

Artificial Intelligence in the Workplace and Employee Wellbeing: A Biblioshiny-Driven Bibliometric Exploration

Nurkaliza Khalid*, Rafiza Kasbun, Siti Zaharah Mohid, Noor Fadzilah Abd Rahman
Faculty of Creative Multimedia and Computing (FMKK), Selangor Islamic University (UIS),
Selangor, Malaysia

nurkaliza@uis.edu.my, rafiza@uis.edu.my, sitizaharah@uis.edu.my, noorfadzilah@uis.edu.my

Abstract

Artificial intelligence (AI) is increasingly integrated into workplaces, shaping recruitment, performance management, task automation, and health monitoring. While AI adoption offers potential benefits such as reduced routine workload, improved engagement, and enhanced wellbeing, it also raises concerns about stress, job insecurity, and ethical risks. Despite growing attention, the literature on AI and employee wellbeing remains fragmented, with limited systematic efforts to map its development. This study applies a bibliometric analysis of publications indexed in Scopus between 2015 and 2025. Using Bibliometrix and Biblioshiny, the analysis examines trends in publication output, influential authors, journals, institutions, and countries, as well as thematic patterns in the field. Findings indicate a sharp growth in research output, with a small number of highly cited studies exerting disproportionate influence. Institutional and geographical analysis shows concentration in leading universities and research clusters, while keyword mapping reveals persistent focus on AI and human-centred issues. The study contributes by providing an integrated overview of how research in this area has evolved and where gaps remain. Future research should extend beyond fragmented case studies and explore contextual conditions that shape whether AI supports or undermines employee wellbeing.

Keywords: *Artificial intelligence, Bibliometric Analysis, Employee wellbeing, Workplace Technology.*

1. Introduction

The integration of artificial intelligence (AI) into the workplace has become one of the defining changes in contemporary employment. Organisations increasingly adopt AI-driven tools for recruitment, performance evaluation, task automation, and decision support. These changes go beyond efficiency, raising critical questions about their effects on employees' wellbeing. Existing studies highlight both positive and negative outcomes. On one side, AI reduces repetitive workloads, supports decision-making, and enables new forms of professional growth, contributing to greater engagement and satisfaction. On the other, it introduces risks of technostress, job insecurity, and surveillance, often eroding trust and creating uncertainty about future employment (Meijerink et al., 2021).

Despite this growing interest, the literature remains fragmented. Researchers such as Čaić et al. (2019) and Meijerink et al. (2021) argued that most studies are industry specific and focus on narrow outcomes, eventually making it difficult to draw general conclusions. Positive outcomes are often studied in isolation, while risks are analysed without considering organisational or cultural contexts. Recent work stresses that AI's impact depends on management practices, industry, and policy frameworks, but there has been little effort to synthesise findings systematically. As a result, the academic conversation lacks an integrated picture of how this area has developed, who the leading contributors are, and which themes dominate the discussion.

This study addresses these gaps by applying a bibliometric analysis of publications from 2015 to 2025, with a focus on AI, workforce transformation, and employee wellbeing. Rather than providing a purely descriptive overview, the analysis examines patterns of growth, influence, and thematic focus. The study is guided by two questions:

RQ1: How has research on AI and employee wellbeing developed over time in terms of publication output, influence, and thematic focus?

RQ2: Which authors, journals, institutions, and countries have been most influential in shaping this field, and what does this reveal about its intellectual and geographical structure?

By addressing these questions, the study contributes to clarifying the state of knowledge, identifying areas of consensus and divergence, and highlighting directions for future research on AI and employee wellbeing.

2. Literature Review

2.1 Positive Impacts of AI on Employee Wellbeing

A growing body of research highlights how AI can support employees by reducing administrative workloads, enhancing productivity, and enabling new forms of learning and development. Dabirian et al. (2017) emphasise AI's role as an augmentation tool that complements rather than replaces human effort. In customer service, Prentice & Nguyen (2020) show that AI-based systems improve decision quality and performance. In healthcare, Doraiswamy et al. (2020) demonstrates that AI-assisted diagnostics reduce clinician strain while improving patient outcomes. Wang et al (2019) extend this argument by showing that hybrid intelligence fosters innovation and increases employee satisfaction in decision-making.

