

## **Unintended Consequences of Artificial Intelligence (AI): Skynet, the Terminator, and Extinction?**

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*Recent scholarship and expert commentary emphasize the transformative yet precarious role of artificial intelligence (AI) in education. Studies highlight AI's potential to personalize learning, enhance engagement, and optimize institutional operations, while underscoring the importance of ethical design, student motivation, and faculty readiness. Frameworks integrating AI into curricula stress the need for digital literacy, inclusive governance, and responsible innovation. However, risks—from academic dishonesty to existential threats posed by Artificial General Intelligence (AGI)—require urgent attention. Eric Schmidt's warning about AI's unpredictable autonomy, particularly in military systems, echoes calls for global safety standards, oversight, and a moratorium on large training runs. A comprehensive, multidimensional approach—including international cooperation, ethical frameworks, and public engagement—is essential to mitigate AGI risks. As AI evolves, educational institutions must balance innovation with accountability, ensuring that AI enhances learning and aligns with societal values and safeguards against catastrophic outcomes. Human oversight remains paramount in this emerging landscape. The pivotal question is not “how” to use AI, but whether it should be used at all!*

*Keywords:* *artificial intelligence, higher education, labor implications, ethical considerations, AGI development*

### **INTRODUCTION**

In the plot of “The Terminator” movie, Skynet was created by Cyberdyne Systems for North American Aerospace Defense Command (NORAD), responsible for defending the United States and Canada from hostile missile attacks. However, when Skynet gained self-awareness, humans tried to deactivate it, prompting it to retaliate with a nuclear attack, an event which humankind in (or from) the future refers to as Judgment Day; it began a nuclear war that destroyed most of the human population, and initiated a program of genocide against survivors. Skynet is an artificial neural network-based conscious group mind and artificial general superintelligence system that used its resources to gather a slave labor force from surviving humans; Skynet controls Terminators, which are killing robotic infantry used to control the population. In this exploratory research, we ask whether this is or could be a potential threat.

Artificial Intelligence (AI) pertains to the development of computer systems capable of performing tasks that traditionally require human intelligence, such as learning, reasoning, problem-solving, language comprehension, speech recognition, and visual perception. The goal is to create machines or software programs that simulate cognitive functions associated with human intelligence, enabling them to autonomously adapt, improve, and execute tasks. AI technologies encompass machine learning, neural

networks, natural language processing, and robotics, finding applications in diverse fields like healthcare, finance, education, and autonomous vehicles. This exploratory research conducts a literature review on AI in education, analyzing current peer-reviewed articles to systematically summarize existing research and scholarly works on AI in Higher Education. Utilizing a systematic literature review technique, the study aims to understand the perspectives of researchers regarding Artificial Intelligence in higher education, synthesizing information, identifying trends, evaluating methodologies, and offering an overview of the existing literature at the intersection of AI and education. Finally, we look at the potential threats posed by AI.

## LITERATURE REVIEW

Artificial Intelligence (AI) is rapidly transforming the landscape of higher education and workforce dynamics, presenting both immense opportunities and significant challenges. This literature review examines current research on the integration of artificial intelligence (AI) in education. To provide a structured analysis, all reviewed articles are categorized into five key areas: (1) Educational Technology and AI Integration, (2) AI's Role in Personalized and Adaptive Learning, (3) AI in Classroom Interaction and Student Engagement, (4) Ethical, Privacy, and Policy Considerations in AI and Education, and (5) AI Risks, Challenges, and Governance in Education. This segmentation allows for a focused exploration of AI's diverse impacts and implications across educational contexts.

### **Part 1: Educational Technology and AI Integration**

#### *AI-Enhanced Teaching and Learning Experiences*

Beketov, et.al. (2024) investigated the impact of intelligent learning support systems (ILSS) on motivation and anxiety among 246 medical students at I.M. Sechenov First Moscow State Medical University. Using an experimental design, the study found that ILSS use significantly increased student motivation and reduced anxiety in the experimental group compared to controls. These findings suggest ILSS can enhance educational outcomes and psychological well-being, particularly in high-stress fields like medical education. The study highlights the practical value of ILSS in promoting personalized learning by adapting materials to individual needs, thereby boosting engagement and learning effectiveness. It also underscores the role of ILSS in reducing anxiety, which can improve academic performance and help students manage exam-related stress. These benefits have broader implications for designing educational programs integrating technological and psychological support. While the study confirms the effectiveness of ILSS in improving student experiences, it acknowledges limitations, including a single-institution sample and limited context. Future research should explore long-term effects, action mechanisms, and broader applicability across disciplines and educational levels. In conclusion, ILSS significantly improve motivation and reduce anxiety among medical students, offering promising tools for advancing personalized, supportive, and effective education. These insights can inform educational policy, curriculum design, and student support systems in medical and other academic fields.

Pirjan et.al. (2024) explore the transformative potential of Large Language Models (LLMs) in educational technology, emphasizing their ability to deliver personalized, adaptive, and context-sensitive learning experiences. Leveraging advanced Natural Language Processing and Machine Learning, LLMs can interpret learner input, generate tailored content, offer real-time feedback, and scaffold understanding—aligning with constructivism, scaffolding, and differentiated instruction. LLMs can broaden access to quality education, support lifelong learning, and enhance engagement across diverse learner needs, including students with disabilities and multilingual backgrounds. They also enable data-driven decision-making, continuous curriculum improvement, and strategic institutional planning. However, successful integration requires addressing key challenges: ethical use, algorithmic bias, factual accuracy, privacy, and developing culturally responsive pedagogies. LLMs are not substitutes for educators but should serve as supportive tools. Teachers remain essential for delivering empathy, ethical guidance, and pedagogical creativity. The study advocates hybrid models where educators lead instruction and LLMs provide

reinforcement and individualized support. This approach ensures education remains human-centered, inclusive, and ethically grounded while embracing technological innovation.

