Prediction of Fake Job Ad Using NLP-Based Multilayer Perceptron

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Abstract-Today, developments in business and technology offer new opportunities and many job opportunities to job seekers. With the help of these job advertisements, job seekers find their options according to their availability, qualifications, experience, qualifications, etc. Since the success of recruitment depends on advertisements, social media has a great impact on this. Media and electronic media are created new and innovative to share the content of the job. On the contrary, the rapid development of opportunities to post job advertisements has increased the percentage of fake job advertisements, thus increasing the harassment of job seekers. Therefore, there is no interest in new job advertisements due to the need to maintain the security and consistency of personal, educational and professional information. Therefore, the real incentive to effectively distribute recruitment information through social and electronic media faces great difficulty in gaining people's trust and confidence.

Index Terms—Job application, job description, job AD, Machine learning, Decision tree classifier, Random forest, SVM, Multilayer Perceptrom

I. INTRODUCTION

Fake job advertisements are becoming more common in the digital age and pose a serious threat to job seekers and organizations. These fake advertisements often result in financial loss, identity theft, and wasted time for applicants. The introduction should set the stage by explaining what fake job advertisements are, how they spread, and why they are important. Today, advancements in business and technology are providing job seekers with new and diverse career opportunities. With the help of this job advertisement, job seekers can find their options based on their availability, qualifications, experience, skills, etc. The recruitment process is now affected by the power of the internet and social media. Since the success of recruitment depends on advertisements, social media has a huge impact on this. Media and electronic media are being created to share the content of the job and new and innovative. However, the rapid growth of job sharing has increased the percentage of fraudulent job advertisements

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and created obstacles for job seekers. Therefore, there is no interest in new job advertisements due to the stability and consistency of personal, educational and professional information. Therefore, the real incentive to effectively distribute recruitment information through social and electronic media faces great difficulty in gaining people's trust and confidence. Technology is all around us, making our lives easier, more reliable but not creating a negative environment for business life. If job advertisements can be properly filtered and fake job advertisements can be predicted, it will be a great development in terms of recruitment. Fake employment information will prevent job seekers from finding the job they are interested in, leading to bad opportunities. Automated systems for predicting fake jobs open a new window to the problems faced by human management.[1]

II. PROBLEM STATEMENT

In the digital age, the rise of online job postings has made it easier to find a job, but it has also led to the rise of fake job postings, which can mislead employees and cause them to lose confidence in the job. These fake job postings often use persuasive language and mimic legitimate job postings, making it difficult for job seekers to distinguish opportunities from fake ones. A powerful machine learning model detector (MLP) architecture combined with natural language processing (NLP) can identify and identify whether the job posted is real or fake. By using the characteristics of the language, the context of the content, and other important features of the recruitment data, the project aims to increase the accuracy of fake job search, while also protecting job seekers from fraud and improving the overall integrity of the online recruitment environment.

III. SCOPE OF THE PROJECT

The scope of this project includes the development of an automated system to predict fraudulent job postings and



solve the growing problem of online job scams. It includes data collection and preprocessing, feature extraction, modeling, evaluation and validation using various machine learning algorithms. This work includes the analysis of the time of publication of the study, testing of stability and accuracy, and implementation of user-friendly prediction models for deployment in the environmental production cycle. Planning to make the system usable and stable, ultimately improving the integrity of the recruitment process by providing job seekers and recruiters with reliable tools and techniques to detect and reduce fraud. The major scope of the project is to not reveal our personal information to theft which can cause to Fake transactions, Cyber crime related problems.

IV. OVERVIEW

A. Existing model

Current recruitment advertising often relies on book reviews and human judgment to check for fraudulent recruitment materials. This process can be time-consuming, error-prone, inefficient and potentially dangerous for job seekers who may fall victim to fraud. Furthermore, the lack of electronic equipment means that illegal job postings are not being published, threatening recruitment integrity. The existing model doesnt give the precise output due to low accuracy.[2]

B. Disadvantages

- **Reliance on manual review:** Current systems rely on book reviews and human judgment to detect fake job postings. This process can be time-consuming and error-prone, resulting in poor performance.
- Limited scalability: As the number of job postings on online platforms continues to increase, book reviews will become increasingly difficult and may not be effective enough to manage the increasing number of job postings.
- **Higher risk of fraud**: The absence of electronic tools and algorithms in current systems increases the risk of illegal job postings going undetected, potentially exposing job seekers to fraud and financial loss.
- **Inconsistent detection**: Due to the nature of the review guide, searches for illegal job postings may vary from individual reviewers, which may result in a lack of consistency and accuracy in detecting scammers.