AI has also been applied directly to wellbeing support. Adaptive learning platforms promote professional growth, consistent with García-Martínez et al. (2023), while Bankins et al. (2024) argue that AI-assisted job redesign strengthens engagement by aligning tasks with employees' skills. Together, these studies show that AI, when deployed thoughtfully, can enhance job satisfaction, learning, and wellbeing.

2.2 Risks and Challenges of AI in the Workplace

Despite these benefits, significant risks accompany AI adoption. Automation threatens job stability, particularly for routine roles. Badri et al. (2018) argue that widespread AI-driven automation displaces workers, while Lee et al. (2022) and Nazareno & Schiff (2021) report that such displacement drives anxiety and insecurity. Brougham and Haar (2020) caution that while reskilling is possible, access is unequal, potentially worsening workforce inequality.

Digital strain also emerges as a recurring concern. Cheng et al. (2022) find that AI-enabled monitoring increases cognitive load, encourages constant performance pressure, and blurs boundaries between personal and professional life. Surveillance applications raise ethical questions, with Meijerink et al. (2021) and McClure (2017) documenting biases in AI recruitment tools and Flores-Viva et al. (2023) warning about declining organisational trust when AI is linked to invasive monitoring. Together, these findings show that AI can erode wellbeing through stress, job insecurity, and reduced autonomy.

2.3 Contextual and Integrative Perspectives

More recent scholarship stresses that AI's impact on wellbeing is shaped by organisational, cultural, and policy contexts. Brougham & Haar (2020) show that outcomes vary across industries and depend on management practices. Khogali & Mekid (2023) argue that AI contributes positively when embedded within ecosystems that promote upskilling, ethical practices, and employee voice.

These findings suggest that AI's effects are not universal but contingent on the conditions under which it is implemented.

Overall, two perspectives dominate the literature. One highlights AI's potential to enhance productivity, learning, and wellbeing; the other underscores risks related to stress, insecurity, and ethics. Current research remains fragmented, with positive and negative outcomes often studied separately. This gap underscores the need for integrative approaches that evaluate both sides systematically, offering guidance for organisations on how to implement AI in ways that balance innovation with employee wellbeing.

3. Methodology

3.1 Data Collection

Bibliometric data were retrieved from Scopus, which was selected for its comprehensive coverage of peer-reviewed journals across management, psychology, and information systems. The search combined keywords related to “artificial intelligence,” “work,” “workplace,” and “employee wellbeing.” Only English-language publications between 2015 and 2025 were included. The final dataset consisted of full records with cited references, exported in CSV format.

3.2 Dataset Overview

The bibliometric search returned 568 publications on AI and employee wellbeing, published between 2015 and 2025. Annual output has grown steadily, with a notable rise after 2019, reflecting the accelerated integration of AI tools in workplaces and the parallel growth of scholarly attention. The dataset covers contributions from multiple disciplines, including management, computer science, psychology, and sociology, which illustrates the interdisciplinary nature of the topic.

3.3 Analytical Approach

Analyses were carried out using the Bibliometrix R package (Aria & Cuccurullo, 2017) and its Biblioshiny web interface. The study applied descriptive bibliometrics to track publication trends, citation patterns, and influential sources. Network analysis was used to examine co-authorship, institutional collaboration, and keyword co-occurrence, providing insight into the intellectual and social structure of the field.

3.4 Validity, Reliability, and Limitations

Reliability was strengthened by using a widely adopted bibliometric tool, reporting parameter settings transparently, and checking stability across different configurations. Limitations include reliance on a single database (Scopus), which may exclude regional studies, and restricting publications to English, which introduces language bias. Bibliometric methods capture structural and thematic patterns but do not provide deep conceptual interpretation, which may require complementary qualitative review.

4. Result

4.1 Publication Trends

Between 2015 and 2025, research on AI and employee wellbeing exhibited a consistent upward trajectory. The number of publications expanded from three articles in 2015 to a peak of 237 in 2025, reflecting the growing scholarly attention to this field (Figure 1a and 1b). Citation dynamics followed a different pattern, with the highest average citations per year recorded in 2018, reaching a mean of 13.21. The second peak for citation dynamics was observed in 2020. The average citations per year for 2020 is 7.24. These peaks were largely driven by a small group of highly cited

publications that introduced influential frameworks on AI in the workplace which shaped the subsequent research and attracting sustained attention (Figure 1c).

