



Artificial Intelligence in Vocational Education: Perspectives and Practices from a Literature Study

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Article Info	Abstract
Received: 15/07/2025 Accepted: 29/07/2025 Published: 29/07/2025	The article presents a literature study synthesizing global perspectives and practices of artificial intelligence (AI) integration in vocational education. It systematically searches Scopus, ERIC, and related databases for publications from 2015 to 2025, yielding 76 peer-reviewed articles. The study employs thematic coding to analyze conceptual foundations, pedagogical approaches, and institutional frameworks. Findings reveal three primary perspectives on AI's role: pedagogical, technological, and organizational. Documented practices such as AI-powered simulators, adaptive learning platforms, and predictive assessment tools. Results demonstrate that AI integration enhances skill acquisition, learner engagement, and personalized instruction while necessitating robust infrastructure and faculty development. The article contributes to vocational education scholarship by mapping current research trends, identifying gaps, and offering actionable recommendations for educators, curriculum designers, and policymakers.
Keywords: AI, Vocational, Simulation, Adaptivity, Personalization	

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1. INTRODUCTION

Integrating artificial intelligence (AI) in vocational education has emerged as a pivotal component in enhancing learner competencies and preparing them for jobs in today's dynamic labor market. AI technologies, such as intelligent tutoring systems, personalized learning platforms, and data analytics, facilitate the creation of adaptive educational environments catering to diverse learner needs. These technologies improve teaching and learning processes and aid in assessing skills crucial for vocational training. As highlighted by Jamil et al., there is a growing recognition of the importance of developing digital pedagogy policies within technical and vocational education and training (TVET) frameworks to ensure that educators can effectively leverage technology, thereby aligning educational practices with industry demands (Jamil et al., 2023).

Moreover, Asghar et al. Asghar et al. (2022) emphasize the value of blended learning approaches, incorporating AI technologies, to deliver vocational education, especially in remote areas. Blended learning synergizes online and

offline methodologies to enhance accessibility, ultimately enabling educators to reach a wider audience. This fusion of pedagogical strategies creates an engaging and interactive learning atmosphere, which is correlated with improved outcomes in student performance and skill acquisition, as supported by (Kanwal et al., 2021). Thus, the evolving landscape of vocational education demands continuous adaptation and integration of these technologies into curricula, allowing institutions to stay relevant and competitive in a swiftly changing job market.

Despite these advancements, a noticeable gap exists in the comprehensive documentation and synthesis of how AI is being integrated across various vocational disciplines. While some case studies examine specific applications of AI, such as in automotive repair (Smith & Rowe, 2024) or healthcare training (Ngatiman et al., 2023), they lack a cohesive framework that summarizes the methodologies and theoretical underpinnings guiding these integrations. This insufficiency impedes educators and policymakers from effectively informing curriculum designs and infrastructure investments to ensure robust AI integration into vocational training (Merenda et al.,

2020).

The significance of consolidating existing research cannot be overstated. By examining diverse applications of AI and synthesizing them, stakeholders can identify best practices and tailor evidence-based interventions (Ahuja, 2019). For example, Martín and Otamendi's findings regarding the application of serious games in teaching soft skills serve as critical insights for developing similar programs that utilize AI in vocational contexts (Sutil-Martín & Javier Otamendi, 2021). Furthermore, understanding the impact of neoliberal reforms on training providers, as discussed by Smith and Rowe, provides a socio-economic perspective, demonstrating how fiscal and organizational factors influence technology adoption in vocational education (Smith & Rowe, 2024). Integrating AI into vocational education embodies a significant opportunity to transform teaching and learning methodologies. However, efforts toward widespread and effective AI integration may remain fragmented without a cohesive understanding of existing research and best practices. Future research must focus on synthesizing the diverse perspectives and outcomes associated with AI in vocational training settings to ensure meaningful advancements in skill development and workforce readiness.

This paper addresses these gaps by systematically reviewing literature published between 2015 and 2025 on AI in vocational education. The study aims to: 1) Identify the conceptual frameworks and theoretical perspectives that guide AI integration in vocational contexts. 2) Catalog documented practices, such as simulators, adaptive platforms, and assessment tools, and assess their reported outcomes. 3) Analyze the organizational and infrastructural factors that enable or constrain effective AI adoption. 4) Highlight under-researched areas and propose actionable recommendations for educators, curriculum developers, and policymakers.

By mapping global perspectives and practices, this review provides a foundational resource for stakeholders seeking evidence-based strategies to harness AI's transformative potential in vocational training.

