

& negative words

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```
library(tm)
```

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

```
library(SnowballC)
```

```
library(wordcloud)
```

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

Read the positive and negative words

```
pos<-read.delim("D:/Surabhi docs/portfolio/positive.txt")
```

```
neg<-read.delim("D:/Surabhi docs/portfolio/negative.txt")
```

Clean the data

```
pos <- pos[-1:-33,]  
head(pos,33)
```

```
## [1] abound          abunds           abundance       abundant
## [5] accessible       accessible      acclaim         acclaimed
## [9] acclamation      accolade        accolades       accommodative
## [13] accomodative     accomplish      accomplished    accomplishment
## [17] accomplishments accurate         accurately      achievable
## [21] achievement      achievements    achievable      acumen
## [25] adaptable        adaptive        adequate        adjustable
## [29] admirable        admirably       admiration      admire
## [33] admirer
## 2031 Levels: ; ... zippy
```

```
str(pos)
```

```
## Factor w/ 2031 levels ";",";", ";",";" (editors: N. Indurkha and F. J. Damerau), 2010.",...: 27
```

```
names(pos) <- c("Positive")
```

```
head(pos)
```

```
## Positive      <NA>      <NA>      <NA>      <NA>      <NA>
## abundant      abunds    abundance    abundant    accessible    accessible
## 2031 Levels: ; ... zippy
```

```
neg <- neg[-1:-33,]
head(neg,33)
```

```
## [1] 2-faces      abnormal      abolish      abominable    abominably
## [6] abominate     abomination  abort        aborted       aborts
## [11] abrade        abrasive      abrupt       abruptly      abscond
## [16] absence       absent-minded absentee     absurd        absurdity
## [21] absurdly      absurdness   abuse        abused        abuses
## [26] abusive       abysmal      abysmally    abyss         accidental
## [31] accost        accursed     accusation
## 4808 Levels: ; ... zombie
```

```
str(neg)
```

```
## Factor w/ 4808 levels ";","; ";","; (editors: N. Indurkha and F. J. Damerau), 2010.",...: 27
```

```
names(neg) <- c("Negative")
```

```
head(neg)
```

```
## Negative      <NA>      <NA>      <NA>      <NA>      <NA>
## 2-faces      abnormal    abolish   abominable abominably abominate
## 4808 Levels: ; ... zombie
```

Process in the MLK Speech and perform text mining operations

```
mlk <- read.delim("D:/Surabhi docs/portfolio/MLK-Speech.txt")
```

```
str(mlk)
```

```
## 'data.frame': 28 obs. of 1 variable:
## $ I.am.happy.to.join.with.you.today.in.what.will.go.down.in.history.as.the.greatest.demonstration.f
```

```
tail(mlk)
```

```
##
## 23
## 24
## 25
## 26
## 27
## 28 And when this happens, when we allow freedom to ring, when we let it ring from every village and c
```

```
#Create a term matrix
```

```
words.vec <- VectorSource(mlk)
words.corpus <- Corpus(words.vec)
words.corpus
```

```
## <<VCorpus>>
## Metadata: corpus specific: 0, document level (indexed): 0
## Content: documents: 1
```

```
tdm <- TermDocumentMatrix(words.corpus)
```

```
m <- as.matrix(tdm)
wordCounts <- rowSums(m)
head(wordCounts)
```

```
words.corpus <- tm_map(words.corpus, content_transformer(tolower))
words.corpus <- tm_map(words.corpus, removePunctuation)
words.corpus <- tm_map(words.corpus, removeNumbers)
words.corpus <- tm_map(words.corpus, removeWords, stopwords("english"))
```

```
wordcloud(words.corpus, max.words = 100, random.order = FALSE, colors=c("orange", "blue", "pink", "green",
```