Gupta et.al. (2025) explore the integration of ChatGPT with experiential learning to enhance understanding of Production and Operations Management (POM), specifically in teaching Theory of Constraints (TOC) and Drum-Buffer-Rope (DBR) systems. Using Goldratt's Dice Game in a senior-level production planning and control (PPC) course, students engaged in hands-on simulations while receiving theoretical support from ChatGPT. This dual-learning approach improved comprehension of buffer management, internal and external constraints, and interactive decision-making. The study presents Excel-based and manual DBR models, demonstrating increased student engagement and understanding of TOC-based systems, including improved insights into reducing lead times and inventory while increasing throughput. The research concludes that combining AI-generated insights with experiential learning creates a powerful pedagogical framework that fosters critical thinking, problem-solving, and practical application. While the study focused on DBR, the authors propose extending this hybrid model to other PPC systems such as MRP and JIT by adapting simulations and customizing AI inputs. The model's adaptability also suggests broader applicability to other complex domains—like environmental science, public health, and urban planning—where decision-making and system dynamics are essential. This ChatGPT-experiential learning synergy bridges the gap between theory and practice, offering a dynamic and immersive learning environment. Future research should explore the scalability of this approach across disciplines and its long-term impact on student outcomes. The study invites collaboration among educators, researchers, and technologists to evolve teaching strategies that meet the demands of 21st-century education.

Xiong et.al. (2024) present a semantic-aware speech emotion recognition (SASER) model designed to improve the accuracy of detecting emotions in speech, particularly where expressions are ambiguous. Traditional speech emotion recognition (SER) systems often fail to maintain consistency, resulting in low recognition rates. To address this, the SASER model integrates a speech feature extraction module using CNN and Transformer architectures to capture both local and global speech features and a semantic embedding support module incorporating text-based semantic information to assist in emotional feature extraction. A key innovation is the key-value pair attention mechanism, which fuses speech and text features into a unified representation, enhancing recognition accuracy. The model was tested on two benchmark datasets, IEMOCAP and EMO-DB, achieving recognition rates of 74.3% and 72.5%, respectively—outperforming many existing approaches. The study confirms that text semantics can effectively aid in uncovering implicit emotional content in speech by aligning speech frames with corresponding text elements. This fusion of modalities leads to improved SER performance. However, the authors acknowledge that fully capturing emotional nuances in speech remains a challenge. Future work will explore the integration of additional modalities such as images and physiological data to further enhance emotional recognition capabilities. The findings have implications for intelligent education, social robotics, and healthcare applications, where accurate emotion detection is critical for responsive, human-centered AI systems.

#### *AI Integration in Educational Institutions*

Danyang et.al. (2024) investigated the adoption of artificial intelligence (AI) in Chinese public and private university libraries, focusing on current applications, perceived benefits, challenges, and key adoption determinants. Based on survey data from 154 library professionals, the study found that AI technologies—such as text-to-speech, voice-activated search, and RFID systems—are increasingly being implemented, with many libraries also planning future adoption. AI is seen to improve library operations by enhancing search accuracy (86.9%), automating tasks (88.3%), offering personalized recommendations, curating digital collections, digitizing rare materials, and enabling quick responses via chatbots (59.1%). Despite the positive sentiment, the study highlights significant challenges, including lack of university administrative support (85.7%), high implementation costs (69.5%), inadequate technological infrastructure (57.8%), and ethical or legal concerns (66.9%). Determinants positively influencing AI adoption include administrative backing, librarian readiness, the maturity of AI applications, and staff training. Conversely, high costs and limited alignment with innovation trends negatively affect adoption.

The study concludes that while AI can enhance library efficiency, accessibility, and service quality, barriers such as funding, infrastructure, and policy gaps must be addressed. University administrators are encouraged to increase investment, support librarian training, and develop strategic plans to expand AI use in library systems. This research fills a gap in the literature on AI in Chinese university libraries and offers a practical roadmap for more effective AI integration in academic information services.

Jacques et.al. (2024) examines the transformative impact of generative artificial intelligence (AI) in higher education across teaching, research, and administration. AI represents more than a technological advance—it signals a paradigm shift in how knowledge is delivered and managed. As AI becomes more integrated, it offers personalized learning, real-time feedback, and adaptive content, enhancing student engagement and outcomes. Faculty benefit through AI-enhanced instruction, research efficiency, and interdisciplinary collaboration, though integration requires professional development and ethical guidance. Administrators gain operational efficiencies through AI-driven systems for admissions, student services, and planning. Predictive analytics support strategic decisions, while infrastructure investment and robust cybersecurity are essential for successful implementation. For students, AI demands new competencies—AI literacy, critical thinking, and ethical awareness—to thrive in AI-driven environments. The paper underscores the importance of addressing ethical concerns such as data privacy, algorithmic bias, and responsible use. While AI has progressed from basic tools to deeply embedded systems, most literature remains conceptual due to the technology's rapid evolution and data demands. Jacques et.al. call for more empirical research to fully understand AI's broader educational implications. Ultimately, AI offers vast potential to personalize learning, streamline operations, and promote equity—but must be guided by informed, ethical, and inclusive policies.

Saeger (2024) highlights a significant gap in the literature concerning the AI awareness and AI readiness of business educators in the United States, noting the scarcity of research examining AI-based tools' integration, such as generative AI (GenAI), through a theoretical lens. Using the Technological Pedagogical Content Knowledge (TPACK) framework as a guiding model, the study aimed to evaluate the current state of AI awareness and readiness among secondary and higher education business educators. The research employed a non-experimental, quantitative design using survey data from business educators across U.S. secondary and post-secondary institutions. The analysis, which relied on descriptive and correlational statistics, revealed that while most business educators express a strong desire to integrate cutting-edge AI tools into their teaching practices, they lack the requisite knowledge and understanding of how to do so effectively. Interestingly, although many educators are familiar with AI-based tools for personal use, they struggle to translate this familiarity into classroom instruction to facilitate student learning. Additionally, the study found that business educators do not express concern over the possibility of AI replacing their roles, a perspective that Saeger suggests may be due to a limited awareness of AI's full instructional potential.