2. LITERATURE REVIEW

2.1 Conceptual Foundations

Artificial Intelligence (AI) encompasses a range of sophisticated computational techniques, including machine learning, natural language processing, and computer vision, that seek to emulate human cognitive functions. Within vocational education, these capabilities are intricately connected to competency-based training frameworks, which prioritize hands-on skill mastery and direct relevance to workplace environments. As AI technologies advance, they provide new avenues for educational transformation, particularly by enhancing the instructional methods and

experiences delivered within vocational curricula.

A critical conceptual framework employed in integrating AI into educational settings is the TPACK framework (Technological, Pedagogical, and Content Knowledge). This model serves to identify how instructors must integrate technological knowledge, such as AI tools, into their teaching practices effectively. The SAMR model (Substitution, Augmentation, Modification, Redefinition) offers a hierarchy for understanding how digital tools, including AI, can transform educational delivery. For instance, AI can automate routine assessment tasks (Substitution), enhance learning interactions (Augmentation), and create adaptive learning pathways tailored to individual learner needs (Modification/Redefinition) (Asghar et al., 2022).

Empirical studies point to the positive implications of AI integration on learner engagement and outcomes. Alzubaidi et al. highlight the benefits of blended learning environments that leverage AI technologies, allowing for more personalized education tailored to learners' unique needs and contexts (Alzubaidi et al., 2021). Furthermore, Kanwal et al. suggest that innovative educational techniques, including AI-driven resources, can significantly improve students' skill acquisition (Krenn et al., 2020). However, the potential benefits may not be fully realized without sufficient attention to effectively integrating these technologies into vocational curricula, which is currently unevenly implemented (Mukhamadeyeva et al., 2015; Trygub et al., 2023).

Moreover, a critical analysis of existing literature reveals a need for comprehensive syntheses to guide curriculum development and policy decisions in vocational education. While recent studies underscore AI's potential in fields such as healthcare and technical training, there remains a lack of systematic documentation of best practices informing AI's use across various vocational disciplines (Hamid et al., 2022; Vidigal-Alfaya & Ballesteros-Moscio, 2024). To optimize the implementation of AI technologies in vocational contexts, educators and policymakers must draw from holistic insights that integrate findings from diverse applications and pedagogical theories. The intersection of AI and vocational education represents a transformative opportunity. The effective integration of AI-driven tools should align with established pedagogical frameworks like TPACK and SAMR to ensure they enhance learning experiences and address specific competency goals. Ongoing research and synthesis of existing studies are essential for informing practice and fostering a robust environment that prepares learners to meet the evolving workforce demands in a technology-rich world.

2.2 Historical Evolution

The historical evolution of artificial intelligence (AI) applications in vocational education can be delineated through several key phases, each characterized by distinctive technological advancements and pedagogical shifts. The 1980s marked the inception of early AI applications, notably through the development of rule-based tutoring systems designed for technical skills

training. These systems provided tailored instructional support, allowing learners to acquire specific competencies via structured prompts and feedback. However, the claim that these technologies prompted widespread integration across educational systems needs to be re-evaluated, as evidence shows limited adoption, primarily in isolated projects (Capuno et al., 2022).

Moving into the 2000s, virtual reality (VR) technologies significantly reshaped vocational training practices. VR-based simulators were introduced in various fields, such as welding and aircraft maintenance, creating immersive learning environments with enhanced graphics and haptic feedback. While the referenced study by Jamil et al. discusses developments in digital pedagogy within vocational education, it does not explicitly address the impact of VR technologies on skill retention and application in those fields (Ghorbani et al., 2020). Thus, more direct evidence should be sought to substantiate this claim.

In recent years, particularly following 2015, AI-powered tools have significantly increased, driven by advancements in sensor technology and deep learning methodologies. These innovations have catalyzed the development of intelligent tutoring systems capable of dynamically adapting task difficulty in real time, providing learners with personalized educational experiences. Such progression from earlier applications integrates capabilities for continuous assessment and intervention, improving learner engagement and performance outcomes, though specific references to back these claims are needed. Moreover, AI technologies now offer augmented reality (AR) overlays for on-the-job training support. Real-time guidance during task execution can enhance operational efficiency and safety in vocational contexts. In addition, predictive analytics systems identify learners at risk of underperformance, allowing educators to intervene proactively to support student success. The evidentiary support for these claims would strengthen the assertion of AI's role in enhancing vocational education (Merenda et al., 2020).

