

Travail individuel 4

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Importation des librairies

```
# Installer les packages  
#install.packages("tm") # for text mining  
#install.packages("SnowballC") # for text stemming  
#install.packages("wordcloud") # word-cloud generator  
#install.packages("RColorBrewer") # color palettes  
#install.packages("syuzhet") # for sentiment analysis  
#install.packages("ggplot2") # for plotting graphs  
# Charger les librairies  
library("tm")
```

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

```
library("SnowballC")  
library("wordcloud")
```

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

```
library("RColorBrewer")  
library("syuzhet")  
library("ggplot2")
```

```
##
```

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

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

```
##
```

```
##      annotate
```

Lecture de texte

```
text <- readLines(file.choose())
```

Chargement des données sous forme de corpus

```
TextDoc <- Corpus(VectorSource(text))
```

Netoyage

```
toSpace <- content_transformer(function (x , pattern ) gsub(pattern, " ", x))
TextDoc <- tm_map(TextDoc, toSpace, "/")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "/"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "200")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "200"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "\\|")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "\\|"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "\231"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "€")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "\200"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "â")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "â"): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, ""): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, ""): transformation drops
## documents
```

```
TextDoc <- tm_map(TextDoc, toSpace, "€")
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, toSpace, "\200"): transformation drops  
## documents
```

Transformation en miniscule, élimination des chiffres, et autres

```
TextDoc <- tm_map(TextDoc, content_transformer(tolower))
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, content_transformer(tolower)):  
## transformation drops documents
```

```
TextDoc <- tm_map(TextDoc, removeNumbers)
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, removeNumbers): transformation drops  
## documents
```

```
TextDoc <- tm_map(TextDoc, removeWords, stopwords("english"))
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, removeWords, stopwords("english")):  
## transformation drops documents
```

```
TextDoc <- tm_map(TextDoc, removeWords, c("s", "company", "team"))
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, removeWords, c("s", "company", "team")):  
## transformation drops documents
```

```
TextDoc <- tm_map(TextDoc, removePunctuation)
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, removePunctuation): transformation drops  
## documents
```

```
TextDoc <- tm_map(TextDoc, stripWhitespace)
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, stripWhitespace): transformation drops  
## documents
```

```
TextDoc <- tm_map(TextDoc, stemDocument)
```

```
## Warning in tm_map.SimpleCorpus(TextDoc, stemDocument): transformation drops  
## documents
```

```
class(TextDoc)
```

```
## [1] "SimpleCorpus" "Corpus"
```

```
inspect(TextDoc[[7]])
```

```
## <<PlainTextDocument>>
## Metadata: 7
## Content: chars: 99
##
## thank pioneer leadership lowest violent crime rate quarter centuri cleanest environ quarter centuri
```

La Matrice “Termes par Document”

```
TextDoc_dtm <- TermDocumentMatrix(TextDoc)
inspect(TextDoc_dtm)
```

```
## <<TermDocumentMatrix (terms: 1263, documents: 156)>>
## Non-/sparse entries: 3552/193476
## Sparsity           : 98%
## Maximal term length: 16
## Weighting           : term frequency (tf)
## Sample              :
##
##      Docs
## Terms  132 137 145 22 32 67 74 76 82 88
## america 1  0  1  0  0  0  0  1  0  0
## american 0  1  0  0  2  0  1  2  0  0
## centuri  0  1  1  0  0  0  0  0  2  1
## must     0  0  0  1  1  0  0  1  1  2
## new      0  0  0  0  2  0  1  2  0  0
## now      1  0  1  1  1  2  0  0  1  1
## secur    0  0  0  3  0  0  0  0  1  0
## will     0  0  0  0  2  0  1  0  2  0
## work     0  0  0  0  0  1  1  0  2  1
## year     0  1  1  4  2  1  3  0  0  0
```

```
dtm_m <- as.matrix(TextDoc_dtm)
```

