

# Sentimental Analysis on Reviews of Amazon Product

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This is the CONTINUATION of the **Amazon Customer Review Analysis** done with Python.

```
#Importing relevant Libraries
library(dplyr)
library(rvest)
library(stringr)
library(tm)
library(wordcloud2)
library(ggplot2)
library(tidytext)
```

```
#Read csv -- and print shape of the data.
df <- read.csv('Cleaned_review_output.csv')
print(dim(df))
```

```
## [1] 14337      3
```

```
#Tokenizing the customer review column and setting the "ngram" to 2
tokens <- df %>%
  unnest_tokens(words, c(Cleaned_Review), token = 'ngrams', n = 2)
```

```
#Selecting only the relevant columns needed
tokens <- tokens %>%
  select(c('Product', 'ReviewStar', 'words'))

print(head(tokens, 4))
```

```
##           Product ReviewStar      words
## 1 boAt Rockerz 255      Neutral doubt great
## 2 boAt Rockerz 255      Neutral  great bass
## 3 boAt Rockerz 255      Neutral  bass great
## 4 boAt Rockerz 255      Neutral great extent
```

After tokenizing the data... the shape of the data has been changed - the data now consist of **170306 rows** compared to **14337** earlier.

```
#Checking the structure of the data
str(tokens)
```

```
## 'data.frame': 170306 obs. of 3 variables:
## $ Product : chr "boAt Rockerz 255" "boAt Rockerz 255" "boAt Rockerz 255" "boAt Rockerz 255" ...
## $ ReviewStar: chr "Neutral" "Neutral" "Neutral" "Neutral" ...
## $ words : chr "doubt great" "great bass" "bass great" "great extent" ...
```

```
# Converting columns to "Factor" data type
tokens$words <- as.factor(tokens$words)
tokens$Product <- as.factor(tokens$Product)
tokens$ReviewStar <- as.factor(tokens$ReviewStar)

print(str(tokens))
```

```
## 'data.frame': 170306 obs. of 3 variables:
## $ Product : Factor w/ 10 levels "boAt Rockerz 255",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ ReviewStar: Factor w/ 3 levels "Negative","Neutral",...: 2 2 2 2 2 2 2 2 2 2 ...
## $ words : Factor w/ 90162 levels "_charger cable",...: 22279 34312 6263 34414 28004 52957 12743 19765 73951 14761 ...
## NULL
```

```
#Checking the order of the product sales to know which product sold most..
print(sort(table(tokens$Product),decreasing = T))
```

```
##
## boAt Rockerz 255 Sennheiser CX 6.0BT JBL T110BT
## 66578 61217 16132
## JBL T205BT PTron Intunes Skullcandy S2PGHW-174
## 14439 3481 2992
## Samsung EO-BG950CBEIN Flybot Wave Flybot Boom
## 2476 1838 858
## Flybot Beat
## 295
```

```
#Checking the Review that has the highest rating.
print(sort(table(tokens$ReviewStar), decreasing = T))
```

```
##
## Positive Negative Neutral
## 107108 42421 20777
```

Checking the Frequency of the tokenize customer reviews - this is to know the number of time each word appears in the data.

```
# Printing the first 6 rows from the Frequency.
word_count = as.data.frame(sort(table(tokens$words),
                                   decreasing = T))
colnames(word_count) <- c('Word','Frequency')
print(head(word_count))
```

```
##              Word Frequency
## 1      sound quality      3192
## 2      quality good       927
## 3      battery life       776
## 4 noise cancellation      639
## 5      good product       628
## 6      good sound        618
```

```
#Printing the Last 6 rows from the Frequency
print(tail(word_count))
```

```
##              Word Frequency
## 90157 zero.one star         1
## 90158 zero.please dont      1
## 90159 zero.this earphone     1
## 90160 zindagi main           1
## 90161 zipped pocket         1
## 90162 zl trash              1
```

Preparing for **WordCloud** GRAPH PLOT ...

```
# Filtering any frequency that is LESS than 40 ...
word_filter <- word_count %>%
  filter(Frequency > 39)

print(tail(word_filter))
```

```
##              Word Frequency
## 219  sound battery         41
## 220   wire long           41
## 221  love product         40
## 222  product jbl          40
## 223  quality decent       40
## 224  speaker working      40
```

```
#WordCloud Plot
wordcloud2(word_filter, size = 1.3)
```



```
##           Product ReviewStar           words    n
## 1    boAt Rockerz 255    Positive sound quality 947
## 2 Sennheiser CX 6.0BT    Positive sound quality 880
## 3    boAt Rockerz 255    Positive  battery life 399
## 4    boAt Rockerz 255    Positive  quality good 276
## 5    boAt Rockerz 255    Positive  good product 228
## 6 Sennheiser CX 6.0BT    Positive  quality good 217
```

```
# Rechecking the shape of the data.
print(dim(output))
```

