# Data Cleaning & Sentiment Score

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## Data Prep

Read tweets from the files & Find the overlapping time frame

```
# Read data from csv files
bernie_original <- read_csv("bernie_tweets.csv")</pre>
warren_original <- read_csv("warren_tweets.csv")</pre>
pete_original <- read_csv("buttigieg_tweets.csv")</pre>
bloomberg_original <- read_csv("bloomberg_tweets_updated.csv")</pre>
amy_original <- read_csv("klobuchar_tweets.csv")</pre>
# Convert ID to character from numeric.
bernie_original$user_id <- as.character(bernie_original$user_id)</pre>
warren_original$user_id <- as.character(warren_original$user_id)</pre>
pete_original$user_id <- as.character(pete_original$user_id)</pre>
bloomberg_original$user_id <- as.character(bloomberg_original$user_id)
amy_original$user_id <- as.character(amy_original$user_id)</pre>
# check the time frame of each data
date_check <- function(df){</pre>
  df %>%
  group_by(1) %>%
  summarise(max = max(created), min = min(created))
}
# apply tthe new function to each data frame
bernie_date <- date_check(bernie_original)</pre>
warren_date <- date_check(warren_original)</pre>
pete_date <- date_check(pete_original)</pre>
bloomberg_date <- date_check(bloomberg_original)</pre>
amy_date <- date_check(amy_original)</pre>
# find overlapping time period
date_frame <- bind_rows(bernie_date, warren_date, pete_date, bloomberg_date, amy_date, .id = "id")[, -2]
```

### Preparing/cleaning the data

The first step of data cleaning process entails removing:

- 1. urls
- 2. character strings between "<" and ">" to deal with smileys and other encoded text.
- 3. retweet marks, @RT
- 4. quotation marks and apostrophes
- 5. any @userid
- 6. punctuation and blank spaces
- 7. stopwords and single letters

```
clean_tweets <- function(df) {</pre>
  # Remove URLs
  df$text <- gsub("http[^[:space:]]*", "",df$text)</pre>
  # Remove retweet entities
  df\$text \leftarrow gsub("(RT|via)((?:\b\\w*@\\w+)+)"," ", df\$text)
  # Remove quotes
  df$text <- gsub("'s|'s|[...]", "", df$text)</pre>
  # Remove at people
  df$text <- gsub("@\\w+", " ", df$text)</pre>
  # Remove punctuation
  df$text <- gsub("[[:punct:]]", " ", df$text)</pre>
  # Remove single letters.
  df$text <- gsub(" *\\b[[:alpha:]]{1}\\b *", "", df$text)</pre>
  # Remove unnecessary spaces
  df$text <- gsub("[ \t]{2,}", " ", df$text)</pre>
  # Remove leading and trailing whitespaces
  df$text <- gsub("^\\s+\\\s+$", "", df$text)</pre>
  ## parsing, tokenizing, and re-grouping text column
  df <- df %>%
    unnest_tokens(output = word, input = text) %>%
    anti_join(stop_words) %>%
    filter(!str_detect(word, "^[0-9]*$")) %>%
    group_by(username, created) %>%
    summarize(text = str_c(word, collapse = " ")) %>%
    ungroup()
    # creating a date column without time
    df$date <- as.Date(df$created, "%Y-%m-%d")</pre>
    # setting date objects
    start <- as.Date("2020-01-02")
    end <- as.Date("2020-02-26")
    # subsetting by date range
    df <- df %>%
      subset(date >= start & date <= end) %>%
      select(-created)
}
bernie_clean <- clean_tweets(bernie_original)</pre>
warren_clean <- clean_tweets(warren_original)</pre>
pete_clean <- clean_tweets(pete_original)</pre>
bloomberg_clean <- clean_tweets(bloomberg_original)</pre>
amy_clean <- clean_tweets(amy_original)</pre>
```