4.2 Productive Authors, Journals, and Institutions

The analysis in Figure 1d) identified Mg Herlina Maria Grace and several unidentified authors as the most productive contributors, each publishing 4 articles in the field. This is followed by AM Sherrill Andrew MB, B Kim Byungjik, CW Wiese Christopher William, G Manoharan Geetha, Zhao Hairong and J Lee Julak with 3 publications each. These authors have played a central role in shaping research on AI implementation and its impact on employee wellbeing. The journals with the highest number of publications were Lecture Notes in Computer Science, International Journal of Environmental Research and Public Health, and Lecture Notes in Network and System (Figure 2a)). These outlets serve as primary platforms for research on AI in organisational settings and employee outcomes.

The institutional analysis in Figure 2b) revealed that research activity was concentrated in a small set of universities. The Lovely Professional University, Amity University, and Bina Nusantara University emerged as the leading contributors in terms of publication output, each producing more than 13 articles during the period of 2015 to 2025. As such, these institutions not only generated substantial output but also played a central role in collaborative networks by linking with partners from countries across multiple regions. Meanwhile, geographically, the majority of publications originated from India, USA, and China, which together accounted for more than half of the total output. The collaboration world map (Figure 2d)) showed strong cross-country partnerships, with particularly dense links between India and USA. Emerging contributions from countries such as Argentina, Azerbaijan, Cameroon, Chile, Colombia, Croatia, Cyprus, Israel, Monaco, Peru and Philippines (Figure 3a)) suggest that the field is gaining traction in developing regions, even though their outputs remain comparatively smaller.

4.3 Conceptual Structure and Thematic Patterns

Keyword co-occurrence analysis (Figure 3b)) revealed a relatively simple conceptual structure, with two dominant clusters. The first cluster centred on “artificial intelligence”, linking terms such as automation, algorithms, and digital transformation. The second cluster was organised around “human”, and included terms associated with wellbeing, stress, engagement, and organisational outcomes. The limited number of clusters suggests that research in this area remains thematically concentrated, with most studies framing discussions around the interaction between AI technologies and human factors at work. This highlights the continuing emphasis on balancing technological adoption with employee wellbeing in scholarly discourse.

4.4 Collaboration Networks

Author-level collaboration networks (Figure 3c)) revealed several tightly knit research groups, often clustered around leading institutions (Figure 3d)). International collaboration in Figure 4a) was strongest between USA and United Kingdom, with multiple co-authored papers reflecting cross-border research partnerships. Country-level analysis demonstrated that developed economies dominated output and collaboration, though contributions from emerging regions including Malaysia have been steadily increasing.

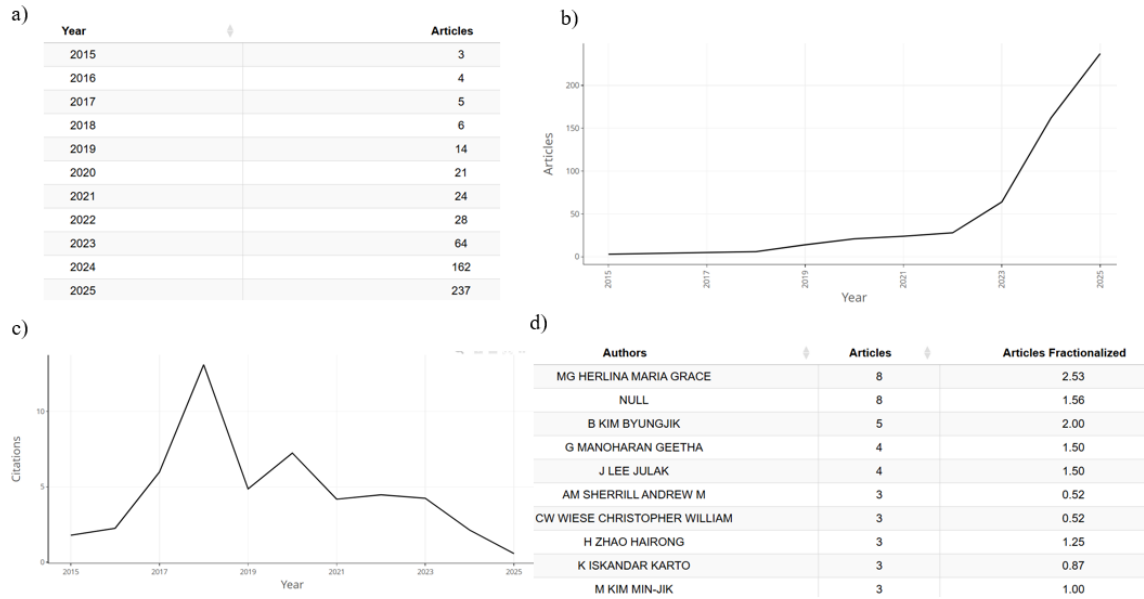


Figure 1a): Annual Scientific Production (Table)

