Naives Bayes Algorithm for Twitter Sentiment Analysis

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Abstract. On 2 March 2020, the Indonesian government, through President Joko 'Jokowi' Widodo, announced the first two cases of COVID-19 in Indonesia. This is the first case of COVID-19 officially confirmed in that country. Several cases have continued to increase since then. President Jokowi began issuing policies on the spread of this virus. This is different from other countries, such as Malaysia and Singapore, which responded from the previous month when the Indonesian government still stated that coronavirus does not exist in Indonesia. Our case study is to find a public opinion through social network analysis of Indonesian public policy during the beginning of the Indonesian COVID-19 pandemic in March 2020. This research implements text mining and document-based sentiments on Twitter data that is reprocessed through machine learning techniques using the Naive Bayes method. We found negative opinions in the period more dominant by 46%, while that was 35% positive sentiment and 20% neutral. This research shows that anticipation, sadness, and anger are very dominant in the emotional analysis.

1. Introduction

The World Health Organization (WHO) has declared the coronavirus (COVID-19) a global pandemic on March 11, 2020[1]. This has a significant impact on the global economy in the short term[2]. This pandemic began in the city of Wuhan, Hubei province, China, at the end of December 2019, and spread rapidly in various countries. In the second week of April 2020, the virus has caused 1.5 million people infected with 88 thousand deaths in 209 countries. Although the vaccine has not been found, in general, COVID-19 is a disease that can heal itself but can also be deadly with a mortality rate of 5%[3][4].

Since the outbreak of the pandemic, several countries, including Indonesia, have issued policy plans to deal with it. Following the announcement by President Joko 'Jokowi' Widodo of the first case of COVID-19 on 2 March 2020[5], the Indonesian government began issuing policies on the spread of this virus. This is considered too late because several other countries, such as Malaysia and Singapore, have responded since the previous month[6][7]. In February, the Indonesian government even declared that there was no coronavirus in Indonesia[8]. The Indonesian government also considers that the opening of positive patient data for COVID-19 is not transparent. A survey conducted by the Indonesian Institute of Sciences (LIPI) in collaboration with the National Disaster Management Agency (BNPB) to find out
public perceptions about information transparency concerning COVID-19 showed 97 percent of 15,101 respondents agreed that the recent travel history of positive patients must be opened transparently by the government[9].

This study analyzes the opinion of the Indonesian people through a social network analysis of the government policy towards COVID-19 during the early pandemic period of March 2020. We used sentiment analysis by crawling Twitter with the keywords Jokowi and Coronavirus. The keyword 'jokowi' is the representation of President Jokowi as the government and the coronavirus is the alias name of COVID-19. This process is done by drawing up Twitter opinions on an event or issue so that positive or negative opinions can be concluded[10]. The classification process of the Naïve Bayes method uses simple statistical probability values that assume strong independence from Bayesian rules[11].

2. Methodology

2.1 Social Network Analysis

Social Network Analysis (SNA) examines user interactions and relationships. The main concept is to study human relationships, such as friendship and family relationships, or the same interests. This relationship is then visualized in such a way that it is easily analyzed[12][13][14]. Social networks have an important role to play in the development of social computing. Some interesting topics, such as semantic analysis, content mining, or sentiment analysis, are widely used[12].

2.2 Sentiment Analysis

Sentiment analysis is the process of determining sentiments and grouping the polarity of texts in documents or sentences so that they can be categorized as positive, negative, or neutral sentiments [15][16]. Currently, sentiment analysis is widely used by researchers as one branch of research in computer science along with the explosion of information on the internet. Social media like Twitter are usually used in sentiment analysis to determine people’s perceptions[17]. Sentiment analysis can also be likened to opinion mining because it focuses on opinions that state positive or negative [18].

2.3 Text Mining

Text mining aims to produce information from the collection of documents. Text mining is a solution for the retrieval of information by processing, grouping and analyzing large amounts of unstructured data[19]. The process of retrieving information in the mining text may result in a sentiment analysis that identifies the emotional nature of a statement, whether positive or negative Text mining is currently widely used in a variety of research areas, such as software development, online media, marketing, academics, and politics. Like data mining, text pre-processing is necessary for the analysis of text mining through several stages, namely: case folding, tokenizing, filtering, stemming, tagging, and analysis[20].

2.4 Naïve Bayes

The Naïve Bayes method is a classification method for text mining sentiment analysis. This method can be classified and has the capability of data accuracy and computation[21]. Naïve Bayes has recently been widely used in classification techniques, particularly in social media networks such as Twitter, using several methods, including Unigram Naïve Bayes, Multinomial Naïve Bayes and Maximum Entropy Classification[22]. Calculation of the probability categories in Naïve Bayes uses the Bayes algorithm approach using the following equations:

\[ P(c|z) = \frac{\frac{P(z|c) \times P(c)}{P(z)}} \]

Equation (1) shows that c is the class label, z is the attribute applied, while P (c) and P (x | c) are the previous probabilities of the class. In the Naïve Bayes classification, the data set that is processed will
be annotated in three classes, positive, negative, and neutral[23]. The main feature of the Naïve Bayes classification is getting strong assumptions from each condition or event [24].

