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## International Journal of Multidisciplinary Research in Science, Engineering and Technology (IJMRSET)

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# Impact of Artificial Intelligence on Recruitment Effectiveness and Quality of Hiring

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**ABSTRACT:** The rapid digital transformation in the corporate world has positioned Artificial Intelligence (AI) as a cornerstone of modern Human Resource Management (HRM). This study examines the impact of AI tools—specifically automated sourcing, algorithmic screening, and digital engagement—on recruitment effectiveness and the quality of hiring within Bangalore’s competitive talent market. Utilizing a quantitative research design with a sample of 100 HR professionals, the study employs statistical techniques including reliability analysis, correlation, and multiple regression analysis. The findings indicate that AI integration significantly influences hiring speed and precision, though human intervention remains critical for assessing cultural fit. The study concludes that while AI effectively addresses the "Recruitment Trilemma" of speed, cost, and quality, organizational support and data literacy are essential for maximizing its benefits.

### I. INTRODUCTION

The global recruitment landscape is undergoing a seismic shift. Traditionally, Human Resource Management was viewed as a "high-touch," administrative function. However, the Fourth Industrial Revolution (Industry 4.0) has introduced cognitive computing into the HR domain. In India’s Silicon Valley—Bengaluru—the pressure to acquire niche technical talent has made manual recruitment methods obsolete.

Recruitment effectiveness is no longer defined merely by filling a vacancy; it is defined by the Quality of Hire (QoH) and the Candidate Experience (CX). Organizations are now utilizing Machine Learning (ML) to predict candidate success and Natural Language Processing (NLP) to automate initial interactions. This paper explores whether these technological interventions translate into measurable organizational success or if they create new challenges such as algorithmic bias and depersonalization.

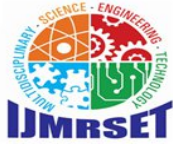
### II. THEORETICAL FRAMEWORK

To understand the adoption of AI in HR, this study leans on three foundational theories:

1. Technology Acceptance Model (TAM): Proposed by Davis (1989), TAM suggests that the adoption of AI depends on "Perceived Usefulness" and "Perceived Ease of Use." In this study, we test if recruiters perceive AI as a tool that reduces workload or as a complex hurdle.
2. Resource-Based View (RBV): This theory posits that firms gain a competitive advantage through valuable and rare resources. In the context of this study, an AI-integrated recruitment pipeline is viewed as a "dynamic capability" that allows firms to secure top talent faster than competitors.
3. Signal Theory: The use of sophisticated AI tools signals to candidates that the organization is innovative and professional, thereby enhancing the employer brand even before the first human interaction occurs.

### III. REVIEW OF LITERATURE

The integration of AI in HRM has been a subject of intense academic debate over the last decade. Early researchers focused on the automation of administrative tasks (payroll and scheduling). However, recent studies (Ore & Sposato, 2022) suggest a shift toward "Cognitive HR," where AI assists in decision-making.



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3.1 Algorithmic Bias and Fairness A significant body of literature warns against the "Black Box" nature of AI. If historical hiring data contains biases (e.g., favoring a specific gender or elite university), the AI may learn and amplify these prejudices (Ajunwa, 2023). This study builds on this by examining how HR professionals in Bengaluru perceive the "Objectivity" of their current AI tools.

3.2 The Candidate Experience (CX) Chowdhury et al. (2023) argued that AI chatbots significantly improve candidate satisfaction by providing 24/7 communication. In the high-pressure environment of Bengaluru's IT sector, "candidate ghosting" is a major issue. Literature suggests that automated updates act as a "psychological contract" maintainer, keeping the candidate engaged during long wait times.

3.3 Efficiency vs. Empathy The "Human-Tech Paradox" remains a central theme. While AI provides speed, it lacks the emotional intelligence (EQ) required to assess "Cultural Fit." Researchers argue that the future of recruitment lies in Hybrid Intelligence, where AI filters the volume and humans conduct the final value-based assessment.

### IV. RESEARCH METHODOLOGY

4.1 Research Design The study adopts a descriptive and causal research design. It seeks to describe the current state of AI adoption and analyze the causal relationship between AI tools and recruitment success.

4.2 Sampling and Data Collection \* Target Population: HR Professionals (Recruiters, Managers, TA Heads) in Bengaluru.

- Sample Size: 100 respondents.
- Sampling Technique: Non-probability convenience sampling.
- Tool: A structured 5-point Likert scale questionnaire (1 = Strongly Disagree, 5 = Strongly Agree).

