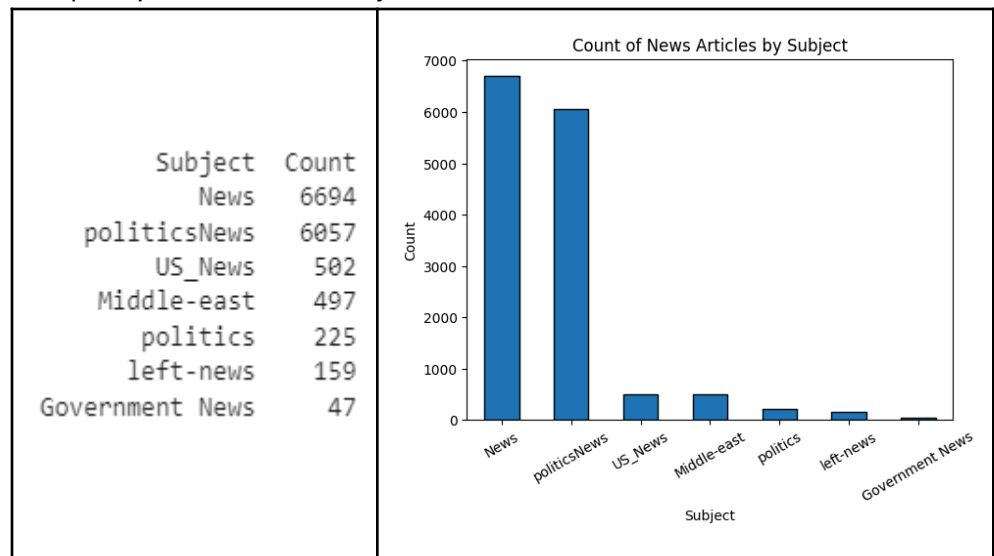
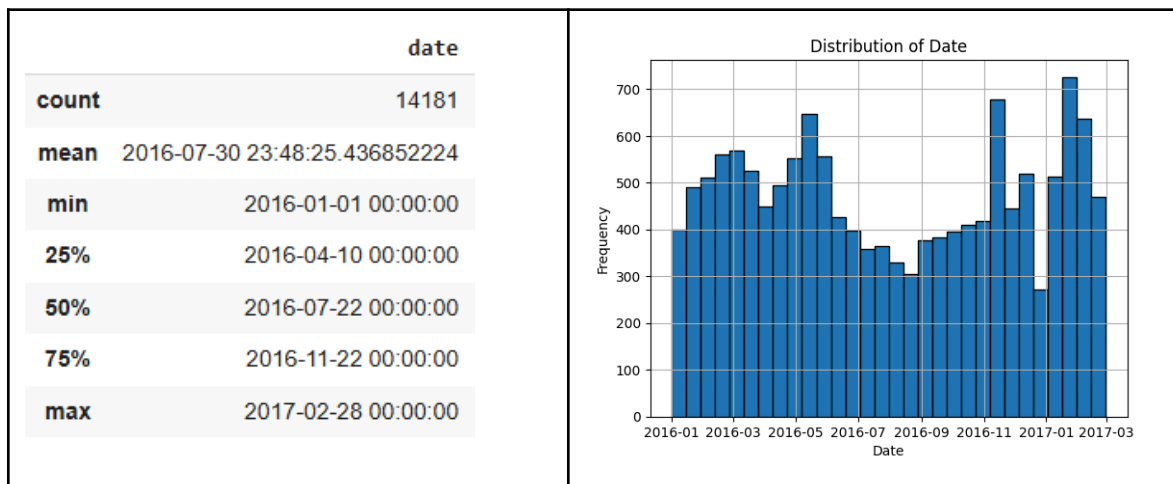


- Possible Responses: [World-News, Politics-News, Government-News, Middle-east, US News, left-news, politics, News]
- Frequency Information of Subject Areas

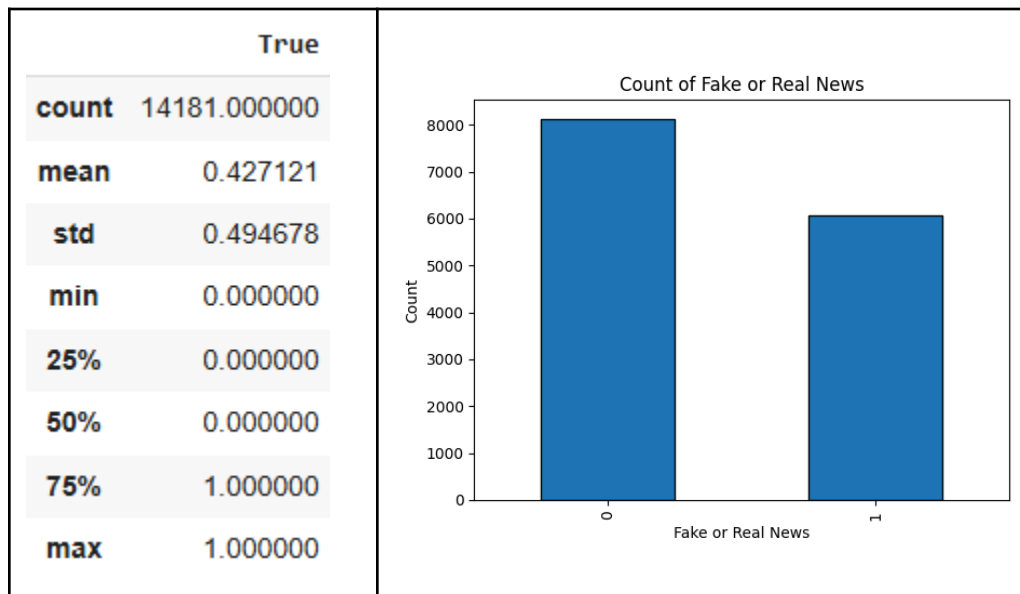


○ date:

- Definition: the publishing date of each article
- Data Type: date object
- Processing steps:
 - a) In order to focus on the hypothesis, which focused on proximity to news near the 2016 election, the data was filtered to only include data entries between January 2016 to March 2017
 - b) In order to perform this, we transformed the initial dates into a uniform date object.
 - c) Then, we filtered the data entries to only include articles published between 01/01/16 and 03/01/2017
- Response: Ranges from January 2016 to March 2017
- Statistics:



- True:
 - Definition: indicates whether an article is fake or real
 - Data type: numerical, categorical
 - Processing steps:
 - a) This column was added from the initial data to merge the fake and real news articles into a single dataset
 - b) To do this, fake new articles were mapped to 0 and real new articles mapped to 1. From there, the two datasets were merged
 - Response: Values are either 0 or 1
 - a) 0 signifies fake news, 1 signifies real news
 - Statistics:



- clean_text:
 - Definition: the processed text of each article after data cleaning function was applied
 - Processing steps:
 - a) In order to create this column, a python function was applied to every entry, transforming the text from the *text* column to the *clean_text* column
 - b) The python function used for clearing removed whitespace, URL formatting, and numbers in order to prepare for sentiment analysis
 - Response: Varies
- sentiment:
 - Definition: the sentiment score produced by the VADER python package for each article
 - Processing steps:
 - a) To create this column, the VADER python package SentimentIntensityAnalyzer was applied to the clean_text column

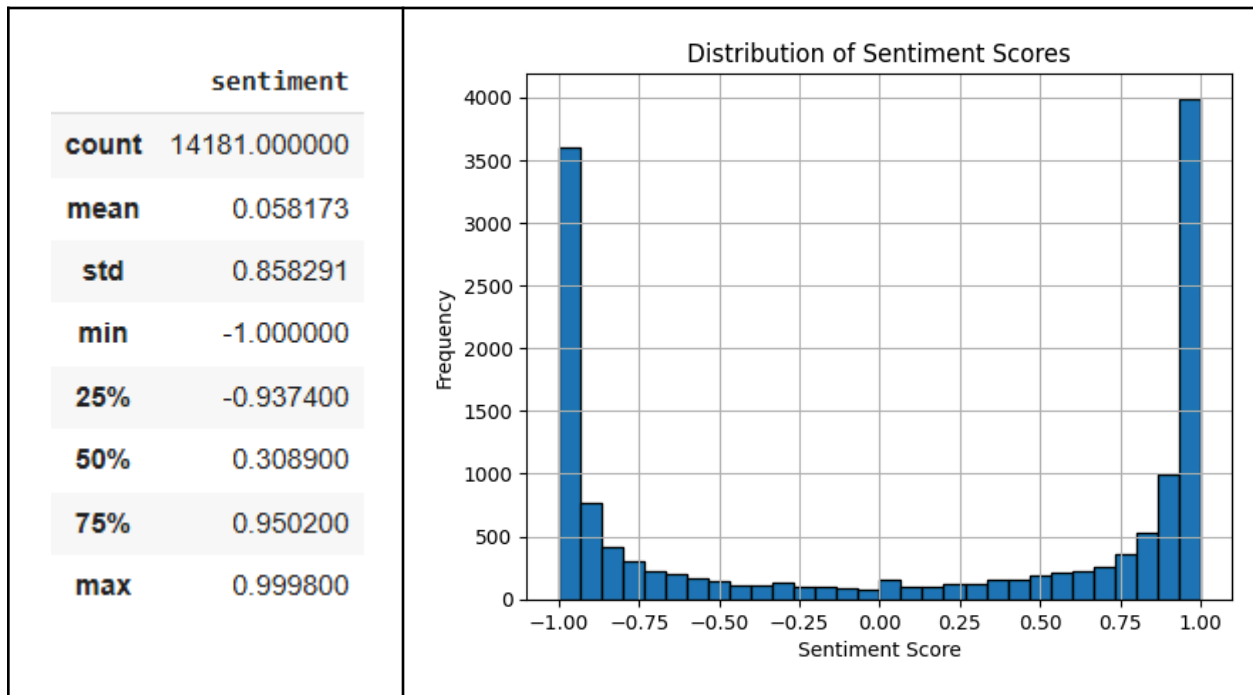
- b) This produced a sentiment score ranging from -1 to 1 for every data entry

```
# Initialize VADER Sentiment Analyzer
sia = SentimentIntensityAnalyzer()

# Apply sentiment analysis
filtered_df['sentiment'] = filtered_df['clean_text'].apply(lambda x: sia.polarity_scores(x)['compound'])
```

c)

- Response: the sentiment scores ranged from -1 to 1, with scores closer to -1 reflecting more negative sentiment and scores closer to 1 reflecting more positive sentiment
- Statistics:



○ words:

- Definition: the list of words from the clean_text column formatted as a list
- Processing steps:
 - a) In order to create this column, the text from clean_text columns were tokenized into separate words, formatted as a list

```
# Tokenize text into words
filtered_df['words'] = filtered_df['clean_text'].str.split()
```

b)

- Response: Varies