Figure 1b): Annual Scientific Production (Plot)

Figure 1c): Average Citations Per Year

Figure 1d): Most Relevant Authors

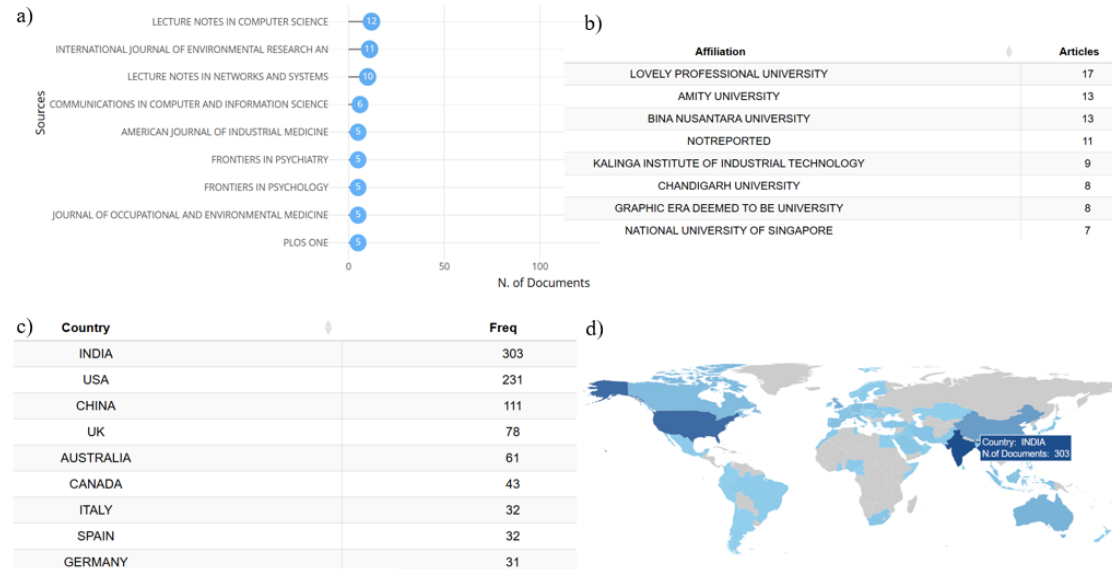


Figure 2a): Most Relevant Sources (Plot)

Figure 2b): Most Relevant Affiliations

Figure 2c): Top Countries' Scientific Production (Table)

Figure 2d): Countries' Scientific Production (Plot)

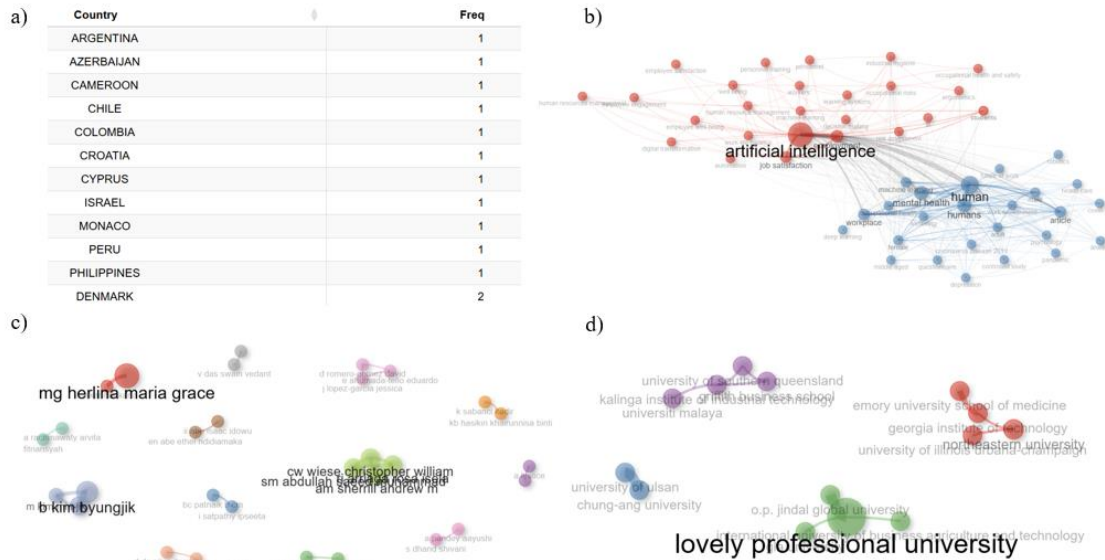


Figure 3a): Bottom Countries' Scientific Production (Table)