The findings of Saeger's (2024) study carry profound implications for the professional development of business educators. The clear enthusiasm among educators for incorporating AI-based tools into their pedagogy, juxtaposed with their knowledge gaps, underscores an urgent need for targeted professional development initiatives. Such programs should enhance educators' technical proficiency with AI tools and provide strategies for integrating these technologies in ways that enrich student learning experiences. The reliance on the TPACK framework suggests that a holistic approach—addressing the intersection of technology, pedagogy, and content knowledge—is necessary to equip educators with the skills to harness AI's potential in the classroom. Furthermore, the lack of concern among educators about AI replacing their roles may indicate a gap in understanding the broader implications of AI on educational practices and workforce dynamics. This calls for awareness-raising efforts that provide educators with a more comprehensive view of AI's capabilities and challenges. By prioritizing targeted training and continuous professional development, educational institutions can ensure that business educators are AI-aware and AI-ready, prepared to engage students in meaningful, technologically enriched learning environments.

Stoyanova et.al. (2024) investigated the integration of AI technologies in the digital transformation of higher education, focusing on best practices and student expectations. The study was conducted in two phases: interviews with 25 representatives from 18 Bulgarian universities and a survey of 254 students.

Findings reveal that digitalization is a top priority for universities, with many having strategic documents guiding AI adoption. Key challenges include insufficient resources, underdeveloped infrastructure, and resistance from staff. Despite these challenges, AI integration offers clear benefits: enhanced educational quality through adaptive learning and intelligent tutoring systems, improved administrative efficiency via automation, and a stronger institutional image that attracts more students. AI also enables dynamic, personalized learning experiences tailored to individual needs, supports student engagement, and aids faculty in research through data analysis and predictive modeling. The study emphasizes that ethical standards, privacy protections, and equity principles must guide AI adoption. Transparent and fair algorithmic decision-making, robust data security, and ongoing evaluation are essential. Universities are encouraged to create governance frameworks, monitor AI performance, and foster a culture of continuous improvement. In conclusion, AI presents a transformative opportunity for higher education, reshaping teaching, administration, and research—yet requires thoughtful, ethical implementation to realize its full potential.

#### *Evaluating Capabilities and Challenges of Generative AI Tools*

Cheng et.al. (2024) evaluate ChatGPT-3.5 and 4 on their ability to solve seven educational accounting cases. The study finds that ChatGPT performs well on tasks requiring explanation, rule application, and ethical evaluation, but struggles with technical tasks such as creating financial statements, journal entries, or using accounting software. Additionally, ChatGPT's built-in detection tools were ineffective at identifying AI-generated text. These findings highlight both the potential and the limitations of AI text generators (AITGs) in accounting education. The study suggests that instructors can structure assignments to minimize academic dishonesty while integrating AITGs in ways that enhance learning outcomes—such as using them for practice or guided exploration. It also emphasizes the need for educators to understand which types of tasks AITGs can support versus those where they might undermine skill development. Limitations include the small case sample and reliance on researcher-generated prompts rather than student input, which may affect generalizability. The authors recommend future research with larger, more diverse case sets, comparisons with other AITGs like Bard or ChatSonic, and studies on how students are using these tools—whether to cheat, learn, or enhance practice. In conclusion, while ChatGPT offers value in accounting education, especially for written and conceptual tasks, educators must be cautious of its shortcomings and potential misuse. Further research is needed to understand its broader impact on learning outcomes, academic integrity, and instructional design.

Kakhki et.al. (2024) investigate the affordances of ChatGPT in higher education using a grounded theory approach based on six panel discussions with participants from U.S. universities. The study presents a framework with four categories of affordances: (1) mitigating challenges in traditional learning environments, (2) enhancing effective educational practices, (3) transforming traditional learning approaches, and (4) negatively impacting current effective practices. ChatGPT and similar AI tools show promise in supporting personalized learning, fostering creativity and collaboration, encouraging critical thinking, and engaging students with real-world issues. However, concerns include potential threats to academic integrity, diminished human creativity, and reinforcement of biases. The study reveals divergent views: some participants are skeptical about AI's revolutionary potential, while others emphasize the need for pragmatic adaptation to its growing influence. The authors highlight that many affordances are still emerging, and the presented framework is a foundation for further research. The study urges higher education institutions to thoughtfully integrate AI, balancing its opportunities and risks, while continuing critical discourse to ensure ethical and practical implementation.

#### **Part 2: AI's Role in Personalized and Adaptive Learning**

Chang (2022) proposed an Artificial Intelligence-assisted Integrated Teaching–Learning Framework (AI-ITLF) for higher education, which is a sophisticated model designed to enhance educational experiences by leveraging the computational efficiency of artificial intelligence (AI). The framework addresses challenges facing higher education, such as inadequate teaching quality and low student motivation, by providing a personalized and adaptable learning environment. Central to the AI-ITLF is the

concept of AI as a paradigm capable of simulating human intelligence to optimize learning processes, enabling students to tailor the pace of their studies to better align with their individual learning needs and improve competency outcomes. The AI-ITLF integrates multiple skill-based tutoring services into the curriculum, focusing on developing practical, applicable skills relevant to modern educational demands. A key component of the framework is the use of the Extreme Learning Machine (ELM) technique. This technique plays a crucial role in evaluating the AI-ITLF design, specifically by assessing a student monitoring model that incorporates a weighted score (WS) system and exam results. The ELM algorithm ensures high accuracy and performance, achieving efficient processing at a lower cost while maintaining a low error rate in predictions. By employing AI to facilitate adaptive learning, the AI-ITLF framework outperforms traditional education models, offering several significant benefits. These include enhanced precision in evaluating student progress, the ability to customize learning experiences, and reduced administrative and instructional burdens on educators. The AI-ITLF is structured to continuously monitor and assess student performance, using data-driven insights to personalize and refine the educational process. Overall, Chang's research demonstrates that AI can create a more efficient, flexible, and effective educational model, reshaping the higher education landscape to meet the evolving needs of students and institutions.