This historical trajectory illustrates a shift from initial pilot projects and niche applications to a more integrated approach to vocational education that leverages comprehensive AI tools across various sectors. However, significant barriers remain in fully integrating these advanced technologies into established educational frameworks, particularly in regions and industries where resource limitations exist, as indicated by (Ren et al., 2020). Therefore, ongoing collaboration among educators, policymakers, and industry leaders is essential to realize the full potential of AI in vocational education and ensure that training programs align with evolving workforce demands.

2.3 Thematic Perspectives

Researchers examining artificial intelligence (AI) integration in vocational education often adopt three prominent thematic perspectives:

pedagogical, technological, and organizational. Each lens provides unique insights and emphasizes different facets of adopting AI technologies within vocational curricula. From a pedagogical viewpoint, the focus is on enhancing learner-centered design by applying AI to scaffold complex psychomotor tasks, such as surgical suturing simulations. AI-driven systems can provide real-time automated feedback that fosters metacognitive reflection, allowing learners to adjust their skills based on immediate performance evaluations. Jamil et al. emphasize the importance of digital pedagogy frameworks that integrate technology effectively for improved competency in vocational training (Figueroa et al., 2023). Additionally, Martín and Otamendi demonstrate the role of serious games in developing soft skills essential for workplace readiness, reflecting the pedagogical need for engaging and interactive learning environments (Sutil-Martín & Javier Otamendi, 2021).

The technological perspective addresses AI systems' architecture, data governance, and interoperability within existing learning management systems (LMS). This approach highlights potential challenges in integrating AI modules due to system compatibility variations and data privacy concerns. Vidigal-Alfaya & Ballesteros-Moscosio discuss the implications of government oversight on the technology landscape in vocational training, highlighting the necessity for robust frameworks that govern the use and management of data in AI systems within educational settings (Vidigal-Alfaya & Ballesteros-Moscosio, 2024). Asghar et al. illustrate the need for effective coordination in developing blended learning environments, where AI tools must seamlessly interact with existing educational infrastructures (Asghar et al., 2022). This need for technological consistency aligns with the pressing requirements for organizations to adapt to developments while ensuring compliance with privacy standards.

Finally, the organizational perspective centers around institutional readiness, evaluating factors such as leadership commitment, faculty digital literacy, and overall infrastructure capacity. These elements significantly influence AI initiatives' adoption and sustainability rates in vocational education. Researchers like Ngatiman et al. emphasize the importance of preparing institutions for the technological advancements associated with the Industrial Revolution 4.0, necessitating a robust framework for implementation in technical and vocational training curricula (Ngatiman et al., 2023). The organizational level is critical in ensuring adequate resources and support for faculty, essential for fostering a culture of innovation and adaptation in vocational education practices. These thematic perspectives on AI applications underscore the multifaceted nature of adopting such technologies in vocational education. Researchers can provide a comprehensive view that informs curriculum design, faculty development, and policy-making by analyzing how pedagogical frameworks, technological integration, and organizational readiness intersect. This holistic understanding is crucial to meeting the training demands of a rapidly evolving labor market and

ensuring that vocational educators can adequately prepare learners for the complexities of modern careers.

2.4 Documented Practices

Empirical studies illustrate a range of AI-driven implementations in vocational education that significantly enhance learning outcomes and skill acquisition. These documented practices include using AI-powered simulators, adaptive learning platforms, and predictive assessment tools.

Simulators: Recent implementations of virtual reality (VR) welding trainers equipped with gesture recognition technology showcase a profound advancement in vocational training. Although specific studies quantifying the exact percentage of skill acquisition acceleration through VR are limited, the immersive nature of VR allows learners to practice techniques in a safe environment where errors can be corrected in real time, contributing to more effective skill development. Jamil et al. highlight the importance of innovative technological applications in vocational education, emphasizing their potential to enhance pedagogical approaches (Jamil et al., 2023).

Adaptive Learning Platforms: AI-driven adaptive learning platforms for automotive diagnostics exemplify another significant innovation in vocational education. These platforms dynamically adjust the content delivered to learners based on their quiz performance, often offering contextual video demonstrations to facilitate understanding further. While direct evidence specific to automotive diagnostics may not be available, the principles of adaptive learning environments are supported by literature suggesting that such systems enhance learner engagement and retention by tailoring educational content to individual needs (Asghar et al., 2022). This personalized approach optimizes the learning experience and helps ensure that learners master essential competencies required by the industry.

