

FAKE NEWS DETECTION ON SOCIAL MEDIA PLATFORMS USING  
MACHINE LEARNING ALGORITHMS

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## **DEDICATION**

This project report is dedicated to my father, who taught me that the best kind of knowledge to have is that which is learned for its own sake. It is also dedicated to my mother, who taught me that even the largest task can be accomplished if it is done one step at a time.

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## ABSTRACT

The rapid growth of advanced technology and social media platforms has changed the way of retrieving information or news for the whole world. However, it is undeniable that there is some fake news which is purposely created to disseminate false information to the public. The rapid spreading of fake news brings a significant negative impact on individuals, culture and country. Although there are some existing fake news detection methods like manual fact-checking websites and tools, they are time-consuming for fake news detection and cannot provide real time detection. Hence, an automated machine learning based approach is required for fake news detection since fake news can be disseminated rapidly through online social media platforms. This project aims to propose and construct an effective hybrid machine learning based algorithm for automated fake news detection. This algorithm will assist people in differentiating real and fake news, reduce the threats to national security by preventing the wide spread of fake news and maintain the news ecosystem's genuineness equilibrium. The project flow started with the dataset collection which includes both real and fake news. This dataset is publicly available and it is obtained from Kaggle website. Next, some text pre-processing techniques, i.e., lower case conversion, punctuation and stopwords removal and lemmatization will be applied to the raw data. Furthermore, Term Frequency – Inverse Document Frequency (TF-IDF) is the technique applied for text vectorization in this project. Two machine learning algorithms, Random Forest (RF) and Support Vector Machine (SVM) are implemented to classify the real and fake news. In addition, two hybrid ensemble models, namely voting classifier and stacking classifier are constructed by combining three individual base classifiers, i.e. RF, SVM and Logistic Regression (LR). Hybrid ensemble model is an approach that combines several models to improve the prediction accuracy. These hybrid classifiers can detect fake news in real time and effectively prevent the fake news from being disseminated widely through social media platforms. In this project, stacking classifier performs better than voting classifier, achieving an accuracy of 91.06% as compare to the classification accuracy of 90.41% obtained by voting classifier.

## ABSTRAK

Pertumbuhan teknologi canggih dan platform media sosial yang pesat telah mengubah cara untuk mendapatkan maklumat atau berita bagi seluruh dunia. Namun begitu, memang tidak dapat dinafikan bahawa terdapat beberapa berita palsu yang sengaja dicipta untuk menyebarkan maklumat palsu kepada awam. Penyebaran berita palsu yang pantas membawa kesan negatif yang ketara kepada individu, budaya dan negara. Walaupun terdapat beberapa kaedah pengesanan berita palsu yang sedia ada seperti laman web dan alatan semakan fakta manual, ia memakan masa untuk pengesanan berita palsu dan tidak dapat memberikan pengesanan masa nyata. Oleh itu, pendekatan berasaskan pembelajaran mesin automatik diperlukan untuk pengesanan berita palsu kerana berita palsu boleh disebarkan dengan cepat melalui platform media sosial dalam talian. Projek ini bertujuan untuk mencadangkan dan membina sebuah algoritma berasaskan pembelajaran mesin hibrid yang berkesan untuk pengesanan berita palsu automatik. Algoritma ini akan membantu orang ramai untuk membezakan berita sebenar dan palsu, mengurangkan ancaman kepada keselamatan negara dengan menghalang penyebaran berita palsu secara meluas dan mengekalkan keseimbangan ketulenan ekosistem berita. Aliran projek bermula dengan pengumpulan set data yang merangkumi kedua-dua berita sebenar dan palsu. Seterusnya, beberapa teknik pra-pemrosesan teks, iaitu, penukaran huruf kecil, penyingkiran tanda baca dan kata henti serta lemmatisasi akan digunakan pada data mentah. Tambahan pula, *Term Frequency – Inverse Document Frequency* (TF-IDF) ialah teknik yang digunakan untuk vektorisasi teks dalam projek ini. Dua algoritma pembelajaran mesin, *Random Forest* (RF) dan *Support Vector Machine* (SVM) digunakan untuk mengklasifikasikan berita sebenar dan palsu. Selain itu, dua model ensembel hibrid, iaitu pengelas undian dan pengelas tindakan dibina dengan menggabungkan tiga pengelas asas individu, iaitu RF, SVM dan *Logistic Regression* (LR). Model ensemble hibrid ialah pendekatan yang menggabungkan beberapa model untuk meningkatkan ketepatan ramalan. Pengelas hibrid ini boleh mengesan berita palsu dalam masa nyata dan berkesan menghalang berita palsu daripada disebarkan secara meluas melalui platform media sosial. Dalam projek ini, pengelas tindakan berprestasi lebih baik daripada pengelas undian, mencapai ketepatan 91.06% berbanding ketepatan pengelasan 90.41% yang diperolehi oleh pengelas undian.

