## **Term work 3**

```
install.packages("tm")
install.packages("wordcloud")
install.packages("e1071")
library(tm)
library(wordcloud)
library(e1071)
#creating a data frame using CSV files
#With base R functions, to avoid conversion of strings to factors you would do, for example:
sms_spam_df <- read.csv(file="E:\\sms_spam.csv",stringsAsFactors=F)</pre>
View(sms spam df)
str(sms spam df)
#head(sms_spam_df)
#table(sms_spam_df$category)
#Data preparation - cleaning and standardizing text data
#creating a corpus
sms corpus <- VCorpus(VectorSource(sms spam df$text))
#sms_corpus
print(sms corpus)
#iew a summary of the first and second SMS messages in the corpus:
inspect(sms_corpus[1:2])
#sms corpus[1:2]
#data pre-processing
#translate all letters to lower case
clean corpus <- tm map(sms corpus, content transformer(tolower))</pre>
#clean_corpus <- tm_map(sms_corpus, tolower)
#as.character(clean_corpus[[1]])
# remove numbers
clean_corpus <- tm_map(clean_corpus, removeNumbers)</pre>
# remove punctuation
clean_corpus <- tm_map(clean_corpus, removePunctuation)</pre>
# remove stopwords
```

```
stopwords()[1:15]
clean corpus <- tm map(clean corpus, removeWords, stopwords())</pre>
#as.character(clean_corpus[[1]])
# remove whitespaces
clean_corpus <- tm_map(clean_corpus, stripWhitespace)</pre>
#inspect, i.e., display detailed information on a corpus, a term-document matrix, or a text
document.
inspect(clean_corpus[1:3])
#tokenize each message into words to build the key structure for the analysis, a sparse matrix
comprising:
sms_dtm <- DocumentTermMatrix(clean_corpus)</pre>
str(sms dtm)
#The which() function will return the position of the elements
#(i.e., row number/column number/array index) in a logical vector which are TRUE.
spam indices <- which(sms spam df$type == "spam")</pre>
ham indices <- which(sms spam df$type == "ham")
#spam indices
#ham indices
wordcloud(clean corpus[ham indices], min.freq=40) # look at the 40 most common words
wordcloud(clean_corpus[spam_indices], min.freq=40)
#Split out training and test sets
# split the raw data
sms raw_train <- sms_spam_df[1:4169,]</pre>
sms raw test <- sms spam df[4170:5559,]
# then split the document-term matrix
sms_dtm_train <- sms_dtm[1:4169,]</pre>
sms_dtm_test <- sms_dtm[4170:5559,]
# and finally the corpus
sms_corpus_train <- clean_corpus[1:4169]</pre>
sms corpus test <- clean corpus[4170:5559]
```

```
#create separate corpuses for spam and ham
spam <- subset(sms raw train, type == "spam")</pre>
ham <- subset(sms_raw_train, type == "ham")
#reducing the DTM
five_times_words <- findFreqTerms(sms_dtm_train, 5)
sms train
              <-
                    DocumentTermMatrix(sms_corpus_train,
                                                                  control=list(dictionary
five_times_words))
sms test
             <-
                    DocumentTermMatrix(sms corpus test,
                                                                  control=list(dictionary
five_times_words))
#Remember that NB works on factors, but our DTM only has numerics.
#Let's define a function which converts counts to yes/no factor, and apply it to our reduced
matrices.
convert count <- function(x) {</pre>
y < -ifelse(x > 0, 1,0)
y <- factor(y, levels=c(0,1), labels=c("No", "Yes"))
У
}
sms_train <- apply(sms_train, 2, convert_count)</pre>
sms_test <- apply(sms_test, 2, convert_count)</pre>
sms_classifier <- naiveBayes(sms_train, factor(sms_raw_train$type))</pre>
sms test pred <- predict(sms classifier, newdata=sms test)</pre>
k=table(sms_test_pred, sms_raw_test$type)
accuracy = sum(diag(k))/sum(k)*100
accuracy
```