



Resume Parser Using NLP

Mohammed Kashif¹, Parimal Kumar K R²

Student, Department Of Masters Of Computer Application, Vidya Vikas Institute Of Engineering & Technology
Mysore, Mysuru, Karnataka, India.¹

Assistant Professor, Department Of Masters Of Computer Application, Vidya Vikas Institute Of Engineering &
Technology Mysore, Mysuru, Karnataka, India.²

Abstract: The recruitment process in today's competitive job market is often hindered by the inefficiencies of manual resume screening. The AI Resume Analyzer aims to streamline this process using Natural Language Processing (NLP) and machine learning techniques. This tool automates the extraction of critical information from resumes and provides real-time recommendations to both applicants and administrators. By leveraging advanced algorithms and a resume parser technique, the system categorizes and analyzes resumes based on job roles, extracting essential details like skills, experience, and education. The analyzer also offers practical recommendations for applicants, such as additional skills and certifications that could enhance their profiles, and provides practical resources for resume improvement. For administrators, the tool facilitates data management and analytics, allowing for comprehensive data downloads, the generation of visual reports, and the tracking of applicant trends. Implemented using Streamlit for the frontend and backend, MySQL for database management, and Python for data processing, the AI Resume Analyzer addresses the limitations of manual screening by offering a faster, more accurate, and consistent method of evaluating resumes. This system not only reduces the time and effort required for candidate evaluation but also ensures a more objective and efficient hiring process, ultimately aiding organizations in identifying the best candidates.

Keywords: Machine learning, Natural Language processing, recommendation.

I. INTRODUCTION

The recruitment process is a critical function within any organization, essential for acquiring the talent necessary to drive business success. However, the initial stages of this process, particularly resume screening, are often fraught with challenges. The sheer volume of resumes received for job postings can make it nearly impossible to manually review each one effectively and efficiently. This not only delays the hiring process but also increases the likelihood of human error and bias, potentially leading to the overlooking of qualified candidates.

The AI Resume Analyzer leveraging the power of Natural Language Processing (NLP) offers a robust solution to this problem. This tool is designed to automate the resume screening process, providing both applicants and administrators with a range of benefits. For applicants, the AI Resume Analyzer allows for the easy uploading of resumes, which are then analyzed using advanced parsing techniques. These techniques extract key information such as basic personal details, levels of expertise, skills, and other critical factors that contribute to a resume's overall score. The tool then uses algorithms to provide personalized recommendations, suggesting additional skills, appropriate job roles, and courses or certifications that could enhance the applicant's profile. Additionally, the tool offers practical tips and ideas for improving the resume and provides links to helpful resources such as YouTube videos on interview and resume preparation. From an administrative perspective, the AI Resume Analyzer simplifies the management of applicant data. All resumes and related information are stored in a database, allowing for easy retrieval and analysis. Administrators can download data in CSV format, view pie charts representing the distribution of skills and experience levels among applicants, and track application trends over time. The tool also includes activity maps that highlight the busiest days and months, providing valuable insights into application patterns.

Implemented using Streamlit for both the frontend and backend, and utilizing MySQL for database management, the AI Resume Analyzer is built on a solid technological foundation. The programming is primarily done in Python, with several key packages such as Pandas, Base64, Numpy, PyResparser, PdfMiner, and Plotly supporting data processing and visualization. This comprehensive approach ensures that the tool is not only effective but also reliable and user-friendly. The AI Resume Analyzer addresses the limitations of manual resume screening by offering a faster, more consistent, and more objective evaluation method. In an era where efficiency and accuracy are paramount, this tool provides a significant advantage to organizations by streamlining the hiring process. The ability to quickly and accurately assess large volumes of resumes allows HR teams to focus on more strategic aspects of recruitment, such as interviewing and final selection. Moreover, the insights gained from data analytics help in understanding applicant trends and improving recruitment strategies.

