036335 0.9194313 1.0624013 1.7053712
tail(sort(colMeans(cluster2)))

##      bush democrat challenge      vote      poll november
## 2.847352 2.850467 4.096573 4.398754 4.847352 10.339564
tail(sort(colMeans(cluster3)))

##      elect      parties      state republican      democrat      bush
## 1.647059 1.665775 2.320856 2.524064 3.823529 4.406417
tail(sort(colMeans(cluster4)))

## campaign voter presided      poll      bush      kerry
## 1.431655 1.539568 1.625899 3.589928 7.834532 8.438849
tail(sort(colMeans(cluster5)))

##      american      presided administration      war      iraq
## 1.090909 1.120393 1.230958 1.776413 2.427518
##      bush
## 3.941032
tail(sort(colMeans(cluster6)))

##      race      bush      kerry      elect democrat      poll
## 0.4579832 0.4887955 0.5168067 0.5350140 0.5644258 0.5812325
tail(sort(colMeans(cluster7)))

## democrat clark edward      poll      kerry      dean
## 2.148325 2.497608 2.607656 2.765550 3.952153 5.803828
table(k_cluster$cluster)

##
## 1 2 3 4 5 6 7
## 146 144 277 2063 163 329 308
tail(sort(colMeans(cluster1_k)))

##      state      iraq      kerry administration      presided
## 1.609589 1.616438 1.636986 2.664384 2.767123
##      bush
## 11.431507
tail(sort(colMeans(cluster2_k)))

## primaries democrat edward clark kerry dean
## 2.319444 2.694444 2.798611 3.090278 4.979167 8.277778
tail(sort(colMeans(cluster3_k)))

## administration      iraqi      american      bush      war
## 1.389892 1.610108 1.685921 2.610108 3.025271
##      iraq
## 4.093863
```

```
tail(sort(colMeans(cluster4_k)))
```

```
##      elect republican      kerry      poll      democrat      bush
## 0.6010664 0.6175473 0.6495395 0.7474552 0.7891420 1.1473582
```

```
tail(sort(colMeans(cluster5_k)))
```

```
##      race      senate      state      parties republican      democrat
## 2.484663 2.650307 3.521472 3.619632 4.638037 6.993865
```

```
tail(sort(colMeans(cluster6_k)))
```

```
## democrat      bush challenge      vote      poll      november
## 2.899696 2.960486 4.121581 4.446809 4.872340 10.370821
```

```
tail(sort(colMeans(cluster7_k)))
```

```
## presided      voter campaign      poll      bush      kerry
## 1.324675 1.334416 1.383117 2.788961 5.970779 6.480519
```

Explanation

From “table(hierGroups, KmeansCluster\$cluster)”, we read that 116 (80.6%) of the observations in K-Means Cluster 2 also fall in Hierarchical Cluster 7.

```
table(clusters_no, k_cluster$cluster)
```

```
##
## clusters_no      1      2      3      4      5      6      7
##           1      3     11     64    1045     32      0    111
##           2      0      0      0      0      0    320      1
##           3     85     10     42     79    126      8     24
##           4     10      5      0      0      1      0    123
##           5     48      0    171    145      3      1     39
##           6      0      2      0    712      0      0      0
##           7      0    116      0     82      1      0     10
```

2.4) K-Means Clustering

Which Hierarchical Cluster best corresponds to K-Means Cluster 3?

Explanation

From “table(hierGroups, KmeansCluster\$cluster)”, we read that 171 (61.7%) of the observations in K-Means Cluster 3 also fall in Hierarchical Cluster 5.

2.5) K-Means Clustering

Which Hierarchical Cluster best corresponds to K-Means Cluster 7?

Explanation

From “table(hierGroups, KmeansCluster\$cluster)”, we read that no more than 123 (39.9%) of the observations in K-Means Cluster 7 fall in any hierarchical cluster.

2.6) K-Means Clustering

Which Hierarchical Cluster best corresponds to K-Means Cluster 6?

Explanation

From “`table(hierGroups, KmeansCluster$cluster)`”, we read that 320 (97.3%) of observations in K-Means Cluster 6 fall in Hierarchical Cluster 2.