

Machine Learning Approach For Fake News Detection

Devansh Singh

Dept. of Artificial Intelligence and Machine Learning
Symbiosis Institute of Technology
Pune, India
singh.devansh.btech2021@sitpune.edu.in

Kedar Hardikar

Dept. of Artificial Intelligence and Machine Learning
Symbiosis Institute of Technology
Pune, India
kedar.hardikar.btech2021@sitpune.edu.in

Chaitanya Jagtap

Dept. of Artificial Intelligence and Machine Learning
Symbiosis Institute of Technology
Pune, India
chaitanya.jagtap.btech2021@sitpune.edu.in

Dr. Preksha Pareek

Dept. of Artificial Intelligence and Machine Learning
Symbiosis Institute of Technology
Pune, India
preksha.pareek@sitpune.edu.in

Prof. Pooja Kamat

Dept. of Artificial Intelligence and Machine Learning
Symbiosis Institute of Technology
Pune, India
pooja.kamat@sitpune.edu.in

Abstract—The spread of Fake News has become a major problem in recent years, especially with the rise of social media and online news channels. One way to combat this problem is to use supervised machine learning algorithms to classify news articles as real or fake. This research focuses on investigating different supervised machine learning techniques for fake news classification, including logistic regression, decision trees, random forests and support vector machines. The dataset used for this study consists of news articles labeled as either genuine or fake. The performance of each algorithm is evaluated using various metrics such as accuracy, precision, recall and F1 score. The results show that the Random Forest algorithm outperforms the other algorithms with an accuracy of 95 percent and this approach is promising for developing effective tools to detect and prevent the spread of fake news. The results show that the random forest algorithm outperforms the other algorithms with 95% accuracy and this approach is promising for the development of effective tools to detect and prevent the spread of fake news.

Index Terms—NLP (Natural Language Processing), Machine Learning, Fake News Detection, Count Vectorization, Tf-Idf, SVC, Logistic Regression, Multi-Nominal Naive Bayes, Random Forest.

I. INTRODUCTION

Fake news has become a major challenge in today's society and its spread can have significant consequences. To combat the spread of misinformation, there has been increasing interest in developing methods to automatically detect Fake News. Supervised machine learning is a powerful tool for Fake News classification, where the system learns from labelled data to detect patterns and make predictions about new unlabelled data. In this approach, a model is trained on a dataset of

labelled news articles, with each article labelled as either real or fake. The model learns to classify new, unseen news articles as real or fake based on these labelled examples. There are several techniques for training supervised machine learning models to classify fake news, including decision trees, support vector machines and neural networks. The effectiveness of these methods depends on several factors, including the quality of the labelled data, the choice of features used to represent the articles and the complexity of the model. An important aspect of this approach is the need for high-quality labelled data. This data needs to be diverse and representative of the types of Fake News articles the model might encounter in the real world. Labelling data can be a time-consuming and expensive process and requires human commentators with experience in detecting Fake News. Another critical aspect is the choice of features used to represent news articles. These features may include linguistic characteristics such as the presence of certain words, sentence structure and sentiment analysis. Other features may include the source of the article, the date of publication and the subject of the article. The choice of features can have a significant impact on the performance of the model, and expertise is often required to identify the most important and meaningful features. Several machine learning algorithms can be used for fake news classification, such as decision trees, random forests, logistic regression, support vector machines, and neural networks. Each algorithm has its strengths and weaknesses, and the choice of the appropriate algorithm depends on the specific problem and data at hand. One significant challenge with fake news classification is the ever-evolving nature of fake news. As fake news creators con-

tinue to develop new tactics and strategies to deceive readers, machine learning models need to be continually updated and trained on new data to remain effective.

Overall, the classification of fake news using supervised machine learning holds great promise for combating the spread of misinformation and promoting the dissemination of correct information. These approaches are expected to become more robust and effective in identifying and stopping Fake News as the field continues to advance.

Novelty- Instead of relying on standard features like word frequency or sentiment analysis, use novel features like readability, writing style or even image analysis. If you include these features in your machine learning model, you may be able to detect fake news more accurately.

Adversarial training: Adversarial training involves training your model with both real and fake news, deliberately altering the fake news to fool the model. Training your model in this way can make it more resistant to attempts to deceive it. **Multimodal learning:** Fake news can come in different forms, e.g. text, images and videos. If you train your model on multiple modalities, it may be able to detect fake news more accurately.