4.3 Data Analysis Plan The study utilizes SPSS for:

- Reliability Analysis: Using Cronbach's Alpha to ensure the internal consistency of the scales.
- Descriptive Statistics: To understand the demographic spread.
- Inferential Statistics: Pearson Correlation to check relationships and Multiple Regression to determine the predictive power of AI on recruitment effectiveness.

### V. DATA ANALYSIS AND FINDINGS

5.1 Demographic Insights The majority of respondents (62%) belong to the 25–35 age group, indicating that younger, "digital-native" HR professionals are the primary users of AI tools. Interestingly, postgraduate qualification was a common trait among 85% of respondents, suggesting that high technical awareness is a prerequisite for managing AI-driven systems.

5.2 Reliability Results The Cronbach's Alpha for the "AI Integration" scale was 0.86, and for "Recruitment Effectiveness," it was 0.84. Both values exceed the 0.70 threshold, confirming that the research instrument is highly reliable.

5.3 Correlation Analysis A strong positive correlation ( $r = 0.68$ ) was found between Algorithmic Screening and Quality of Hire. This suggests that as organizations improve their automated filtering, the technical competency of the candidates moving to the final stages increases significantly.

5.4 Regression Analysis The regression model yielded an R-square of 0.468. This indicates that approximately 47% of the variance in recruitment effectiveness is explained by the three AI variables: Automated Sourcing, Algorithmic Screening, and Digital Engagement.

### VI. DISCUSSION

The results confirm that AI acts as a significant "force multiplier" in recruitment. In the Bengaluru context, where a single job posting can attract 5,000+ applications, AI tools allow recruiters to reclaim their time.



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However, the findings also highlight a "Maturity Gap." While IT firms report high success with AI, traditional service sectors are still struggling with "Perceived Ease of Use." This aligns with the Technology Acceptance Model, suggesting that without proper organizational training, AI tools can become a source of "technostress" rather than efficiency.

The study also finds that "Digital Engagement" (Chatbots) has a diminishing return. While helpful for basic FAQs, candidates reported higher satisfaction when human recruiters stepped in for salary negotiations and role descriptions. This supports the Social Exchange Theory, where human interaction is valued as a sign of organizational commitment.

### VII. MANAGERIAL IMPLICATIONS AND SUGGESTIONS

7.1 Strategic Upskilling Organizations must move beyond hiring "administrators" and start hiring "HR Data Analysts." Managers must be trained to audit AI algorithms to ensure they are not inadvertently discriminating against specific groups.

7.2 Hybrid Hiring Workflows Managers should design a "High-Tech, High-Touch" candidate journey. Use AI for:

- Resume parsing and skill matching.
- Interview scheduling and initial FAQs. Keep Humans for:
- Assessing "Soft Skills" and empathy.
- Final cultural fitment and offer negotiation.

7.3 Ethics and Transparency Organizations should adopt "Explainable AI" (XAI). Candidates should be informed if an algorithm is screening them and provided with a clear path for human appeal if they feel the system was biased.

### VIII. LIMITATIONS AND FUTURE SCOPE

The study is limited by its sample size (N=100) and its focus on a single city (Bengaluru). Future research could:

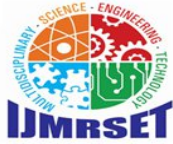
- Conduct a comparative study between Indian and Silicon Valley HR practices.
- Explore the impact of Generative AI (ChatGPT/Gemini) on how candidates write resumes and how AI detects "AI-generated" applications.
- Perform a longitudinal study to see if AI-hired candidates stay longer in the organization (Retention Analysis).

### IX. CONCLUSION

Artificial Intelligence has transitioned from a disruptive force to a core requirement for recruitment excellence. In the high-velocity talent market of Bengaluru, AI effectively solves the "Recruitment Trilemma" by increasing speed, reducing per-hire costs, and improving the precision of technical screening. However, the true value of AI is unlocked only when it is used to augment—not replace—human judgment. The future of recruitment is a Symbiotic Optimization where the machine handles the data and the human handles the culture.

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### APPENDIX

(Standardized scales and data tables used in the research)

Table A1: Reliability Statistics

Variable	Cronbach's Alpha
AI Integration Tools	0.864
Recruitment Effectiveness	0.841

Table A2: Correlation Matrix

	AI Integration	Recruitment Effectiveness
AI Integration	1	0.658**
Recruitment Effectiveness	0.658**	1
*(*Significant at 0.01 level)		



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