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Resume Screen Using Natural Language Processing

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ABSTRACT – A typical job posting on the Internet receives a massive number of applications within a short window of time. Manually filtering out the resumes is not practically possible as it takes a lot of time and incurs huge costs that the hiring companies cannot afford to bear. In addition, this process of screening resumes is not fair as many suitable profiles don't get enough consideration which they deserve. This may result in missing out on the right candidates or selection of unsuitable applicants for the job. In this project, we describe a solution that aims to solve these issues by automatically suggesting the most appropriate candidates according to the given job description. Our system uses Natural Language Processing to extract relevant information like skills, education, experience, etc. from the unstructured resumes and hence creates a summarized form of each application. With all the irrelevant information removed, the task of screening is simplified and recruiters are able to better analyse each resume in less time. After this text mining process is completed, the proposed solution employs a vectorization model and uses cosine similarity to match each resume with the job description. The calculated ranking scores can then be utilised to determine best-fitting candidates for that particular job opening.

Index Terms – Information Retrieval, Natural Language Processing, Resume Parser, Resume Analyzer, Resume Summarization, Ranking Candidates, e-recruitment.

I. INTRODUCTION

In today's competitive job market, the task of efficiently matching job seekers with relevant job openings is crucial for both employers and candidates. Traditional resume screening methods often suffer from inefficiencies due to manual processing, biased decision-making, and time-consuming procedures. To address these challenges, automated systems leveraging Natural Language Processing (NLP) and Machine Learning (ML) techniques have gained prominence in resume ranking and selection processes.

The aim of this project is to develop a robust and intelligent system for ResumeRanking using NLP and Machine Learning algorithms. By harnessing the power of NLP, the system will extract meaningful information from resumes, such as skills, experience, and qualifications, to create a comprehensive profile for each candidate. Subsequently, ML models will be employed to analyze and rank resumes based on their relevance to specific job descriptions, company requirements, and industry standards.

This project holds significant potential in revolutionizing the recruitment process by enhancing efficiency, reducing bias, and improving the quality of candidate selections. Moreover, the insights gained from analyzing large volumes of resumes and job data can provide valuable feedback to recruiters, leading to continuous optimization and refinement of the ranking algorithms.

In this report, we present the methodology, implementation details, experimental results, and implications of our Resume Ranking system. The goal is to demonstrate the effectiveness and scalability of using NLP and Machine Learning in automating resume evaluations and facilitating smarter hiring decisions.

II. LITERATURE SURVEY

A) *First Generation Hiring Systems*

In this System the Hiring team would publish their vacancies and invite applicants. Methods of publishing were newspaper, television and mouth. The interested candidates would then apply by sending their resumes. These resumes were then received and sorted by the hiring team and shortlisted candidates were called for further rounds of interviews. The whole process would take lot of time and human efforts to find right candidate suitable for their job roles.

B) *Second Generation Hiring Systems*

As the industries have grown, their hiring needs have rapidly grown. To serve these hiring needs certain consultancy units have come into existence. They offered a solution in which the candidate has to upload their information in a particular format and submit it to the agency. Then these agencies would search the candidates based on certain keywords. These agencies were middle level organizations between the candidate and company. These systems were not flexible as the candidate has to upload their resume in a particular format, and these formats changed from system to system.

C) Third Generation Hiring Systems

This is our proposed system, which allow the candidates to upload their resumes in flexible format. These resumes are then analyzed by our system, indexed and stored in a specific format. This makes our search process easy. The analyzing system works on the algorithm that uses Natural Language Processing, sub domain of Artificial Intelligence. It reads the resumes and understands the natural language/format created by the candidate and transforms it into a specific format. This acquired knowledge is stored in the knowledge base. The system acquires more information about candidate from his social profiles like LinkedIn and Github and updates the knowledge base. Ranking. Intelligent searching

Put simply, Artificial Intelligence or "AI" is an add-on to system, complementing to provide the online recruitment solution. As the name suggests, AI enables a combination of an applicant-tracking system and an artificial intelligence resume parsing, searching and matching engine. The result is a supercharged tool giving incredibly accurate candidate matching to jobs, and 'talent pool' searching that makes other systems look like they're from the stone-age.

