

AI Powered Resume Screening System

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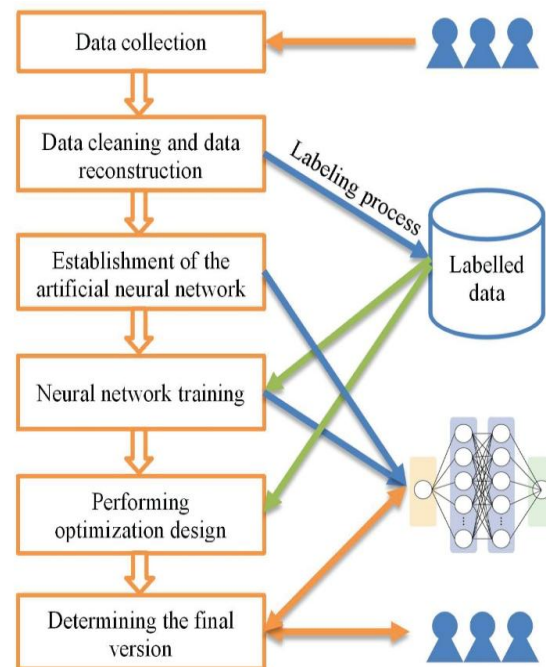
Abstract- Resume screening is the most time-intensive stage of recruitment, especially in large-scale hiring environments. Traditional manual evaluation often results in delays, inconsistency, and unintentional bias. With advancements in Artificial Intelligence (AI) and Natural Language Processing (NLP), automated systems now offer reliable alternatives to manual screening. This paper presents an overview of an AI-powered resume screening system capable of parsing resumes, extracting key candidate information, comparing profiles with job descriptions, and generating ATS-based relevance scores. The system improves fairness, accuracy, and efficiency while reducing recruiter workload. The paper discusses system architecture, algorithms, advantages, limitations, and future scope of AI-driven recruitment.

I. INTRODUCTION

Modern organizations receive hundreds to thousands of applications for a single job role, making resume screening the most critical and time-consuming step of recruitment. Traditional manual screening is prone to human error, evaluator fatigue, and inconsistency. Early Applicant Tracking Systems (ATS) attempted to automate screening but relied primarily on keyword matching, resulting in inaccurate candidate evaluation.

The integration of AI, NLP, and machine learning provides new opportunities for smart hiring tools. AI-based resume screening systems intelligently extract information, understand semantic meaning, and compare resumes with job requirements. This improves the speed and fairness of hiring decisions and optimizes the recruiter's workflow.

This paper provides a structured overview of how AI-powered resume screening systems operate and their significance in modern recruitment.



II. PROBLEM STATEMENT

Organizations struggle with the inefficiency and subjectivity associated with manual resume screening:

High volume of applicants overwhelms recruiters.

Inconsistent judgments occur across different reviewers.

Bias and human subjectivity influence hiring decisions.

Keyword-dependent ATS systems overlook qualified candidates with non-standard resume formats.

Lack of scalability during bulk hiring.

Therefore, there is a need for an intelligent, automated, and fair resume screening system capable of understanding context, reducing human involvement, and improving decision accuracy.

III. OBJECTIVES

The main objectives of the AI-powered resume screening system are:

1. To automatically parse resumes in various formats (PDF, DOCX, TXT).
2. To extract structured information such as skills, education, certifications, and work experience.
3. To match candidate profiles with job descriptions using NLP-based similarity analysis.
4. To generate ATS scores reflecting candidate-job relevance.
5. To reduce manual screening effort and ensure fairness by minimizing human bias.
6. To provide feedback to applicants to improve their resume quality.
7. To offer a scalable and secure system suitable for organizations of all sizes.

IV. LITERATURE REVIEW (SUMMARY)

Existing research highlights increasing adoption of AI in recruitment:

NLP-based resume parsing improves extraction accuracy over traditional ATS.

Machine learning models enable candidate ranking and skill matching.

Deep learning enhances classification and context understanding.

Explainable AI is emerging to improve transparency and reduce bias.

Semantic embeddings (e.g., BERT, MPNet) significantly improve job-resume similarity measurement.

V. SYSTEM ARCHITECTURE

The AI-powered resume screening system typically includes the following layers:

5.1 Input Layer

Resume upload (PDF, DOCX, TXT).

Job description input from recruiter.

5.2 NLP Parsing Layer

Text extraction

Section detection (skills, experience, education)

Named Entity Recognition (NER)

Skill extraction using AI models

5.3 Screening & Scoring Engine

Semantic similarity using embeddings (e.g., MPNet, BERT).

Keyword & skill matching.

Experience evaluation.

Bias detection (optional).

ATS score computation.

5.4 Database Layer

Candidate profiles

Parsed data storage

Job metadata

Screening results

5.5 Interface Layer

Recruiter dashboard

Candidate portals

Higher Accuracy: Semantic similarity ensures better job-resume alignment.

Export to ATS / HRMS

Consistency: Decisions are uniform and unaffected by human fatigue.

6. Ranking

Candidates are sorted from highest to lowest match score.

Improved Fairness: Bias detection reduces discriminatory filtering.

7. Recruiter Review

Dashboard displays ranked results with insights.

Scalability: Can process thousands of resumes simultaneously.

8. Feedback Generation (Optional)

Candidate receives suggestions for improvement.

Security: Authentication and encrypted storage protect candidate data.

VI. METHODOLOGY / ALGORITHM

Recruiters receive clear dashboards, rankings, and analytics, enabling faster hiring decisions.

The resume screening workflow consists of:

VIII. ADVANTAGES

1. Resume Upload

Candidate submits resume or HR uploads bulk resumes.

Reduces manual workload

2. Text Extraction

Parse resume using NLP libraries.

Faster recruitment cycles

3. Data Structuring

Identify and categorize skills, education, experience, projects, etc.

Eliminates human errors and fatigue

4. Semantic Analysis

Compare candidate data with job description using embedding models.

Ensures fair and unbiased evaluation

5. Scoring

Generate ATS score based on:

Scales for large hiring drives

Supports modern HR platforms

Skill match

Improves candidate experience with feedback

Experience relevance

IX. LIMITATIONS

Keyword coverage

May struggle with poorly formatted resumes

Semantic similarity

Requires training data for better accuracy

Initial setup cost may be high

VII. RESULTS AND DISCUSSION

Implementation of the AI-powered system yields significant improvements:

Over-reliance on AI may risk missing unique candidate qualities

Screening Time Reduced: Automated processing reduces analysis time from hours to seconds.

Bias mitigation depends on model quality

X. FUTURE SCOPE

Future enhancements may include:

Multi-language resume parsing

Integration with chatbot-based interviews

Predictive analytics for candidate performance

Personalized job recommendations for candidates

Real-time integration with LinkedIn and recruitment platforms

Voice-based resume summaries

Advanced explainable AI for transparency

XI. CONCLUSION

AI-powered resume screening systems represent a major advancement in recruitment technology. By integrating NLP, machine learning, and semantic analysis, the system reduces manual screening effort, accelerates hiring, and ensures consistency in candidate evaluation. While some limitations remain, the system delivers substantial benefits, making recruitment faster, fairer, and more intelligent. As AI continues to evolve, these systems will become essential tools for modern HR departments and organizations worldwide.

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