Literature Study: Highway Traffic Management with Sentiment Analysis and Data Mining

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ABSTRACT

In today's era, technology is growing rapidly, many of the latest technologies are in great demand by the Indonesian people, one of which is social media. Various social media such as Facebook, Twitter, Instagram, have become very popular applications for various ages, including teenagers, adults, and the elderly. Social media has a positive impact that can help people convey the latest information through posts on their respective accounts. Social media can disseminate information in a short time, this is why social media is an interesting application to research. The problem of road traffic congestion is strongly influenced by the number of vehicles that pass every day. A large number of private vehicles and public vehicles that pass greatly confuses the atmosphere of highway traffic. Congestion often occurs during working hours. Road congestion also often occurs when an unwanted incident occurs. Sentiment analysis algorithms and data mining algorithms can be combined to find information on traffic jams through social media such as Facebook, Twitter, Instagram, and other social media. The results show that sentiment analysis methods and data mining algorithms can be used to find information about current traffic jams through social media. The conclusion from this literature study can be seen that the K-Nearest Neighbor data mining algorithm is the best choice to overcome road traffic congestion, which will then be further developed in the form of highway traffic management modeling.

INTRODUCTION

Medan city is one of the largest provincial capitals in Indonesia. Medan City is a metropolitan city that has a fairly large population and reaches 2 million inhabitants per the year 2015 [1]. The number of private vehicles and public transportation often triggers traffic jams on the highway during working hours. According to Kartika Ratnasari, Medan is the 6th most congested city in Indonesia. The length of traffic jams in Medan in a year reach 42 hours with a percentage of 20%. The percentage of time drivers are stuck in traffic jams reaches 25% during working hours and 23% outside working hours [2].

Natural disasters such as earthquakes, landslides, strong winds, tsunamis, and so on can also disrupt the flow of highway traffic. In addition, events such as demonstrations, student brawls, going home on National holidays, road repairs, Covid-19 prevention checks, zebra operations, traffic light damage, traffic accidents, and so on can also cause long traffic jams and take up time for drivers. According to Rozie Winata, the number of traffic accidents in 2019 has increased in 2018. Throughout 2019 there were 6,100 cases, while previously in 2018 there were only 5,990 cases. In addition, Kapolda also said that material losses due to traffic accidents in 2019 decreased, which was only Rp. 12,625,000,000, while in 2018 it previously reached Rp. 13,696,830,000 [3].

Minister of Transportation Budi Karya Sumadi said that traffic jams in urban areas had a major impact on the Indonesian economy. One of them inhibits economic growth[4]. In 2016, SITRAMP (Study on Integrated Transportation Master Plan) also estimated that losses due to congestion could reach Rp. 12.8 trillion per year[5]. Until now, the Medan City Government continues to strive to overcome road traffic congestion, one of which is by building toll roads and flyovers[6]. In this study, the researcher plans to conduct a literature review of the sentiment analysis algorithm and data mining algorithm to find alternative solutions for handling traffic jams in terms of the field of informatics engineering which will also involve social media [7][8]. Several theories will be analyzed and tested to help the government overcome road traffic congestion.

LITERATURE REVIEW
theory of Traffic Demand Management [10]. Laif Satoïnong et al conducted a study that focused on performance analysis and traffic management at the ITS Roundabout and the Mulyosari Roundabout in Surabaya using the MKJI theory [11]. Popy Purnamasari Wahid Suyitno et al conducted a study by detecting traffic congestion in Jakarta with the Data Mining C4.5 algorithm and offering solutions to overcome traffic congestion [12]. Putu Amelia Vennanda Widyaswari et al conducted a study on the classification of traffic congestion in the city of Malang using the Neighbor Weighted K-Nearest Neighbor method [13]. Siti Nurhayuni researched the form of sentiment analysis on the KRL Access Online Transportation Application using the Naïve Bayes method [14]. Myrna Ermawati and Joko Lianto Buliali conducted a study using a text-based approach to find out similar traffic incidents through social media Twitter[15].