D) Identifying "best" applicants in recruiting using data envelopment analysis

Selecting the most promising candidates to fill an open position can be a difficult task when there are many applicants. Each applicant achieves certain performance levels in various categories and the resulting information can be overwhelming. We demonstrate how data envelopment analysis (DEA) can be used as a fair screening and sorting tool to support the candidate selection and decision-making process. Each applicant is viewed as an entity with multiple achievements. Without any a priori preference or information on the multiple achievements, DEA identifies the nondominated solutions, which, in our case, represent the "best" candidates. A DEA aided recruiting process was developed that (1) determines the performance level of the "best" candidates relative to other applicants; (2) evaluates the degree of excellence of "best" candidates' performance; (3) forms consistent tradeoff information on multiple recruiting criteria among search committee members, and, then, (4) clusters the applicants.

III. PROPOSED SYSTEM

The overview of our proposed system is shown in the below figure.

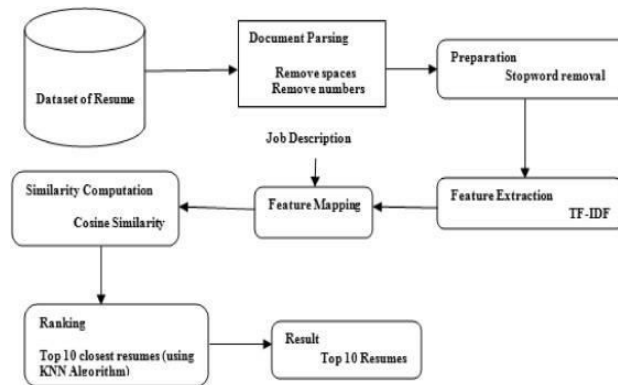


Fig. 1: System Overview

Implementation Modules

Load Dataset

- In this phase, load the dataset into program and extract the data from the .csv file. This data can be analyzed and extract the best features to preprocess the data.

Preprocess

- For the given data set, there are quite a few 'NA' values which are filtered in python. Furthermore, as the data set consists of numeric data, we used robust scaling, which is quite similar to normalization, but it instead uses the interquartile range whereas normalization is something which normalization shrinks the data in terms of 0 to 1.

Train and Test Model

- In this module, the service provider split the Used dataset into train and test data of ratio 70 % and 30 % respectively. The 70% of the data is consider as train data which is used to train the model and 30% of the data is consider as test which is used to test the model.

Resume Analyze

- In this module NLP model is used then it will analyze resume and then suggest the best resume.

Graph analyze

- In this module the graph analyzer analyzes the resume and shows the better resume.

Implementation Algorithms

NLP Model

- NLP models work by finding relationships between the constituent parts of language for example, the letters, words, and sentences found in a text dataset. NLP architectures use various methods for data preprocessing, feature extraction, and modeling. Some of these processes are:
 - Data preprocessing: Before a model processes text for a specific task, the text often needs to be preprocessed to improve model performance or to turn words and characters into a format the model can understand. Data-centric AI is a growing movement that prioritizes data preprocessing. Various techniques may be used in this data preprocessing:
 - Stemming and lemmatization: Stemming is an informal process of converting words to their base forms using heuristic rules. For example, “university,” “universities,” and “university’s” might all be mapped to the base universe. (One limitation in this approach is that “universe” may also be mapped to universe, even though universe and university don’t have a close semantic relationship.) Lemmatization is a more formal way to find roots by analyzing a word’s morphology using vocabulary from a dictionary. Stemming and lemmatization are provided by libraries like spaCy and NLTK.
 - Sentence segmentation breaks a large piece of text into linguistically meaningful sentence units. This is obvious in languages like English, where the end of a sentence is marked by a period, but it is still not trivial. A period can be used to mark an abbreviation as well as to terminate a sentence, and in this case, the period should be part of the abbreviation token itself. The process becomes even more complex in languages, such as ancient Chinese, that don’t have a delimiter that marks the end of a sentence.
 - Stop word removal aims to remove the most commonly occurring words that don’t add much information to the text. For example, “the,” “a,” “an,” and so on.
 - Tokenization splits text into individual words and word fragments. The result generally consists of a word index and tokenized text in which words may be represented as numerical tokens for use in various deep learning methods. A method that instructs language models to ignore unimportant tokens can improve efficiency.