La Matrice “Termes par Document” (suite)

```
dtm_v <- sort(rowSums(dtm_m),decreasing=TRUE)
dtm_d <- data.frame(word = names(dtm_v),freq=dtm_v)
```

Afficher les mot les plus fréquents , ici on a choisit 15

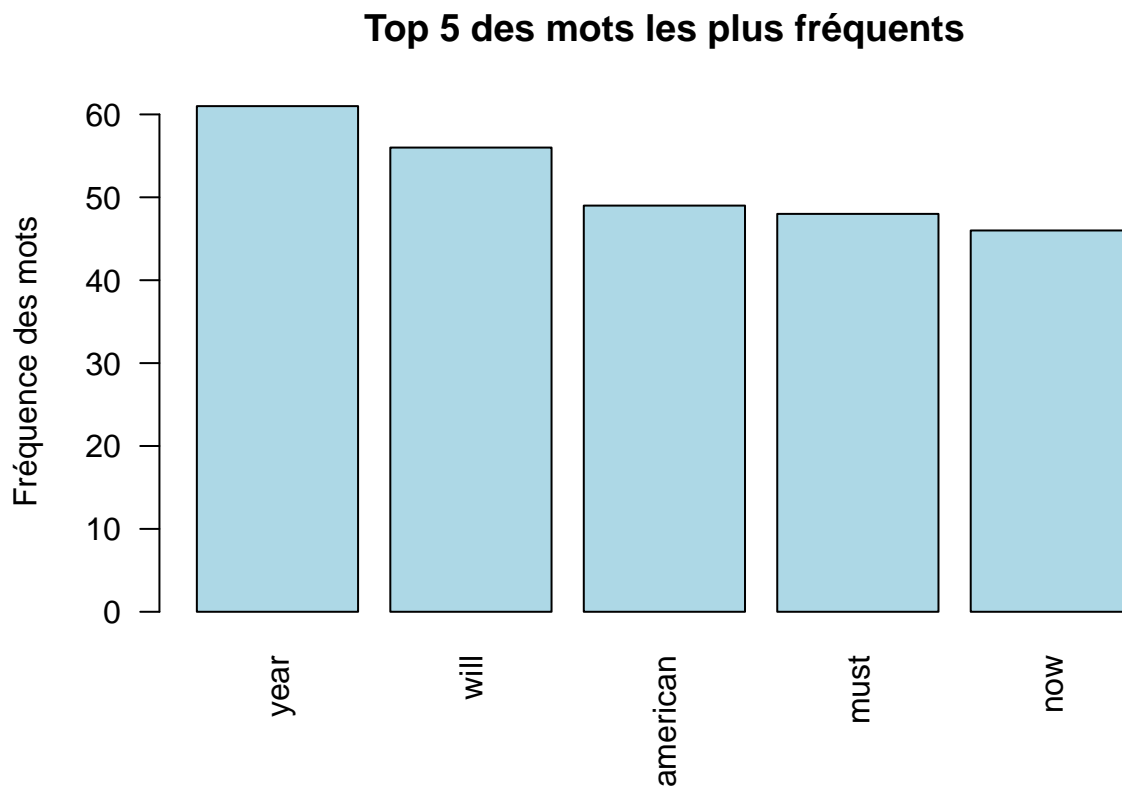
```
head(dtm_d, 15)
```

```
##      word freq
## year   year  61
## will   will  56
## american american 49
## must   must  48
```

```
## now          now  46
## work         work  42
## america     america 37
## centuri     centuri 36
## secur       secur  33
## new         new   32
## school      school 30
## nation      nation 28
## support     support 28
## help        help   27
## congress    congress 26
```

Tracer ces mots par un barplot()

```
barplot(dtm_d[1:5,]$freq,
        las = 2,
        names.arg = dtm_d[1:5,]$word,
        col = "lightblue",
        main = "Top 5 des mots les plus fréquents",
        ylab = "Fréquence des mots"
)
```

