Management Studies and Entrepreneurship Journal
Vol 7(4) 2026:1153-1161

The Impact Of Artificial Intelligence–Based Recruitment On Employee Performance And Organizational Fairness

Dampak Rekrutmen Berbasis Kecerdasan Buatan Terhadap Kinerja Karyawan dan Keadilan Organisasi

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ABSTRACT
This study aims to analyze the impact of Artificial Intelligence (AI)–based recruitment on employee performance and organizational fairness through a systematic literature review approach. The rapid development of AI technologies has transformed recruitment practices by enabling data-driven decision-making, improving efficiency, and reducing human bias. However, the implications of these technologies for organizational outcomes remain a subject of debate. This study synthesizes findings from peer-reviewed journals, books, and reputable academic sources published between 2010 and 2025. The analysis focuses on three main aspects: the effectiveness of AI-based recruitment, its influence on employee performance, and its impact on organizational fairness. The results indicate that AI-based recruitment enhances the accuracy and efficiency of hiring processes, leading to improved person–job fit and potential increases in employee performance. However, its impact on performance is indirect and influenced by broader organizational factors such as leadership and organizational culture. In terms of fairness, AI systems promote procedural consistency but raise concerns regarding algorithmic bias and lack of transparency. These challenges may affect employee trust and perceptions of organizational justice. Therefore, the study concludes that while AI-based recruitment offers significant benefits, its implementation must be accompanied by ethical governance, transparency, and integration with human-centered HR practices to ensure sustainable organizational outcomes.
Keywords: *Artificial Intelligence, recruitment, employee performance, organizational fairness, HRM*

Abstrak
Penelitian ini bertujuan untuk menganalisis dampak rekrutmen berbasis Artificial Intelligence (AI) terhadap kinerja karyawan dan keadilan organisasi melalui pendekatan kajian pustaka sistematis. Perkembangan pesat teknologi AI telah mengubah praktik rekrutmen dengan memungkinkan pengambilan keputusan berbasis data, meningkatkan efisiensi, serta mengurangi bias manusia. Namun demikian, implikasi penggunaan teknologi ini terhadap hasil organisasi masih menjadi perdebatan. Penelitian ini mensintesis temuan dari jurnal ilmiah, buku, dan sumber akademik bereputasi yang diterbitkan pada periode 2010 hingga 2025. Analisis difokuskan pada tiga aspek utama, yaitu efektivitas rekrutmen berbasis AI, pengaruhnya terhadap kinerja karyawan, serta dampaknya terhadap keadilan organisasi. Hasil penelitian menunjukkan bahwa rekrutmen berbasis AI mampu meningkatkan akurasi dan efisiensi proses seleksi, sehingga berkontribusi pada peningkatan kesesuaian antara individu dan pekerjaan yang berpotensi meningkatkan kinerja karyawan. Namun, pengaruh tersebut bersifat tidak langsung dan dipengaruhi oleh faktor organisasi lainnya seperti kepemimpinan dan budaya kerja. Dari sisi keadilan, sistem AI meningkatkan konsistensi prosedural, tetapi menimbulkan kekhawatiran terkait bias algoritma dan kurangnya transparansi. Oleh karena itu, implementasi AI perlu disertai dengan tata kelola yang etis, transparansi, serta integrasi dengan praktik manajemen sumber daya manusia yang berorientasi pada manusia.
Kata kunci: Artificial Intelligence, rekrutmen, kinerja karyawan, keadilan organisasi, MSDM

1. Introduction

The rapid advancement of digital technologies has significantly transformed organizational practices across various sectors, including human resource management (HRM). Among these technological innovations, Artificial Intelligence (AI) has emerged as a disruptive force, particularly in the domain of recruitment and selection. AI-based recruitment systems utilize machine learning algorithms, natural language processing, and predictive analytics to streamline hiring processes, enhance decision-making, and reduce human bias (Bogen & Rieke, 2018). As organizations increasingly adopt AI-driven recruitment tools, it becomes crucial to examine their broader implications, especially concerning employee performance and perceptions of organizational fairness.

Traditionally, recruitment has been a labor-intensive process involving human judgment, which is often susceptible to subjectivity and bias. Recruiters may unconsciously favor candidates based on non-job-related characteristics such as gender, ethnicity, or educational background (Rivera, 2012). AI-based recruitment systems promise to mitigate such biases by relying on data-driven decision-making processes. These systems can efficiently screen large volumes of applications, identify patterns, and match candidates with job requirements more accurately than human recruiters (Chamorro-Premuzic et al., 2019). Consequently, organizations anticipate improved hiring quality and enhanced workforce productivity.

