NLP with Disaster Tweets

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Natural Language Processing with Disaster Tweets

• Predict which Tweets are about real disasters and which ones are not

```
library(dplyr)
library(tidyverse)
library(tidytext)
library(tm)
library(SnowballC)
library(readr)
train_data <- read.csv("train.csv")</pre>
test_data <- read.csv("test.csv")</pre>
complete_data <- bind_rows(train_data, test_data)</pre>
glimpse(complete_data)
## Rows: 10,876
## Columns: 5
## $ id
           <int> 1, 4, 5, 6, 7, 8, 10, 13, 14, 15, 16, 17, 18, 19, 20, 23, 24,~
<chr> "Our Deeds are the Reason of this #earthquake May ALLAH Forgi~
## $ text
head(complete_data)
    id keyword location
## 1 1
## 2 4
## 3 5
## 4 6
## 5 7
## 6 8
##
## 1
                                                       Our Deeds are the Reason of this #
```

4

3 All residents asked to 'shelter in place' are being notified by officers. No other evacuation or si

13,000 people receive #wildfir

```
## 5
                                                    Just got sent this photo from Ruby #Alaska as smoke f
## 6
                             #RockyFire Update => California Hwy. 20 closed in both directions due to La
## target
## 1
## 2
## 3
## 4
## 5
         1
## 6
#----- test Corpus
#Train dataset text Processing
#creating corpus
Corpus <- VCorpus(VectorSource(complete_data$text))</pre>
#Transformation to lowercase
Corpus <- tm_map(Corpus, content_transformer(tolower))</pre>
#removing numbers
Corpus <- tm_map(Corpus,removeNumbers)</pre>
#removing punctuation
Corpus <- tm_map(Corpus, removePunctuation)</pre>
#removing stopwords
Corpus <- tm_map(Corpus, removeWords, stopwords())</pre>
#stemming
Corpus <- tm_map(Corpus, stemDocument)</pre>
#remove whitespaces
Corpus <- tm_map(Corpus, stripWhitespace)</pre>
as.character(Corpus[[1]])
## [1] "deed reason earthquak may allah forgiv us"
#creating dtm (bag-of-words) model
Corpus_dtm <- DocumentTermMatrix(Corpus)</pre>
Corpus_dtm
## <<DocumentTermMatrix (documents: 10876, terms: 23920)>>
## Non-/sparse entries: 101048/260052872
                      : 100%
## Sparsity
## Maximal term length: 51
## Weighting
                   : term frequency (tf)
Corpus_dtm <- removeSparseTerms(Corpus_dtm, 0.99)</pre>
Corpus_dtm
```

```
## <<DocumentTermMatrix (documents: 10876, terms: 106)>>
## Non-/sparse entries: 19142/1133714
## Sparsity
                    : 98%
## Maximal term length: 10
## Weighting
                : term frequency (tf)
#converting as matrix
Corpus_dtm <- as.data.frame(as.matrix(Corpus_dtm))</pre>
Corpus_dtm$id <- as.factor(complete_data$id )</pre>
Corpus_dtm$target <- as.factor(complete_data$target)</pre>
# Fixing incomplete cases
incomplete.cases <- which(!complete.cases(Corpus_dtm))</pre>
Corpus_dtm[incomplete.cases,] <- rep(0.0, ncol(Corpus_dtm))</pre>
#----- Classification algorithm implementation -----
library(caret)
set.seed(3456)
trainIndex <- createDataPartition(Corpus_dtm$target , p = .70,</pre>
                                 list = FALSE,
                                 times = 1)
train <- Corpus_dtm[ trainIndex,]</pre>
test <- Corpus_dtm[-trainIndex,]</pre>
#-----Model Building-----
set.seed(123)
library(e1071)
# nb.model <- naiveBayes</pre>
nb.model \leftarrow naiveBayes(x = train[,-108],
                       y = train$target )
summary(nb.model)
##
            Length Class Mode
## apriori 2 table numeric
## tables 107 -none- list
## levels
            2 -none- character
## isnumeric 107 -none- logical
## call 3 -none- call
# Predicting the Test set results
nb_pred <- predict(nb.model, newdata = test[,-108])</pre>
# Making the Confusion Matrix
cm <- table(test$target , nb_pred)</pre>
##
     nb_pred
##
        0
```

```
##
   0 1948 333
##
   1 417 564
error_metric <- function(cm){</pre>
  TN = cm[1,1]
  TP = cm[2,2]
  FP = cm[1,2]
  FN = cm[2,1]
  accuracy=(TP+TN)/(TP+FP+TN+FN)
  precision =(TP)/(FP+TP)
  recall=(TP)/(FN+TP)
  F_score = 2* ((precision*recall)/(precision+recall))
  print(paste("Naive Bayes Accuracy", round(accuracy, 2)))
  print(paste("Naive Bayes Precision", round(precision, 2)))
  print(paste("Naive Bayes Recall", round(recall, 2)))
  print(paste("Naive Bayes F_score",round(F_score,2)))
rf_results <- error_metric(cm)</pre>
## [1] "Naive Bayes Accuracy 0.77"
## [1] "Naive Bayes Precision 0.63"
## [1] "Naive Bayes Recall 0.57"
## [1] "Naive Bayes F_score 0.6"
rf_results
## [1] "Naive Bayes F_score 0.6"
# Fitting Kernel SVM to the Training set
# install.packages('e1071')
library(e1071)
svm.model <- svm(formula = target ~ .,</pre>
                 data = train,
                 type = 'C-classification',
                 kernel = 'radial')
svm.model
##
## Call:
## svm(formula = target ~ ., data = train, type = "C-classification",
       kernel = "radial")
##
##
##
## Parameters:
      SVM-Type: C-classification
## SVM-Kernel: radial
##
        cost: 1
##
## Number of Support Vectors: 4500
```

```
##
## Call:
## svm(formula = target ~ ., data = train, type = "C-classification",
       kernel = "radial")
##
##
##
## Parameters:
      SVM-Type: C-classification
##
## SVM-Kernel: radial
##
          cost: 1
##
## Number of Support Vectors: 4500
##
## ( 2204 2296 )
##
## Number of Classes: 2
##
## Levels:
## 0 1
# Predicting the Test set results
svm_pred <- predict(svm.model, newdata = test[,-108])</pre>
# Making the Confusion Matrix
cm2 <- table(test$target , svm_pred)</pre>
cm2
##
      svm_pred
##
         0
              1
##
     0 2259
              22
##
     1 755 226
error_metric <- function(cm2){</pre>
  TN = cm2[1,1]
  TP = cm2[2,2]
  FP = cm2[1,2]
  FN = cm2[2,1]
  accuracy=(TP+TN)/(TP+FP+TN+FN)
  precision =(TP)/(FP+TP)
  recall=(TP)/(FN+TP)
  F_score = 2* ((precision*recall)/(precision+recall))
  print(paste("SVM Accuracy", round(accuracy, 2)))
  print(paste("SVM Precision", round(precision, 2)))
  print(paste("SVM Recall",round(recall,2)))
  print(paste("SVM F_score",round(F_score,2)))
}
SVM_results <- error_metric(cm2)</pre>
```

summary(svm.model)