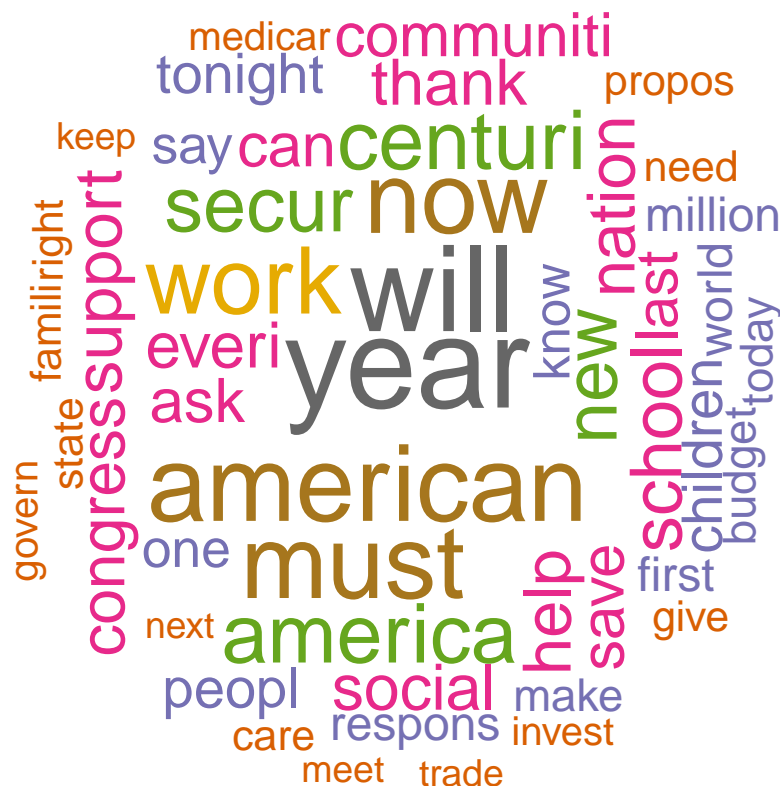


Génération d'un nuage de mots (word cloud)

```

set.seed(1234)
wordcloud(words = dtm_d$word,
  freq = dtm_d$freq,
  min.freq = 5,
  max.words=50,
  random.order=FALSE,
  rot.per=0.40,
  colors=brewer.pal(8, "Dark2")
)

```



Génération des associations

```

# Trouver des associations
findAssocs(TextDoc_dtm, terms = c("war", "peace", "peopl"), corlimit = 0.25)

```

```

## $war
##      cold      answer    depress dispossess    largest    overcom    prejudic
##      0.75      0.71      0.71      0.71      0.71      0.71      0.71
##      struggl    twilight      win      racial      class      barrier      long
##      0.71      0.71      0.71      0.60      0.49      0.49      0.46
##      middl      lift      generat    percent    point      arsenal    framework
##      0.39      0.39      0.38      0.36      0.35      0.35      0.35
##      height      iii      attack      bomber    captain      desert      execut
##      0.35      0.35      0.35      0.35      0.35      0.35      0.35
##      flawless    flew      fox      jeff      machin      oper      superb
##      0.35      0.35      0.35      0.35      0.35      0.35      0.35

