# Sentiment-Analysis (Corvera, Paclibar, Sabarillo)"

Rotciv Corvera, Jhon Albert Paclibar, Kirk Axl Dend Sabarillo

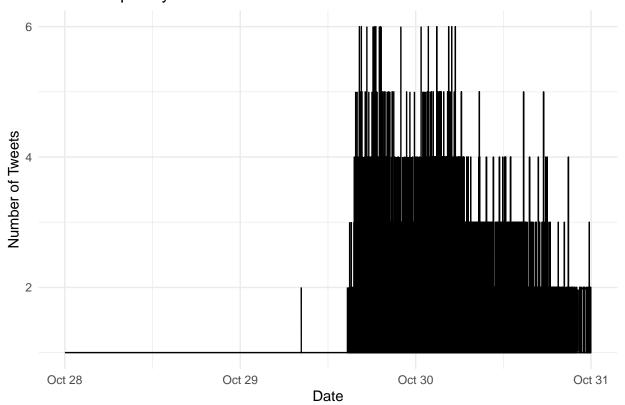
#### 2024-12-10

```
#installing packages
install.packages("lubridate")
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
## (as 'lib' is unspecified)
library(lubridate)
##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
       date, intersect, setdiff, union
install.packages("ggplot2")
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
## (as 'lib' is unspecified)
library(ggplot2)
install.packages("tidyverse")
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
## (as 'lib' is unspecified)
install.packages("tidytext")
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
## (as 'lib' is unspecified)
library(tidytext)
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.4 v stringr 1.5.1
## v forcats 1.0.0
                     v tibble 3.2.1
## v purrr
           1.0.2
                      v tidyr
                               1.3.1
## v readr
           2.1.5
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
# Read the CSV file
tweet_data <- read.csv("tweetsDF.csv", stringsAsFactors = FALSE)</pre>
```

```
# Convert the created column to a datetime format
tweet_data$created <- ymd_hms(tweet_data$created)</pre>
```

```
# Create a time series plot
ggplot(tweet_data, aes(x = created)) +
  geom_line(stat = "count") +
  labs(x = "Date", y = "Number of Tweets", title = "Tweet Frequency Over Time") +
  theme_minimal()
```

### Tweet Frequency Over Time



```
#By analyzing the time series data, we can identify the specific
#date when user activity on the platform peaked,
#revealing when people were most active.
#Here, it shows that most people tweet during October 30.

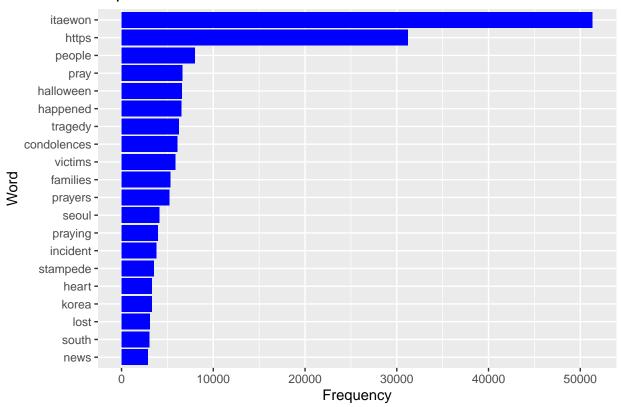
# Create a tokenized data frame
tokens <- tweet_data %>%
    unnest_tokens(word, text)

# Remove stop words
tokens <- tokens %>%
    anti_join(stop_words)
```

```
## Joining with `by = join_by(word)`
# Count the frequency of each word
word_counts <- tokens %>%
    count(word, sort = TRUE)
```

```
# Visualize the top 20 words
word_counts %>%
  head(20) %>%
  ggplot(aes(x = reorder(word, n), y = n)) +
  geom_col(fill = "blue") +
  labs(x = "Word", y = "Frequency", title = "Top 20 Words Used") +
  coord_flip()
```

## Top 20 Words Used

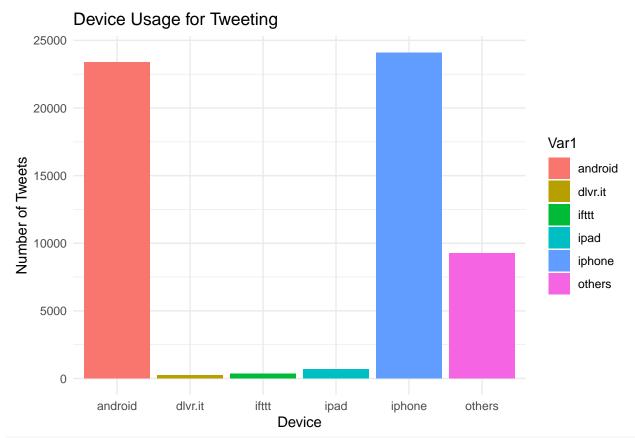


```
#Using the bar chart, we can pinpoint the most frequently
#used word in the tweets during that period.
#This allows us to infer and identify the recent event that captured people's attention.
#Here, the data reveals that the most tweeted word was
#"Itaewon," a district located in Seoul, South Korea.

# Count the frequency of each device
device_counts <- table(tweet_data$tweetSource)

# Convert the table to a data frame for ggplot2
device_df <- as.data.frame(device_counts)

# Create a bar plot using ggplot2
ggplot(device_df, aes(x = Var1, y = Freq, fill = Var1)) +
    geom_bar(stat = "identity") +
    labs(x = "Device", y = "Number of Tweets", title = "Device Usage for Tweeting") +
    theme_minimal()</pre>
```



#The graph illustrates the devices most commonly used for posting tweets.

#The results indicate that the majority of users tweeted from iPhones,

#with Android devices coming in second, showing only a small gap between them.

### **#USE CASE OF THE ANALYSIS**

#Understand how people reacted to and engaged with the event over time.

#Detect key topics or hashtags that shaped the narrative.

#Identify peak engagement times to plan content releases.

#Tailor campaigns based on trending topics or popular keywords.

#Analyze device usage to infer user demographics and preferences.

#Target campaigns for specific devices (e.g., iOS vs. Android).

#For emergencies or tragedies (as suggested by the Itaewon mention), monitor public sentiment and respo

#Provide actionable insights to policymakers or organizations for timely interventions.