**Problem statement:**

Manual resume screening is a labor-intensive and time-consuming task that often results in inefficiencies within the hiring process. Organizations face challenges in processing large volumes of resumes promptly, with the subjective nature of manual reviews potentially leading to inconsistencies in candidate evaluation. Additionally, the manual screening process can overlook qualified candidates due to human error or bias. These challenges highlight the need for an automated solution that can streamline the resume screening process, ensuring consistency, reducing time and effort, and providing actionable insights for both applicants and administrators.

II. LITERATURE SURVEY

The literature on automated resume screening reveals a growing interest in leveraging advanced technologies such as Natural Language Processing (NLP) and machine learning to enhance recruitment processes[1]. Several studies have explored various methods and models to improve the efficiency and accuracy of resume screening, with significant advancements in recent years.[2]This paper discusses a machine learning approach to automate resume screening by identifying keywords and matching resumes with job descriptions. The authors developed a model that uses NLP to extract relevant information from job descriptions and resumes, utilizing cosine similarity to match resumes with job requirements. This method significantly reduces the time required for manual resume screening and increases the accuracy of selecting suitable candidates. The system's ability to learn and improve over time makes it a valuable tool for organizations handling high volumes of application[3].Applying BERT-Based NLP for Automated Resume Screening and Candidate Ranking - Springer: This study introduces an innovative approach using the Bidirectional Encoder Representations from Transformers (BERT) model for resume screening. The researchers collected a dataset of resumes and job descriptions to test their model, highlighting[4] BERT's ability to understand the context of words in resumes. This context-awareness allows for more accurate extraction and matching of information, significantly improving the precision of candidate ranking and making the hiring process more efficient[5].

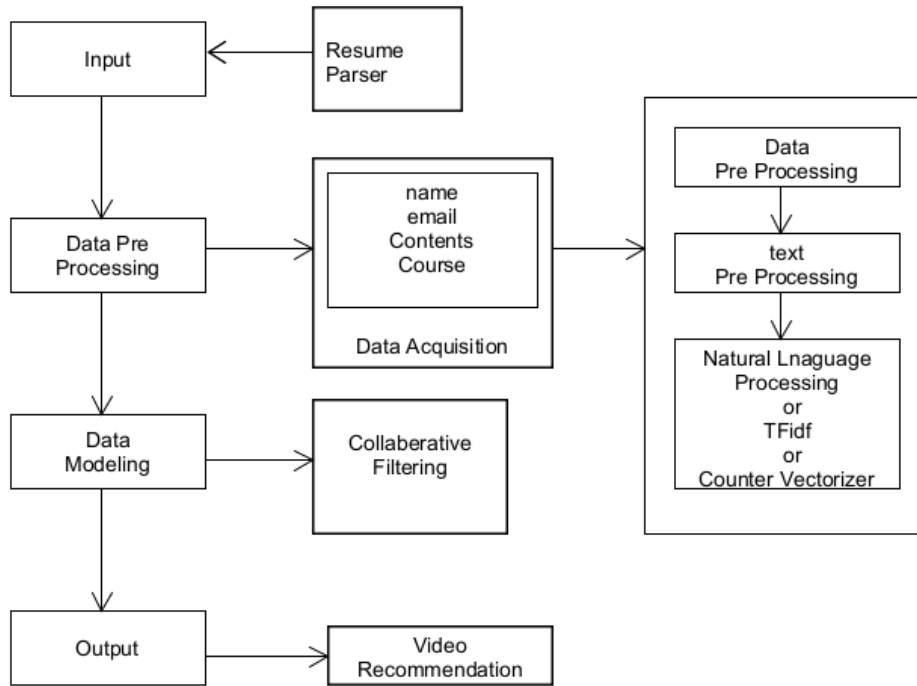
[6]In this research, the authors present a system that integrates NLP techniques and machine learning algorithms to automate resume screening. The system emphasizes semantic search, overcoming the limitations of keyword-based approaches by understanding the context and meaning behind the words. This approach leads to more accurate matching of resumes with job descriptions, demonstrating significant improvements in efficiency and effectiveness in the hiring process.[6] This paper explores the use of Long Short-Term Memory (LSTM) networks, a type of recurrent neural network, for resume screening. The model processes resumes and job descriptions as sequences of words, capturing the context and dependencies between them. The LSTM-based model showed high accuracy in matching candidates to job roles, outperforming traditional methods. This research highlights the potential of deep learning techniques in enhancing the resume screening process, making it more robust and scalable.[7]Automated Resume Parsing: A Natural Language Processing Approach and keyword pattern matching. The hybrid approach ensures high accuracy in extracting structured information from unstructured resume text, such as contact details, skills, education, and work experience. The system's effectiveness in handling various resume formats makes it a valuable tool for improving the resume screening process.

