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Sentiment Analysis Using Text Mining Techniques On Social Media Using the Support Vector Machine Method Case Study Seagames 2023 Football Final

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Abstract

This thesis aims to analyze sentiment on text data from social media related to the 2023 SEA Games, especially in the final match of the soccer sport. The method used is the Text Mining Technique with the SVM (Support Vector Machine) algorithm to classify user sentiment as positive or negative regarding the match. Text data is retrieved from various social media platforms during and after the game. The results of the sentiment analysis are expected to provide insight into the public's view of the sporting event. This research can contribute to the understanding of public sentiment towards the 2023 SEA Games final football match through the analysis of text data from social media.

Keywords: Sentiment Analysis, Text Mining techniques, SVM, SEA Games, Football.

1. Introduction

Sentiment analysis on social media is becoming increasingly important for understanding public opinion and consumer behavior in today's digital era. Text mining techniques can be used to process and classify large text data from social media into positive and negative sentiment categories. The Support Vector Machine (SVM) method is one of the popular machine learning algorithms for sentiment analysis. The Support Vector Machine (SVM) method can take into account the subjectivity factor in text data and can assist in identifying sentiment patterns and trends. By using text mining techniques and SVM methods, companies and organisms can process and analyze text data from social media efficiently and accurately to assist in decision making [1].

Current technological developments make it easier for many people to express opinions on online social media as a form of response and response to something. Social Media has also become a very important platform for spreading information and views.

Text mining techniques are used in sentiment analysis to process text data from social media, such as Tiktok, tweets, Facebook statuses, or Instagram comments. This technique involves data collection, data cleaning, data processing, and data classification [2].

Based on the background of the problems above, the formulation of the problem in this study is as follows:

- 1. How to do sentiment analysis using text mining techniques on text data from social media?
- 2. How to get large enough text data to train and test the Support Vector Machine model in recognizing and classifying sentiments from text data on social media?
- 3. How to make a sentiment analysis system from text data on social media using the Support Vector Machine method?

In solving a problem it is necessary to make a problem boundary so as not to deviate from the existing problem, therefore the problem boundaries in this writing are:

- 1. The SVM(Support Vector Machine) method is a machine learning algorithm that is widely used for data classification in sentiment analysis.
- 2. The program used is Python software.
- 3. The variables used are: positive and negative language.
- 4. The media used is Twitter.
- 5. This study uses data taken from the Twitter platform related to seagames 2023, especially focusing on the finals of the Football competition.
- 6. Text sentiment is analyzed into several categories such as positive and negative. Aims to understand the reactions and opinions of social media users.
- 7. The data taken is in the form of comments related to the case study. This data will be in the form of text in Indonesian.
- 8. Using a database with excel format.

The objectives of this study are as follows:

- 1. Finding and collecting large enough text data sources to train and test Support Vector Machine models.
- 2. Using text mining techniques to process and analyze text data in different languages or non-standard languages.
- 3. To determine the test results of the Support Vector Machine method with positive and negative resulting data.
- 4. to analyze the sentiments of social media users regarding the SeaGames 2023 final in the football sport using text mining techniques and the SVM method.
- 5. This study aims to understand the views and responses of people on social media and evaluate the effectiveness of the SVM model in classifying texts into positive and negative sentiment categories.
- 6. The results of the research are expected to provide insight into the public's response to the sporting event and the potential for improvement in interpreting sentiment on social media.

The benefits of this research are as follows:

- 1. Helping decision making in the results of sentiment analysis of various things, such as understanding a person's response, and knowing the views of the public on a particular topic or event.
- 2. Able to group sentiment data using the Support Vector Machine method.
- 3. Using the Support Vector Machine method can make it easier to group sentiment data more effectively and efficiently.
- 4. The benefit of this research is to understand public opinion and sentiment towards the 2023 SEA Games final in sports football through text analysis on social media.

2. Research Methodology

2.1. Data Mining

Data mining is a process of finding meaningful relationships, patterns, and trends by examining large sets of data stored in storage using pattern recognition techniques such as statistical and mathematical techniques. Extracting data information is carried out based on the classification of the data mining method to be used. There are several methods and data mining algorithms that are used to extract data information, including: association methods, clustering methods, classification methods, prediction methods, and estimation methods [3].

