

## MACHINE LEARNING

# Fake News detection using Python and Machine Learning

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## Abstract

Online social networks have seen a surge in fake news, leading to widespread dissemination for commercial and political gain. Fake news manipulates information cleverly, making it contagious among social media users and impacting offline society. To create a trustworthy online environment, this project focuses on timely identification of fake news articles, creators, and subjects. It also evaluates the performance of detection mechanisms. The proposed approach involves a machine learning model to predict whether news articles are real or fake, utilizing various techniques to enhance accuracy. By leveraging machine learning, we can effectively combat misinformation and promote an informed digital society.

## Introduction

*We are entering a new world. The technologies of machine learning, speech recognition, and natural language understanding are reaching a nexus of capability. The end result is that we'll soon have artificially intelligent assistants to help us in every aspect of our lives.* –Amy Stapleton

Fake news online is a pressing challenge, undermining the reliability of information. Deceptive narratives flood social media, eroding trust and manipulating opinions. Researchers leverage machine learning and data analytics to develop robust detection methods, combating misinformation and restoring trust. Challenges include evolving tactics, diverse datasets, and ethical considerations. Progress in this field offers hope for a resilient information landscape.

This project explores approaches, methodologies, and advancements in fake news detection. By understanding the complexities of this issue, we aim to create a more trustworthy information ecosystem. Our research focuses on leveraging machine learning and data analytics to identify and address fake news effectively. Despite challenges such as evolving tactics and ethical dilemmas, we are dedicated to mitigating the impact of misinformation and safeguarding the integrity of our society.

## Methodology

- i. **Data collection** This stage refers to the process of gathering and obtaining relevant data to train a machine learning model. We can gather data required for our project from various web scraping, articles, headlines, metadata, and associated contents. We efficiently use these data sets to train and test our model using its diverse data.
- ii. **Data cleaning and pre-processing** In this step we remove irrelevant or duplicate data, such as advertisements, boilerplate content, or non-textual elements. Standardize the text by converting to lowercase, removing punctuation, and handling special characters. Handle misspellings, abbreviations, and stemming to reduce noise in the text. Remove stop words (commonly occurring words with little semantic value). Perform text normalization, such as dates, and numerical values.

v. **Training, Evaluation and Improving machine learning models** We Split the dataset into training and testing sets for model evaluation and selecting appropriate ML algorithms, such as logistic regression, naive Bayes classification model, decision tree and passive aggressive classifier. The ML models are trained based on the labeled training dataset. The models are then evaluated using performance metrics such as accuracy, precision, recall, and F1 score. Based on the evaluation we further fine-tune the models by adjusting hyperparameters or using techniques like cross-validation.

## Implementation

For getting a better understanding of the data we visualise it using seaborn and matplotlib. It creates count plots to display the distribution of fake and real news, as well as the distribution of news across different subjects.

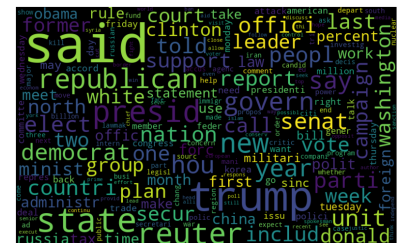


Figure 3. Caption for Figure 3

Finally we define a function called `fake_news_det` that uses the saved model to predict the label of new news articles and classify news articles as fake or real. The code and the readme file for this project are available at the GitHub repository: [https://github.com/jobint001/Fake\\_news\\_detection](https://github.com/jobint001/Fake_news_detection). Results of this implementation is discussed in the below section.

## Result and Discussion

The project has been successful in achieving the goal of creating a fake news detection facility to overcome the proliferating fake news in this digital age. We have trained various Machine Learning models and evaluated them to select the best preferred model. The Machine Learning models used in this project are Logistic Regression, Naive Bayes classification model, Decision Tree, and Passive-Aggressive Classifier.

We have used confusion matrices and ROC graphs to help us further evaluate the models. The confusion matrix provides detailed information about the model's performance, giving insights into the true positive, true negative, false positive, and false negative predictions of the models. while the ROC curve illustrates the trade-off between the true positive rate and the false positive rate at different classification thresholds. The confusion matrices and ROC graphs obtained from evaluating the models are given below,