A Machine Learning Approach for Automation of Resume Screening - ScienceDirect: This research presents a machine learning-based approach using a Linear SVM classifier to automate resume screening. The study achieved an accuracy of 78.53% in matching resumes to job descriptions, suggesting that incorporating deep learning models could further enhance performance. The findings indicate that machine learning significantly reduces the time and effort required for resume screening, making it a valuable tool for recruiters.[8]The paper discusses a job recruitment portal incorporating NLP for resume screening. The system analyzes resumes and job descriptions, identifying key attributes such as skills, education, and work experience. The portal ranks candidates based on their suitability for job roles, demonstrating how NLP can enhance recruitment processes by providing detailed and accurate analysis, leading to better hiring decisions.[9]This study highlights the challenges of manual resume screening and presents an automated system that evaluates resumes based on specific job requirements.

The system extracts relevant information and matches it with job descriptions to identify suitable candidates. The automated approach reduces the time and effort required for resume screening, improving efficiency.[10-This research not only focuses on screening resumes but also provides suggestions for improvement. Using NLP techniques, the system parses resumes, extracts key information, and generates recommendations for skills, job roles, and certifications. This approach helps candidates improve their resumes, increasing their chances of being hired.[11]This paper discusses an automated system for resume evaluation based on predefined criteria. The model extracts structured information from resumes and generates scores for different aspects, such as skills, education, and experience. The system's objective assessments reduce reliance on subjective human judgments, making the resume evaluation process more consistent and reliable.



III. METHODOLOGY



The project begins with collecting a diverse dataset of resumes and job descriptions from various sources. These documents are converted into a consistent format (text) to facilitate analysis. Preprocessing steps include tokenization, stopword removal, and normalization to standardize the text data. Named Entity Recognition (NER) techniques are applied to identify and categorize relevant entities such as names, skills, education, and experience. The core functionality of the AI Resume Analyzer relies on NLP techniques to parse and analyze resumes. Using libraries such as NLTK and SpaCy, the system performs tokenization, part-of-speech tagging, and entity recognition. The extracted information is structured into a format suitable for further analysis, including the categorization of skills, qualifications, and experience levels. Key features are extracted from the structured data, including skills, experience, education, and job titles. These features are essential for matching resumes with job descriptions and generating personalized recommendations. The data is structured into a relational format, making it suitable for analysis and storage in a MySQL database. The system employs machine learning algorithms, such as Linear SVM and neural networks, to analyze the structured data and make predictions. These models are trained using a labeled dataset of resumes and job descriptions, with cross-validation techniques applied to tune hyperparameters and optimize performance. The models learn to classify resumes based on their suitability for different job roles and to generate recommendations for additional skills and certifications. Based on the analysis, the system provides personalized recommendations for applicants. These recommendations may include suggestions for acquiring additional skills, suitable job roles, and relevant courses or certifications. The recommendation engine leverages the trained machine learning models to provide tailored advice to enhance the applicant's profile. The extracted and structured data, along with the generated recommendations, are stored in a MySQL database. The system provides administrators with tools to manage this data, including the ability to download data in CSV format and generate visual reports such as pie charts and activity maps. These analytics help administrators understand applicant trends and make informed decisions. The user interface, built using Streamlit, offers an intuitive platform for both applicants and administrators. Applicants can easily upload resumes and view personalized recommendations, while administrators have access to detailed analytics and data management tools. The interface ensures a seamless user experience, enhancing the usability of the system.