2.2. Sentiment Analysis

According to[2]Sentiment analysis is also known as opinion extraction, opinion mining, sentiment mining, and subjectivity analysis. Sentiment analysis is a computational study of people's opinions, appraisals and emotions through entities, events and attributes.

2.3. Text Mining

Text mining is a data mining technique in the form of text. Text taken from social media which of course does not use standard language really needs to be processed using text mining. Typical text mining processes include text categorization, text clustering, concept/entity extraction, granular taxonomy production, sentiment analysis, document inference, and entity relationship modeling [4].

2.4. Twitter

Twitter is a website, owned and operated by Twitter Inc., which offers a social network in the form of a microblog that enables its users to send and read tweets.

2.5. Pythons

Python is an interactive object-oriented programming language that provides high-level data structures. Python is a multipurpose interpretive programming language with a design that focuses on code readability so that the syntax used is easier to understand. Python is said to be a language that combines capabilities, capabilities, with a very clear code syntax [5].

2.6. Steps to the Support Vector machine(SVM) method

Create Word Weighting

$$W_{ij\,ij} = tf . idf$$
$$w_{permit} = tf . \log ()$$
$$df$$

Where:

- Wij : term weight (t_j) to document (d_1) ,
- tfij: The number of occurrences of the term (tj) in (d1),
- *n* : The number of all documents in the database,
- df : The number of documents that contain the term

3. Flow Chart

The flowchart system to be designed is as follows:



Figure 1: Flowchart flow

Case in point: @FaktaSepakbola @Changsuek_TH @ThaiFootball @TL_Central Asean football level role model Preprocessing steps [6], [7]:

- 1. Case folding: The process of this step is to reduce letters or words (lowercase)
- 2. Tokenizing: To remove certain characters such as punctuation marks. While the space character is used as a delimiter to break sentences into groups of words.
- 3. Filtering: Used to select stopword words. If there is a stopword, then the word will be automatically deleted.
- 4. Stemming: the process of filtering words that are changed into basic word forms by using certain rules.
- 5. If the sentence contains joy, appreciation, pleasure, it is a positive sentence
- 6. If the sentence contains disappointment, frustration, dissatisfaction including negative sentences

| | | Table 1: Example of Trai | ning Data | | | | | | |
|----|----------------------|--|---|------------------|--------------|--|--|--|--|
| No | Keyword | Tweets | date | Source | Sentiment | | | | |
| 1 | @MhmmdRauf13 | @FaktaSepakbola @Changsuek_TH @ThaiFootball @TL_Cent A role model for ASEAN football | ral 18 june 2023 | ticktock | positive | | | | |
| No | | Table 2: Example of Test Tweets | sting Data | | | | | | |
| 1 | 1 weeds | | | | | | | | |
| 1 | @FaktaSepakbola @ | Changsuek_TH @ThaiFootball @ | TL_Central A ro | le model for ASI | EAN football | | | | |
| | | Table 3: Case Folding | Process | | | | | | |
| No | Т | weets | | Casefolding | | | | | |
| | @FaktaSepakbola @Cha | angsuek_TH @ThaiFootball | @FactsFootball @changsuek_th @thaifootball | | | | | | |
| 1 | @TL_Central A role r | nodel for ASEAN football | @tl_central a role model for ASEAN football | | | | | | |
| | | Table 4: Tokenizing F | Process | | | | | | |
| | No | tokenizing | | | | | | | |
| | | | @ | factsfootball | | | | | |
| | | | @changsuek_th | | | | | | |
| | @EastsEast | all Ochomografi th Otherfoothal | @thaifootball | | | | | | |
| | 1 @tl_central | a role model for ASEAN football | @tlcentral | | | | | | |
| | | | ro | role model | | | | | |
| | | | levels | | | | | | |
| | | | football | | | | | | |
| | | | asean | | | | | | |

| No | Tweets | filtering |
|-----------|---|---|
| | role model | |
| 1 | @tl_control a role model for ASEAN football | levels |
| 1 | WIL_central a role model for ASEAN football | football |
| | | asean |
| | | |
| | Table 6: Stemming Process | |
| No | Table 6: Stemming Process Tweets | filtering |
| No | Table 6: Stemming Process Tweets | filtering role model |
| No 1 | Table 6: Stemming Process Tweets @FactsFootball @changsuek_th @thaifootball @tl_central a role model for ASEAN football | filtering role model levels |
| No | Table 6: Stemming Process Tweets @FactsFootball @changsuek_th @thaifootball @tl_central a role model for ASEAN football | filtering role model levels football |

