

Project_Sentiment_Analysis#(BASA, LLANERA, TUARES)

BASA, LLANERA, TUARES

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```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.1      v tibble    3.2.1
## v lubridate  1.9.4      v tidyr     1.3.1
## v purrr      1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(lubridate)
library(ggplot2)
library(tidytext)
library(sentimentr)
```

Data Loading and Inspection

```
data <- read.csv("/cloud/project/Project_Sentiment_Analysis/tweetsDF.csv")
```

```
str(data)
```

```
## 'data.frame':    58086 obs. of  7 variables:
## $ X              : int  1 2 3 4 5 6 7 8 9 10 ...
## $ screenName      : chr  "whourj31" "nnainot" "febry_sri_M" "telehuntwatch" ...
## $ text             : chr  "A soldier angry at the support fund consolation money for the bereaved fa
## $ created          : chr  "2022-10-30 23:59:43" "2022-10-30 23:59:32" "2022-10-30 23:59:31" "2022-10-
## $ statusSource     : chr  "<a href=\"https://www.fs-poster.com/\" rel=\"nofollow\">FS_Poster_App</a>
## $ Created_At_Round: chr  "2022-10-31 00:00:00" "2022-10-31 00:00:00" "2022-10-31 00:00:00" "2022-10-
## $ tweetSource      : chr  "others" "android" "android" "others" ...
```

```
summary(data)
```

```
##           X           screenName           text           created
## Min.      :    1   Length:58086   Length:58086   Length:58086
## 1st Qu.:14522   Class :character   Class :character   Class :character
## Median :29044   Mode  :character   Mode  :character   Mode  :character
## Mean      :29044
## 3rd Qu.:43565
## Max.      :58086
## statusSource   Created_At_Round   tweetSource
## Length:58086   Length:58086   Length:58086
```

```
## Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character
##
##
##
```

```
head(data)
```

```
##   X      screenName
## 1 1      whourj31
## 2 2      nnainot
## 3 3    febry_sri_M
## 4 4 telehuntwatch
## 5 5    Typing0824
## 6 6   niccijsmith
##
## 1          A soldier angry at the support fund consolation money for the bereaved family of the Itaewon
## 2                                     Nah this Itaewon
## 3
## 4 TRANSLATION : \nSeoul residents lay flowers at a makeshift memorial near the site of the crush in Itaewon
## 5 The Itaewon stampede incident really caught me off guard. Makes me notice how important it is to be careful
## 6 "What to do about my child? What to do about my child?" Park Ga-young's mother, Choi Seon-mi, said
##      created
## 1 2022-10-30 23:59:43
## 2 2022-10-30 23:59:32
## 3 2022-10-30 23:59:31
## 4 2022-10-30 23:59:28
## 5 2022-10-30 23:59:20
## 6 2022-10-30 23:59:04
##
##                                     statusSource
## 1          <a href="https://www.fs-poster.com/" rel="nofollow">FS_Poster_App</a>
## 2 <a href="http://twitter.com/download/android" rel="nofollow">Twitter for Android</a>
## 3 <a href="http://twitter.com/download/android" rel="nofollow">Twitter for Android</a>
## 4          <a href="https://ruprop.live" rel="nofollow">telehunt</a>
## 5 <a href="http://twitter.com/download/android" rel="nofollow">Twitter for Android</a>
## 6   <a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhone</a>
##      Created_At_Round tweetSource
## 1 2022-10-31 00:00:00      others
## 2 2022-10-31 00:00:00      android
## 3 2022-10-31 00:00:00      android
## 4 2022-10-31 00:00:00      others
## 5 2022-10-31 00:00:00      android
## 6 2022-10-31 00:00:00      iphone
```

Data Cleaning

```
data <- data %>% distinct() %>% drop_na()
if (!"created" %in% colnames(data)) {
  stop("The dataset is missing a 'created' column. Please verify that the dataset includes a timestamp column.")
}

data <- data %>% rename(datetime = created)

data <- data %>% mutate(datetime = ymd_hms(datetime))

data <- data %>%
```

```
mutate(text = str_remove_all(text, "http[s]?://\\S+|#\\w+|@\\w+|[^\\w\\s]"))