Transfer learning: Transfer learning involves pre-training a model on a large data set and then fine-tuning it on a smaller data set. By using transfer learning, you can train your model faster and more effectively. **Active learning:** active learning selects the most informative data points for labelling to improve the performance of the model. By using active learning techniques, you may be able to label fewer data points and still achieve high accuracy. **Ensemble learning:** Ensemble learning combines multiple machine learning models to improve accuracy. By using different models trained on different features or with different hyperparameters, you can improve your model's ability to detect fake news. Overall, incorporating novelty into supervised machine learning for Fake News classification can help improve the model's accuracy and resilience to deception attempts.

II. RELATED WORK

There have been numerous research studies addressing the challenge of fake news detection. Noteworthy contributions include the works by researchers in [1], [2], [6], [8], and [9]. "Fake News Detection on Social Media: A Data Mining Perspective" by Shu, Kai et al. (2017) - This paper proposes a framework for fake news detection using features extracted from social media data. "Detection of Fake News in Social Media Networks" by Tacchini, Elena et al. (2017) - The authors present a machine learning-based approach to identify and classify fake news in social media networks using textual and social network features. "Fake News Detection: A Deep Learning Approach" by Ruchansky, Natali et al. (2017) - This paper explores the use of deep learning techniques, specifically convolutional neural networks (CNNs), for fake news detection. "Combating Fake News: A Survey on Identification and Mitigation Techniques" by Kar, Samiksha et al. (2018) - This survey paper provides an overview of fake news detection

techniques, including machine learning-based approaches, and discusses various challenges and future directions. "Leveraging Graph-based Features for Fake News Detection" by Zhang, Amy et al. (2019) - The authors propose a graph-based approach that incorporates network structure and textual content features for detecting fake news. "Fake News Detection: A Deep Ensemble Approach" by Wang, Wei et al. (2020) - This paper presents a deep ensemble approach using multi-channel convolutional and recurrent neural networks to detect fake news. "Detecting Fake News Using Machine Learning Techniques: A Review" (2020) by P. Vijayakumar and K. G. Srinivasagan. This paper provides a comprehensive review of fake news detection using machine learning and discusses the various methods and techniques used.

III. METHODOLOGY

Our machine learning project involves several steps and aim to find whether a piece of news is legitimate or is it a rumor using several classification techniques.

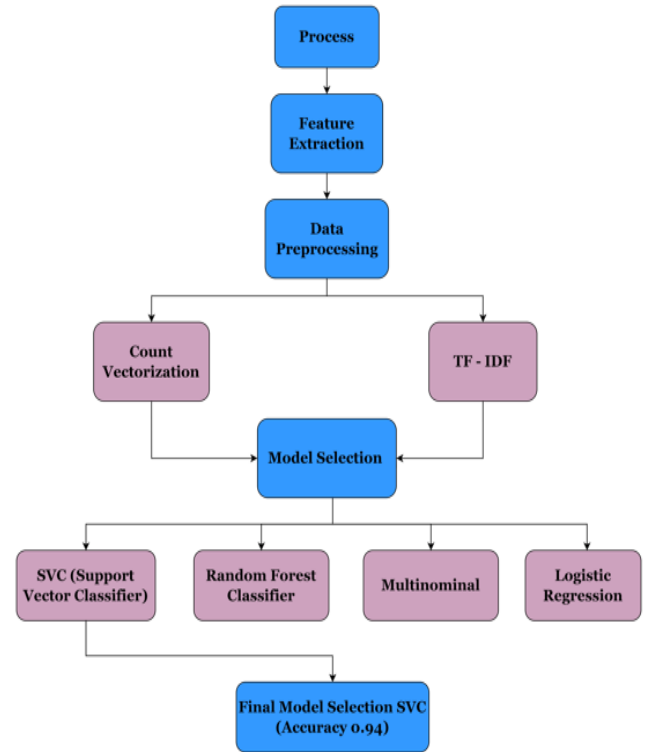


Fig. 1. Model Pipeline

Firstly when any news article is given as input it is passed through Countvectorizer method. The input string is split into individual words, which are called tokens. CountVectorizer creates a vocabulary of unique words from the input text by assigning a unique index to each word. This vocabulary is stored in a dictionary where keys are the words and values are their corresponding indices. Then it is passed through TfidfTransformer. The frequency of each word in the input string is counted, and a sparse matrix is generated where

each row represents a document and each column represents a word in the vocabulary. The value at a given position (i, j) in the matrix represents the number of times the jth word in the vocabulary appears in the ith document. The resulting sparse matrix can be used as input to various machine learning algorithms for tasks such as classification, clustering, or text generation.