```

```
## taliaferro   advisori   alabama   board     bus       goe       journey
##      0.35      0.35      0.35      0.35      0.35      0.35      0.35
##      other      rosa      sens      sought throughout      sinc      start
##      0.35      0.35      0.35      0.35      0.35      0.31      0.30
##      end      great
##      0.29      0.29
##
## $peace
## numeric(0)
##
## $peopl
##      news      podium      pride      ago      welfar      hire
##      0.52      0.52      0.52      0.49      0.48      0.40
##      tonight      roll      choos      lose      digniti      move
##      0.35      0.35      0.35      0.35      0.35      0.35
## partnership      republ      real      access      beyond      coverag
##      0.35      0.35      0.35      0.34      0.34      0.34
## jefford      kennedi      moynihan      offer      roth      bought
##      0.34      0.34      0.34      0.34      0.34      0.34
## expens      advisori      alabama      board      bus      goe
##      0.34      0.34      0.34      0.34      0.34      0.34
## journey      other      rosa      sens      sought throughout
##      0.34      0.34      0.34      0.34      0.34      0.34
## china      good      thousand      anoth      five      bring
##      0.32      0.31      0.31      0.31      0.31      0.30
## health      insur      longer      get      hundr      realli
##      0.29      0.29      0.27      0.27      0.27      0.27
##      past      liberti
##      0.27      0.27
```

```
# Trouver des associations pour des mots qui se produisent au moins 50 fois
findAssocs(TextDoc_dtm,
            terms = findFreqTerms(TextDoc_dtm, lowfreq = 50),
            corlimit = 0.25)
```

```
## $year
##      six      last      fulfil      reserv      wise      next      knew      sound
##      0.48      0.43      0.39      0.39      0.39      0.37      0.37      0.35
##      anoth      surplus      improv      spend      five      grant      enact      joint
##      0.35      0.34      0.33      0.33      0.30      0.29      0.28      0.28
## patient fifthgrad fiveyear      hurt      less      literaci      mount      rapid
##      0.28      0.28      0.28      0.28      0.28      0.28      0.28      0.28
## skill      team      train      corpor      opic      untap      felon      fugit
##      0.28      0.28      0.28      0.28      0.28      0.28      0.28      0.28
## murder      schedul      stalker      straight      now      pass      bill      oversea
##      0.28      0.28      0.28      0.28      0.26      0.26      0.26      0.26
##
## $will
##      asid      divis      heal      love      anoth      hope      time      look      reach      hundr
##      0.44      0.42      0.42      0.42      0.39      0.39      0.37      0.37      0.33      0.33
## exhaust      full      older      suffici      unabl      educ      said      cover      payment      decis
##      0.31      0.31      0.31      0.31      0.31      0.28      0.28      0.27      0.27      0.27
## listen      five      hour      shape      found      ideal
##      0.27      0.27      0.27      0.27      0.27      0.27
```

Score des sentiments

```
syuzhet_vector <- get_sentiment(text,method="syuzhet")
head(syuzhet_vector)
```

```
## [1] 0.9 1.0 3.1 1.0 -1.0 0.0
```

```
summary(syuzhet_vector)
```

```
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
## -4.5000  0.3875  1.6250  1.5506  2.7125  7.4500
```

```
#par la methode bing
bing_vector <- get_sentiment(text, method="bing")
head(bing_vector)
```

```
## [1] 0 1 1 2 0 1
```

```
summary(bing_vector)
```

```
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
## -4.000   0.000   1.000   1.026   2.000   7.000
```

```
#par la methode affinn
afinn_vector <- get_sentiment(text, method="afinn")
head(afinn_vector)
```

```
## [1] 2 2 4 -1 -3 -1
```

```
summary(afinn_vector)
```

```
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
## -16.000   0.000   2.000   2.744   6.000  18.000
```