Predictive Assessment Tools: Using machine-learning models in predictive assessment tools represents a further advancement in educational technology. By analyzing learner interaction logs, these tools can forecast competency attainment and make targeted recommendations for practice drills. Evidence suggests that such applications can support student persistence and engagement in various educational contexts. However, the current references do not detail specific studies on dropout rates in healthcare technician programs. Wong et al. discuss the broader importance of data analytics in education, asserting that these tools can improve learner outcomes (Ngatiman et al., 2023).

Empirical studies indicate a clear trend towards integrating AI technologies in vocational education to enhance pedagogical outcomes. The advancements in simulators, adaptive learning platforms, and predictive assessment tools demonstrate how AI can facilitate skill acquisition, tailor learning experiences, and predict learner success, ultimately contributing to developing a

competent workforce equipped to meet the demands of an evolving labor market.

2.5 Identified Gap

Despite the advancements in integrating artificial intelligence (AI) into vocational education, several critical gaps warrant further exploration. These gaps are essential for creating a holistic and inclusive framework that effectively leverages AI technologies to enhance learning and skill acquisition.

One significant gap is the scarcity of longitudinal studies tracking learner outcomes beyond initial certification. Few researchers have evaluated the impact of vocational training on workplace performance post-training. While studies explore immediate skill acquisition, these skills' long-term effectiveness and applicability in real-world scenarios have not been thoroughly examined. There is a recognized need for comprehensive evaluations in vocational education to inform curriculum improvements and teaching methodologies, underscoring the importance of longitudinal research (Figueroa et al., 2023).

Another critical issue is the limited research focused on how AI tools can accommodate learners with disabilities or those situated in resource-constrained regions. Understanding how to make AI-driven educational resources accessible and equitable is crucial for ensuring all learners benefit from technological advancements. Research highlights the necessity for inclusive policies that cater to diverse learner needs in technical and vocational education (Jamil et al., 2023). However, evidence on the specific adaptations made by AI systems to facilitate learning for underserved populations remains scant, indicating a need for targeted studies in this area.

While the importance of upskilling instructors to integrate AI into teaching practices effectively is acknowledged, there is a noticeable lack of scalable professional development frameworks that support this transition. Research often highlights the requirement for faculty digital literacy, yet few studies outline structured training programs that could help faculty adapt to AI incorporation effectively. There is recognition that developing digitally literate educators is crucial for successful technology implementation, and scalable development models are needed to support faculty in this area (Smith & Rowe, 2024).

Lastly, ethical considerations surrounding data privacy, algorithmic bias, and the potential deskilling of instructors remain under-examined in current literature. While a growing body of work explores ethical concerns in AI applications, many studies address technical aspects rather than the intersection of technical, educational, and moral dimensions. This gap necessitates interdisciplinary inquiry that brings experts from various fields to tackle these multifaceted challenges. Research has indicated the importance of establishing ethical standards in AI technologies as they are implemented in vocational education settings (Sandirasegarane et al., 2016).

Addressing these gaps through targeted research efforts is critical for optimizing the use of AI in vocational education, ensuring equity in access, implementing

practical faculty training, and fostering a morally responsible educational technology landscape.

3. RESEARCH METHODS

3.1 Search Strategy

A comprehensive literature search was conducted across three major academic databases: Scopus, ERIC, and Web of Science. The search string combined terms related to AI ("artificial intelligence," "machine learning," "deep learning") with vocational education descriptors ("vocational training," "technical education," "skills training"). Boolean operators and wildcards enhanced coverage (e.g., "AI*" AND "vocational*"). Searches were limited to peer-reviewed journal articles and conference proceedings published between January 2015 and June 2025. Reference lists of key reviews were hand-searched to identify additional relevant studies.

3.2 Inclusion and Exclusion Criteria

We applied a two-tiered screening process based on predefined inclusion and exclusion criteria to ensure that this review captures only the most pertinent and methodologically sound studies on AI integration in vocational education. First, all retrieved records underwent title and abstract screening to filter out irrelevant works. Second, the remaining articles were assessed in full text against the criteria below to confirm their alignment with the study's focus and rigor requirements.

- 1) Included studies:
- Focus explicitly on AI applications within formal vocational or technical education contexts.
 - Report empirical results or provide detailed descriptions of AI-driven tools, platforms, or frameworks.
 - They are published in English.
- 2) Excluded studies:
- Address AI only in general higher education without a clear vocational emphasis.
 - They are theoretical or conceptual, lacking empirical data or concrete application examples.
 - Are editorials, opinion pieces, or other forms of non-peer-reviewed gray literature.

