# Data wrangling - Project Report

The relationship between twitter usage and the occurrence of world events/news.

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## 1 Research Question

Is there a correlation between the number of tweets posted and the occurrence of world events/news.

- 1. one.
- 2. two
- 3. three

## 2 Data Source.

A non-profit ordination runs a web-scrapper that collects data from all tweets created on the social media platform called twitter. This data is collected into packets which can be accessed by using the following website, from which the raw tweet data of a specific month can be downloaded via a torrent. Link: https://archive.org/search?query=collection

## 3 Data Wrangling methods

### 3.1 Twitter Data

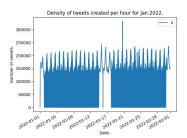
#### 3.1.1 Getting the Data.

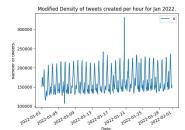
Using the link provided in the Data Source part, the raw twitter data for the given can be downloaded. This download file contains multiply tar files, which inside each then contains approximately 1440 gz compressed json files. The json files held more information than we required for this project, hence we need to extract only the two following fields 'created\_at' and 'lang', the first giving us the time the tweet was posted to the website and the second gives us information about the user language it was written in. Note that twitter supports xx languages, with more details can be found via Link xx. In order to process all this downloaded information the python script TwitterZip2DataFrame.py was created, which can be found via our GitHub page Link. Due to the sheer size of the files being process extra care is needed to process them. Hence the python script unzips one of the tar files into a temporary file, from which a multiprocess function is called to process the gz jsons files. This is to speed up the process. The json are convert in pandas data frames then the data frames are stripped for the desired information, which are then concatenated with the other json files and saved to the disk as csv files. Once all json files are processed in the .tar file, the temporary file is cleared to allow space for the next .tar file, until all are done. The result is that approximately 100Gb of raw twitter data can be filtered to under 7Gb of data, which is more user friendly. This processed data can be found on our GitHub page Link:dataset/jan22All.

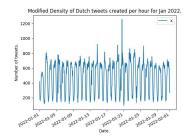
#### 3.1.2 Processing the data.

The first task here was to group all the tweet data into 1 hour bin of posting time, this was achieved by using pandas to convert the timeformat into timestamp data types for the 'created\_at' column and then using the pandas groupby function with specific options. The result of which is we have a density of tweets per the hour. This data was plot by using pandas line plot function. It become clear from the plot that the data contained outliers, namely data points that contained less than the mean number of tweets for that time. There are a number of reason for these outliers, such as the web-scrapper may not have been working correctly for that time

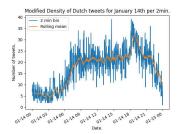
period or that the website twitter was down itself. The outlining data was removed which resulted in a clear plot of the tweets. Furthermore, using the pandas groupby function we can split the data by the twitter supported language, in our case we shall be looking at the Dutch tweets 'nl'. Again removing outliers we can generate a good plot for all Dutch tweets in January 2022.

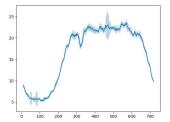






Since we are looking at discrete time events, we would like to have a better resolution of the density of tweets per a time period. We achieve this by selecting again only the Dutch tweets and then choosing a smaller time bin, in this case of '2min', while doing so we also choose a smaller range of data, as we have index the dataframe by a Timestamp data type, hence we can focus on 14-01-2022. Plotting this we see large fluctuations in the data points, making the plot hard to read. The readability can be improved by using the pandas rolling function in alignment of the mean function, the result is a smooth rolling mean plot of the data.





# 4 Conclusion