```
## <<TermDocumentMatrix (terms: 561, documents: 1)>>
## Non-/sparse entries: 561/0
## Sparsity           : 0%
## Maximal term length: 15
## Weighting          : term frequency (tf)
##
##                      Docs
```

## Terms	1
## 'tis	1
## able	8
## again	2
## ago,	1
## ahead.	1
## alabama,	3
## all	7
## alleghenies	1
## allow	2
## almighty,	1
## alone.	1
## also	1
## always	1
## america	5
## american	3
## american,	1
## and	41
## architects	1
## are	8
## areas	1
## asking	1
## autumn	1
## awakening	1
## back	8
## back.	1
## bad	1
## bank	1
## bankrupt.	1
## basic	1
## battered	1
## beacon	1
## beautiful	1
## become	1
## been	2
## beginning.	1
## believe	2
## believes	1
## bitterness	1
## black	4
## blow	1
## bodies,	1
## bound	1
## boys	2
## bright	1
## brotherhood.	3
## brothers,	1
## brothers.	1
## brutality.	2
## business	1
## but	6
## california!	1
## came	2
## can	4

##	cannot	6
##	capital	1
##	captivity.	1
##	carolina,	1
##	cash	2
##	catholics,	1
##	cells.	1
##	chains	1
##	changed.	1
##	character.	1
##	check	3
##	check,	1
##	check.	1
##	children	3
##	children,	1
##	children.	1
##	cities,	1
##	cities.	1
##	citizens	1
##	citizenship	1
##	city,	1
##	civil	1
##	color	2
##	colorado!	1
##	come	10
##	community	1
##	concerned.	1
##	condition.	1
##	conduct	1
##	constitution	1
##	content	2
##	continue	2
##	cooling	1
##	corners	1
##	country,	1
##	created	1
##	creative	2
##	creed:	1
##	crippled	1
##	crooked	1
##	cup	1
##	curvaceous	1
##	dark	1
##	day	9
##	day,	1
##	day.	1
##	daybreak	1
##	declaration	1
##	decree	1
##	deeds.	1
##	deeply	1
##	defaulted	1
##	degenerate	1
##	demand	1

##	democracy.	1
##	desolate	1
##	despair	1
##	despair.	1
##	destiny	1
##	destiny.	1
##	devotees	1
##	died,	1
##	difficulties	1
##	dignity	2
##	discipline.	1
##	discontent	1
##	discords	1
##	discrimination.	1
##	distrust	1
##	down	3
##	dramatize	1
##	dream	9
##	dream.	2
##	drinking	1
##	dripping	1
##	drug	1
##	emancipation	1
##	emerges.	1
##	end	1
##	end,	1
##	engage	1
##	engulfed	1
##	equal.	1
##	equality.	1
##	even	2
##	every	10
##	evidenced	1
##	exalted,	1
##	exile	1
##	face	1
##	faith	5
##	fall	1
##	fatal	1
##	fathers	1
##	fatigue	1
##	fierce	1
##	finds	1
##	five	1
##	flames	1
##	flesh	1
##	for	8
##	force	1
##	force.	1
##	forever	1
##	former	2
##	foundations	1
##	four	1
##	free	4

##	free.	1
##	freedom	18
##	freedom.	1
##	fresh	1
##	friends,	1
##	from	18
##	funds	1
##	funds.	1
##	gain	1
##	gaining	1
##	gentiles,	1
##	georgia	1
##	georgia!	1
##	georgia,	1
##	ghetto	1
##	ghettos	1
##	girls	2
##	give	1
##	given	1
##	glory	1
##	god	1
##	god's	3
##	governor	1
##	gradualism.	1
##	granted	1
##	great	5
##	guaranteed	1
##	guilty	1
##	had	1
##	hallowed	1
##	hamlet,	1
##	hampshire.	1
##	hands	2
##	happens,	1
##	happiness.	1
##	has	5
##	hatred.	1
##	have	20
##	having	1
##	heat	2
##	heavy	1
##	heightening	1
##	heights	1
##	heir.	1
##	her	1
##	here	3
##	hew	1
##	high	1
##	highways	1
##	hill	2
##	hills	1
##	hilltops	1
##	himself	1
##	his	3

##	hold	1
##	honoring	1
##	hope	2
##	hope.	2
##	horrors	1
##	hotels	1
##	hundred	4
##	independence,	1
##	inextricably	1
##	injustice	1
##	injustice,	1
##	injustice.	1
##	insofar	1
##	instead	1
##	insufficient	2
##	interposition	1
##	into	4
##	invigorating	1
##	island	1
##	its	3
##	jail	2
##	jangling	1
##	jews	1
##	join	2
##	joyous	1
##	judged	1
##	justice	4
##	justice.	