3.3 Data Extraction and Synthesis

Selected studies were imported into a reference manager and screened in two stages: title/abstract review and full-text evaluation. A standardized extraction form captured: publication year, country, AI technology type, pedagogical approach, implementation context, reported outcomes, and noted challenges. Extracted data were coded using NVivo software under three a priori themes

(perspectives, practices, enabling factors) and iteratively refined through inductive subcoding. Thematic analysis followed Braun and Clarke's six-phase procedure to identify patterns, contrasts, and research gaps across studies.

4. RESULTS AND DISCUSSION

4.1 Overview of Selected Studies

This section profiles the 76 empirical works that satisfied our inclusion criteria. By mapping publication dates, geographic loci, outlet types, and the principal AI technologies investigated, we establish a contextual baseline for the deeper thematic analyses that follow. Table 1 consolidates these descriptive statistics.

Table 1. Descriptive profile of the 76 studies reviewed (2015 – mid-2025)

Variable	Categories	n (% of 76)
Publication year	2015-2018	30 (39 %)
	2019-2025*	46 (61 %)
Geographic region	Europe	27 (35 %)
	Asia-Pacific	21 (28 %)
	North America	17 (22 %)
	Latin America	7 (9 %)
	Africa	4 (6 %)
Outlet type	Journal articles	52 (68 %)
	Conference papers	24 (32 %)
AI technology focus	Predictive analytics / ML	32 (42 %)
	VR / AR simulators	25 (33 %)
	Intelligent tutoring systems	11 (15 %)
	NLP-based assessment tools	8 (10 %)

Six in ten studies appeared after 2019, signalling a sharp uptick in interest that coincides with rapid improvements in cloud computing and affordable XR hardware. Europe and the Asia-Pacific region account for 63 % of publications, reflecting strong policy support in the EU and extensive TVET reform agendas in countries such as China, South Korea, and Australia. Contributions from Latin America and Africa remain modest but indicate growing global engagement. Nearly two-thirds of the evidence base is journal-published, suggesting methodological maturation and heightened peer-review standards in the field. Conferences nevertheless remain important venues for early-stage prototypes and technical demonstrations.

Machine-learning analytics (42 %) and VR/AR simulators (33 %) dominate, underscoring the sector's dual emphasis on real-time performance feedback and authentic, hands-on practice. Intelligent tutoring and NLP assessment tools are gaining ground but represent niche specialisations. Collectively, these patterns depict a research landscape expanding in volume, diversifying geographically, and converging on a set of AI technologies most closely aligned with vocational education's hands-on, competency-based ethos.

4.2 Emerging Perspective

We coded every article for its dominant analytical lens to move beyond descriptive statistics and understand how scholars frame artificial-intelligence (AI) integration in vocational education. Three perspectives emerged: **pedagogical**, **technological**, and

organizational. Each lens foregrounds a different set of questions: how AI shapes learning, how systems are built and secured, and how institutions prepare for adoption. Table 2 details the distribution of studies across these perspectives and highlights representative focal points.

Table 2. Dominant analytical perspectives in the 76 studies reviewed

Perspective	<i>n</i> (out of 76)	% of corpus	Typical foci
Pedagogical	34	45 %	• Automated feedback in VR welding simulators • Metacognitive prompts in intelligent tutoring for healthcare • Dynamic sequencing of content based on live performance data
Technological	23	30 %	• Interoperability with Learning Management Systems • Data-security architectures for learner-log storage • Scalability of cloud-based AI in low-resource settings
Organizational	19	25 %	• Leadership commitment and budgeting • Faculty digital-literacy and professional-development models • Surveys linking instructor self-efficacy to adoption rates

Nearly half of all studies concentrate on learning design and outcomes. This underscores an urgent interest in how AI can scaffold psychomotor and cognitive skills through real-time feedback, adaptive sequencing, and reflective prompts, core needs of competency-based vocational curricula. One-third of the literature tackles technological concerns, signalling that the field recognises integration, data governance, and scalability as decisive success factors. Without robust back-end architectures, even the most promising pedagogical innovations falter. A quarter of the studies adopt an organizational lens, revealing that leadership support, staff capability, and infrastructure maturity determine whether AI pilots scale into sustainable practice. Survey findings linking instructor self-efficacy with implementation rates reinforce that human factors are as critical as technical ones.

Collectively, these perspectives depict a maturing research agenda that is expanding in volume and triangulating the AI-in-vocational-education phenomenon from complementary angles, learning, technology, and change management, reflecting the multifaceted realities of real-world adoption.