**METHOD**

This literature review study was conducted to determine whether sentiment analysis and data mining algorithms can prevent road traffic congestion through road traffic management. The literature review was conducted on three studies [13] [15] [16] by analyzing the algorithms and cases to be solved as can be seen in Table 1.

| Author                        | Title                                                                 | Algorithm                     | Case               |
|-------------------------------|-----------------------------------------------------------------------|-------------------------------|--------------------|
| Putu Amelia Vennanda Widyaswari | Malang City Traffic Congestion Classification Using the Neighbor Weighted K-Nearest Neighbor (NW-KNN) Method | Traffic congestion           |                    |
| Indriati                      | Through Twitter Media Using the Neighbor Weighted K-Nearest Neighbor (NW-KNN) Method                      | Traffic congestion           |                    |
| Rizal Setya Perdana           |                                                                                       | Traffic congestion           |                    |
| Myrna Ermawati, Joko Lianto Buliali | Text Based Approach For Similar Traffic Incident Detection from Twitter                | Traffic Incidents            |                    |
| Amani AlFarasani, Tahani AlHarthi, Sarah AliHumoud | ATAM: Arabic Traffic Analysis Model for Twitter | Sentiment Road Incident Prediction |                    |

**RESULT**

The results of the literature review carried out by researchers can be in the form of comparative analysis of research results, research gaps, and the advantages and disadvantages of the three studies [13][15][16] that have been described in Table 1. The results of the literature review can be seen in Table 2.

| Algorithm                     | Object of Research          | Research Result                                                                 |
|-------------------------------|----------------------------|---------------------------------------------------------------------------------|
| Neighbor                     | Traffic congestion         | The average value of precision is 0.7336507, recall is 0.2210526, and accuracy is 0.665 |
| Weighted K-Nearest Neighbor (NW-KNN) |                           |                                                                                  |
Approach: Incident
70% was achieved by retrieval system that tested using jaccard coefficient

Sentiment Analysis, Support Vector Machines: Road Incident Prediction
The overall accuracy was 82% for all four classes

| Algorithm                        | Object of Research                  | Research Gap                                                                 |
|----------------------------------|-------------------------------------|------------------------------------------------------------------------------|
| Neighbor Weighted K-Nearest Neighbor (NW-KNN) | Traffic congestion                  | Optimizing the performance of the Neighbor Weighted K-Nearest Neighbor (NW-KNN) algorithm to overcome traffic jams |
| Text Based Approach               | Traffic Incident                    | Optimizing traffic incident detection with semantic analysis                  |
| Sentiment Analysis, Support Vector Machines | Road Incident Prediction             | Build a web service for live streaming and classifications.                 |

After determining the results from the three articles [13][15][16], their strengths and weaknesses were reviewed, as shown in Table 4.

| Title                                                            | Strength                                                                 | Weakness                                                                 |
|-----------------------------------------------------------------|--------------------------------------------------------------------------|--------------------------------------------------------------------------|
| Malang City Traffic Congestion Classification Through Twitter Media Using the Neighbor Weighted K-Nearest Neighbor (NW-KNN) Method | Able to detect congestion by providing information in the form of time, place, road congestion conditions, and weather conditions | In some trials there is still low accuracy, this is due to a large number of test data with incorrect conditions, such as test data that should be in the crash category are included in the non-jammed category. |
| Text Based Approach For Similar Traffic Incident Detection from Twitter | Able to filter sentences without symbols/punctuation from Twitter, can extract hashtags, conditions, locations, and objects | Unable to distinguish the presence of different words with the same meaning in a traffic incident, for example, the terms 'consecutive collision' and 'accident' streak'. |
A TAM: Arabic Traffic Analysis Model for Twitter

Able to detect traffic incidents in the form of safe conditions, need attention, danger, and neutral

Still need Arabic speakers to measure the accuracy of word classification

| DISCUSSION |
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| After conducting a literature study on sentiment analysis and data mining algorithms in the three studies listed in Table 2, Table 3, and Table 4, the researcher has a further research plan to build a highway traffic management model, as detailed in Figure 1 below: |

![Figure 1. Planning for Further Research](#)

| CONCLUSION |
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| The literature study is an initial study to explore how sentiment analysis and data mining algorithms work, especially in traffic management, which will be useful for overcoming traffic congestion. After the researchers analyzed the three studies in Table 2, Table 3, and Table 4, it can be seen that the sentiment analysis method and data mining algorithm can be used to detect traffic congestion through social media. From the literature study, it can also be seen that the right data mining algorithm to use is K-Nearest Neighbor. For further research, researchers will build a traffic management model to avoid congestion, this is based on the results of literature studies on sentiment analysis methods and data mining algorithms. |

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