4
##	knowing	2
##	land	3
##	land.	1
##	languishing	1
##	larger	1
##	last!	3
##	later,	4
##	lead	1
##	leads	1
##	left	1
##	legitimate	1
##	let	13
##	liberty,	2
##	life	1
##	life,	1
##	lift	1
##	light	1
##	like	2
##	lips	1
##	little	3
##	live	2
##	lives	1
##	lodging	1
##	lonely	1
##	long	6
##	lookout	1

##	lord	1
##	louisiana,	1
##	low,	1
##	luxury	1
##	made	3
##	magnificent	1
##	majestic	1
##	make	3
##	manacles	1
##	many	1
##	march	1
##	marked	1
##	marvelous	1
##	material	1
##	meaning	1
##	meaning,	1
##	meeting	1
##	men	3
##	men,	3
##	midst	1
##	mighty	2
##	militancy	1
##	millions	1
##	mississippi	1
##	mississippi,	2
##	mississippi.	1
##	mobility	1
##	molehill	1
##	moment.	1
##	momentous	1
##	motels	1
##	mountain	4
##	mountains	1
##	mountainside,	2
##	must	8
##	narrow	1
##	nation	8
##	nation's	1
##	nation.	1
##	needed	1
##	negro	13
##	negro's	2
##	neither	1
##	never	3
##	new	5
##	night	1
##	nineteen	1
##	no,	2
##	nor	1
##	northern	1
##	not	13
##	note	3
##	nothing	1
##	now	5

##	now.	1
##	nullification;	1
##	oasis	1
##	obligation,	1
##	obvious	1
##	ocean	1
##	off	2
##	old	1
##	one	12
##	one.	1
##	only	1
##	only.	1
##	opportunity	1
##	oppression,	1
##	our	16
##	out	3
##	overlook	1
##	own	1
##	owners	1
##	palace	1
##	pass	1
##	path	1
##	pennsylvania!	1
##	people	2
##	people,	1
##	persecution	1
##	physical	2
##	pilgrim's	1
##	place	1
##	places	2
##	plain,	1
##	plane	1
##	pledge	1
##	police	2
##	poverty	1
##	pray	1
##	presence	1
##	pride,	1
##	process	1
##	proclamation.	1
##	prodigious	1
##	promise	1
##	promises	1
##	promissory	2
##	prosperity.	1
##	protest	1
##	protestants	1
##	pursuit	1
##	quest	1
##	quick	1
##	racial	2
##	racists,	1
##	real	1
##	reality	1

##	realize	2
##	red	1
##	redemptive.	1
##	refuse	2
##	remind	1
##	republic	1
##	rest	1
##	returns	1
##	revealed,	1
##	revolt	1
##	riches	1
##	right	1
##	righteousness	1
##	rightful	1
##	rights	1
##	rights,	1
##	rights.	1
##	ring	9
##	ring,	1
##	ring.	2
##	rise	3
##	robbed	1
##	rock	1
##	rockies	1
##	rolls	1
##	rooted	1
##	rough	1
##	rude	1
##	sacred	1
##	sadly	1
##	sands	1
##	satisfied	5
##	satisfied,	2
##	satisfied?	1
##	satisfy	1
##	say	2
##	score	1
##	seared	1
##	security	1
##	see	1
##	seek	1
##	segregation	2
##	self-evident:	1
##	selfhood	1
##	sense	1
##	shadow	1
##	shake	1
##	shall	5
##	shameful	1
##	signed	1
##	signing	1
##	signs	1
##	sing	2
##	sing.	1

##	sisters	1
##	sit	1
##	situation	1
##	sixty-three	1
##	skin	1
##	slave	1
##	slaves	2
##	slopes	1
##	slums	1
##	smaller	1
##	snowcapped	1
##	society	1
##	solid	1
##	some	3
##	somehow	1
##	something	1
##	sons	2
##	soul	1
##	south	2
##	speed	1
##	spiritual,	1
##	spot	1
##	staggered	1
##	stand	3
##	state	3
##	stating	1
##	steam	1
##	still	4
##	stone	2
##	storms	1
##	straight,	1
##	stream.	1
##	stripped	1
##	struggle	2
##	suffering	1
##	suffering.	1
##	summer	1
##	sunlit	1
##	sweet	1
##	sweltering	3
##	symbolic	1
##	symphony	1
##	table	1
##	take	1
##	tennessee!	1
##	thank	1
##	that	23
##	that;	1
##	the	100
##	thee	1
##	thee,	1
##	their	8
##	there	6
##	these	1

##	they	3
##	thirst	1
##	this	19
##	those	2
##	though	1
##	threshold	1
##	tied	1
##	time	5
##	today	3
##	today,	3
##	today.	2
##	together	1
##	together,	5
##	together.	1
##	tomorrow,	1
##	tranquility	1
##	tranquilizing	1
##	transform	1
##	transformed	1
##	travel,	1
##	trials	1
##	tribulations.	1
##	true	1
##	true.	1
##	truths	1
##	turn	1
##	unalienable	1
##	unearned	1
##	unmindful	1
##	unspeakable	1
##	until	4
##	upon	1
##	urgency	2
##	usual.	1
##	valley	3
##	vast	1
##	vaults	1
##	veterans	1
##	vicious	1
##	victim	1
##	village	1
##	violence.	1
##	vote	1
##	vote.	1
##	walk	1
##	walk,	1
##	wallow	1
##	warm	1
##	was	2
##	waters	1
##	well	1
##	were	1
##	when	7
##	where	3