### Check for duplicates (Sample Data: Bernie Sanders)

# creating a data frame only containing the text column of the clean data

```
bernie_text <- as.tibble(bernie_clean$text)

# groupingn by frequency of the same texts/words
grouped_bernie_text <- aggregate(bernie_text, by = list(bernie_clean$text), FUN = length);
colnames(grouped_bernie_text) <- c("Text","TweetCount")

# reordering by frequency rate
grouped_bernie_text <- arrange(grouped_bernie_text, desc(TweetCount))

# duplicated words across the tweets</pre>
```

```
bernie_text_duplicates <- subset(grouped_bernie_text, grouped_bernie_text$TweetCount > 1)

# un-commen the code below to take a look
# head(bernie_text_duplicates, n = 20)

# finding any tweet that contains 4 words below as it's likely to be a meaningless tweet
# check Tweetcount for how many times they appear in the dataset
duplicates_bernie <- bernie_text_duplicates[grep("stock|retweet|follow|update", bernie_text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplicates$Text_duplic
```

Still looking for a way to remove the duplicates in the data in an efficient manner.

#### Get sentiment scores

```
# function to calculate the sentiment score
sentiment_score <- function(df) {</pre>
  df %>%
    unnest %>%
    get_sentences() %>%
    # get sentiment score for each tweet
    sentiment() %>%
    mutate(characters = nchar(stripWhitespace(text))) %>%
    filter(characters > 1 )
    # same number of obs, hence no error
}
# get sentiment scores for all our data frames
bernie_sent <- sentiment_score(bernie_clean)</pre>
warren_sent <- sentiment_score(warren_clean)</pre>
pete_sent <- sentiment_score(pete_clean)</pre>
bloomberg_sent <- sentiment_score(bloomberg_clean)</pre>
amy_sent <- sentiment_score(amy_clean)</pre>
# quick summary of the result
skim(bernie_sent$sentiment)
```

Table 1: Data summary

Name	bernie_sent\$sentiment
Number of rows	51567
Number of columns	1
Column type frequency:	
numeric	1
Group variables	None

### Variable type: numeric

skim_variable	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100	hist
data	0	1	-0.03	0.35	-2.3	-0.22	0	0.17	2.87	

Table 3: Data summary

Name	warren_sent\$sentiment
Number of rows	52736
Number of columns	1
Column type frequency:	
numeric	1
Group variables	None

## Variable type: numeric

skim_variable	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100	hist
data	0	1	-0.03	0.34	-2.76	-0.22	0	0.18	1.88	

## skim(pete\_sent\$sentiment)

Table 5: Data summary

Name	pete_sent\$sentiment
Number of rows	50601
Number of columns	1
Column type frequency: numeric	1
Group variables	None

## Variable type: numeric

skim_variable	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100	hist
data	0	1	0.01	0.36	-2.65	-0.15	0	0.21	1.96	

## skim(bloomberg\_sent\$sentiment)

Table 7: Data summary

Name Number of rows	bloomberg_sent\$sentiment 55134
Number of columns	1
Column type frequency: numeric	1
Group variables	None

## Variable type: numeric

skim_variable	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100	hist
data	0	1	-0.03	0.31	-1.92	-0.19	0	0.13	1.53	

### skim(amy\_sent\$sentiment)

Table 9: Data summary

Name	amy_sent\$sentiment
Number of rows	51568
Number of columns	1
Column type frequency:	
numeric	1
Group variables	None

### Variable type: numeric

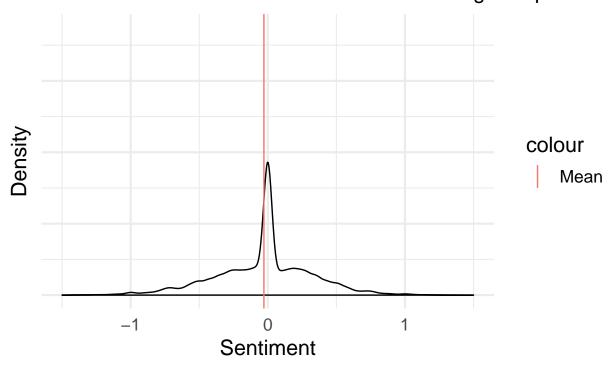
skim_variable	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100	hist
data	0	1	0.02	0.35	-2.3	-0.14	0	0.21	2.06	