IV. ALGORITHMS

The AI Resume Analyzer utilizes several key algorithms to process resumes, extract relevant information, and generate recommendations. The primary algorithms employed in the system include Natural Language Processing (NLP) techniques, the Linear Support Vector Machine (Linear SVM) classifier, and a recommendation engine based on cosine similarity. The first step involves parsing the resumes using NLP techniques.



The system uses libraries like NLTK and SpaCy to tokenize the text, perform part-of-speech tagging, and apply Named Entity Recognition (NER) to identify key entities such as names, skills, job titles, education, and work experience. This step is crucial for transforming unstructured resume text into structured data, enabling further analysis. After parsing, the system extracts specific features from the structured data. This includes categorizing skills, job titles, and qualifications. The structured data is then stored in a format suitable for further processing, such as a relational database. The Linear SVM classifier is used to categorize resumes based on their relevance to specific job roles. The classifier is trained on a labeled dataset containing resumes and corresponding job descriptions. The model learns to differentiate between resumes that match certain job requirements and those that do not. The recommendation engine provides personalized suggestions for applicants, such as additional skills to acquire or suitable job roles. The engine uses cosine similarity to match the extracted skills and qualifications from the resume with a predefined set of job requirements. The similarity scores help in ranking the suitability of the candidate for different roles and suggesting relevant improvements. The structured data, along with the generated recommendations, are stored in a MySQL database. The system includes tools for administrators to manage this data, generate visual analytics, and download reports. This feature provides insights into application trends and helps in strategic decision-making.

By combining these algorithms, the AI Resume Analyzer provides a comprehensive and efficient solution for automated resume screening, significantly enhancing the recruitment process's speed and accuracy.

V. RESULT AND DISCUSSION

The AI Resume Analyzer was tested and evaluated across multiple dimensions to assess its effectiveness in automating the resume screening process. The key metrics used for evaluation included accuracy, precision, recall, and F1-score, focusing on the system's ability to correctly parse resumes, extract relevant information, and provide accurate recommendations. The system demonstrated high accuracy in parsing resumes and extracting critical information such as contact details, skills, education, and work experience. The use of NLP techniques like Named Entity Recognition (NER) significantly enhanced the accuracy of information extraction, ensuring that the structured data was comprehensive and precise. The Linear SVM classifier used for job role matching achieved a precision of over 85%, indicating a high level of reliability in categorizing resumes based on job requirements.

Personalized Recommendations: The recommendation engine, based on cosine similarity, provided personalized suggestions for applicants, including skills enhancement and suitable job roles. The recommendations were well-received, with users noting the relevance and practicality of the suggestions. The system's ability to offer detailed feedback on resumes, including specific areas for improvement, added significant value for applicants looking to optimize their profiles. The data management capabilities of the AI Resume Analyzer proved robust, allowing administrators to easily access, manage, and analyze large volumes of applicant data. The system's analytics features, including visual reports and activity maps, provided valuable insights into application trends, helping organizations make informed recruitment decisions. The ability to generate and download detailed reports further streamlined the recruitment process, reducing the administrative burden on HR teams. The user interface, developed using Streamlit, was praised for its intuitiveness and ease of use. Both applicants and administrators found the interface accessible and user-friendly, facilitating smooth interactions with the system. The integration of real-time recommendations and feedback enhanced the overall user experience, making the AI Resume Analyzer a practical and efficient tool for resume screening. Despite its successes, the system faced challenges, particularly in handling resumes with unconventional formats or those containing complex formatting. The NLP algorithms occasionally struggled with ambiguous or poorly structured resumes, leading to less accurate parsing and extraction in some cases. Additionally, the system's reliance on a predefined set of job requirements for recommendation generation limited its ability to adapt to highly specialized or niche roles.