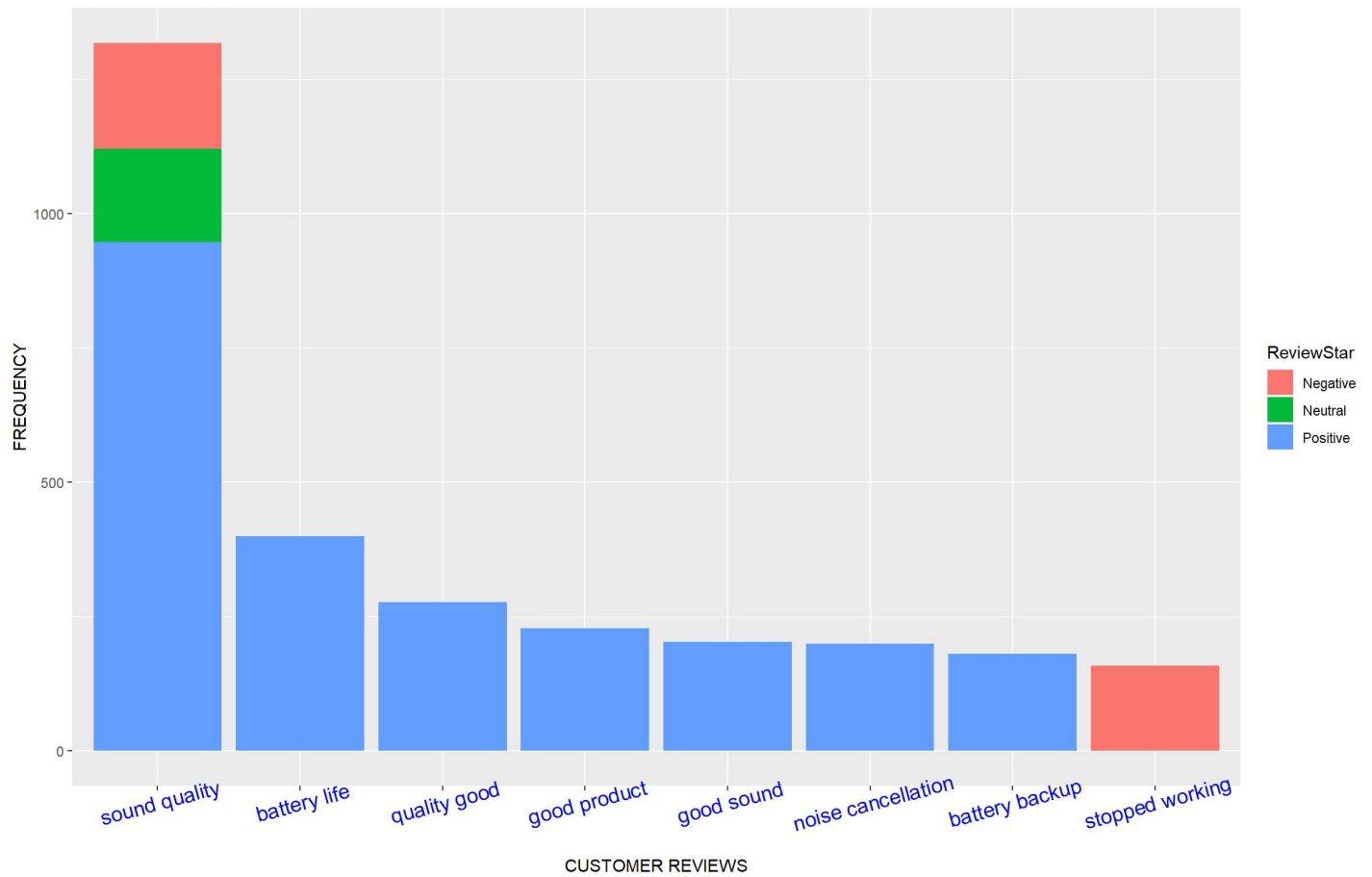
```
## [1] 2821    4
```

## GRAPH PLOT

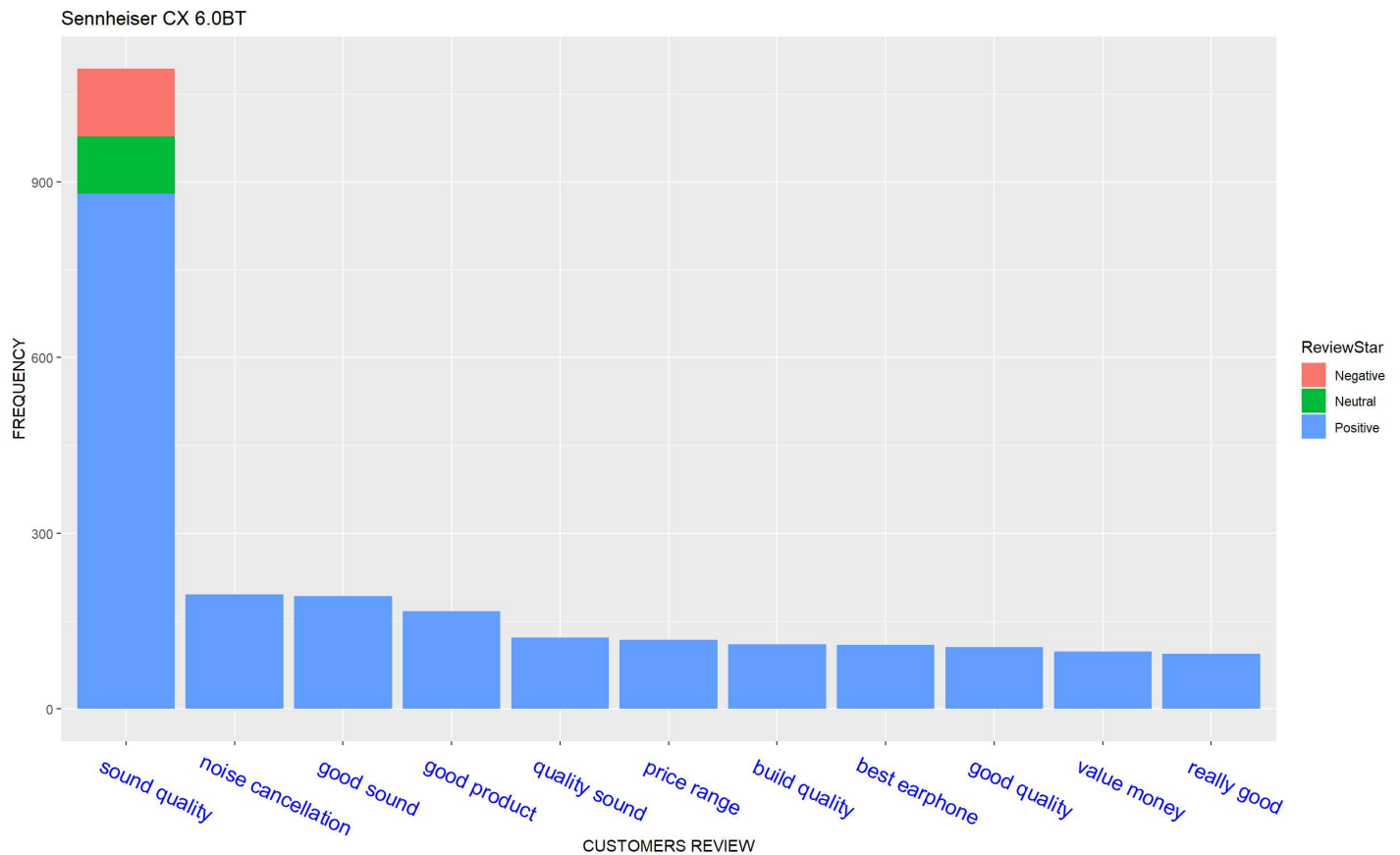
The graphical plot analysis will only be performed on the two products that has the highest sales which are:

1. boAt Rockerz 255
2. Sennheiser CX 6.0BT

```
# ggplot graph for First Product
output %>%
  filter(Product == "boAt Rockerz 255" & n > 150) %>%
  ggplot(aes(x = reorder(words, -n), y = n)) +
  geom_col(aes(fill = ReviewStar)) +
  labs(y = 'FREQUENCY', x = 'CUSTOMER REVIEWS',
        title = "boAt Rockerz 255") +
  theme(axis.text.x = element_text(angle = 15, colour = 'blue',
                                     size = 13))
```



```
# ggplot graph for Second Product
output %>%
  subset(words!='quality good') %>%
  filter(Product == 'Sennheiser CX 6.0BT' & n > 90) %>%
  ggplot(aes(x = reorder(words,-n), y = n)) +
  geom_col(aes(fill = ReviewStar)) +
  labs(x = 'CUSTOMERS REVIEW', y = 'FREQUENCY',
        title = 'Sennheiser CX 6.0BT') +
  theme(axis.text.x = element_text(angle = -25,
                                    color = 'blue',size = 14, hjust = 0.2))
```



## Summary

1. From the ggplot graph, **boAt Rockerz 255** has much positive review text because most of the **Users** comment very well in terms of the (*sound, noise, battery, quality*) while only few **Users** express their dissatisfaction concerning the *quality* and also how it *stopped working*.
2. Likewise also the user of **Sennheiser CX 6.0BT** has a much positive review about the product. Only a few of the users were not satisfied in terms of the *sound quality*.