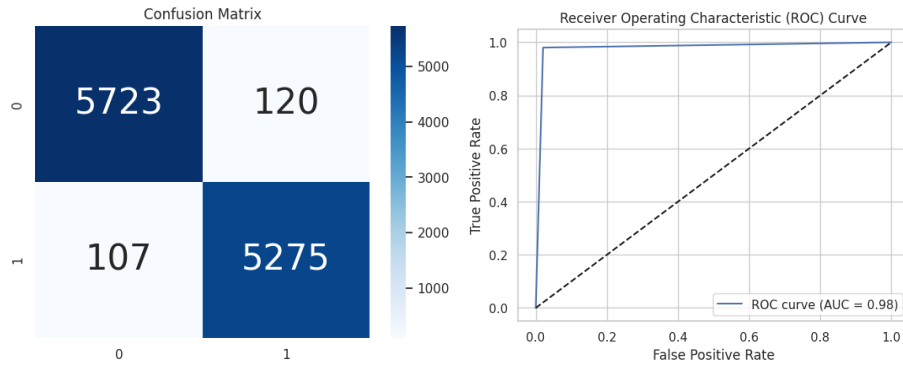


Figure 4. Logistic Regression

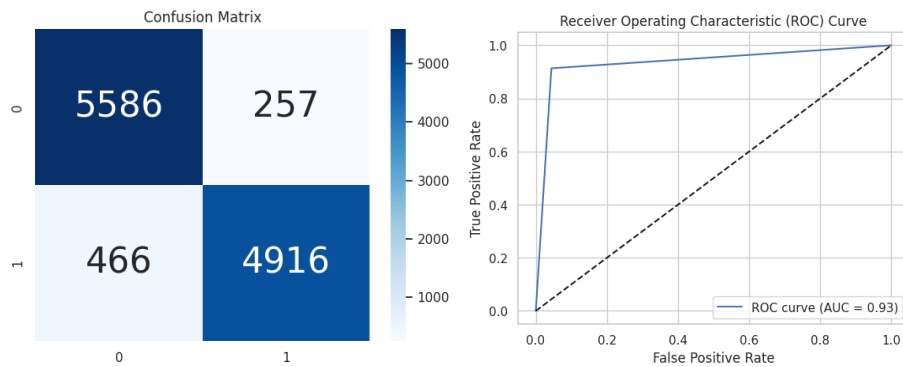


Figure 5. Naive Bayes classification model

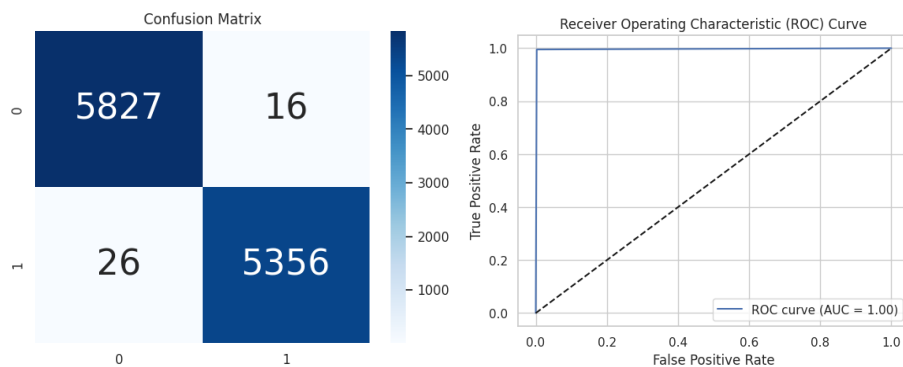


Figure 6. Decision Tree

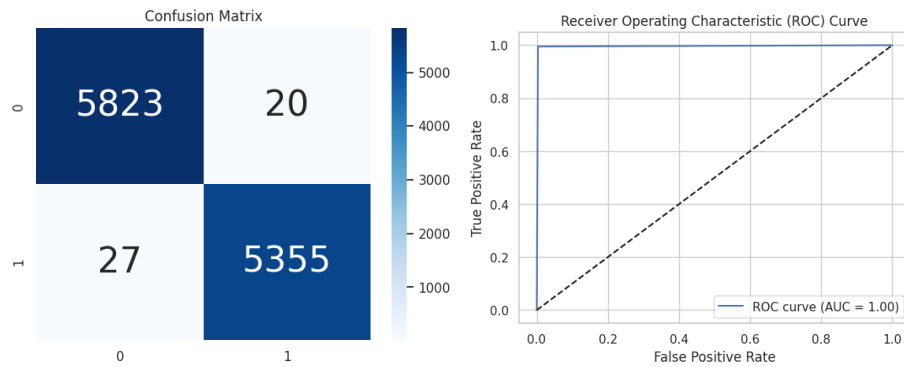


Figure 7. Passive-Aggressive Classifier

The summary of the of the various evaluations and results of the confused matrices and ROC curves of the Machine Learning models have been compiled and shown in Table 1.