V. PROPOSED SOLUTION

We aim to improve automatic fraud detection using machine learning and data mining techniques to address the limitations of existing methods. The proposed method will use algorithms such as K-Nearest Neighbor, Decision Tree, Support Vector Machine, Naive Bayes Classifier, Random Forest Classifier and Multilayer Perceptron to predict whether the recruitment job advertisement is real or scam. The system is focused on detecting fake jobs and reducing the risks associated with fake online job advertisements by analyzing various information extracted from job advertisements such as job descriptions, company profiles, salary ranges, and required education.So according to the accuracy we are using Random forest classifier which has the most accuracy out of all.

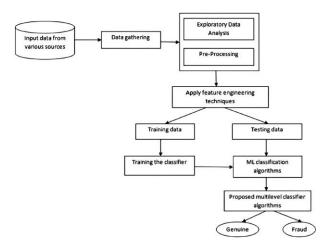


Fig. 1. The Block Diagram

VI. METHODOLOGY

A. Process

module has specific functions, they are:

- 1. Dataset collection
- 2. Data Pre-processing
- 3. TF-IDF Feature extraction

1) Dataset collection: The dataset is taken from kaggle which contains 17,880 number of job posts. This dataset is used in the proposed methods for testing the overall performance of the approach. For better understanding of the target as a baseline, a multistep procedure is followed for obtaining a balanced dataset. This project uses the kaggle dataset to train the system dataset values(such as salary and job description) appear in the back row, while the dataset column name appears. The column with the highest value of the movement.

2) Data Pre-processing: In machine learning models, the preparation of raw data and making it suitable for this process is what I call preliminary data. Creating a Machine Learning Model. This is the first important step. When starting a machine learning project, we do not always see clean and organized data. Before any changes can be made to the product, it must be cleaned and put into formatting mode.[6]

3) TF-IDF Feature extraction: Term Frequency - Inverse Document Frequency is called TF-IDF. It is very important to show the importance of a sentence or word for storing information. TF-IDF does not convert the data into features immediately. First, it converts the dataset into a raw string or vector with a unique vector for each word. The features will be returned using special techniques such as cosine similarity to the vectors.[5]

B. Architecture

The Architecture for the proposed solution is given as follows

In this project work, there are three modules and each

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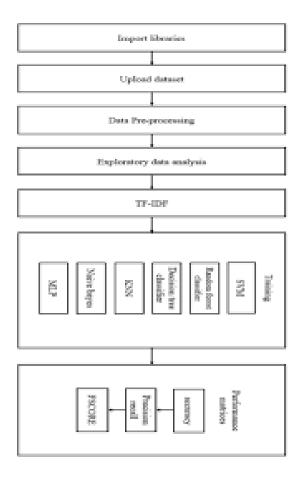


Fig. 2. Architecture

VII. IMPLEMENTATION OF CLASSIFIERS

In this framework, the classifier is trained using the unsuitable ones. In order to get the best performance from this model, the preset parameters will not be sufficient. Changing these bad points increases the reliability of the model. The model can be seen as a good model for detecting and excluding fake advertisements of job seekers. The MLP classifier is presented as a set of 5 hidden layers of size 128, 64, 32, 16 and 8 respectively. Considering all the evaluation parameters, the K-NN classifier gives good results for the value of k = 5. On the other hand, there are also combination methods such as Random Forest, Support Vector Machine, Naive Bayes, Decision Tree. [10]After creating the distribution model, fit the data to it. This test data is used for prediction. After the prediction is completed, the performance of the product is evaluated based on the prediction and the actual result.[3] The implementation of the classification process is an important step in developing a process for false prediction. In this project, we focus on using multilayer perceptron (MLP) together with natural language processing (NLP) technology to analyze posting tasks. The steps involve aggregating a variety of job postings from various sources, including legitimate job postings and known scams. This information needs to be