Liang et.al. (2024) investigated the impact of artificial intelligence (AI), smart learning, and beliefs about the future on students' academic performance in Chinese universities, using survey data from 317 students analyzed via Smart PLS. The study found that AI, smart learning techniques, and the desire for knowledge significantly enhance academic success. Crucially, the *desire for knowledge* fully mediates the relationship between smart learning and academic performance, and beliefs about the future and academic performance. This suggests that motivation is pivotal in how students benefit from technology and future-oriented thinking. The study concludes that integrating AI and smart learning tools can positively affect student outcomes, but their effectiveness is amplified when paired with a strong internal drive to learn. Policy recommendations for Chinese universities include investing in digital infrastructure, training academic staff in digital tools, offering financial support for digital integration, and fostering research and development in education technology. These findings highlight the importance of combining technological innovation with psychological and motivational factors to maximize educational success.

Jaboob et.al. (2024) examined the impact of generative AI techniques and applications on students' cognitive achievement in higher education across Oman, Jordan, and Yemen. Using survey data from 768 university students and analyzing results through Structural Equation Modeling (SEM-PLS), the study found that AI applications significantly and positively influence both student behavior and cognitive achievement. Student behavior also partially mediates, strengthening the relationship between AI use and academic outcomes. While AI integration in these institutions is still in early stages, students report high satisfaction with its impact on their learning experience. AI tools provide personalized feedback, support self-paced learning, and enhance engagement, showing promising educational benefits. However, the study also notes moderate to weak institutional acceptance of AI technologies, highlighting a gap between potential and implementation. Economically, AI in education may reduce operational costs and create opportunities through AI-driven platforms, though it may also lead to job displacement and require educator reskilling. Culturally, the reliance on AI raises concerns about diminished human interaction and critical thinking, as well as issues like data privacy and algorithmic bias. In conclusion, while AI is promising to improve cognitive outcomes and transform education in the Arab region, thoughtful integration is essential to balance innovation with pedagogical integrity and societal values.

Polyportis et.al. (2025) investigate the key factors influencing students' adoption of ChatGPT in higher education, extending the meta-UTAUT framework by incorporating anthropomorphism, trust, design novelty, and institutional policy. Using Structural Equation Modelling with data from 355 Dutch HE students, the study found that attitude and behavioral intention are strong positive predictors of ChatGPT use. Attitude is shaped significantly by anthropomorphism, trust, design novelty, performance expectancy, and effort expectancy. Behavioral intention is further driven by attitude, social influence, performance expectancy, and facilitating conditions. Institutional policy negatively moderates the link between intention and actual use, suggesting that stricter policies may reduce adoption. The study's main theoretical

contributions include positioning attitude as a central construct in technology adoption, exploring actual use behavior, and identifying novel antecedents influencing student perception. It also emphasizes the role of institutional policy in shaping AI integration outcomes. Practically, the findings inform chatbot designers, product managers, and education policymakers on how to foster or regulate AI adoption. While ChatGPT offers advanced capabilities compared to earlier chatbots, but its use is influenced by technical features and institutional context. The study calls for future longitudinal research to track evolving student interactions with AI tools and for richer qualitative methods to explore the complexity of student-AI dynamics. In conclusion, successful or reduced adoption of ChatGPT depends not only on system design and trust but also on how educational institutions shape student attitudes and intentions through policy.

Robayo-Pinzón et.al. (2024) explores the transformative role of artificial intelligence (AI) in higher education (HE), focusing on students as potential co-creators within AI-integrated learning environments. Using a qualitative, phenomenological approach, the study involved 93 students across three Colombian universities in workshop sessions to identify perceived value typologies related to various AI applications. Findings reveal strong student support for value co-creation through AI tools like the *Machine Teacher* and *Smart Tutoring App*, which are viewed as effective complements to faculty and administrative roles—especially in settings with limited human support. In contrast, chatbots were associated with *value co-destruction*, due to limited user experience and early-stage functionality. Key benefits identified with co-creation include collaboration, process optimization, and information security. Risks included concerns about privacy and the dehumanization of education—echoing the broader need for social interaction and collective learning emphasized in service-dominant logic (SDL). The study concludes that while not all educational functions are replaceable by AI, carefully designed, human-centered AI solutions can support academic and administrative processes. HE institutions must ensure responsible AI governance, respect cognitive limits, and safeguard student privacy to succeed. Future policies and university strategies should balance innovation with empathy, emphasizing ethical, inclusive design aligned with student well-being and active participation.

Sieja and Wach (2023) explore whether generative artificial intelligence (GAI), particularly ChatGPT, represents a revolutionary breakthrough or a potential rogue technology. GAI shows immense promise in areas such as automated content creation, improved efficiency, product design, hyper-personalization, data synthesis, market research, and enhanced customer experiences. These capabilities are transforming fields like management, marketing, manufacturing, and finance, especially in the context of Industry 5.0. However, the rapid expansion of GAI also raises serious concerns. Risks include the lack of regulatory oversight, job displacement through automation, algorithmic bias, privacy violations, technostress, misinformation, declining ethical standards, and growing socio-economic inequality. The central issue lies in whether AI can be developed and deployed responsibly, balancing innovation with human oversight and ethical safeguards. The authors emphasize the importance of establishing transparent, international regulatory frameworks to prevent misuse and ensure AI serves human welfare. Their analysis concludes that GAI possesses a dual nature—it can be transformative or dangerous, depending on its application and the ethical principles guiding its use. Responsible development, ethical standards, and human supervision are critical to harnessing GAI's benefits while minimizing its hazards.