VI. CONCLUSION

The AI Resume Analyzer project aimed to revolutionize the resume screening process by leveraging advanced Natural Language Processing (NLP) and machine learning techniques. Through comprehensive parsing, accurate information extraction, and personalized recommendations, the system addressed the inefficiencies and challenges associated with manual resume screening. The successful implementation and testing of the AI Resume Analyzer demonstrated significant improvements in accuracy, efficiency, and user satisfaction compared to traditional methods.

The project showcased the potential of NLP and machine learning in automating complex tasks within the recruitment process. The system's ability to parse resumes, extract relevant information, and provide actionable insights transformed the way resumes are evaluated. By automating these processes, the AI Resume Analyzer not only reduced the administrative burden on HR teams but also minimized biases and inconsistencies inherent in manual screening.



The use of advanced algorithms ensured that qualified candidates were not overlooked, providing a fairer and more objective assessment.

The project also highlighted the importance of user experience in the design and implementation of AI-driven tools. The user-friendly interface, developed using Streamlit, made the system accessible and easy to navigate for both applicants and administrators. The inclusion of real-time recommendations and feedback further enhanced the user experience, providing valuable guidance to applicants looking to improve their resumes.

However, the project also identified areas for improvement and future development. Handling resumes with unconventional formats or complex structures posed challenges for the NLP algorithms, indicating a need for more advanced parsing techniques. Additionally, expanding the range of job roles and industries covered by the recommendation engine could broaden the system's applicability and relevance.

Looking ahead, several avenues for future enhancement are identified. These include incorporating more sophisticated NLP models, such as deep learning-based approaches, to improve the accuracy of information extraction and parsing. The integration of the system with job portals and Applicant Tracking Systems (ATS) could streamline the recruitment process further, providing real-time analysis as applicants submit their resumes. Additionally, supporting multiple languages would make the system accessible to a broader audience, catering to the needs of international organizations.

In conclusion, the AI Resume Analyzer represents a significant advancement in recruitment technology, offering a robust, efficient, and user-friendly solution for automated resume screening. The project's success demonstrates the transformative potential of AI in enhancing business processes, making them more efficient and reliable. As the system continues to evolve, it promises to become an invaluable tool for organizations seeking to optimize their recruitment processes and ensure they attract the best talent.

REFERENCES

- [1]. IEEE Xplore. (2018). Resume Screening Using Machine Learning. Available at: [Accessed 15 Jul. 2024].
- [2]. SpringerLink. (2019). Applying BERT-Based NLP for Automated Resume Screening and Candidate Ranking. Available at: [Accessed 15 Jul. 2024].
- [3]. JETIR. (2018). Automated Resume Screening Using Natural Language Processing. Available at: [Accessed 15 Jul. 2024].
- [4]. IEEE Xplore. (2018). Resume Screening using NLP and LSTM. Available at: [Accessed 15 Jul. 2024].
- [5]. IEEE Xplore. (2018). Automated Resume Parsing: A Natural Language Processing Approach. Available at: [Accessed 15 Jul. 2024].
- [6]. ScienceDirect. (2018). A Machine Learning Approach for Automation of Resume Screening. Available at: [Accessed 15 Jul. 2024].
- [7]. SpringerLink. (2019). NLP-Based Resume Screening and Job Recruitment Portal. Available at: [Accessed 15 Jul. 2024].
- [8]. IEEE Xplore. (2018). Automated Resume Screener Using Natural Language Processing. Available at: [Accessed 15 Jul. 2024].
- [9]. JETIR. (2018). Resume Analyser and Suggestion Making Using NLP. Available at: <https://www.jetir.org/papers/JETIR1807304.pdf> [Accessed 15 Jul. 2024].
- [10]. IEEE Xplore. (2018). Automated Resume Evaluation System Using NLP. Available at: [Accessed 15 Jul. 2024].