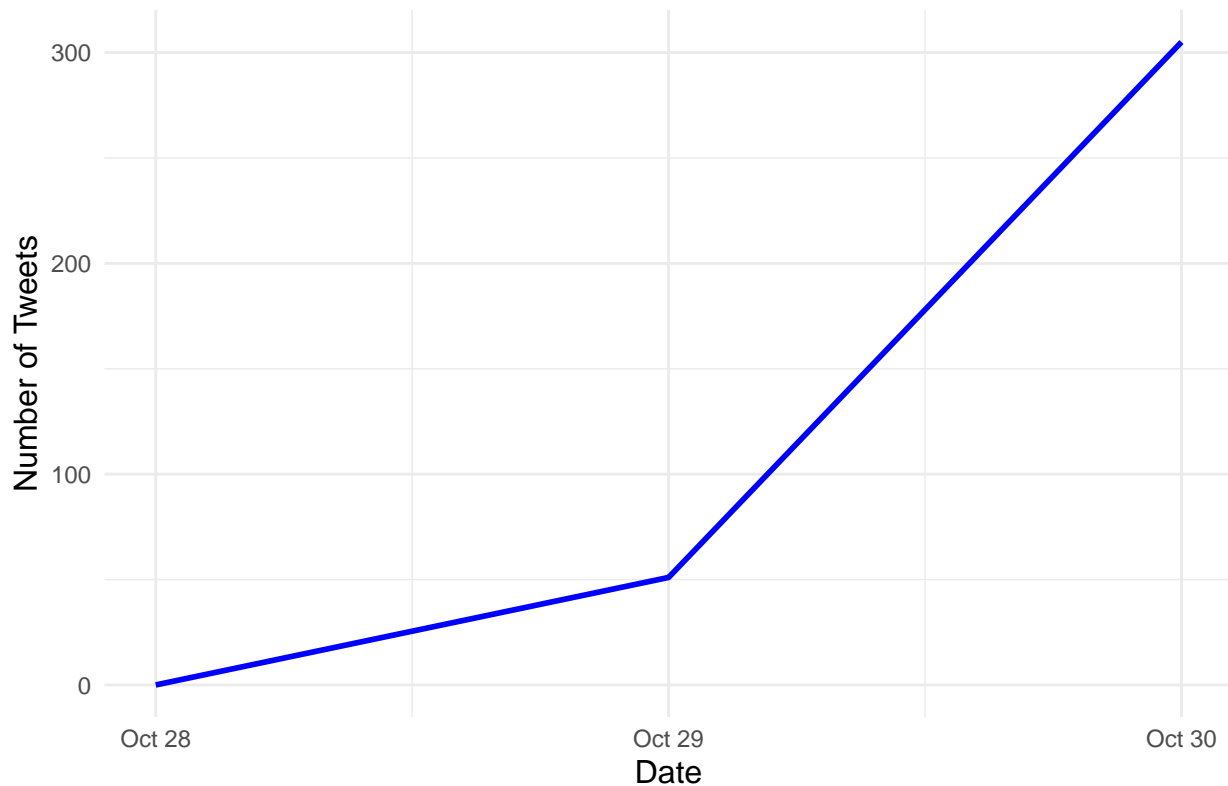
keywords <- c("Blackpink", "Rose", "Lisa", "Jennie", "Jisoo", "concert")
data_filtered <- data %>% filter(str_detect(tolower(text), paste(keywords, collapse = "|")))
```

Trend Analysis

```
# Restrict data to October 28-30, 2022
data_filtered <- data_filtered %>%
  filter(datetime >= as.POSIXct("2022-10-28 00:00:00") & datetime <= as.POSIXct("2022-10-30 23:59:59"))
# Count tweets over time
time_trend <- data_filtered %>%
  mutate(date = as.Date(datetime)) %>%
  group_by(date) %>%
  summarise(tweet_count = n())
# Ensure all dates in the range are included, even with zero counts
date_range <- seq(as.Date("2022-10-28"), as.Date("2022-10-30"), by = "day")
time_trend <- time_trend %>%
  complete(date = date_range, fill = list(tweet_count = 0))
# Plot tweet trends over time
ggplot(time_trend, aes(x = date, y = tweet_count)) +
  geom_line(color = "blue", size = 1) +
  labs(title = "Tweet Trends for Blackpink and Members (Oct 28-30)", x = "Date", y = "Number of Tweets") +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5, size = 16), axis.title = element_text(size = 12))
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