```
##   which           5
##   whirlwinds     1
##   white           6
##   whites          1
##   who             4
##   whose           1
##   will            25
##   winds           1
##   with            14
##   with.           1
##   withering       1
##   words            3
##   work             2
##   would            2
##   wrongful         1
##   wrote            1
##   years            5
##   yes,             1
##   york             1
##   york.            1
##   you              7
##   your             1
```

```
m = as.matrix(tdm)
wordCounts = rowSums(m)

wordCounts <- sort(wordCounts, decreasing=TRUE)
head(wordCounts)
```

```
## the and will that have this
## 100 41 25 23 20 19
```

```
str(wordCounts)
```

```
## Named num [1:561] 100 41 25 23 20 19 18 18 16 14 ...
## - attr(*, "names")= chr [1:561] "the" "and" "will" "that" ...
```

```
total <- sum(wordCounts)
words <- names(wordCounts)
str(words)
```

```
## chr [1:561] "the" "and" "will" "that" "have" "this" ...
```

```
#Step 3: Determine how many positive words were in the speech
```

```
countPos <- match(words,pos,nomatch=0)
countPos
```

```
## [1] 0 0 0 0 0 0 0 765 0 0 0 0 0 0 0
## [15] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [29] 0 0 0 0 0 0 0 0 672 856 0 0 0 1565 0
## [43] 0 0 0 0 0 0 0 763 0 0 0 0 0 0 0
## [57] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [71] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [85] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [99] 0 0 378 438 0 0 0 0 0 0 0 0 0 0 0
```

```
## [113] 0 0 0 0 1087 0 1178 0 0 0 0 0 0 0
## [127] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [141] 0 0 0 0 0 0 1986 0 0 0 0 0 0 0
## [155] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [169] 0 173 0 0 0 0 0 0 0 234 0 0 0 0
## [183] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [197] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [211] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [225] 0 0 0 0 0 0 430 0 0 0 0 0 0 0
## [239] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [253] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [267] 0 0 0 0 0 0 767 0 0 0 782 786 0 0
## [281] 0 0 0 0 0 0 821 0 0 0 0 0 0 0
## [295] 873 0 0 0 0 0 0 0 0 0 0 0 0 0
## [309] 0 0 0 0 0 922 0 0 0 0 0 0 0 0
## [323] 0 0 1026 0 0 0 1045 0 0 0 0 1072 1074 0
## [337] 0 0 0 0 0 0 0 0 0 0 0 0 0 1127
## [351] 1134 1136 0 0 0 0 1144 0 0 0 0 0 0 0
## [365] 0 0 0 0 0 1188 0 0 0 0 0 0 0 0
## [379] 0 0 0 0 0 0 1216 0 0 0 0 0 0 0
## [393] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [407] 0 0 0 0 0 0 0 0 1355 1371 1373 0 0 0
## [421] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [435] 0 1532 1536 1537 0 0 0 0 0 0 0 0 0 0
## [449] 0 0 0 0 0 1567 0 0 0 0 0 0 0 0
## [463] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [477] 0 0 0 0 1642 0 0 0 0 0 0 0 0 0
## [491] 0 0 0 0 0 0 0 0 1768 0 0 0 0 0
## [505] 1791 0 0 0 0 0 0 0 0 0 0 0 1829 0
## [519] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [533] 0 0 0 0 0 0 0 0 0 0 0 0 0 1928
## [547] 0 1935 0 0 0 0 0 0 0 0 0 0 0 0
## [561] 0
```