IV. RESULTS

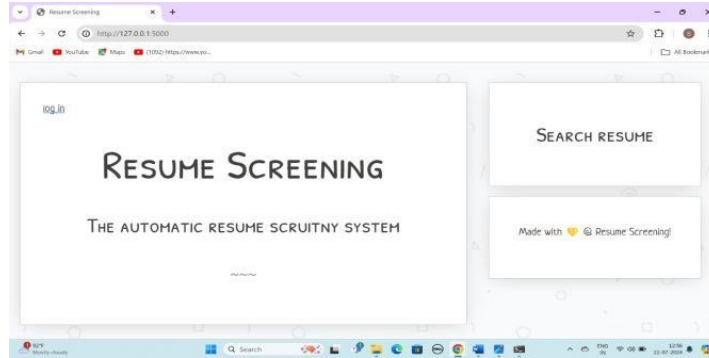


Fig. 2: Home Page

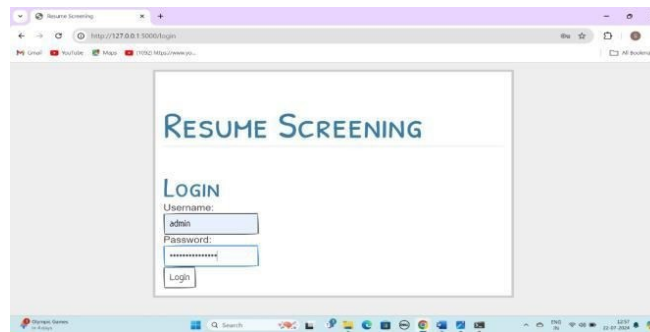


Fig. 3: Login Page

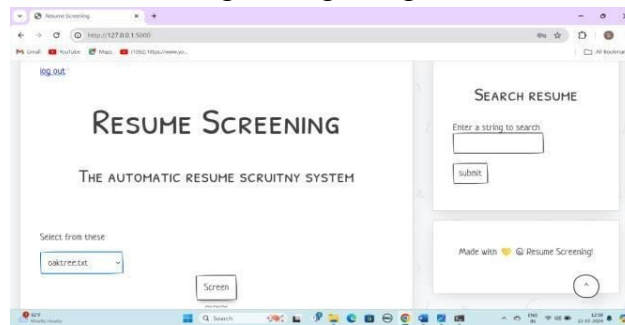


Fig. 4: Select Job Description

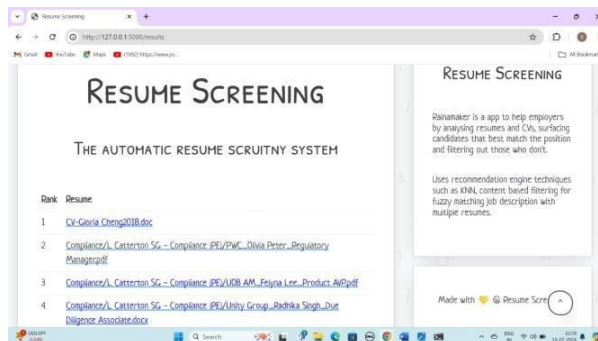


Fig. 5: Screening

V. CONCLUSION

The process of classifying the candidate's resume is manual, time consuming, and waste of resources. To overcome this issue, we have proposed an automated machine learning based model which recommends suitable candidate's resume to the HR based on given job description. The proposed model worked in two phases: first, classify the resume into different categories. Second, recommends resume based on the similarity index with the given job description. The proposed approach effectively captures the resume insights, their semantics and yielded best accuracy.

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