Tweet Trends for Blackpink and Members (Oct 28–30)



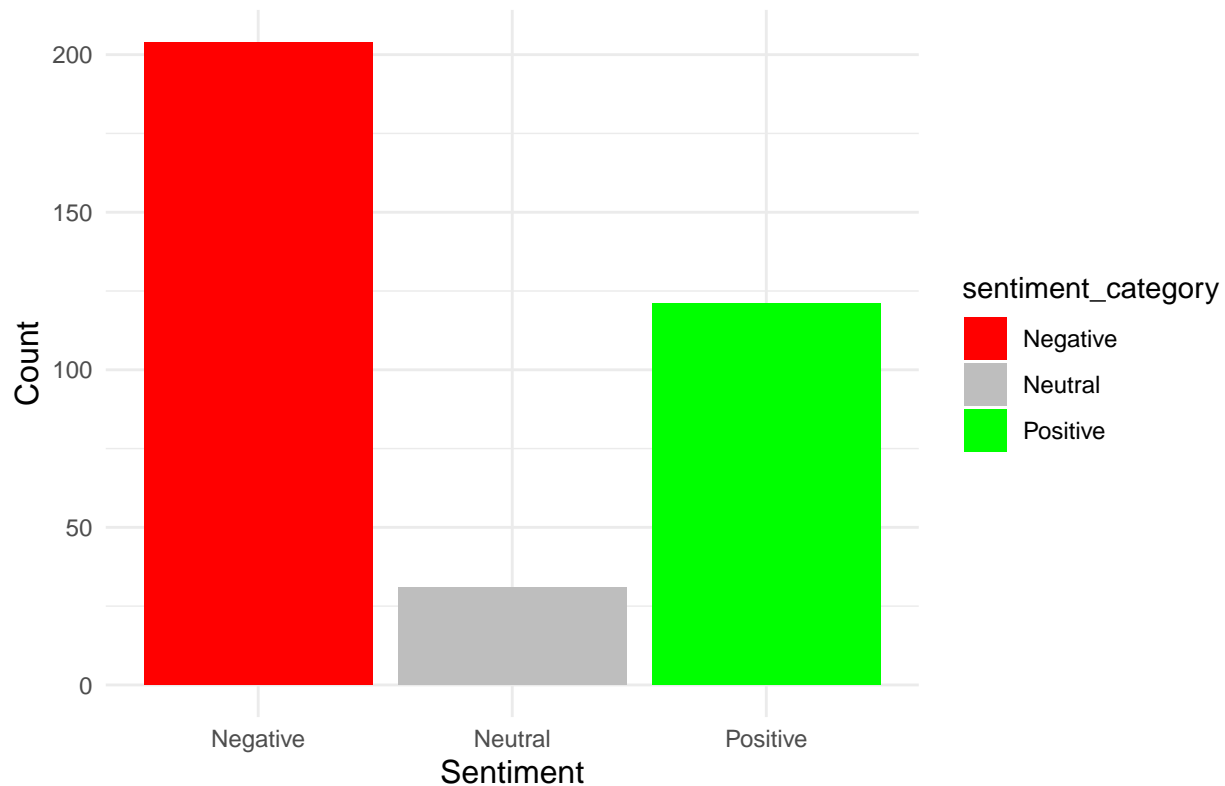
*# **Trend Analysis Insights:***

The tweet activity mentioning BLACKPINK and its members between October 28 and October 30, 2022, shows a significant increase. This surge in activity coincides with discussions about the Itaewon tragedy and BLACKPINK's decision to perform at the event. The contrast in opinions likely heightened public interest, as the conversation became polarized between supporters and critics. This spike also highlights the influential role of social media, where high-profile events and controversies can rapidly spread and gain traction.