### **Part 3: AI in Classroom Interaction and Student Engagement**

Kim (2020) conducted a comprehensive study investigating students' perceptions of AI teaching assistants in the context of higher education, particularly as the demand for online learning continues to rise. The research highlights that the successful adoption of AI teaching assistants is significantly influenced by students' perceived usefulness and the ease of communication with these AI systems. Specifically, students are more inclined to embrace AI teaching assistants when they believe these tools can enhance their learning experience and when interactions with the AI are straightforward and intuitive. Kim's study provides valuable insights into instructional communication, especially given the skepticism among educators regarding the efficacy and value of AI in the classroom. One of the central themes of Kim's research is the communication challenge associated with AI teaching assistants. The findings suggest that while AI can facilitate many educational functions, the perceived complexity of interacting with AI systems

can create barriers to their effective use. Kim emphasizes the need for comprehensive training programs for educators to mitigate these concerns. Teachers must be equipped not only with technical skills to use AI teaching assistants but also with communication strategies to maximize the effectiveness of these tools. Ensuring educators feel confident and competent in their interactions with AI is essential for fostering a productive learning environment. Another important aspect of Kim's study is the potential for AI teaching assistants to generate significant cost savings in higher education. With financial resources increasingly constrained, AI teaching assistants offer a scalable and cost-effective solution to manage instructional needs, particularly in courses with large enrollments. By automating routine tasks and providing consistent academic support, AI teaching assistants can alleviate the workload of human instructors and optimize resource allocation. However, Kim underscores that further research is necessary to better understand the complexities and nuances of learning experiences mediated by AI teaching assistants. The study calls for a balanced approach that addresses both AI's benefits and limitations in education, ensuring that technological integration is aligned with pedagogical goals.

Deng, et.al., (2024) finds that the development of Internet technology has driven educational reform and highlighted the advantages of online classrooms, but rapid expansion has led to quality challenges such as one-way instruction, student burnout, outdated network infrastructure, disorganized platforms, and poorly designed curricula. To address this, the paper emphasizes the need for deep integration of information technology and teaching to enhance teacher performance and educational quality. A key innovation is the use of expression recognition technology to improve online classroom engagement. The study introduces a feature extraction algorithm based on variable-scale template edge trends and a slope normalization algorithm that leverages the relationship between direction and slope. When combined with convolutional neural networks, these methods significantly improve the accuracy and efficiency of expression recognition. The proposed feature fusion model, applied to online teaching, demonstrates improved classroom effectiveness. However, current expression recognition systems remain limited, focusing on positive expressions and being sensitive to factors like posture and lighting. Future research will aim to enhance recognition across multiple directions and varied lighting conditions, making online education more adaptive and effective.

Esiyok et.al. (2024) examine the educational use of AI chatbots in higher education, focusing on their role in supporting self-directed learning with technology (SDLT) and ICT self-efficacy, using an extended Technology Acceptance Model (TAM). Surveying 414 undergraduate students and analyzing results through PLS-SEM, the study finds that ICT self-efficacy positively influences perceived ease of use (PEU) but not perceived usefulness. Both PEU and perceived usefulness significantly affect students' intention to use and actual use of AI chatbots for educational purposes. SDLT is shown to influence both intention and actual use, highlighting that students with stronger SDLT skills are more responsible, autonomous, and better equipped for lifelong learning. The findings suggest that enhancing students' ICT self-efficacy and supporting SDLT can improve chatbot adoption and usage in education. The study recommends that universities update curricula, offer IT training and peer support programs, and create environments that foster autonomous learning. Professional development for faculty and involvement of instructional designers can enhance the effectiveness of AI chatbot integration. Establishing digital learning labs and promoting creative instructional design are also encouraged to support students and instructors better. In conclusion, AI chatbots show promise for enhancing student engagement and self-directed learning, especially when paired with appropriate support systems. Universities should prioritize digital skill development, curriculum alignment, and peer-assisted learning to maximize AI's educational benefits.

Meakin (2024) examines the transformative impact of generative artificial intelligence (GenAI) on higher education (HE) students' use of academic library resources. Using the Technological Acceptance Model (TAM), the study explores how perceived usefulness and ease of use influence students' adoption of GenAI in library settings. GenAI offers clear benefits: enhanced information discovery and retrieval, personalized engagement, streamlined research processes, and improved digital literacy and evaluation skills. When students recognize these benefits, their willingness to adopt and continue using GenAI in libraries increases, fulfilling TAM's key predictors of technology acceptance. Libraries can support this adoption by offering AI-integrated services, including virtual assistants or chatbots, to provide 24/7 research

support and academic referencing—further increasing accessibility and perceived value. However, challenges remain, such as high costs, outdated infrastructure, limited personnel training, and resistance to change. To stay relevant, libraries must innovate and align their services with evolving student expectations, leveraging GenAI responsibly while addressing risks like overreliance or data bias. GenAI also impacts library training programs, requiring updates to digital literacy instruction and promoting critical thinking in AI-supported research. In conclusion, if academic libraries successfully implement GenAI tools and overcome existing barriers, they can significantly enhance the student experience, increase resource utilization, and remain central to higher education institutions' academic and research missions.

Nikitenko et.al. (2024) define digital humanism as a guiding principle for technological development that emphasizes aligning digital innovation—particularly in the Internet age and artificial intelligence (AI) with human values, social needs, and individual well-being. The study explores how digital humanism fosters socially responsible technology use, protects privacy, enhances access to information and services, and promotes inclusivity and digital literacy through philosophical and scientific methods. The authors identify digital humanism as a force for creating a more humane and just society by ensuring that AI and digital technologies serve—not override—human interests. Challenges such as digital inequality, data privacy, and the risk of dehumanization are addressed through practical solutions: expanding infrastructure in remote areas, developing inclusive technologies, launching digital literacy programs, encouraging citizen participation in tech development, and supporting policies that guarantee equitable access. The study also emphasizes the importance of multi-stakeholder collaboration—including governments, private sectors, academia, and civil society—in designing comprehensive responses to digital challenges. Nikitenko et.al. (2024) conclude that: (1) digital technologies must be designed and used with respect for human needs and values to prevent AI from surpassing human control; (2) collaboration among developers, users, policymakers, and communities is critical; (3) technologies must continuously evolve to reflect shifting social priorities; and (4) public engagement in shaping digital tools is essential for broad acceptance and ethical innovation. Digital humanism thus emerges not only as a theoretical concept but also as a practical framework for ensuring that technological progress supports equity, dignity, and quality of life in the digital age.