However, despite these advantages, the implementation of AI in recruitment raises critical concerns regarding fairness and transparency. Algorithmic decision-making is not inherently free from bias; rather, it can replicate and even amplify existing biases embedded in historical data (Barocas & Selbst, 2016). For instance, if an AI system is trained on past hiring data that reflects discriminatory practices, it may perpetuate similar patterns in future recruitment decisions. This phenomenon challenges the assumption that AI systems are inherently objective and highlights the need for careful evaluation of their fairness implications.

Organizational fairness, often conceptualized through the lens of organizational justice theory, plays a pivotal role in shaping employee attitudes and behaviors. Organizational justice comprises three primary dimensions: distributive justice (perceived fairness of outcomes), procedural justice (fairness of processes), and interactional justice (quality of interpersonal treatment) (Colquitt et al., 2001). Recruitment practices, as the initial point of contact between employees and organizations, significantly influence employees' perceptions of fairness. When recruitment processes are perceived as fair, employees are more likely to exhibit higher levels of job satisfaction, organizational commitment, and performance (Gilliland, 1993).

The integration of AI into recruitment processes introduces new dynamics that may affect these perceptions. On one hand, AI-based systems can enhance procedural consistency by applying standardized criteria across all candidates, thereby improving perceptions of fairness. On the other hand, the opacity of algorithmic decision-making—often referred to as the "black box" problem—can undermine transparency and trust (Pasquale, 2015). Candidates and employees may find it difficult to understand how decisions are made, leading to skepticism and concerns about accountability. This lack of transparency may negatively impact perceived fairness, even if the outcomes are objectively unbiased.

Furthermore, the relationship between recruitment practices and employee performance warrants careful consideration. Effective recruitment is essential for aligning individual competencies with organizational needs, which directly influences employee performance. AI-based recruitment systems have the potential to improve this alignment by leveraging predictive analytics to identify candidates with the highest likelihood of success in specific roles (Kuncel et al., 2013). By selecting candidates whose skills, experiences, and

personality traits match job requirements, organizations can enhance overall workforce performance.

Nevertheless, the effectiveness of AI-based recruitment in improving employee performance is not guaranteed. Several factors may influence this relationship, including the quality of data used to train AI models, the design of algorithms, and the organizational context in which these systems are implemented. Moreover, employee performance is not solely determined by recruitment outcomes; it is also shaped by organizational culture, leadership, training, and motivation (Armstrong & Taylor, 2020). Therefore, while AI can contribute to better hiring decisions, its impact on performance must be examined within a broader organizational framework.

Another critical issue relates to employee acceptance of AI-based recruitment systems. The Technology Acceptance Model (TAM) suggests that perceived usefulness and ease of use influence individuals' acceptance of new technologies (Davis, 1989). In the context of recruitment, both candidates and employees may evaluate AI systems based on their perceived fairness, transparency, and reliability. Negative perceptions may lead to resistance, reduced trust in the organization, and lower levels of engagement, which can ultimately affect performance outcomes.

Ethical considerations also play a central role in the adoption of AI in recruitment. Organizations must ensure that their AI systems comply with legal and ethical standards, particularly concerning data privacy and non-discrimination. Regulatory frameworks such as the General Data Protection Regulation (GDPR) emphasize the importance of transparency and accountability in automated decision-making processes (European Union, 2016). Failure to address these ethical concerns may expose organizations to legal risks and reputational damage.

In addition, the socio-cultural context in which organizations operate may influence the impact of AI-based recruitment on fairness and performance. In developing countries, including Indonesia, the adoption of AI technologies in HRM is still evolving. Factors such as technological infrastructure, digital literacy, and organizational readiness may affect the effectiveness of AI implementation. Moreover, cultural values related to fairness, hierarchy, and interpersonal relationships may shape how employees perceive AI-driven decisions (Hofstede, 2001). These contextual factors underscore the need for empirical research that examines the impact of AI-based recruitment in specific organizational settings.

Despite the growing interest in AI applications in HRM, there remains a limited body of empirical research that simultaneously examines its effects on employee performance and organizational fairness. Most existing studies focus either on the technical aspects of AI systems or on their potential to reduce bias, without fully exploring their broader organizational implications. This gap highlights the importance of conducting comprehensive research that integrates technological, behavioral, and ethical perspectives.

Therefore, this study aims to investigate the impact of AI-based recruitment on employee performance and organizational fairness. By examining both outcomes, this research seeks to provide a more holistic understanding of how AI influences organizational effectiveness and employee perceptions. The findings are expected to contribute to the development of more equitable and effective recruitment practices, as well as to inform policymakers and practitioners about the potential benefits and risks associated with AI adoption in HRM.