Figure 3b): Co-occurrence Network by Author Keywords

Figure 3c): Authors Collaboration Network

Figure 3d): Institutions Collaboration Network

5. Conclusion and Future Recommendations

This study addressed two research questions. For RQ1, the analysis showed that research themes and collaboration patterns in AI and employee wellbeing have developed steadily over the past decade. Publication trends demonstrated a consistent rise in output, reflecting the growing importance of the topic. Keyword co-occurrence revealed that research remains concentrated around two clusters; artificial intelligence and human factors. This indicates that the field is still largely framed through technology adoption and wellbeing impacts. Collaboration networks further highlighted that a small number of institutions and countries dominate knowledge production, while many remain peripheral. These findings justify that the field is still in a developmental stage, with opportunities for broader thematic integration and stronger international collaboration.

For RQ2, the results identified leading authors, sources, and references that have consistently shaped the debate. Citation analysis showed that influential works primarily come from management, psychology, and information systems journals, with highly cited authors bridging technical and organisational perspectives. Institutional and geographical mapping underscored that research activity is concentrated in universities from high-income countries. This concentration suggests that the knowledge base is still uneven, with limited perspectives from developing regions. Addressing this gap will be crucial for achieving a more inclusive understanding of how AI adoption affects wellbeing across different cultural and economic contexts.

Based on these insights, future research should expand thematically beyond the two dominant clusters to explore underdeveloped intersections such as ethical governance, fairness in algorithmic management, and sector-specific wellbeing outcomes. More cross-disciplinary work that integrates organisational behaviour, computer science, and occupational health will enrich conceptual depth. Researchers should also pursue collaborations across regions to diversify perspectives and avoid geographical concentration. Complementing bibliometric evidence with qualitative or mixed-

method approaches could further contextualise the findings, helping to bridge structural patterns with current workplace experiences.

Acknowledgement

The authors would like to acknowledge the use of Bibliometrix R package and its web interface Biblioshiny for conducting the bibliometric analysis. We also thank ChatGPT (OpenAI, 2025) for language refinement support, which was used to improve clarity and readability. The final interpretation and conclusions are solely the responsibility of the authors.