Figure 1. Research Method

Figure 1 above shows the research method used in this study. Naïve Bayes is used in opinion-mining as a method. Crawling data uses Twitter data by providing keywords within a given timeframe. The data collected are labeled to determine feelings and are transformed into structured data. The cleaning process is carried out at this stage to reduce noise. The last step is to remove the word 'stop' and delete meaningless words like me and, or[25]. The next process is the identification of words in the text by breaking up sentences into terms based on spaces and punctuation. This process is called tokenization, which ends with the process of stemming changing the affix words into basic words[19][26].

The process of creating a feature is carried out in the extraction feature to facilitate the classification of Naïve Bayes. The final stage is the stage at which the classification process begins based on the sentiments expressed in the document. This stage produces a model and is used to show the accuracy of the results of the classification. The level of accuracy at this stage is measured by the value of the performance of the classification by performing model tests and model assessments.

3. Result and Discussion

Research data was drawn from Twitter using the keyword 'Jokowi' at the beginning of the period of the COVID-19 pandemic in Indonesia. We used the public streaming of the Twitter API Drone Emprit to search[27] and collected data from 1 March to 31 March 2020. Crawler data successfully collected 306,284 tweets from the sample in Table 1.

| Table 1. Crawling Data |
|------------------------|
| @PermanaRikie          | Selamat pagi. Perkenalkan saya pasien 10 di Cirebon. Dinyatakan positif Covid-19 14 Maret lalu. Semoga surat ini bisa sampai ke Pak Presiden @jokowi & Pak Menkes @KemenkesRi https://t.co/fp8Gt2RmvA |
| @zenrs                 | Pak @jokowi, pejabatmu dusta soal pasien Cianjur? Anak istri pasien tsb kini positif. Brp lg yg terkena [or hrs diracking] krm pejabat bilang alm negatif? Stop menyimpan [apalagi jk berdusta] soal data/fakta. Kalian berbahaya betul main2 dg nasib kami https://t.co/cKPotOH6aZ |
| @sandalista1789        | Penulis twit yang tak direspons admin akun @jokowi ini telah meninggal dunia sebagai pasien PDP corona R.I.P. |
| @anandabadudu         | Pak Menkes Terawan dan pemerintahan @jokowi baiknya menunjuk sesorang/tim yang kompeten di dunia medis untuk menjadi juru bicara publik dlm menghadapi krisis #CoronaVirusIndonesia. Sudah statement2 denial, perbanyak info mitigasi |
| @jansen_jsp            | Sejak awal kami sudah menunda politik pak @jokowi. Tapi anda malah menghadapi wabah corona ini dgn pendekatan politik. Darurat sipil ini kebijakan "cuci tangan" dan power oriented. Anda ingin mengendalikan publik tapi tidak mau menangguh hidup mereka. |

Preprocessing steps were carried out to remove noise, clarify features, tokenize, and convert data[28]. Data cleaning needs to be done at this stage to determine the class and to delete unneeded attributes such as URL, mention, username, RT, hashtag, and punctuation, as shown in Table 2.
Table 2 Preprocessing Data

| Category   | Preprocessing Data                                                                 |
|------------|------------------------------------------------------------------------------------|
| Positive   | saya pasien 10 di Cirebon dinyatakan positif Covid-19 14 Maret lalu. Semoga surat ini bisa sampai ke Pak Presiden |
| Negative   | Pak jokowi pejabatmu dusta soal pasien Cianjur istri pasien tsb kini positif. Brrp lg yg terkena Stop menyimpan data/fakta. |
| Negative   | Penulis twit yang tak direspons admin akun jokowi ini telah meninggal dunia sebagai pasien PDP |
| Positive   | Pak Menkes Terawan dan pemerintahan jokowi baiknya menunjuk seseorang tim yang kompeten di dunia medis untuk menjadi juru |
| Negative   | Sejak awal kami sudah menunda politik pak jokowi. Tapi anda malah menghadapi wabah corona ini dgn pendekatan politik |

The next stage calculates the frequency of words that often appears in both categories to produce positive or negative and visualize them using word cloud as shown in Figure 2 and Figure 3.

Figure 2 Wordcloud Positive

Figure 3 Wordcloud Negative

The results of the word cloud were analyzed for emotions using Plutchik's Wheel of Emotions model. This model divides emotions into eight categories: joy, trust, fear, surprise, sadness, disgust, anger, and anticipation[29]. Figure 4 shows the dominant emotions in aggregate in March 2020 are anticipation, sadness, and anger. The most dominant emotion is anticipation, indicating the public hopes for readiness and anticipation in the future[30].

Figure 4 Emotion Analysis

Figure 5 Analysis Sentiment

Sentiment analysis generates 35% positive, 46% negative, and 20% neutral as shown in Figure 5. Negative sentiment is stronger because it produces more negative mentions than positive mentions.

4. Conclusion

This study applies the Naïve Bayes algorithm for sentiment classification. This sentiment model can detect the value of polarity and emotion. The research data was drawn from Twitter using the keyword 'Jokowi' at the beginning of the COVID-19 pandemic period in Indonesia in March 2020. This keyword considered a representation of President Jokowi as the government. Research in that period showed 35% positive sentiment, 46% negative sentiment, and 20% neutral. The negative opinion generated allegedly
due to public dissatisfaction with President Jokowi’s policies in the early days of the start of the COVID-19 pandemic in Indonesia.

In that period, we found dominant emotions in anticipation, sadness, and anger in the emotional analysis. The most dominant emotion is anticipation, which indicates public hopes for readiness and anticipation in the future. Further research is expected to use different algorithms to obtain more accurate results in the analysis of opinions.

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