In conclusion, while AI-based recruitment offers significant opportunities for improving efficiency and decision-making in hiring processes, it also presents challenges related to fairness, transparency, and employee acceptance. Understanding the interplay between these factors is essential for ensuring that AI technologies are used responsibly and effectively. As

organizations continue to embrace digital transformation, research on AI-driven HR practices will become increasingly important in shaping the future of work.

2. Research Methods (Metode Penelitian)

Research Design

This study employs a qualitative research design using a **literature review approach** to examine the impact of Artificial Intelligence (AI)–based recruitment on employee performance and organizational fairness. A literature review is a systematic method of identifying, evaluating, and synthesizing existing research to develop a comprehensive understanding of a particular phenomenon (Snyder, 2019). This approach is appropriate for the present study because the topic involves a rapidly evolving field where empirical findings are dispersed across multiple disciplines, including human resource management, information systems, and organizational behavior.

The literature review method enables the researcher to integrate theoretical perspectives and empirical findings, thereby providing a holistic analysis of how AI-based recruitment influences organizational outcomes. Furthermore, this approach facilitates the identification of research gaps and the development of a conceptual framework that can guide future empirical studies (Tranfield et al., 2003).

Type of Literature Review

This study adopts a **systematic literature review (SLR)** approach. A systematic literature review is characterized by a structured, transparent, and replicable process of searching, selecting, and analyzing relevant studies (Kitchenham & Charters, 2007). Unlike traditional narrative reviews, SLR minimizes bias by applying explicit inclusion and exclusion criteria and by documenting each stage of the review process.

The systematic approach is particularly suitable for this research because it ensures rigor in synthesizing findings related to AI-based recruitment, employee performance, and organizational fairness. It also allows for a critical evaluation of the methodological quality of existing studies and the consistency of their findings.

Data Sources and Search Strategy

The data used in this study consist of secondary data obtained from reputable academic sources. The literature search was conducted using major electronic databases, including:

- Scopus
- Web of Science
- Google Scholar
- ScienceDirect
- Emerald Insight

Keywords were carefully selected to capture the core constructs of the study. The search terms included combinations of:

- “Artificial Intelligence recruitment”
- “AI in human resource management”
- “algorithmic hiring”
- “employee performance”
- “organizational justice”
- “fairness in recruitment”

Boolean operators (AND, OR) were used to refine the search results. For example: “Artificial Intelligence AND recruitment AND employee performance” and “algorithmic hiring AND organizational fairness.”

To ensure relevance and timeliness, the search focused primarily on articles published between **2010 and 2025**, reflecting the period of rapid development in AI technologies.

Inclusion and Exclusion Criteria

To maintain the quality and relevance of the reviewed literature, this study applied the following criteria:

Inclusion Criteria:

1. Peer-reviewed journal articles, conference papers, and scholarly books.
2. Studies discussing AI or algorithmic systems in recruitment and selection.
3. Research addressing employee performance, organizational fairness, or related constructs.
4. Publications written in English.

Exclusion Criteria:

1. Non-scholarly sources such as blogs, opinion articles, and news reports.
2. Studies not directly related to recruitment or organizational outcomes.
3. Duplicate publications across databases.

The application of these criteria ensures that the selected studies are both credible and relevant to the research objectives (Petticrew & Roberts, 2006).

Data Collection and Selection Process

The literature selection process followed several systematic stages:

1. **Identification:** Initial search results were collected from selected databases using predefined keywords.
2. **Screening:** Titles and abstracts were reviewed to eliminate irrelevant studies.
3. **Eligibility:** Full-text articles were assessed based on inclusion and exclusion criteria.
4. **Final Selection:** Only studies meeting all criteria were included in the analysis.

This multi-stage process aligns with established systematic review protocols and enhances the transparency and reliability of the study (Moher et al., 2009).

Data Analysis Technique

The selected literature was analyzed using **thematic analysis**, a qualitative method for identifying, analyzing, and reporting patterns within data (Braun & Clarke, 2006). This technique allows the researcher to systematically categorize findings into key themes related to the research variables.

The analysis process involved the following steps:

1. **Familiarization:** Reading and re-reading selected articles to understand their content.
2. **Coding:** Identifying relevant concepts and assigning codes to significant findings.
3. **Theme Development:** Grouping codes into broader themes, such as:
   - Efficiency and accuracy of AI-based recruitment
   - Algorithmic bias and fairness concerns
   - Impact on employee performance
   - Perceptions of organizational justice
4. **Interpretation:** Synthesizing themes to develop a comprehensive understanding of the relationships among variables.