References

- Aria, M. & Cuccurullo, C. (2017) bibliometrix: An R-tool for comprehensive science mapping analysis, *Journal of Informetrics*, 11(4), 959-975, Elsevier.
- Aroles, J., Mitev, N., & Vaujany, F. X. (2019). Mapping themes in the study of new work practices. *New Technology, Work and Employment*, 34(3), 285–299. <https://doi.org/10.1111/ntwe.12146>
- Badri, A., Boudreau-Trudel, B., & Souissi, A. S. (2018). Occupational health and safety in the industry 4.0 era: A cause for major concern?. *Safety science*, 109, 403-411.
- Bankins, S., Ocampo, A. C., Marrone, M., Restubog, S. L. D., & Woo, S. E. (2024). A multilevel review of artificial intelligence in organizations: Implications for organizational behavior research and practice. *Journal of organizational behavior*, 45(2), 159-182.
- Brougham, D., & Haar, J. (2020). Technological disruption and employment: The influence on job insecurity and turnover intentions: A multi-country study. *Technological Forecasting and Social Change*, 161, 120276.
- Čaić, M., Odekerken-Schröder, G., & Mahr, D. (2019). Service robots: Value co-creation and co-destruction in elderly care networks. *Journal of Service Management*, 30(2), 209–232. <https://doi.org/10.1108/JOSM-07-2018-0199>
- Cheng, J., Zhang, L., Lin, Y., Guo, H., & Zhang, S. (2022). Enhancing employee wellbeing by ethical leadership in the construction industry: the role of perceived organizational support. *Frontiers in Public Health*, 10. <https://doi.org/10.3389/fpubh.2022.935557>
- Dabirian, A., Kietzmann, J., & Diba, H. (2017). A great place to work!? Understanding crowdsourced employer branding. *Business horizons*, 60(2), 197-205.
- Doraiswamy, P. M., Blease, C., & Bodner, K. (2020). Artificial intelligence and the future of psychiatry: insights from a global physician survey. *Artificial intelligence in medicine*, 102, 101753.
- Flores-Viva, J. M., & García-Peñalvo, F. J. (2023). Reflections on the ethics, potential, and challenges of artificial intelligence in the framework of quality education (SDG4). *Comunicar: Media Education Research Journal*, 31(74), 35-44.
- Goswami, M., Jain, S., Alam, T., Deifalla, A. F., Ragab, A. E., & Khargotra, R. (2023). Exploring the antecedents of ai adoption for effective hrn practices in the indian pharmaceutical sector. *Frontiers in Pharmacology*, 14. <https://doi.org/10.3389/fphar.2023.1215706>
- García-Martínez, I., Fernández-Batanero, J. M., Fernández-Cerero, J., & León, S. P. (2023). Analysing the impact of artificial intelligence and computational sciences on student performance: Systematic review and meta-analysis. *Journal of New Approaches in Educational Research*, 12(1), 171-197.
- Jarrahi, M. H., Newlands, G., Lee, M. K., Wolf, C. T., Kinder, E., & Sutherland, W. (2021). Artificial intelligence and the future of work: Human-AI symbiosis in organizational decision making. *Journal of Business Research*, 129, 912–920. <https://doi.org/10.1016/j.jbusres.2020.11.037>
- Khogali, H. O., & Mekid, S. (2023). The blended future of automation and AI: Examining some long-term societal and ethical impact features. *Technology in Society*, 73, 102232.

- Lee, J., Kim, S., & Park, Y. (2022). Research trends and opportunities of artificial intelligence in human resources management: A text-mining based literature review. *Sustainability*, 14(5), 2722. <https://doi.org/10.3390/su14052722>
- Maden, W. v. d., Lomas, D., & Hekkert, P. (2023). A framework for designing ai systems that support community wellbeing. *Frontiers in Psychology*, 13. <https://doi.org/10.3389/fpsyg.2022.1011883>
- Malik, A., Budhwar, P., Patel, C., & Srikanth, N. R. (2020). May the bots be with you! delivering hr cost-effectiveness and individualised employee experiences in an mne. *The International Journal of Human Resource Management*, 33(6), 1148-1178. <https://doi.org/10.1080/09585192.2020.1859582>
- McClure, P. K. (2017). “You’re Fired,” Says the Robot: The Rise of Automation in the Workplace, Technophobes, and Fears of Unemployment. *Social Science Computer Review*, 36(2), 139-156. <https://doi.org/10.1177/0894439317698637> (Original work published 2018)
- Meijerink, J., Bondarouk, T., & Lepak, D. P. (2021). Employees as active consumers of HRM: Linking employees’ HRM competences with their perceptions of HRM service value. *Human Resource Management*, 60(4), 567–581. <https://doi.org/10.1002/hrm.22023>
- Nazareno, L., & Schiff, D. S. (2021). The impact of automation and artificial intelligence on worker well-being. *Technology in Society*, 67, 101679.
- OpenAI. (2025). ChatGPT [Large language model]. OpenAI. <https://chat.openai.com/>
- Prentice, C., & Nguyen, M. (2020). Engaging and retaining customers with AI and employee service. *Journal of retailing and consumer services*, 56, 102186.
- Spagnoletti, P., Resca, A., & Lee, G. (2022). The duality of artificial intelligence in organizations: Managing the tension between automation and augmentation. *Information Systems Journal*, 32(1), 5–34. <https://doi.org/10.1111/isj.12332>
- Tarafdar, M., Cooper, C. L., & Stich, J. F. (2019). The technostress trifecta—Techno eustress, techno distress and design: Theoretical directions and an agenda for research. *Information Systems Journal*, 29(1), 6–42. <https://doi.org/10.1111/isj.12169>
- Wang, D., Weisz, J. D., Muller, M., Ram, P., Geyer, W., Dugan, C., ... & Gray, A. (2019). Human-AI collaboration in data science: Exploring data scientists' perceptions of automated AI. *Proceedings of the ACM on human-computer interaction*, 3(CSCW), 1-24.