SentimentAnalysis

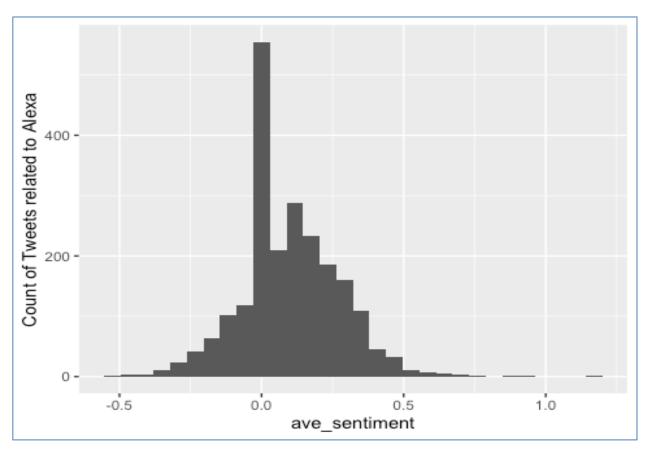
Project – 2

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2020-04-09

```
# Import all the libraries
library(rtweet)
library(sentimentr)
library(ndjson)
## Attaching package: 'ndjson'
## The following object is masked from 'package:rtweet':
##
##
     flatten
library(ggplot2)
library(tidyverse)
## — Attaching packages
tidyverse 1.3.0 —
## √ tibble 3.0.0 √ dplyr
                           0.8.5
## √ tidyr 1.0.2
                 √ stringr 1.4.0
                  √ forcats 0.5.0
## √ readr 1.3.1
## √ purrr 0.3.3
## Warning: package 'tibble' was built under R version 3.6.2
## — Conflicts
tidyverse_conflicts() —
## x dplyr::filter() masks stats::filter()
## x purrr::flatten() masks ndjson::flatten(), rtweet::flatten()
## x dplyr::lag() masks stats::lag()
library(tidytext)
```

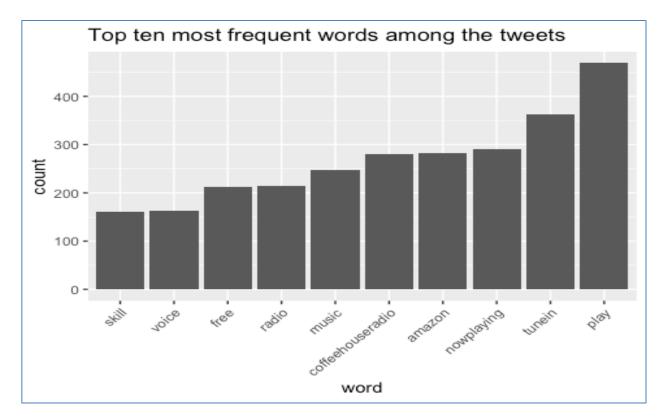
```
Sentiment Analysis using package 'sentimentr'
# Get ALEXA related tweets using search_stream()"
alexa_tweets <- search_tweets(q="#alexa", n = 10000, include_rts = FALSE,</pre>
lang="en")
# Sentiment Extraction
alexa_sentiment <- sentiment_by(alexa_tweets$text)</pre>
## Warning: Each time `sentiment_by` is run it has to do sentence boundary
disambiguation when a
## raw `character` vector is passed to `text.var`. This may be costly of time
and
## memory. It is highly recommended that the user first runs the raw
`character`
## vector through the `get sentences` function.
head(alexa sentiment)
##
    element_id word_count
                          sd ave_sentiment
## 1:
           1
                  51 0.1740237 -0.03625007
                  18 0.1767767 -0.14949425
## 2:
           2
## 3:
          3
                 16 0.0000000 0.00000000
                          NA 0.00000000
## 4:
           4
                  15
## 5:
           5
                          NA 0.00000000
                  14
           6
                          NA
                             0.21380899
## 6:
                  14
summary(alexa_sentiment$ave_sentiment)
     Min.
         1st Qu.
                 Median
                         Mean 3rd Qu.
                                      Max.
## -0.54569 0.00000 0.08175 0.09621 0.20750 1.14945
```



Observation: We can see most of the sentiments are either neutral or towards positive and few of them are on the negative side.