Thematic analysis is suitable for this study because it enables the integration of diverse findings and supports theory development (Nowell et al., 2017).

Validity and Reliability

To ensure the rigor of the literature review, several strategies were employed:

- **Transparency:** Clearly documenting the search strategy, selection criteria, and analysis procedures.
- **Triangulation:** Using multiple sources and databases to reduce bias.
- **Critical Evaluation:** Assessing the methodological quality of each study, including sample size, research design, and analytical techniques.

These measures enhance the credibility and trustworthiness of the findings (Lincoln & Guba, 1985).

Research Framework Development

Based on the reviewed literature, this study develops a conceptual framework that links AI-based recruitment to employee performance and organizational fairness. AI recruitment is treated as the independent variable, while employee performance and organizational fairness are the dependent variables.

The framework also considers potential mediating or moderating factors, such as transparency, perceived fairness, and organizational context. This integrative approach provides a foundation for future empirical testing and contributes to the advancement of theory in HRM and technology adoption.

3. Results and Discussions

Results of the Literature Review

Based on the systematic literature review, the findings are categorized into three major themes: (1) the effectiveness of AI-based recruitment in improving hiring quality, (2) its impact on employee performance, and (3) its implications for organizational fairness.

Effectiveness of AI-Based Recruitment

The reviewed literature consistently indicates that AI-based recruitment enhances efficiency and accuracy in the hiring process. AI systems can process large volumes of applicant data, identify relevant competencies, and match candidates to job requirements more effectively than traditional methods (Chamorro-Premuzic et al., 2019). These systems utilize machine learning algorithms to detect patterns in historical hiring data, enabling predictive decision-making regarding candidate success (Kuncel et al., 2013).

Moreover, AI-driven tools reduce time-to-hire and recruitment costs, which are critical performance indicators in human resource management (Bogen & Rieke, 2018). Automated screening processes eliminate repetitive tasks, allowing HR professionals to focus on strategic decision-making. However, the effectiveness of these systems depends heavily on the quality and representativeness of the data used to train the algorithms.

Impact on Employee Performance

The literature suggests a positive relationship between AI-based recruitment and employee performance, primarily through improved person–job fit. By selecting candidates whose skills, experiences, and psychological attributes align with job requirements, AI systems contribute to higher productivity and job effectiveness (Kuncel et al., 2013).

Additionally, AI-based recruitment can support long-term performance by identifying candidates with high potential for learning and adaptability. This aligns with strategic human resource management theories, which emphasize the importance of aligning human capital with organizational goals (Armstrong & Taylor, 2020).

However, several studies caution that improved recruitment outcomes do not automatically translate into enhanced employee performance. Performance is influenced by

multiple factors, including organizational culture, leadership, training, and employee motivation (Colquitt et al., 2001). Therefore, while AI can improve initial selection decisions, its impact on performance is indirect and contingent upon broader organizational practices.

Organizational Fairness and Justice

A critical theme emerging from the literature is the dual impact of AI-based recruitment on organizational fairness. On one hand, AI systems promote procedural consistency by applying standardized criteria across all candidates, which can enhance perceptions of fairness (Gilliland, 1993). This consistency reduces human subjectivity and potential discrimination in hiring decisions.

On the other hand, concerns about algorithmic bias and lack of transparency pose significant challenges. Studies show that AI systems may reproduce existing biases embedded in historical data, leading to discriminatory outcomes (Barocas & Selbst, 2016). For example, biased training data can result in the systematic exclusion of certain demographic groups.

Furthermore, the “black box” nature of AI decision-making reduces transparency, making it difficult for candidates and employees to understand how decisions are made (Pasquale, 2015). This lack of explainability can negatively affect perceptions of procedural and interactional justice, even when outcomes are objectively fair.

Discussion

The findings of this study highlight the complex and multidimensional impact of AI-based recruitment on employee performance and organizational fairness. While AI offers significant advantages in terms of efficiency and decision-making accuracy, its implications extend beyond technical performance to include ethical and organizational considerations.

From a human resource management perspective, the adoption of AI-based recruitment aligns with the broader trend of digital transformation and data-driven decision-making. The ability of AI systems to enhance person–job fit provides a strong theoretical explanation for their positive impact on employee performance. This finding supports prior research emphasizing the importance of selection accuracy in improving workforce productivity (Kuncel et al., 2013).