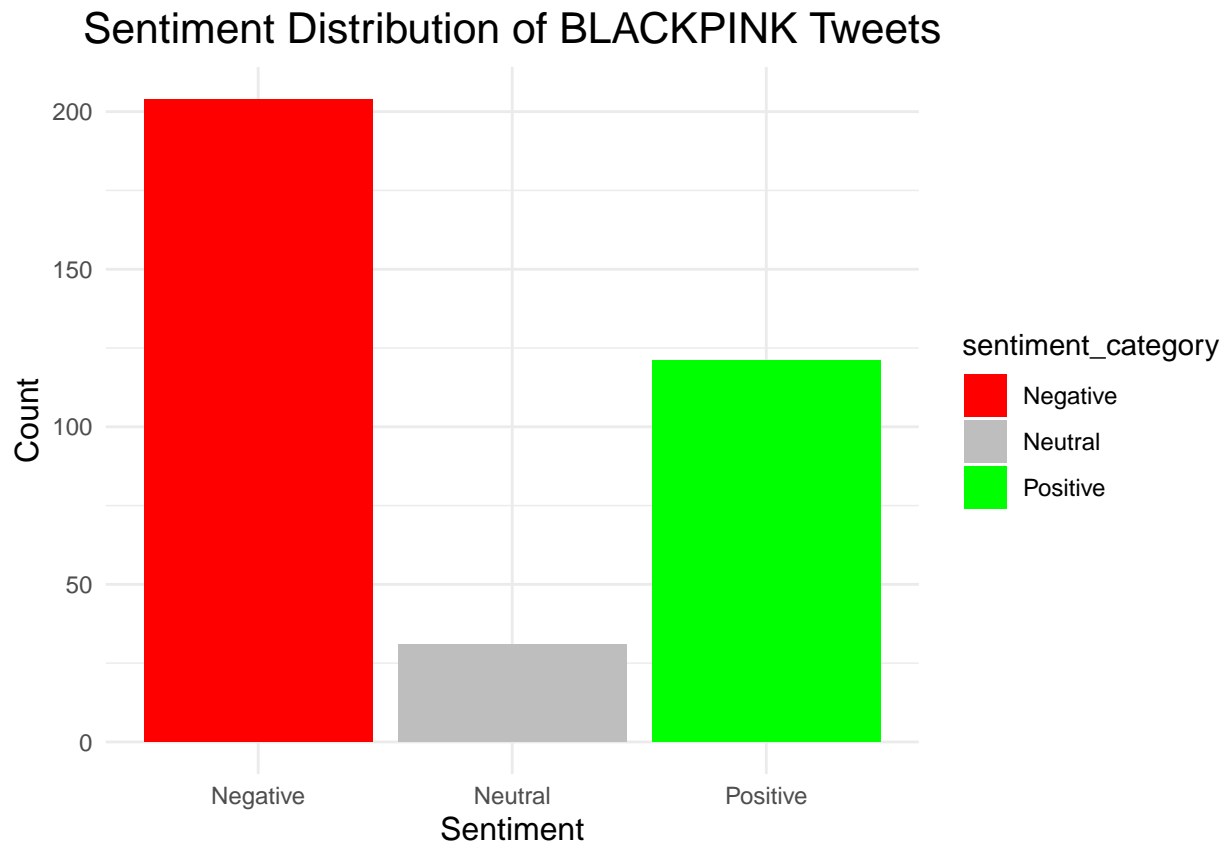
Sentiment Analysis

```
# Perform sentiment analysis on the filtered tweets
sentiment_scores <- sentiment_by(data_filtered$text)
data_filtered$sentiment <- sentiment_scores$sentiment
# Categorize sentiment
data_filtered <- data_filtered %>%
  mutate(sentiment_category = case_when(
    sentiment > 0 ~ "Positive",
    sentiment < 0 ~ "Negative",
    TRUE ~ "Neutral"
  ))
# Sentiment Distribution
ggplot(data_filtered, aes(x = sentiment_category, fill = sentiment_category)) +
  geom_bar() +
  labs(title = "Sentiment Distribution of BLACKPINK Tweets", x = "Sentiment", y = "Count") +
  scale_fill_manual(values = c("Positive" = "green", "Neutral" = "gray", "Negative" = "red")) +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5, size = 16), axis.title = element_text(size = 12))
```