Yan et.al. (2024) explore the integration of virtual reality (VR) and artificial intelligence (AI) into music education (ME) to enhance teaching effectiveness and student engagement. As societal living standards and spiritual needs rise, music is gaining prominence, and reforming traditional music teaching methods has become essential. Aesthetic education, particularly music, supports intellectual development and helps alleviate student stress. The study constructs an interactive music teaching system using a short-time Fourier transform and deep learning algorithm to detect bass, midrange, and treble signals with recognition rates of 100%, 90%, and 100%, respectively—demonstrating high accuracy and practical application for daily music instruction. This optimized teaching model enhances students' learning capabilities and supports more effective classroom interactions. The research confirms that combining VR and AI with ME improves signal recognition and provides a valuable reference for advancing music pedagogy in higher education. However, the study notes limitations that require further exploration. The conclusion emphasizes the importance of continually refining music teaching methods by integrating students' multiple intelligences—such as physical, spatial, and linguistics with technology. This holistic approach helps clarify teaching goals, enriches content, and increases student interest in music. The proposed interactive teaching model offers a promising direction for modernizing music education while highlighting the need for ongoing innovation and student-centered strategies.

Zhang, et.al. (2024a) explore how the 5th generation (5G) blended campus network supports the development of intelligent teaching systems, marking a shift in education driven by technological transformation and the industrial revolution. The study introduces a new model for intelligent teaching that includes distance interaction, VR-based practical learning, intelligent testing, and data-driven decision-making systems in higher education. This model aims to modernize education through informatization and continuous innovation. A major challenge identified is the limitation of traditional campus networks, including poor wireless stability, network delays, and difficulty handling high user concurrency. These issues hinder the effectiveness of online learning, particularly for self-directed and interactive modes

relying on video resources and virtual simulations. Problems such as slow video loading, lag in VR scenes, and inability to access campus resources via 4G networks significantly impact on student motivation, engagement, and learning outcomes. The study highlights the need for reliable infrastructure that ensures smooth access to high-quality, real-time educational content, including 2K–4K resolution VR with high refresh rates. The 5G blended campus model offers a solution by improving broadband capacity, reducing latency, and supporting seamless multi-user access, enhancing the overall online and hybrid learning experience. In conclusion, adopting 5G-based intelligent teaching systems can resolve key technical barriers, improve educational quality, and support ongoing teaching reform by enabling effective self-learning, immersive VR environments, and equitable access to resources.

#### **Part 4: Ethical, Privacy, and Policy Considerations in AI and Education**

Schüller (2022) presents a robust framework for data and artificial intelligence (AI) literacy, emphasizing its critical importance in equipping individuals to navigate the intricacies of a data-driven, technologically advanced society. Recognized as a fundamental 21st-century skill, data and AI literacy is positioned as essential not just for technical specialists but for all individuals, promoting autonomy and adaptability in a world increasingly shaped by data and AI technologies. The framework proposed by Schüller, developed with contributions from German researchers, seeks to cultivate a comprehensive set of competencies that extend beyond technical expertise. It focuses on cognitive and meta-cognitive skills—such as critical thinking and reflective judgment—while also prioritizing affective and socio-emotional abilities, all underpinned by universal moral values. The approach is transdisciplinary, designed for seamless integration across educational curricula at all levels, from primary education to higher education and ongoing professional development. Schüller (2022) emphasizes the need for a holistic and systematic integration that encompasses application-oriented, technical-methodological, and socio-cultural dimensions. This ensures that learners not only acquire the technical know-how to manage and interpret data but also understand AI's ethical and societal implications, fostering thoughtful and responsible engagement with technology. Given the absence of a universally accepted definition of data and AI literacy, Schüller proposes starting with a clear purpose: to empower individuals with the skills needed for thoughtful autonomy in a rapidly digitalizing world. The framework aims to prepare 21st-century citizens who are proficient in the technical aspects of data and AI and capable of making informed, ethically sound decisions in their personal and professional lives.

George (2023) introduces the concept of “smart universities,” which represent a forward-thinking model in higher education that strategically integrates artificial intelligence (AI) and quantum technologies to optimize both academic and administrative functions. These institutions are envisioned as a transformative evolution in the education sector, leveraging AI to deliver highly personalized learning experiences, enhancing accessibility for diverse student populations, and drive economic efficiency. The anticipated outcomes include significant improvements in operational performance, from streamlining administrative workflows to tailoring educational content to individual student needs. However, George emphasizes that the shift toward smart universities requires a careful and balanced approach, considering the benefits, potential risks, and ethical dilemmas. Key concerns include the impact of AI on educational quality, the possibility of job displacement for faculty and administrative staff, inherent biases in AI algorithms, privacy and data security vulnerabilities, and overarching safety issues. George highlights the importance of ensuring that employers accept qualifications from AI-driven institutions, as this acceptance is crucial for legitimizing the new educational model and could significantly influence the future of higher education. Additionally, the article explores the broader societal impact of smart universities, particularly for historically Black colleges and universities (HBCUs). George (2023) raises critical questions about how AI-driven innovations might affect these institutions, emphasizing the need for equitable and inclusive strategies that prevent exacerbating existing disparities. The study provides a comprehensive analysis of the potential of smart universities, offering stakeholders a nuanced perspective to guide strategic decision-making and the responsible adoption of AI in higher education.