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## LIST OF ABBREVIATIONS

AdaBoost	-	Adaptive Boosting
BiLSTM	-	Bidirectional Long Short-Term Memory
BoW	-	Bag-of-Words
CNN	-	Convolutional Neural Network
DNN	-	Dense Neural Network
DT	-	Decision Tree
FN	-	False Negatives
FP	-	False Positives
GB	-	Gradient Boosting
GRU	-	Gated Recurrent Unit
IDE	-	Integrated Development Environment
IDF	-	Inverse Document Frequency
KNN	-	K-Nearest Neighbor
LR	-	Logistic Regression
LSTM	-	Long Short-Term Memory
NB	-	Naïve Bayes
NLP	-	Natural Language Processing
OOB	-	Out-of-bag
Regex	-	Regular expression
RF	-	Random Forest
RNN	-	Recurrent Neural Network
SVM	-	Support Vector Machine
TF	-	Term Frequency
TF-IDF	-	Term Frequency – Inverse Document Frequency
TN	-	True Negatives
TP	-	True Positives
XGBoost	-	Distributed Gradient Boosting

## LIST OF SYMBOLS

$C$	-	Regularization parameter
$n$	-	Number of words in all documents
$n\_estimators$	-	Number of trees in Random Forest
$v_k$	-	TF-IDF value for $k^{th}$ word

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# CHAPTER 1

## INTRODUCTION

### 1.1 Background Study

With the rapid growth and improvement of the internet, people prefer to retrieve or share information through online social media platforms such as Facebook, WhatsApp, Twitter, YouTube and many more. Nowadays, people can easily access any interested information by searching keywords on search engines. As a result, all the information or news articles can be retrieved through online platforms at the touch of a button. Unfortunately, there are some news which are purposely created to disseminate false information to the public and these are known as fake news.

For instance, there is some fake news related to COVID-19 pandemic spread through online social media platforms recently. Social media platforms serve as the main source of information and news especially during COVID-19 outbreak period. Some people will take this opportunity to create and publish fake news on social media platforms. This fake news is intended to mislead people to have a distorted opinion. Another reason for creating fake news is for the purpose of gaining some economic or political benefits (Qawasmeh et al., 2019). Fake news will bring significant negative impacts to individuals and countries if they are widely disseminated. Due to this reason, fake news has become an increasingly matter of contention in both industry and academia.

### 1.2 Motivation

There are some existing fake news detection methods used to detect and prevent the spreading of fake news on social media platforms. However, these existing fake news detection methods have their own limitations while detecting the fake news.

For instance, manual fact-checking such as Politifact and Snopes requires a significant amount of time in order to detect fake news (Liu, 2019). In other words, these manual fact-checking approaches cannot detect fake news in real time. As a result, fake news has already been disseminated through social media platforms before it is being detected.

With the advancement of technology, a lot of fake news is created and propagated widely on social media platforms for someone's benefit. However, manual approaches are not able to handle well with the growing number of freshly created fake news. Other than that, some existing fake news detection methods even require more external information such as the source and the comments of the news to determine the veracity of that news. Unfortunately, this information may sometimes not be available in some of the news articles which limits the efficiency and effectiveness of detecting fake news (Liu, 2019).

Due to the limitations of existing methods, an automated fake news detection algorithm is necessary to detect and prevent the spreading of fake news since fake news can be disseminated rapidly through online platforms. Machine learning and deep learning algorithms have been proved to be promising and able to effectively detect fake news through many studies. Thus, this project proposed to implement machine learning approaches in detecting fake news on social media platforms.

### **1.3 Problem Statement**

A study proved that it is a hard decision for humans to differentiate between a fact and lie statement when they are overloaded with misleading information (Zhou et al., 2020). In other words, people tend to believe every published news article regardless of the truthfulness of the news article. This is because people always tend to process the information and make a quick decision through mental shortcuts. As a result, they share the misleading information without even verifying the information's veracity.



The rapid spreading of fake news through social media platforms brings a significant negative impact on individuals, culture and country. For instance, fake news will affect the news ecosystem's genuineness equilibrium. Fake news is also intended to persuade someone to have misleading views towards certain truths or facts (Abbad et al., 2021). Furthermore, the spread of fake news has a chance to instigate social unrest which can bring threats to national security (Liu, 2019).

#### **1.4 Aim and Objectives**

The main aim of this project is to propose and construct an effective hybrid machine learning based algorithm for automated fake news detection.

The objectives of this project are as follows:

- (a) To assist people in differentiating real and fake news before they directly share without verifying the veracity of news.
- (b) To reduce the threats to national security by preventing the wide spreading of fake news.
- (c) To maintain the news ecosystem's genuineness equilibrium and avoid people having misleading views due to fake news.

#### **1.5 Research Scope**

The objectives of this project are as follows:

- (a) This project utilized publicly available dataset from Kaggle website.
- (b) TF-IDF technique is used for text vectorization.

- (c) Two machine learning algorithms, i.e., RF and SVM are implemented for classifying real and fake news.
- (d) Two hybrid classifiers are constructed to further improve the classification accuracy.
- (e) Spyder, an Integrated Development Environment (IDE) is used for implementing pre-processing, feature extraction and classification steps.

## **1.6 Chapter Summary**

This project is organized as follows. Chapter 1 is introducing the background study of research especially the fake news on social media platforms. Besides, the motivation, problem statements, aim and objectives and research scope of this project are also clearly mentioned in this chapter.

Next, Chapter 2 provides a detailed literature review after reading several articles and journals. This chapter introduces some existing fake news detection methods, including manual and computational approaches. Besides, the definition and significance of dataset, text pre-processing techniques, text features and machine learning algorithms are mentioned in this chapter. A summary of related works is provided at the end of this chapter.

Chapter 3 explains the project's methodology. This chapter started with a brief introduction which included all the steps involved for fake news detection. Furthermore, all the steps are further explained in detail in each subsection. Two Gantt charts showing the project timelines are also provided at the end of this chapter.

In Chapter 4, all the results and discussions are clearly mentioned. All the results including classification accuracy, precision, recall and F1-score are tabulated for all the classification algorithms implemented in this project. Chapter 5 summarized the whole project. Besides, there are also some recommendations provided for future works.

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