Extraction des émotions

```
d<-get_nrc_sentiment(text)
head (d,10)
```

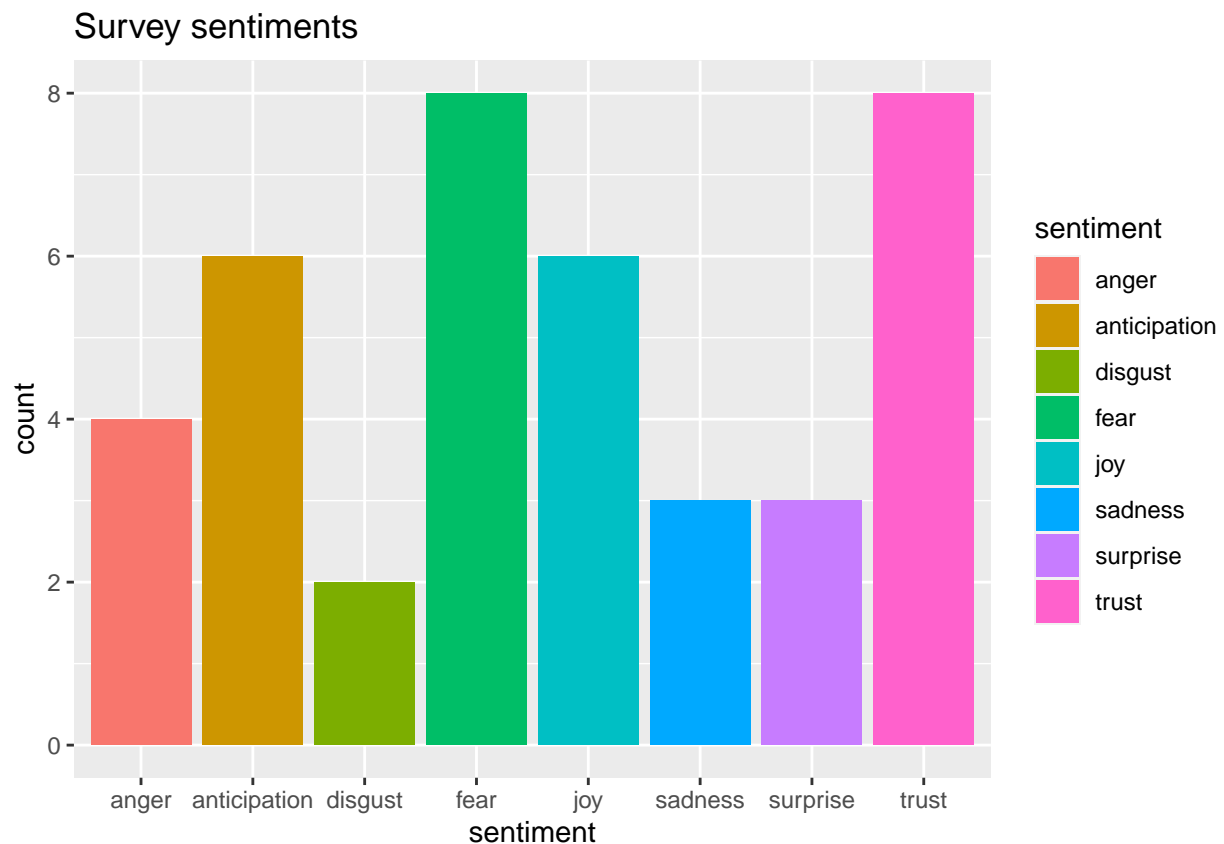
```
##      anger anticipation disgust fear joy sadness surprise trust negative positive
## 1         0             0       1   0   0         0         0     3         1         2
## 2         0             0       0   0   0         0         0     1         0         1
## 3         0             0       0   2   1         0         0     1         0         3
## 4         0             0       0   0   0         0         0     0         0         2
## 5         0             2       0   2   3         1         1     0         3         3
## 6         0             1       0   0   0         0         0     1         1         1
## 7         2             0       1   1   0         1         1     0         3         0
## 8         1             1       0   1   1         0         0     1         1         1
## 9         1             2       1   2   1         1         1     3         3         8
## 10        0             0       0   0   0         0         0     1         0         1
```

Classification des émotions (suite)


```
td<-data.frame(t(d))
td_new <- data.frame(rowSums(td[2:10]))
names(td_new)[1] <- "count"
td_new <- cbind("sentiment" = rownames(td_new), td_new)
rownames(td_new) <- NULL
td_new2<-td_new[1:8,]
```