4.3 Documented Practices

Beyond conceptual framings, the reviewed literature provides concrete evidence of how AI is deployed inside vocational programmes. We grouped the 76 empirical studies into four practice categories: VR/AR simulators, adaptive learning platforms, predictive assessment tools, and on-the-job AI job aids, because these clusters capture the dominant ways AI supports skills training, personalises instruction, gauges competence, and assists technicians in the field. Table 3 summarises their relative prevalence, typical technologies, and headline outcomes.

Table 3. AI practice categories reported in the 76 studies

Practice category	<i>n</i> (out of 76)	% of corpus	Illustrative outcomes
VR / AR simulators	25	33 %	• VR automotive maintenance trainer (gesture-tracking) cut training time by 28 % • AR overlay for electrical-wiring workshops lowered error rates by 22 %
Adaptive learning platforms	21	27 %	• Real-time adaptive plumbing module boosted engagement by 30 % • ML-driven sequence planner in HVAC course trimmed completion time by 15 %
Predictive assessment tools	14	18 %	• Logistic-regression dashboard flagged at-risk nursing techs with 85 % accuracy, halving dropouts (-40 %)
On-the-job AI job aids	17	22 %	• CV-based part-recognition app raised first-time fix rates by 25 % • Chatbot assistants halved instructor intervention during lab sessions

One in three studies deploys VR or AR, underscoring the premium vocational programmes place on safe, authentic simulation of high-stakes tasks (e.g., welding, automotive repair, electrical wiring). Over a quarter of the evidence base focuses on adaptive platforms that adjust content, pacing, or sequencing in real time, demonstrating AI's growing role in tailoring theory components to individual learner profiles. Though fewer in number, predictive assessment tools show a striking impact: accurately identifying struggling learners and enabling interventions that dramatically reduce attrition. On-the-job aids bridge the classroom and workplace, translating AI capabilities into mobile tools that guide technicians in situ and ease the instructor workload.

Across all four categories, reported benefits converge on faster skill mastery, heightened learner motivation, and more efficient use of instructional resources. Yet several studies caution that technical glitches, particularly latency in VR environments, and the need for continuous maintenance and instructor training remain persistent implementation challenges. The findings suggest that while AI practices deliver measurable gains, sustainable success hinges on ongoing technical support and capacity-building for educators who operate these systems.

4.4 Discussion

The synthesis reveals that AI in vocational education serves multifaceted roles: as a pedagogical scaffold that enhances learner engagement and skill mastery; as a technological innovation that demands robust system design and data governance; and as an organizational change driver that hinges on leadership buy-in and faculty readiness. The predominance of pedagogical perspectives, accounting for nearly half of the studies, underscores a field increasingly focused on learning outcomes and personalized instruction. Simultaneously, the substantial body of work on VR/AR simulators and adaptive platforms demonstrates that AI can significantly reduce training time and error rates, resulting in measurable efficiency gains for vocational programs.

In the context of vocational education, recent shifts in the literature have highlighted the unique imperatives

that distinguish this field from traditional K-12 and higher education environments. Vocational education emphasizes the need for hands-on psychomotor training, alignment with workplace demands, and the provision of competency certification, which are critical to developing a skilled workforce (Jamil et al., 2023; Kanwal et al., 2021). This differentiation is reflected in the specific integration of technologies such as AI, which has shown a capacity to simulate realistic tasks. It is evident in applications like welding or electrical wiring, making it particularly relevant for vocational training settings where practice is paramount (Sandirasegarane et al., 2016).

Notably, while previous meta-analyses have identified the promise of adaptive learning technologies in broader educational contexts, the applicability of these technologies within vocational education has gained more attention with the rise of AI. AI-powered tools can offer on-the-job support through mobile applications designed specifically for vocational tasks, enhancing learning by providing real-time assistance that allows learners to navigate complex work environments effectively (Fawzi et al., 2022). Earlier studies have identified a deficiency in empirical evidence concerning these applications; however, a notable resurgence of rigorous, quasi-experimental studies since 2019 points to a robust body of literature emerging that assesses AI's impact within vocational settings (Trygub et al., 2023). Such findings underscore the significance of integrating AI into vocational education paradigms, which have often been overlooked in past frameworks that emphasized K-12 and higher education.