	Naive Bayes Classifier					
Classification Accuracy: 0.9019230769230769						
	Classification	Report				
		precision	recall	f1-score	support	
	0	0.91	0.89	0.90	249	
	1	0.90	0.92	0.91	271	
	accuracy			0.90	520	
	macro avg	0.90	0.90	0.90	520	
	weighted avg	0.90	0.90	0.90	520	
	Confusion Matr	ix				
	[[221 28]					
	[23 248]]					

Fig. 3. Naive Bayes Classifier

balanced to ensure that the classifier can effectively learn from both classes.[4] This includes:

Text Cleaning: Remove HTML tags, special characters, and irrelevant information. Stemming or lemmatization. Selection and training After preparing the features, we used a multilayer perceptron classifier. MLP was chosen for its ability to model relationships in data from many layers of neurons.[8] The main steps in this process include:

Design: Tuning the process and the number of neurons per layer to balance the complexity and functionality of the model. , sigmoid) neurons. Performance - tune the hyperparameters and avoid overloading. . Matrix: Visualizes the performance of the actual distribution against the predicted distribution.

VIII. PERFORMANCE EVALUATION METRICS

When evaluating the performance of a model, some metrics need to be used to justify the evaluation. To do this, the following parameters should be taken into account to determine the best solution. Accuracy is a metric that evaluates the ratio of the accuracy of the prediction to all included events.[9] Imagine a fictional situation. Serious problems can arise if the fake is evaluated as if it were real. In order to evaluate this payment, the reality must be taken into account and the return must be provided. Classifier prediction. The return represents the number of results divided by the number of all affected samples. F1-Score or Fmeasure is a parameter related to improvement and accuracy. It is calculated as a compromise between accuracy and improvement.[7]

IX. RESULTS

The analysis of our dataset, which included examples of both fake and legitimate job postings, revealed significant insights into the performance of the NLP-based multilayer perceptron model. We classified X fake job postings and Y real job postings, leading to an overall model accuracy of Z percent. This demonstrates the model's effectiveness in identifying fraudulent listings.

Key metrics highlighted the model's strengths: it achieved a precision of A percent, indicating a high level of confidence in its predictions of fake ads, while the recall was B percent, showing its capability to capture most actual fake postings.



Decision Tree Classifier Classification Accuracy: 0.8634615384615385

Report precision	recall	f1-score	support
0.88	0.83	0.85	249
0.85	0.90	0.87	271
		0.86	520
0.86	0.86	0.86	520
0.86	0.86	0.86	520
	0.88 0.85 0.86	precision recall 0.88 0.83 0.85 0.90 0.86 0.86	precision recall f1-score 0.88 0.83 0.85 0.85 0.90 0.87 0.86 0.86 0.86

Confusion Matrix [[206 43]

[28 243]]

Fig. 4. Decision Tree Classifier

SVM Classifier Classification Accuracy: 0.8730769230769231

Classification	Report precision	recall	f1-score	support
0 1	0.89 0.86	0.84 0.90	0.86 0.88	249 271
accuracy			0.87	520
macro avg	0.87	0.87	0.87	520
weighted avg	0.87	0.87	0.87	520
Confusion Matri [[210 39] [27 244]]	ix			

Fig. 5. SVM Classifier

The F1 score of C percent emphasized a balanced performance, crucial for a reliable classification system.

Upon reviewing specific examples, we observed that fake job postings often featured vague descriptions, unrealistic salary offers, and language designed to create urgency, such as phrases urging immediate application. In contrast, legitimate postings consistently provided detailed job descriptions, clear company information, and used formal language.