#### Plot the result

```
# function to plot the distribution of sentiment scores
sentiment_plot <- function(df){</pre>
  ggplot(df, aes(sentiment)) +
    geom_density() +
    scale_y_continuous(limits = c(0,7.5)) +
    scale_x_continuous(limits = c(-1.5, 1.5)) +
    theme_minimal(base_size = 16) +
    geom_vline(aes(xintercept = mean(sentiment), color = "Mean"))
}
# customizing the labels
sentiment_plot(bernie_sent) +
  labs(x = "Sentiment",
       y = "Density",
       title = "Bernie Sanders",
       subtitle = "The Distribution of Sentiment Across R Package Help Docs") +
  theme(plot.title = element_text(hjust = 0.5),
        axis.text.y=element_blank())
```

## **Bernie Sanders**

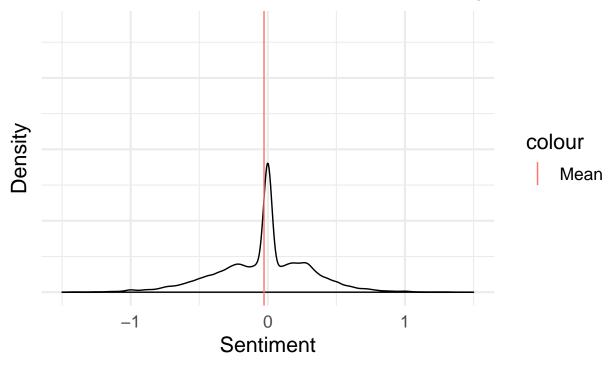
The Distribution of Sentiment Across R Package Help Docs



```
sentiment_plot(warren_sent) +
labs(x = "Sentiment",
    y = "Density",
    title = "Elizabeth Warren",
    subtitle = "The Distribution of Sentiment Across R Package Help Docs") +
theme(plot.title = element_text(hjust = 0.5),
    axis.text.y=element_blank())
```

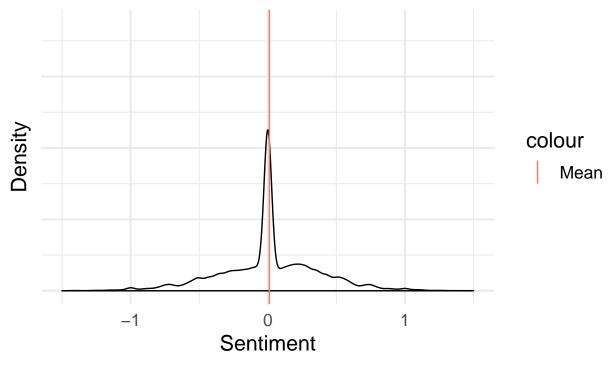
## Elizabeth Warren

The Distribution of Sentiment Across R Package Help Docs



# Pete Buttigieg

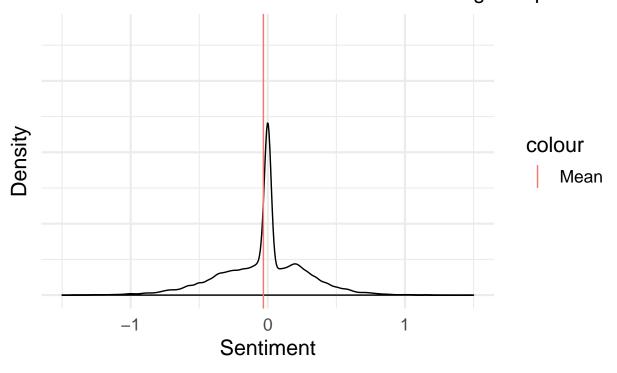
The Distribution of Sentiment Across R Package Help Docs



```
sentiment_plot(bloomberg_sent) +
labs(x = "Sentiment",
    y = "Density",
    title = "Mike Bloomberg",
    subtitle = "The Distribution of Sentiment Across R Package Help Docs") +
theme(plot.title = element_text(hjust = 0.5),
    axis.text.y=element_blank())
```

# Mike Bloomberg

The Distribution of Sentiment Across R Package Help Docs



```
sentiment_plot(amy_sent) +
labs(x = "Sentiment",
    y = "Density",
    title = "Amy Klobuchar",
    subtitle = "The Distribution of Sentiment Across R Package Help Docs") +
theme(plot.title = element_text(hjust = 0.5),
    axis.text.y=element_blank())
```

# Amy Klobuchar

The Distribution of Sentiment Across R Package Help Docs