Table 5: Filtering Process

3.1. Weighting

In this process, count the number of terms or words that appear in the tweet (tf), count the number of tweets that contain the term (df), calculate the inverse document frequency (idf), and multiply tf by idf as the weight of the term in each tweet [8], [9].

| FormulaTf.idf | Wij | : term weight (tj) to document $(d1)$, |
|---|--------------|--|
| $W_{ijij} = tf . idf$ | <i>tf</i> ij | : The number of occurrences of the term (tj) in $(d1)$, |
| $m = t \int \log(x)$ | n | : The number of all documents in the database, |
| $w_{\text{permit}} = i f \cdot \log(f)$ | df | : The number of documents that contain the term |

Example of sentences :Asean football level role model

| Table 7: Word Weighting Results | | | | | | | | | | | |
|---------------------------------|----|----|----|----|----|------|-------|--------|----|------|------|
| Torm | tf | | | | đf | n/df | idf | tf-idf | | | |
| 1011 | t1 | t2 | t3 | t4 | ui | n/ui | iui - | t1 | t2 | t3 | t4 |
| role model | 1 | 0 | 0 | 0 | 1 | 4 | 0.60 | 0.60 | 0 | 0 | 0 |
| levels | 1 | 0 | 0 | 0 | 1 | 4 | 0.60 | 0.60 | 0 | 0 | 0 |
| football | 1 | 0 | 1 | 1 | 3 | 1,3 | 0.11 | 0.60 | 0 | 0.60 | 0.60 |
| asean | 1 | 0 | 0 | 0 | 1 | 4 | 0.60 | 0.60 | 0 | 0 | 0 |

4. Program Discussion

4.1. Display Sentiment Analysis

As for the view to analyze using Anaconda Python Jupyter Notebook



Figure 2: Initial View of the Jupyter Project

4.2. Data Used in Sentiment Analysis

Before starting to enter the syntax, first upload the data that will be used in sentiment analysis.

| 🔁 Jupyter | Quit | Logout |
|--|---------------------------|-----------|
| Files Running Clusters | | |
| Select items to perform actions on them. | Upload | New - |
| 🗆 0 🕞 🖿 / crawling | Name 🗣 🛛 Last Modified | File size |
| | beberapa detik yang lalu | |
| 🖸 🤗 analisis sentimen.ipynb | Running 2 menit yang lalu | 72 B |
| E Sentiment Analysis with Python-Skripsi.jpynb | 6 hari yang lalu | 861 kB |
| Sentiment Analysis with Python.ipynb | 11 hari yang lalu | 67.2 kB |
| G df_test10.csv | 7 hari yang lalu | 930 B |
| C df_train90.csv | 7 hari yang lalu | 10.6 kB |
| Sentimen.csv | 11 hari yang lalu | 16.5 kB |
| sentimen1.csv | 11 hari yang lalu | 9.88 kB |
| sentimen2.csv | 11 hari yang lalu | 11.1 kB |
| sentimen_cleaned.csv | 7 hari yang lalu | 18.9 kB |
| D sentimen_preprocessed.csv | 7 hari yang lalu | 45.9 kB |
| D sentimen_stemmed.csv | 7 hari yang lalu | 19 kB |
| Sentiment Analysis with Python.py | 12 hari yang lalu | 2.95 kB |
| test prediction 9010.csv | 7 hari yang lalu | 930 B |

Figure 3: Display of Data Filled in the Sentiment Analysis folder

4.3. Calling Libraries in Python Jupyter Notebook

To load the library in Jupyter Notebook's Python environment, you can use the import command. Here are some examples of calling the library used.



Figure 4: Library view on jupyter

4.4. Preprocessing Stage

Following are the 4 stages of preprocessing syntax and the results of calling data that has gone through 4 stages of processing.