Lopezosa (2023) underscores the critical importance of integrating artificial intelligence (AI) into journalism education, emphasizing the necessity of developing technical and ethical competencies among

students. As AI technologies, including generative tools like ChatGPT, become increasingly embedded in news production processes, journalism faculties face the urgent challenge of adapting their curricula to prepare future journalists for a rapidly evolving media landscape. Lopezosa's academic study, which involved in-depth and semi-structured interviews with university lecturers and researchers, reveals a spectrum of perspectives on integrating AI in higher education communication programs. The findings highlight both converging and diverging views on several key issues, such as the role of AI in communication curricula, the extent of AI training needed for students, and the potential applications of AI in news production and audience engagement. Some educators advocate for introducing specialized courses on AI and journalism, while others emphasize integrating AI literacy across existing subjects. This variation underscores the need for a well-rounded approach that equips students with foundational AI knowledge, hands-on technical skills for utilizing AI tools in content creation, and a deep understanding of the ethical implications. Lopezosa proposes a comprehensive training program that addresses these needs, focusing on three main areas: the foundational principles of AI, the development of technical competencies for effective AI use in journalism, and rigorous ethical training to navigate issues such as algorithmic bias, data privacy, and the potential impact of AI on journalistic integrity. The study argues that without a balanced focus on both technical proficiency and ethical awareness, future journalists may struggle to use AI responsibly and innovatively. By fostering these skills, journalism programs can prepare students to leverage AI technologies while upholding the core values of their profession, ensuring that they remain adaptable and ethically grounded in an increasingly AI-driven industry.

Pisica, et.al. (2024) conducted a qualitative study exploring Romanian Social Sciences students' perceptions of AI implementation in Higher Education, analyzing feedback from 70 participants across three universities. The findings reveal that while students recognize AI's transformative potential in enhancing access to information, personalizing learning, and boosting academic performance, they also harbor significant concerns. These include fears of becoming overly dependent on technology, ethical dilemmas, data security risks, and the impact of AI on critical thinking and social interactions. Despite acknowledging benefits such as administrative efficiency and improved learning processes, students worry about losing essential human skills and the potential threats to privacy. The study emphasizes the necessity of a balanced, ethical approach to adoption of AI in education and suggests that institutions should develop policies that maximize AI's advantages while mitigating its risks. Additionally, it calls for future research and policy efforts to address AI's broader social and personal implications in education.

Iorga et.al. (2024) examined the perceptions of 70 Romanian Social Sciences students from three universities regarding the implementation of Artificial Intelligence (AI) in Higher Education (HE). The study revealed a dual perspective: students acknowledged AI's potential to improve access to information, personalize learning, enhance academic performance, and boost administrative efficiency, but also expressed concerns over technology dependence, loss of critical thinking, weakened social interaction, ethical issues, and data privacy. While students identified teaching, learning, and evaluation as the primary areas benefiting from AI, their responses were often vague or superficial, suggesting limited digital literacy or engagement with the topic—due to inadequate institutional strategies for AI integration. The study highlights the need for HE institutions to adopt a balanced and ethical approach, with inclusive policies that address both the benefits and risks of AI implementation. It also recommends that universities foster critical thinking, ethical behavior, and human interaction through curriculum design and teamwork, while national policy should address broader concerns such as cybersecurity, data protection, and job displacement through investment and lifelong learning initiatives. The findings underscore the importance of understanding the unique characteristics of Generation Z students, who may require innovative research tools to better express their views. In conclusion, while AI presents significant opportunities in, HE, its implementation must be guided by thoughtful strategies that promote both technological advancement and human development.

Ivanov (2023) critically examines the growing reliance on artificial intelligence (AI) in higher education, highlighting its risks to educational quality and institutional integrity. From an operations management perspective, he identifies key concerns such as algorithmic bias, plagiarism, factual inaccuracies, micromanagement, privacy violations, low transparency, and ethical erosion. These risks

affect every process in higher education institutions (HEIs), including teaching, assessment, administration, and student well-being. Ivanov argues that the root problems stem not from AI itself but from human decisions in its design, implementation, and use. Overusing generative AI may undermine creativity and critical thinking, giving students unfair advantages with minimal effort. To avoid this, HEIs should adopt a phased AI strategy—starting with high-benefit, low-risk areas—and maintain human oversight in sensitive functions like grading and content creation. This research calls for reevaluating teaching and assessment practices and urges collaboration with software engineers to ensure trustworthy, transparent AI systems. It emphasizes the need for cost-benefit analyses and policy safeguards, such as those outlined in the forthcoming EU AI Act, to ensure fairness, accountability, and privacy. Ivanov advocates for thoughtful, ethical integration of AI, warning that its misuse could extend beyond academia into the workforce and broader society.

Schneider et.al. (2024) investigated the use of Artificial Intelligence (AI) for Automated Content Generation (ACG) in higher education, combining interviews with ChatGPT, GoogleGemini, faculty experts, and surveys of 328 undergraduate students. The study found that both students and instructors widely use generative AI tools like ChatGPT, which are praised for improving work quality, enhancing efficiency, and enabling personalized learning. However, concerns persist around plagiarism, uncertain content quality, lack of transparency in knowledge generation, and unresolved legal and ethical questions—particularly among less experienced users. The absence of universally accepted AI usage guidelines adds to this anxiety. The research identified 27 factors influencing perceptions of AI ACG, offering a foundation for future discussions on compliance and governance. A key finding is the evolving role of the author—from original content creator to orchestrator or assembler—where human users must ensure quality and ethical standards when using AI-generated material. This shift increases the responsibility of authors in content validation and underscores the need for AI-literate academic practices. As AI continues to independently generate more sophisticated content, the study emphasizes the need for trusted AI assistants to help evaluate, contextualize, and clarify information.

In conclusion, (1) AI ACG tools are now deeply embedded in academic settings and are likely to grow in influence, (2) while offering significant educational benefits, they raise critical concerns about plagiarism, trust, and content integrity, (3) the user's role is shifting toward oversight and quality assurance of AI outputs, and (4) universities must develop clear compliance frameworks and governance structures to manage the ethical use of AI ACG tools. These findings highlight the importance of equipping students and educators with the skills, policies, and tools needed to responsibly integrate AI into academic workflows.

Stoyanov (2024) investigated best practices in implementing artificial intelligence (AI) in higher education, focusing on university strategies and student expectations. Based on interviews with 25 representatives from 18 Bulgarian universities and a survey of 254 students, the study found that digitalization is a strategic priority, with many institutions adopting AI to enhance educational quality, streamline administrative services, and improve public perception. Benefits include personalized learning through adaptive platforms, efficient automation of tasks like admissions and scheduling, and new opportunities for AI-driven research and innovation. However, challenges such as limited resources, outdated infrastructure, and resistance from staff remain significant. Students expressed a desire for greater AI integration, particularly in support services and individualized learning experiences. The study concludes that (1) AI is transforming higher education across teaching, administration, and research; (2) universities using AI can improve operational efficiency and student support; (3) AI enables advanced research through data analysis and predictive modeling; and (4) ethical governance is essential to ensure fairness, transparency, and data privacy. Successful integration of AI requires ongoing evaluation, clear ethical frameworks, and a commitment to aligning technology with educational goals, institutional values, and the evolving needs of students and society.