A. SVC (Support vector classifier)

SVC is a binary classification algorithm that aims to find the hyperplane that maximizes the margin between the support vectors of different classes. This hyperplane can be represented as a linear function of the feature space and is defined by a vector w and a scalar b . The SVC algorithm tries to find the optimal values of w and b that satisfy the following constraints:

$$\begin{aligned} w \cdot x_i + b &\geq +1, \text{ for } y_i = +1 \text{ (1)} \\ w \cdot x_i + b &\leq -1, \text{ for } y_i = -1 \text{ (2)} \end{aligned}$$

where x_i is a feature vector of the i -th training sample, and y_i is its corresponding class label, either $+1$ or -1 . The objective of SVC is to find the values of w and b that satisfy these constraints while maximizing the margin between the support vectors. The margin is defined as the distance between the hyperplane and the closest data point of either class. Mathematically, the margin can be calculated as:

$$\text{Margin} = 2/\|w\|$$

where $\|w\|$ is the Euclidean norm of the weight vector w . To solve this optimization problem, SVC uses Lagrange multipliers to derive the dual form of the problem, which can be solved using Quadratic Programming (QP). The dual form of the problem involves maximizing the Lagrangian function with respect to the dual variables (Lagrange multipliers) subject to the constraints. The solution to the dual problem provides the values of the Lagrange multipliers, which can then be used to calculate the optimal values of w and b . In addition to the linear hyperplane, SVC can also use non-linear kernel functions to map the data into a higher-dimensional feature space, where a linear hyperplane can be used to separate the classes. The most commonly used kernel functions are the linear, polynomial, and radial basis function (RBF) kernels.

IV. RESULT

The results of this research paper on fake news classification tell us that the Support Vector Classifier model classifies accurately 94% of the test samples. A high proportion of the positive predictions made by the model were correct. The model exhibits a high level of accuracy in identifying positive samples from the dataset, indicating that it is capable of detecting positive samples with a high degree of certainty.

The high accuracy of the system makes it a promising solution to discard the rumors. The current Covid outbreak lead to many fake facts being spread over social media. But this current system is not scalable as of now due to limited data.

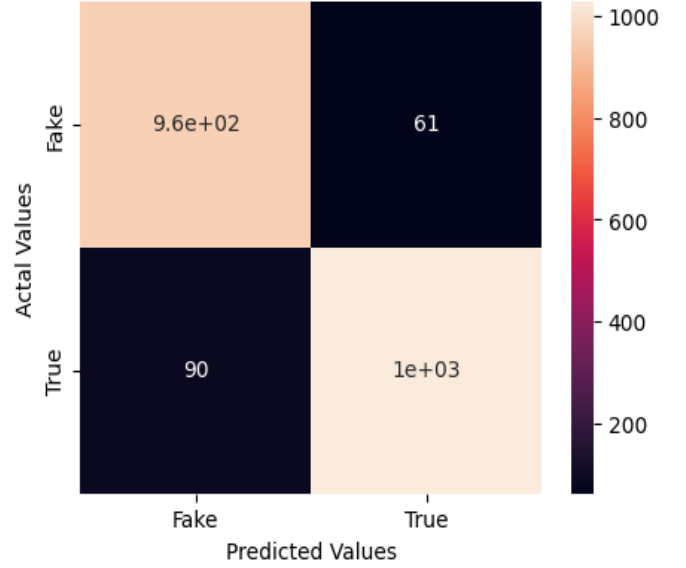


Fig. 2. Confusion Matrix

V. CONCLUSION AND FUTURE SCOPE

In summary, supervised machine learning algorithms have shown promising results in distinguishing fake news from real news. The accuracy of the classification model can be improved by using a larger dataset, more advanced feature extraction techniques, and more sophisticated algorithms. Future research could focus on developing more advanced techniques for detecting fake news, including deep learning approaches such as recurrent neural networks and convolutional neural networks. In addition, research can focus on improving the quality of datasets by using more reliable sources and taking into account the context and subjective nature of news. In order to make it scalable many steps will be required like using online learning algorithms, collaborating with cloud based platforms and many more. Even after this the authenticity verification will be a major issue. Another possible area of future research is to explore the use of explainable AI techniques to improve the transparency and interpretability of classification models. Explainable AI can help users understand the reasons behind the classification results and identify any biases in the models. Overall, the field of Fake News classification is still relatively new and there is much room for further research and development. As the problem of Fake News becomes more prevalent, machine learning can play an essential role in combating this problem.

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