Classification des émotions - nombre de mots associés à chaque sentiment

```
quickplot(sentiment,
  data=td_new2,
  weight=count,
  geom="bar",
  fill=sentiment,
  ylab="count")+
ggtitle("Survey sentiments")
```



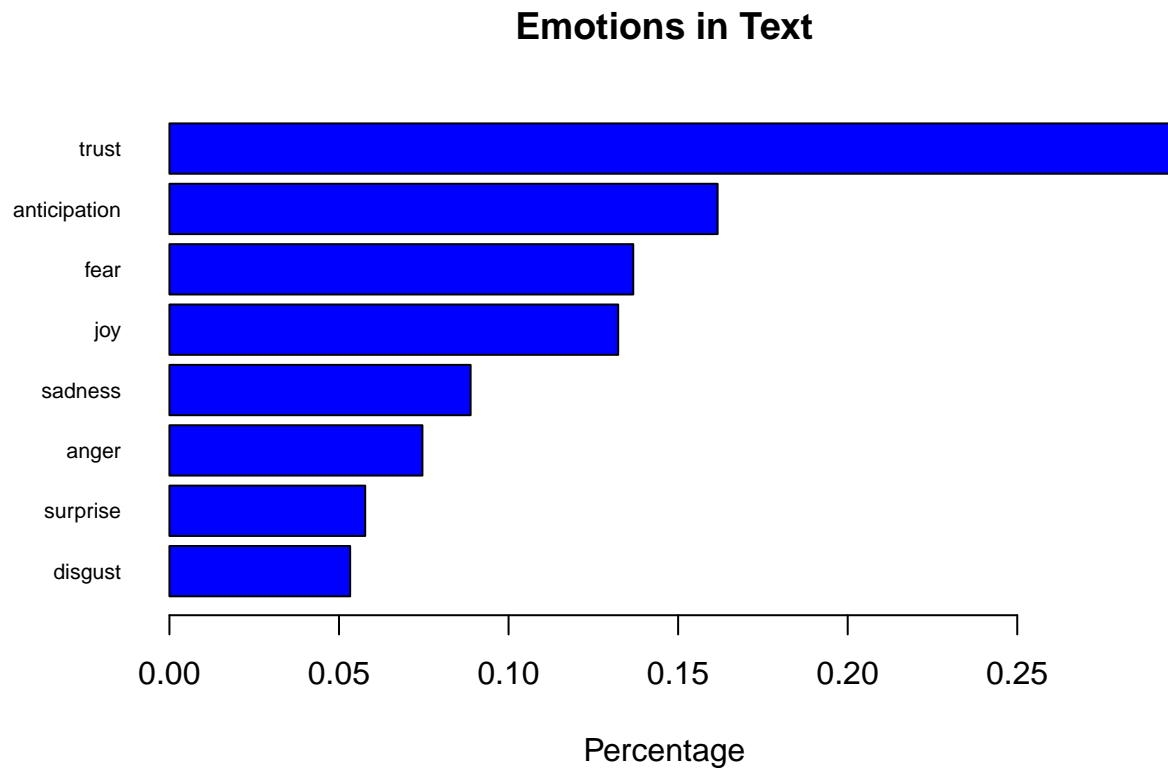
Classification des émotions (suite)

```
barplot(
  sort(
    colSums(prop.table(d[, 1:8]))
  ),
  horiz = TRUE,
  cex.names = 0.7,
```

```

las = 1,
main = "Emotions in Text",
xlab="Percentage",
col="blue"
)

```



Commentaires

Cette analyse de text du président Bill Clinton, lors de son discours annuel tant que president des etats unis, nous montre plein de confiance avec esprit d'anticipation et un sentiment de peur. Il a cité,les mots 'americans', 'must' , 'year' , 'will' , qui peut signifier son intention de faire quelques chose cette année pour les americans.