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### I. Big Data problem

Given tweets regarding pineapple on pizza, how many users believe that pineapple should or should not belong on pizza?

### II. Description of source dataset

- A. The following phrases were used as a datasource:
  - 1. Positive:
    - a) "I like pineapples on pizza"
  - 2. Negative:
    - a) "I don't like pineapples on pizza"
    - b) "I do not like pineapples on pizza"
- B. Tweets will be scraped and the following data will be taken from each tweet:
  - 1. Username and Full Name
  - 2. Tweet-id
  - 3. Tweet text
  - 4. Tweet timestamp
  - 5. No. of likes
  - 6. No. of replies
  - 7. No. of retweets
- C. Each like and each retweet of a tweet may also be recorded to see how many people agree with the tweet
- D. How to Obtain the Data Set
  - 1. At first, the group wanted to use the API of twitter to retrieve the data set, however, twitter's API allows the user to scrape tweets within 7 days only.
  - 2. To acquire the data set, we used a twitterscraper found on github by taspinar. It can be found here: <a href="https://github.com/taspinar/twitterscraper">https://github.com/taspinar/twitterscraper</a>.
  - 3. We then went to twitter and searched for the queries: "I like pineapples on pizza", "I don't like pineapples on pizza", and "I do not like pineapples on pizza". We then copied a part of the url of each of the searched queries

on the terminal and ran the twitterscraper. The data for each query was saved as a json file.

### E. Sample query codes:

1. twitterscraper

"i%20like%20pineapples%20on%20pizza"%20since%3A2017-01-01%20until %3A2017-11-30 like2017.json

twitterscraper

"i%20like%20pineapples%20on%20pizza"%20since%3A2016-01-01%20until %3A2016-12-31 like2016.json

# III. Description of the output, explanation and discussion of the output in relation to the Big Data problem

### A. Output:

#### I LIKE PINEAPPLES ON PIZZA

### Yearly

|      | •      |
|------|--------|
| YEAR | COUNT  |
| 2016 | 112347 |
| 2017 | 270910 |

### Same month

| MONTH     | COUNT |
|-----------|-------|
| January   | 19375 |
| February  | 41314 |
| March     | 66418 |
| April     | 60365 |
| May       | 16221 |
| June      | 15105 |
| July      | 33079 |
| August    | 19075 |
| September | 17879 |

| October  | 16063 |
|----------|-------|
| November | 73242 |
| December | 5121  |

# Monthly (2016-2017)

| YEAR | MONTH     | COUNT |
|------|-----------|-------|
| 2016 | January   | 659   |
| 2016 | February  | 361   |
| 2016 | March     | 1108  |
| 2016 | April     | 483   |
| 2016 | May       | 771   |
| 2016 | June      | 872   |
| 2016 | July      | 969   |
| 2016 | August    | 2512  |
| 2016 | September | 1586  |
| 2016 | October   | 1278  |
| 2016 | November  | 46236 |
| 2016 | December  | 1646  |
| 2017 | January   | 4718  |
| 2017 | February  | 9535  |
| 2017 | March     | 6764  |
| 2017 | April     | 7346  |

| 2017 | May       | 3793  |
|------|-----------|-------|
| 2017 | June      | 4135  |
| 2017 | July      | 4691  |
| 2017 | August    | 3362  |
| 2017 | September | 5475  |
| 2017 | October   | 2484  |
| 2017 | November  | 14125 |

### I DON'T LIKE PINEAPPLES ON PIZZA

# Yearly

| YEAR | COUNT |
|------|-------|
| 2016 | 58481 |
| 2017 | 66246 |

### Same month

| MONTH    | COUNT |
|----------|-------|
| January  | 5377  |
| February | 9714  |
| March    | 7872  |
| April    | 7829  |
| May      | 4564  |
| June     | 5007  |

| July      | 5660  |
|-----------|-------|
| August    | 5874  |
| September | 7061  |
| October   | 3762  |
| November  | 60361 |
| December  | 1646  |

# Monthly (2016-2017)

| YEAR | MONTH     | COUNT |
|------|-----------|-------|
| 2016 | January   | 2401  |
| 2016 | February  | 2185  |
| 2016 | March     | 23298 |
| 2016 | April     | 1978  |
| 2016 | May       | 2085  |
| 2016 | June      | 2719  |
| 2016 | July      | 3380  |
| 2016 | August    | 7571  |
| 2016 | September | 4586  |
| 2016 | October   | 5016  |
| 2016 | November  | 52367 |
| 2016 | December  | 5121  |

| 2017 | January   | 17334 |
|------|-----------|-------|
| 2017 | February  | 39129 |
| 2017 | March     | 43120 |
| 2017 | April     | 58387 |
| 2017 | May       | 14136 |
| 2017 | June      | 12386 |
| 2017 | July      | 29699 |
| 2017 | August    | 11504 |
| 2017 | September | 13293 |
| 2017 | October   | 11047 |
| 2017 | November  | 20875 |

### B. Analysis:

- 1. The trend of when people post a lot about it (high point and low point)
  - a) Compare by year (2016 vs 2017), month
    - (1) In 2017, there were generally more tweets about the pineapples on pizza debate, with the peak month being April of that year. For 2016, more people tweeted about pineapples belonging or not belonging on pizza in November.

The first surge in tweets occurred in February, and the number of tweets continued to increase until it reached it's peak in April. This is possibly because articles were being produced at the time talking about the pineapples on pizza debate. With more articles, comes more discussion,

- therefore the debate really sparked at this time perhaps because news outlets were giving it more attention.
- (2) Together, there were more people participating in this debate during the month of **November**.
- b) Compare the trends
  - (1) Check if there is a direct relationship
    - (a) In almost all the months, the tweets of those who liked pineapples on pizza, and the tweets of those who disliked pineapples on pizza had a direct relationship. As seen in the data, for every increase or decrease in tweets from the previous month for one side of the users, the other side will also increase or decrease depending on the trend.

This could have been made possible because of the conflicting sides always trying to gain the upperhand in the debate. If one party tweets about liking pineapples on pizza, this will in turn cause those who dislike pineapples on pizza to tweet as well.

# IV. Visualization of the output

