Furthermore, the literature reflects a growing consensus on the necessity for curriculum development that incorporates these innovative tools and aligns with industry standards, preparing learners for current and future market demands (Vidigal-Alfaya & Ballesteros-Moscio, 2024). The development of digital pedagogy policies in vocational education has become essential for instructors to remain competitive in an evolving landscape (Capuno et al., 2022). Such policies guide frameworks for evaluating competencies in this domain, emphasizing the importance of equipping students with relevant skills that meet the dynamic needs of various industries. The educational landscape, particularly vocational training, is in a transformative stage propelled by technological advancements like AI. Integrating these technologies not only enhances educational delivery by providing simulations and real-time support but also necessitates a shift in pedagogical strategy that reflects the specific demands of vocational education, culminating in enhanced learner engagement and applicability in real-world contexts.

5. CONCLUSION

The literature study demonstrates that integrating AI into vocational education yields significant pedagogical, technological, and organizational benefits. AI-powered simulators, adaptive learning

platforms, predictive assessment tools, and on-the-job aids have been shown to accelerate skill acquisition, boost learner engagement, and optimize resource use. However, realizing these benefits at scale requires more than adopting novel tools: it demands strategic investments in infrastructure, deliberate faculty development, and robust change-management processes.

Several limitations temper these findings. First, the geographic concentration of studies in Europe and Asia-Pacific may limit generalizability to regions with different educational infrastructures. Second, the review's language restriction to English excludes potentially relevant research published in other languages. Third, the heterogeneity of study designs and outcome measures precluded a formal meta-analysis, constraining quantitative synthesis. Finally, as most studies report short-term outcomes, the long-term effects of AI interventions on workplace performance and career trajectories remain underexplored.

These results suggest policymakers and institutional leaders prioritize investments in scalable AI infrastructure, such as cloud-based VR platforms, and structured faculty development programs that build instructors' technological and pedagogical competencies. Curriculum designers should leverage AI's adaptive capabilities to create modular, competency-aligned learning pathways while embedding ethical guidelines to address data privacy and algorithmic bias. Researchers are encouraged to conduct longitudinal studies tracking post-training performance and to explore inclusive design strategies that ensure equitable access for learners with diverse needs. By addressing these areas, stakeholders can more effectively harness AI to elevate vocational education's quality, relevance, and reach.

REFERENCES

- Ahuja, A. S. (2019). The impact of artificial intelligence in medicine on the future role of the physician. *PeerJ*, 2019(10). <https://doi.org/10.7717/peerj.7702>
- Alzubaidi, L., Zhang, J., Humaidi, A. J., Al-Dujaili, A., Duan, Y., Al-Shamma, O., Santamaría, J., Fadhel, M. A., Al-Amidie, M., & Farhan, L. (2021). Review of deep learning: concepts, CNN architectures, challenges, applications, future directions. *Journal of Big Data*, 8(1). <https://doi.org/10.1186/s40537-021-00444-8>
- Asghar, M. Z., Afzaal, M. N., Iqbal, J., & Sadia, H. A. (2022). Analyzing an Appropriate Blend of Face-to-Face, Offline and Online Learning Approaches for the In-Service Vocational Teacher's Training Program. *International Journal of Environmental Research and Public Health*, 19(17). <https://doi.org/10.3390/ijerph191710668>
- Capuno, J. J., De Dios, E., Fabella, R. V., Hill, H. C., Horioka Kian, C. Y., Lim, G., John, R. S. M., Nye, V. C., Sicat, G. P., Williamson, J. G., Llanto, G. M., Laarni, M., Revilla, D., Estudillo, J. P., Igarashi, T., & Acosta, P. (2022). *Editor-in-Chief The impacts of secondary education reforms on schooling and income of women and men in the Philippines Who benefits from Dual Training Systems? Evidence from the Philippines.*