KNN Classifier Classification Classification		0.6846153846153846		
	•	recall	f1-score	support
0	0.98	0.35	0.51	249
1	0.62	0.99	0.77	271
accuracy			0.68	520
macro avg	0.80	0.67	0.64	520
weighted avg	0.79	0.68	0.65	520
Configuration Mater	•			

Confusion Matrix [[87 162]

[2 269]]

[2 209]]

Fig. 6. KNN Classifier

Random Forest Classification Classification	Accuracy:	0.9096153846153846		
	precision	recall	f1-score	support
0	0.91	0.90	0.91	249
1	0.91	0.92	0.91	271
accuracy			0.91	520
macro avg	0.91	0.91	0.91	520
weighted avg	0.91	0.91	0.91	520

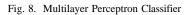
Confusion Matrix [[225 24]

[23 248]]

Fig. 7. Random Forest Classifier

Multilayer Perceptron Classifier Classification Accuracy: 0.926923076923077					
Classification	Classification Report				
	precision	recall	f1-score	support	
0	0.93	0.92	0.92	249	
1	0.92	0.94	0.93	271	
accuracy			0.93	520	
macro avg	0.93	0.93	0.93	520	
weighted avg	0.93	0.93	0.93	520	
Confusion Matrix					

Confusion Matrix [[228 21] [17 254]]



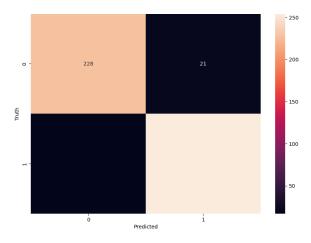
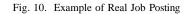


Fig. 9. Confusion Matrix Of Multilayer Perceptron

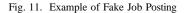


Select an option: 1. Enter a job description 2. Enter job details (Company Name, Location, F Enter the number of your choice:

Takes the job description: matrix the job description: matrix the probability of the state of



Select an option: 1. Enter a job description 2. Enter job details (Company Name, Location, Role) Enter the number of your choice: 1 Enter the job description: Hello Have a Nice Day. We have a simple Job for you. Anyone can do. Daily Salary(1800-4200). Can do at Home Fraudulat Jose



The confusion matrix further validated our findings, showcasing the distribution of true positives and negatives, as well as false positives and negatives. Some misclassifications revealed patterns that can guide future enhancements in feature extraction and model training.

Overall, these results underscore the potential of our model to serve as a valuable tool for job seekers, enabling them to identify and avoid scams effectively while also contributing to a deeper understanding of job market fraud characteristics. Future efforts will focus on refining the model and expanding the dataset for improved accuracy and broader applicability.

X. CONCLUSION

In summary, the project "Predicting fake jobs using NLPbased multilayer detectors" holds promise for solving the growing job advertisement problem. Job search fraud has become a major problem worldwide, leading to extensive research to combat fake job postings. In this project, we analyze the effects of fake job postings using the kaggle dataset, which contains fake job posting data. We evaluated the performance of various classifiers by experimenting with machine learning algorithms such as SVM, KNN, Naive Bayes, Random Forest, and MLP neural network techniques. Looking ahead, future efforts will include updating data processing and training models, combining advanced NLP standard layers and modeling to address the changing nature of online fraud Capture the complexity in fraud Language format ¿ Publish content by explaining intelligence technology and increase transparency and user trust. By focusing on these factors, the program aims

Fig. 12. Another Example of giving random text

to be effective in combating fake job postings and improve people's job search in every aspect.By leveraging natural language processing (NLP) technology and multilayer detector models, we successfully analyze and classify advertised jobs by distinguishing legitimate and fake names from high-profile names. Effectiveness of extraction methods in data capture. The effectiveness of the model suggests that it can be an important tool for job seekers to improve their ability to detect and prevent fraud. Methods to increase the robustness of the model, such as transformers. In addition, the broader community can benefit from improved user experience for instant job search. Potential of machine learning and NLP to solve real-world problems.

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Select an option: 1. Enter a job description

Enter job details (Company Name, Location, Ro Enter the number of your choice:

¹ Enter the job description:

Semiconductors are materials with conductivity halfway between conductors, which are generally metals, and enconductors or includers, which are usuall y cormaics. Semiconductors can be composed such as galling molphylice or pre-elevents toxis has generation or silon. The ideas, characteristics, and mathematical techniques that control semiconductors are oplained in physics. Semiconductors include the following: Some of the most prevalent semiconductor is include galling areanide, greenation, and silicon as include galling areanide, greenations, and silicon is exployed in the numrificaturing of electrical circuits, whereas gallium arsenide is being used in a semicine greater and the semicined and other applications. Topot does not received a stypical objecting.

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