```
matchCounts <- wordCounts[which(countPos != 0)]
length(matchCounts) #42 positive words
```

```
## [1] 42
```

```
#Step 4: Determine how many negative words were in the speech
```

```
countNeg <- match(words,neg,nomatch=0)
countNeg
```

```
## [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [15] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [29] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [43] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [57] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [71] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [85] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [99] 0 0 0 0 0 0 0 0 0 0 0 0 2476
## [113] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [127] 0 0 0 0 3483 0 0 0 0 0 3857 0 0
## [141] 0 4081 0 0 0 0 0 0 0 0 0 0 0
```

```
## [155] 0 0 0 0 0 0 0 0 0 0 245 0 0 0 273
## [169] 0 0 0 0 0 339 376 0 0 0 0 0 0 0
## [183] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [197] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [211] 0 759 773 0 0 825 0 0 0 0 0 0 0 0
## [225] 903 0 0 989 992 0 0 0 0 0 1070 0 1146 0
## [239] 0 1322 0 0 1394 0 0 0 0 0 0 0 0 0
## [253] 0 0 1530 0 1574 1612 0 1621 1664 0 0 0 0 0
## [267] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [281] 0 0 1841 0 0 0 0 0 0 0 0 0 1921 0
## [295] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [309] 0 0 0 0 0 0 0 0 0 0 2380 2431 0 0 0
## [323] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [337] 0 0 0 0 0 0 0 0 0 2738 0 0 0 0
## [351] 0 0 0 0 0 0 0 0 0 0 0 0 0 2857
## [365] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [379] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [393] 0 0 3166 0 0 0 0 0 0 0 3263 0 0 0
## [407] 0 0 3341 0 0 0 0 0 0 0 0 0 3393 0
## [421] 0 0 0 0 0 0 0 0 0 0 0 0 0 3579
## [435] 0 0 0 0 0 0 0 0 0 0 0 0 0 3615
## [449] 3618 0 3648 0 0 0 0 0 0 0 0 0 0 0
## [463] 0 3768 3774 0 0 0 0 0 0 0 0 0 3856 0
## [477] 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [491] 0 0 0 0 4145 0 0 0 0 0 0 0 0 0
## [505] 0 0 0 0 0 4239 0 0 0 0 0 0 0 0
## [519] 0 0 0 0 0 0 0 0 0 0 0 0 4541 0
## [533] 0 0 0 0 4631 0 0 0 0 0 0 0 4671 0
## [547] 0 0 0 0 0 0 0 0 0 4772 0 0 0 0
## [561] 0
```

```
matchCounts <- wordCounts[which(countNeg != 0)]
length(matchCounts) #46 negative words
```

```
## [1] 46
```

```
#Step 5: Redo the 'positive' and 'negative' calculations for each 25% of the speech
```

```
#dividing speech into 4 equal parts
```

```
newdataframes = split(words, sample(rep(1:2:3:4, 459)))
```

```
## Warning in 1:2:3: numerical expression has 2 elements: only the first used
```

```
## Warning in 1:2:3:4: numerical expression has 3 elements: only the first
## used
```

```
## Warning in split.default(words, sample(rep(1:2:3:4, 459))): data length is
## not a multiple of split variable
```