- $-1 \rightarrow$ negative
- $+1 \rightarrow$ positive

```
# Alexa tweets analysis
# Most frequent words
alexa_tokens <- alexa_tweets %>%
 select(text) %>%
 unnest tokens(word, text)
alexa tokens %>%
 group_by(word) %>%
 summarise(count = n()) %>%
 anti join(stop words) %>%
 filter(!word %in% c('https', 't.co', 'alexa99', 'alexa', 'amp', 'gt')) %>%
 arrange(desc(count)) %>%
 top_n(10) %>%
 mutate(word = reorder(word, count)) %>%
 ggplot(aes(x = word, y = count)) +
 geom col() +
 theme(axis.text.x = element text(angle=45, hjust=1)) +
 labs(title = "Top ten most frequent words among the tweets")
```



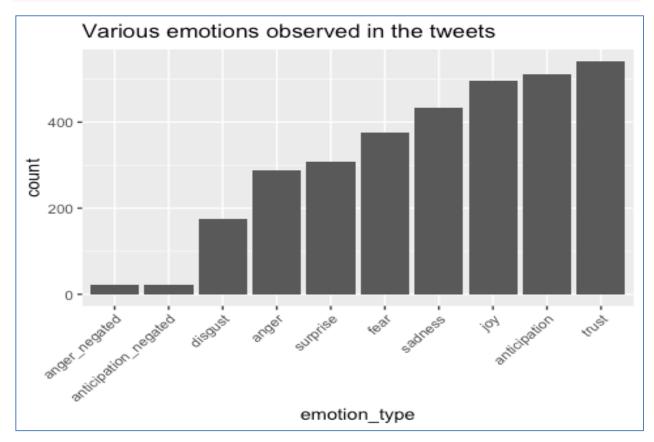
Top 10 words that are used in the tweets related to Alexa is shown above.

Understand Alexa tweets emotions

alexa_emotions <- emotion_by(get_sentences(alexa_tweets\$text))</pre>

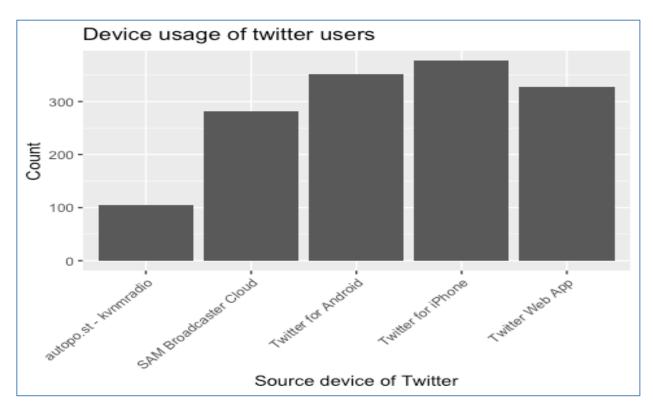
```
alexa_emotions <- alexa_emotions %>%
  group_by(element_id) %>%
  filter(emotion_count == 1)

alexa_emotions %>%
  group_by(emotion_type) %>%
  summarise(count = n()) %>%
  top_n(10) %>%
  mutate(emotion_type = reorder(emotion_type, count)) %>%
  ggplot(aes(x=emotion_type, y = count)) +
  geom_col() +
  theme(axis.text.x = element_text(angle=45, hjust=1)) +
  labs(title = "Various emotions observed in the tweets")
```



We have used **emotion_by()** function that is available with '**sentimentr**' package and assigns an emotion to each tweet.

A plot showing top 10 emotions to which the tweets belong. We can see a mixture of positive and negative emotions that the tweets show.



Twitter users who posted about Alexa vary almost equally among Iphone and Android.

plot time series of tweets

```
alexa_tweets %>%
  ts_plot("3 hours") +
  labs(title = "Frequency of #alexa Twitter statuses from past few days")
```