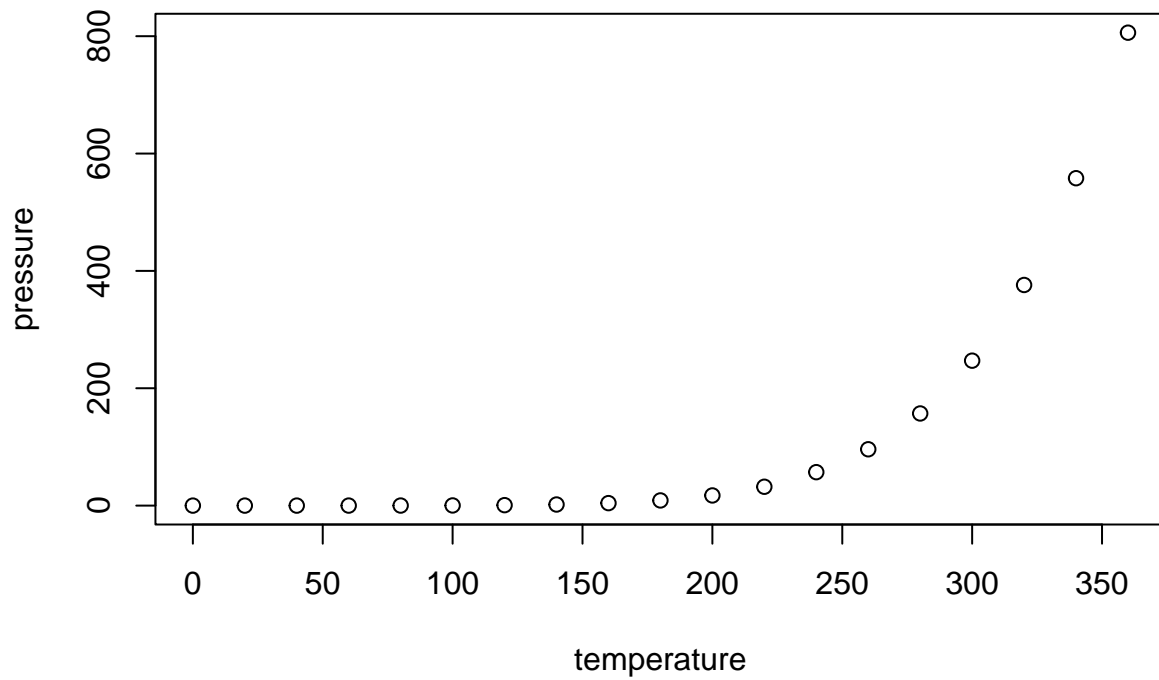
Sentiment Distribution of BLACKPINK Tweets



```
# Perform sentiment analysis on the filtered tweets
sentiment_scores <- sentiment_by(data_filtered$text)
data_filtered$sentiment <- sentiment_scores$ave_sentiment
# Categorize sentiment
data_filtered <- data_filtered %>%
  mutate(sentiment_category = case_when(
    sentiment > 0 ~ "Positive",
    sentiment < 0 ~ "Negative",
    TRUE ~ "Neutral"
  ))
# Sentiment Distribution
ggplot(data_filtered, aes(x = sentiment_category, fill = sentiment_category)) +
  geom_bar() +
  labs(title = "Sentiment Distribution of BLACKPINK Tweets", x = "Sentiment", y = "Count") +
  scale_fill_manual(values = c("Positive" = "green", "Neutral" = "gray", "Negative" = "red")) +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5, size = 16), axis.title = element_text(size = 12))
```



*# **Sentiment Analysis Insights:***
Sentiment analysis of the tweets reveals that negative sentiment was the most dominant during this period.
A significant number of tweets expressed disapproval of BLACKPINK and their management for continuing to release new music.
Upon further examination, some of the negative sentiment appeared to be driven by online trolls amplifying negativity.
While positive sentiments were less common, they reflected fan support for BLACKPINK and sympathy for the group's members.
Neutral sentiment was the least common, as most tweets conveyed strong emotional responses regarding the group's actions.



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.