Figure 5: Display of the Preprocessing Stage

4.5. Converting Text Sets to Number Variables

Converting a set of text to a numerical representation is usually necessary in text analysis, especially when you want to apply machine learning algorithms that require numeric input.

| The state | Marrie Incode | Coll Marrie Midania Hala | | |
|-----------|----------------------------------|--|-----------------------------|----|
| Pile Edit | view inser | | Trasted Pya | |
| 🖺 🕂 🛰 (| 2 6 1 | Run C H Markdown V C | | |
| | | | | |
| In [64]: | # TF-IDF | | | |
| | from sklearn. | .feature_extraction.text import TfidfVectorizer | | |
| | tfidf vect 90 | 010 - TfidfVectorizer(max features - 5000) | | |
| | tfidf_vect_90 | 010.fit(bipolarsentiment['Tweet']) | | |
| | train_X_tfidf | <pre>f_9010 = tfidf_vect_9010.transform(df_train90['Tweet']) 9010 = tfidf_vect_9010.transform(df_tet10['Tweet'])</pre> | | |
| | cese_n_eran_ | | | |
| | TfidfVectorizer y | yang digunakan untuk mengubah kumpulan teks menjadi representasi vektor fitur berdasarkan skema pembobota | n TF-IDF (Term Frequency | - |
| | Inverse Docume | ent Frequency). Penggunaan max_features=5000 menentukan bahwa hanya akan diambil 5000 fitur (kata-kata) d | engan nilai TF-IDF tertingg | gi |
| | dari seluruh kun | mpulan teks. | | |
| In [65]: | tfidf_vect_90 | 210 | | |
| Out[65]: | TfidfVectoriz | zer(max features=5000) | | |
| | | | | |
| | Hasil dari print(t | train_X_tfidf) akan mencetak nilai dari variabel train_X_tfidf ke layar atau konsol. Jika train_X_tfidf adalah array № | umPy atau DataFrame | |
| | Pandas, maka h | hasilnya akan berupa representasi data dalam bentuk tabel atau array sesuai dengan format yang ditentukan oleh | NumPy atau Pandas. | |
| In [66]: | print(train) | x tfidf 9010) | | |
| | (0 547) | a 2445869462169782 | | |
| | (0, 539) | 0.2811946967422275 | | |
| | (0, 477) | 0.2811946967422275 | | |
| | (0, 468) | 0.281174070/4222/5 0.16514802048731533 | | |
| | (0, 420) | 0.2811946967422275 | | |
| | (0, 413) | 0.2811946967422275 | | |
| | (0, 361) | 0.2811946967422275 | | |
| | (0, 328) | 0.2811946967422275 | | |
| | (0, 235) | 0.281194696/4222/5 | | |
| | | 0.2811946967422275 | | |
| | (0, 229) | | | |
| | (0, 229) (0, 209) | 0.19619204280184144 | | |
| | (0, 229) (0, 209) (0, 202) | 0.19619204280184144 0.15454957394236712 | | |
| | (0, 229) (0, 209) (0, 202) | 0.19619204200184144 0.15454957394236712 | | |

4.6. Splitting Document Sentences into Features/Words

Splitting document sentences into features or words is an important step in text preprocessing. You can use a variety of techniques to do this, including tokenization, vocabulary creation, and vector representation.



5. Conclusion

While the sentiment analysis results provide a snapshot of how social media users respond, we recognize that data from the Twitter social media platform may have limitations. Therefore, these results need to be interpreted with caution and should not be taken as representing absolute truth. Therefore, the core conclusions of the research are as follows:

- 1. This study succeeded in conducting a sentiment analysis of the responses of social media users regarding the SEA Games 2023 final in the sport of football.
- 2. The application of text mining techniques allows the identification and extraction of sentiments from social media texts.
- 3. The SVM method is used as a classification algorithm to predict sentiment from text data.
- 4. SVM models are evaluated using accuracy, precision, recall, and F1-score metrics to measure model performance.
- 5. In addition to sentiment analysis, this research also identifies related entities that are often discussed by social media users.
- 6. The results of the analysis provide valuable insights for stakeholders in understanding social media responses to major sporting events.

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