```
#now matching to get the positive words
```

```
#first 25 percent
```

```
matchedpos1 = match(newdataframes$`1`, pos, nomatch = 0)
```

```
#count of match words
```

```
matchcountspos1 = wordCounts[which(matchedpos1 != 0)]
```

```
pos1 = length(matchcountspos1)
```



```

#second 25 percent
matchedpos2 = match(newdataframes$`2`,pos,nomatch = 0)

#count of match words
matchcountspos2 = wordCounts[which(matchedpos2 != 0)]
pos2 = length(matchcountspos2)

#third 25 percent
matchedpos3 = match(newdataframes$`3`,pos,nomatch = 0)

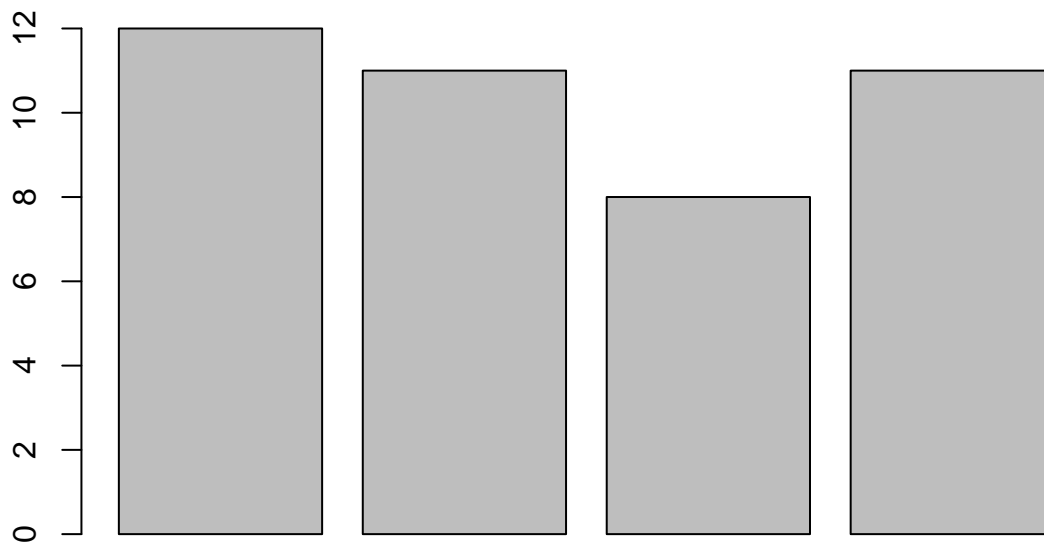
#count of match words
matchcountspos3 = wordCounts[which(matchedpos3 != 0)]
pos3 = length(matchcountspos3)

#fourth 25 percent
matchedpos4 = match(newdataframes$`4`,pos,nomatch = 0)

#count of match words
matchcountspos4 = wordCounts[which(matchedpos4 != 0)]
pos4 = length(matchcountspos4)

#barchart for comparing the result
h = c(pos1,pos2,pos3,pos4)
barplot(h)

```



```

#similarly for the negative words

#first 25 percent
matchedneg1 = match(newdataframes$`1`,neg,nomatch = 0)

#count of match words
matchcountsneg1 = wordCounts[which(matchedneg1 != 0)]
neg1 = length(matchcountsneg1)

#second 25 percent
matchedneg2 = match(newdataframes$`2`,neg,nomatch = 0)

#count of match words
matchcountsneg2 = wordCounts[which(matchedneg2 != 0)]
neg2 = length(matchcountsneg2)

#third 25 percent
matchedneg3 = match(newdataframes$`3`,neg,nomatch = 0)

#count of match words
matchcountsneg3 = wordCounts[which(matchedneg3 != 0)]
neg3 = length(matchcountsneg3)

#fourth 25 percent
matchedneg4 = match(newdataframes$`4`,neg,nomatch = 0)

#count of match words
matchcountsneg4 = wordCounts[which(matchedneg4 != 0)]
neg4 = length(matchcountsneg4)

#barchart for seeing the result
J = c(neg1,neg2,neg3,neg4)
barplot(J)

```