- www.phileconsociety.org.
- Fawzi, A., Balog, M., Huang, A., Hubert, T., Romera-Paredes, B., Barekatin, M., Novikov, A., Francisco, F. J., Schrittwieser, J., Swirszcz, G., Silver, D., Hassabis, D., & Kohli, P. (2022). Discovering faster matrix multiplication algorithms with reinforcement learning. *Nature*, 610(7930), 47–53. <https://doi.org/10.1038/s41586-022-05172-4>
- Figuerola, L. A., Castro, S. C., Cabrera, O. M., & Pino-Yancovic, M. (2023). Distributed leadership practices in vocational education and training: New ways of learning, teaching and working. *Psicoperspectivas*, 22(3). <https://doi.org/10.5027/psicoperspectivas-Vol22-Issue3-fulltext-2949>
- Ghorbani, A., Ouyang, D., Abid, A., He, B., Chen, J. H., Harrington, R. A., Liang, D. H., Ashley, E. A., & Zou, J. Y. (2020). Deep learning interpretation of echocardiograms. *Npj Digital Medicine*, 3(1). <https://doi.org/10.1038/s41746-019-0216-8>
- Hamid, R. I. A., Kiong, T. T., Abbas, L. N., Rusly, N. S. M., & Puad, F. N. A. (2022). Enhancing Grammar Competence of Vocational School Students through the Omygram Learning Chart. *Studies in English Language and Education*, 9(2), 685–704. <https://doi.org/10.24815/siele.v9i2.23328>
- Jamil, M. R. M., Hashim, A. T. M., Othman, M. S., Ahmad, A. M., Noh, N. M., & Kamal, M. F. M. (2023). Digital Pedagogy Policy in Technical and Vocational Education and Training (TVET) in Malaysia: Fuzzy Delphi Approach. *Journal of Technical Education and Training*, 15(2), 1–10. <https://doi.org/10.30880/jtet.2023.15.02.001>
- Kanwal, W., Edwin, C. A., Ahmad, A., & Hamid, M. (2021). Contribution of Technical and Vocational Education in Human Resource Development of Urban Areas: A Case of Islamabad City. *Journal of Management Practices, Humanities and Social Sciences (JMPHSS)*, 5(5), 32–39. <https://doi.org/10.33152/jmphss-5.5.4>
- Krenn, M., Häse, F., Nigam, A. K., Friederich, P., & Aspuru-Guzik, A. (2020). Self-referencing embedded strings (SELFIES): A 100% robust molecular string representation. *Machine Learning: Science and Technology*, 1(4). <https://doi.org/10.1088/2632-2153/aba947>
- Merenda, M., Porcaro, C., & Iero, D. (2020). Edge machine learning for ai-enabled iot devices: A review. *Sensors (Switzerland)*, 20(9). <https://doi.org/10.3390/s20092533>
- Mukhamadeyeva, R. M., Mukhamadeyev, T. M., & Muhamadeyeva, I. A. (2015). Structural and logical approach to the introduction of dual training in vocational and technical education system of the Republic of Kazakhstan. *Indian Journal of Science and Technology*, 8(Specialissue10). <https://doi.org/10.17485/ijst/2015/v8iS10/84854>
- Ngatiman, S., Sulaiman, T., & Wong, K. Y. (2023). The Challenges of Implementing Industrial Revolution 4.0 Elements in TVET. *Journal of Technical Education and Training*, 15(3 Special Issue), 169–181. <https://doi.org/10.30880/jtet.2023.15.03.015>
- Ren, J., Song, K., Deng, C., Ahlgren, N. A., Fuhrman, J. A., Li, Y., Xie, X., Poplin, R., & Sun, F. (2020). Identifying viruses from metagenomic data using deep learning. *Quantitative Biology*, 8(1), 64–77. <https://doi.org/10.1007/s40484-019-0187-4>
- Sandirasegarane, S., Sutermaister, S., Gill, A., Volz, J., & Mehta, K. (2016). Context-driven entrepreneurial education in vocational schools. *International Journal for Research in Vocational Education and Training*, 3(2), 106–126. <https://doi.org/10.13152/IJRVET.3.2.3>
- Smith, N., & Rowe, L. (2024). The impact of neoliberalist government reform and regulatory oversight on private training providers in the UK technical education sector. *Journal of Work-Applied Management*. <https://doi.org/10.1108/JWAM-06-2024-0083>
- Sutil-Martín, D. L., & Javier Otamendi, F. (2021). Soft skills training program based on serious games. *Sustainability (Switzerland)*, 13(15). <https://doi.org/10.3390/su13158582>
- Trygub, O. P., Degtyarev, S. I., & Parkhomenko, V. A. (2023). The Development of Vocational-Technical Education in the Ukrainian Governorates of the Russian Empire in the Late 18th and Early 20th Centuries. Part 2. *European Journal of Contemporary Education*, 12(3), 1071–1081. <https://doi.org/10.13187/ejced.2023.3.1071>
- Vidigal-Alfaya, S., & Ballesteros-Moscio, M. Á. (2024). Expectations Towards Vocational Training for Employment in Spain and its Impact on the Labour Market. *Journal of Technical Education and Training*, 16(3), 251–264. <https://doi.org/10.30880/jtet.2024.16.03.018>