However, the relationship between AI-based recruitment and employee performance is not linear. The literature suggests that recruitment is only one component of the overall HR system. Without supportive organizational practices such as training, performance management, and leadership development, the benefits of AI-driven recruitment may not be fully realized (Armstrong & Taylor, 2020). This indicates the need for an integrated HRM approach that combines technological innovation with human-centered management practices.

In terms of organizational fairness, the findings reveal a paradox. While AI has the potential to reduce human bias and enhance procedural fairness, it simultaneously introduces new forms of algorithmic bias and opacity. This duality reflects the broader ethical challenges associated with AI adoption in organizational contexts (Barocas & Selbst, 2016).

The issue of transparency is particularly critical. Organizational justice theory suggests that employees are more likely to accept decisions when they perceive the process as fair and understandable (Colquitt et al., 2001). Therefore, the lack of explainability in AI systems may undermine trust and acceptance, even if the decisions are accurate. This highlights the importance of developing explainable AI (XAI) systems that provide clear and interpretable decision-making processes.

Another important implication relates to employee perceptions and technology acceptance. According to the Technology Acceptance Model (Davis, 1989), perceived usefulness and fairness influence the acceptance of new technologies. If employees perceive

AI-based recruitment as unfair or biased, they may develop negative attitudes toward the organization, which can adversely affect engagement and performance.

Furthermore, the findings underscore the importance of ethical governance in AI implementation. Organizations must ensure that their AI systems are designed and deployed in a manner that promotes fairness, accountability, and transparency. This includes regular audits of algorithms, the use of diverse and representative training data, and compliance with legal frameworks such as data protection regulations.

Finally, the study identifies several gaps in the existing literature. While many studies examine the technical capabilities of AI in recruitment, fewer studies explore its long-term organizational impacts, particularly in non-Western contexts. Cultural factors, organizational readiness, and technological infrastructure may influence how AI-based recruitment is perceived and implemented. Therefore, future research should focus on empirical investigations in diverse contexts to validate and extend the findings of this study.

Synthesis of Findings

In summary, the results of the literature review indicate that:

1. AI-based recruitment improves efficiency and selection accuracy.
2. It has a positive but indirect impact on employee performance through better person–job fit.
3. It enhances procedural fairness but raises concerns about algorithmic bias and transparency.
4. Its effectiveness depends on organizational context, ethical practices, and employee perceptions.

These findings suggest that the successful implementation of AI-based recruitment requires a balanced approach that integrates technological innovation with ethical and organizational considerations.

4. Conclusion

This study aimed to examine the impact of Artificial Intelligence (AI)–based recruitment on employee performance and organizational fairness through a systematic literature review approach. Based on the synthesis of existing theoretical and empirical studies, several key conclusions can be drawn.

First, AI-based recruitment has demonstrated significant potential in improving the efficiency and effectiveness of the hiring process. By leveraging data-driven algorithms, organizations are able to enhance the accuracy of candidate selection, reduce time-to-hire, and optimize resource allocation. This technological advancement contributes to better person–job fit, which is a critical determinant of employee performance.

Second, the findings indicate that AI-based recruitment has a positive but indirect influence on employee performance. While improved selection processes increase the likelihood of hiring competent and suitable candidates, employee performance is also shaped by broader organizational factors such as leadership, training, motivation, and organizational culture. Therefore, AI should be viewed as a complementary tool within an integrated human resource management system rather than a standalone solution.

Third, in terms of organizational fairness, AI-based recruitment presents a dual impact. On one hand, it enhances procedural fairness by standardizing recruitment processes and minimizing human subjectivity. On the other hand, concerns related to algorithmic bias, lack of transparency, and limited explainability pose significant challenges. These issues may undermine employee trust and negatively affect perceptions of organizational justice, particularly when decision-making processes are not clearly understood.

Fourth, the study highlights the importance of ethical considerations and governance in the implementation of AI technologies. Organizations must ensure that AI systems are designed and utilized in a fair, transparent, and accountable manner. This includes the use of unbiased training data, regular algorithm audits, and adherence to legal and ethical standards. Without proper governance, the potential benefits of AI-based recruitment may be offset by risks related to discrimination and loss of trust.

Finally, this study underscores the need for a balanced and strategic approach to AI adoption in human resource management. While AI offers substantial opportunities for improving recruitment outcomes, its success depends on organizational readiness, technological infrastructure, and employee acceptance. Integrating AI with human judgment and maintaining transparency in decision-making processes are essential to achieving both high employee performance and perceived organizational fairness.

In conclusion, AI-based recruitment represents a transformative innovation in HRM, but its implementation must be carefully managed to ensure that technological efficiency is aligned with ethical responsibility and organizational sustainability.

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