The four different machine learning models were implemented and evaluated as follows,

- The Logistic Regression model achieved an accuracy score of 98.90%, showcasing its effectiveness in predicting the authenticity of news articles. It utilized a linear regression algorithm to classify the data.
- The Naive Bayes classification model achieved an accuracy score of 93.55%, indicating its ability to classify news articles as real or fake. It showed promising performance in distinguishing between the two categories.
- The Decision Tree model achieved an accuracy score of 99.62%, providing another viable approach to fake news detection. It demonstrated its capability to create decision rules based on features derived from the dataset.
- Lastly, the Passive-Aggressive Classifier model achieved an accuracy score of 99.58%, highlighting its ability to adapt to new data and make online updates to its model. It demonstrated good performance in classifying news articles.

Table 1. Performance of Machine Learning Models in testing the fake news data

Model No.	Model Name	RMSE value	Accuracy	Precision	f1-score	Recall	Processing Time
1	Logistic Regression	0.085	98.90	0.99	0.99	0.99	4.735 sec
2	Naive Bayes	0.249	93.55	0.92	0.94	0.96	0.169 sec
3	Decision Tree	0.061	99.62	1.00	1.00	1.00	23.042 sec
4	Passive Aggressive	0.064	99.58	1.00	1.00	1.00	0.427 sec

From the above results we see that Passive-Aggressive Classifier has a high accuracy of 99.58% with a response time of 0.427 seconds, displaying its superior speed while providing a high accuracy rate as compared to the other models. This model shows promise for identifying and detecting fake news accurately at high speeds. Future development could focus on incorporating additional features and exploring more advanced machine learning techniques to further enhance the accuracy and reliability of fake news detection models.

## Conclusions

In this project, we investigated the effectiveness of machine learning models (Logistic Regression, Naive Bayes, Decision Tree, and Passive-Aggressive Classifier) for fake news detection. The Passive-Aggressive Classifier model demonstrated the highest precision, recall, and F1-scores among the tested models, while the Decision Tree model showed comparable performance it had a much longer processing time. These findings contribute to the field of fake news detection and highlight the potential of these models in combating misinformation and promoting reliable information. However, further research is needed to explore ensemble methods, advanced feature engineering techniques, and domain-specific knowledge integration to address the challenges associated with detecting nuanced forms of fake news.

In conclusion, our study demonstrates the effectiveness of Logistic Regression, Naive Bayes, Decision Tree, and Passive-Aggressive Classifier models for fake news detection. The Passive-Aggressive Classifier model in particular, shows promise with its high accuracy and precision. Enhancements through advanced techniques and continued research are necessary to improve the models' performance and tackle the complexities of identifying various forms of fake news. Developing robust and accurate fake news detection systems is crucial for preserving information integrity and enabling informed decision-making in the digital era.

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## References

1. Vijayarani S, Janani R, et al. Text mining: open source tokenization tools—an analysis. *Advanced Computational Intelligence: An International Journal (ACIJ)* 2016;3(1):37–47.
2. Singh J, Gupta V. Text stemming: Approaches, applications, and challenges. *ACM Computing Surveys (CSUR)* 2016;49(3):1–46.
3. Wilbur WJ, Sirotkin K. The automatic identification of stop words. *Journal of information science* 1992;18(1):45–55.
4. Sial AH, Rashdi SYS, Khan AH. Comparative analysis of data visualization libraries Matplotlib and Seaborn in Python. *International Journal* 2021;10(1).
5. Teoh TT, Rong Z. Text Mining. In: *Artificial Intelligence with Python* Springer; 2022.p. 227–237.
6. Zou X, Hu Y, Tian Z, Shen K. Logistic regression model optimization and case analysis. In: *2019 IEEE 7th international conference on computer science and network technology (ICCSNT)* IEEE; 2019. p. 135–139.
7. Singh M, Bhatt MW, Bedi HS, Mishra U. Performance of bernoulli's naive bayes classifier in the detection of fake news. *Materials Today: Proceedings* 2020;.
8. Lyu S, Lo DCT. Fake news detection by decision tree. In: *2020 SoutheastCon IEEE*; 2020. p. 1–2.
9. Gupta S, Meel P. Fake news detection using passive-aggressive classifier. In: *Inventive Communication and Computational Technologies: Proceedings of ICICCT 2020* Springer; 2021. p. 155–164.

## Project Code In Implementing Fake News Detection

### Github repo

[https://github.com/jobint001/Fake\\_news\\_detection](https://github.com/jobint001/Fake_news_detection)