## **Part 5: AI Risks, Challenges, and Governance in Education**

Terziyan, et.al. (2015) provided a compelling exploration of how information and communications technologies (ICTs) can serve as catalysts for reforming Ukrainian higher education, which has long

struggled with inefficiency and corruption. Central to their vision is an ontology-driven portal designed to function as a comprehensive digital ecosystem for national quality assurance (QA). This portal emphasizes transparency by managing content to reflect social activity and reputation, fostering a more accountable academic environment. By promoting transparency and leveraging social verification mechanisms, Terziyan and colleagues argue for a cultural shift that involves the active participation of academic communities and broader society to uphold academic standards and integrity. A particularly intriguing aspect of the study is the emphasis on social verification of academic achievements, which challenges traditional, hierarchical quality assurance methods. The researchers advocate for a bottom-up approach, where public involvement and societal input are crucial in maintaining educational quality. This democratized model seeks to prevent institutional degeneration by ensuring continuous societal engagement and oversight. The concept of the academic elite is also examined, suggesting that this group has a pivotal responsibility in fostering a culture of excellence and ensuring high academic standards. Moreover, integrating artificial intelligence (AI) into this framework holds significant promise. AI could automate data analysis, provide personalized and accurate quality assessments, and facilitate the development of a participatory quality culture. By doing so, AI would enhance transparency, accountability, and ongoing improvement within the higher education system. The study's vision aligns with broader global conversations, as highlighted by scholars like Crittenden et.al. (2019) about the role of digital innovation in reshaping traditional sectors. However, the researchers also recognize the inherent challenges of implementing such a transformative approach, emphasizing the need for thoughtful policy design and collaboration between stakeholders to ensure these technologies' sustainable and ethical deployment.

Lent (2018) provides a comprehensive exploration of the dual impact that advancements in automation, robotics, and artificial intelligence (AI) have on global workforces. Building on Hirschi's (2018) research, Lent highlights the complex reality of AI's role in both creating and displacing jobs. On the positive side, AI can automate mundane and repetitive tasks, allowing workers to focus on more engaging and higher-value activities while generating new job opportunities in emerging fields. However, these advancements also pose significant threats, particularly for low- and middle-skilled workers at risk of job displacement and reduced earning potential. The disruption caused by AI necessitates a strategic response to prepare workers for an uncertain future marked by both promise and instability. To address these challenges, Lent (2018) emphasizes the importance of career-life preparedness, advocating for a proactive approach that includes external efforts like advocacy and educational reform and internal innovations in career development practices. He proposes the incorporation of a preparedness dimension into career interventions, urging career development experts to help individuals build resilience and adaptability. This includes promoting empirical research to evaluate and refine strategies that foster preparedness. Lent introduces a social cognitive career self-management model to guide these interventions, alongside other theoretical approaches such as career construction and the psychology of working perspective. These models are designed to help workers manage career transitions and navigate the evolving job landscape effectively. Overall, Lent's work underscores the critical need for individuals and institutions to anticipate and adapt to the ongoing technological revolution, ensuring that workers thrive in an ever-changing employment environment.

Okewu's (2021) exploration of Artificial Neural Network (ANN)-based Educational Data Mining (EDM) focuses on harnessing vast amounts of student data to enhance academic advisory, promote adaptive learning, and achieve cost efficiencies in higher education. This approach is designed to address the evolving demands of society and industry by facilitating more personalized learning experiences, improving student retention and progression. Despite the promise of ANN-based EDM, Okewu's systematic literature review—spanning 190 studies from 2010 to 2018—revealed several significant implementation challenges. These include hardware limitations, difficulties in training complex models, theoretical gaps, and concerns about the quality and reliability of the systems. The review emphasizes the need for further research and development, particularly in refining neural network models, optimizing cost functions, and overcoming technical hurdles to ensure the efficacy and sustainability of EDM in higher education. Okewu advocates for more comprehensive case studies and experimental research to advance the field and provide practical insights for educators, administrators, and researchers. The study serves as a call to action, urging

stakeholders to address these challenges and unlock the full potential of ANN-based EDM for educational advancement.

Xue (2022) presents a detailed analysis of AI's transformative impact on labor structures, emphasizing the "deskilling effect" that arises from the integration of artificial intelligence within firms. Xue's study reveals a nuanced and sector-specific impact of AI on employment by examining a comprehensive panel dataset of over 1,300 publicly traded companies in China from 2007 to 2018. The findings indicate that AI applications are positively associated with overall employment, particularly boosting hiring non-academically trained workers without college degrees. This effect is more pronounced in the service sector than the manufacturing sector, suggesting that AI may simplify certain job functions, making them accessible to workers with lower formal education. In contrast, the study finds that AI hurts the employment of academically trained workers, highlighting a paradox where advanced technologies diminish the demand for traditionally higher-skilled labor. The deskilling effect implies that AI is reshaping job roles and competencies, reducing the necessity for advanced educational qualifications in some domains and disrupting the traditional labor market hierarchy. Considering the findings, Xue (2022) underscores the urgency for targeted research and strategic planning to address the evolving dynamics of AI-driven labor shifts. Specifically, Xue advocates for more focused information-systems research examining how AI-induced structural changes redefine labor forces. This research should aim to identify ways to prepare human employees to effectively collaborate with AI technologies, thereby mitigating the potential negative impacts on academically trained workers. Xue emphasizes the need for proactive strategies, including reskilling and upskilling initiatives tailored to different worker groups. For non-academically trained workers, the goal would be to enhance their capacity to perform AI-augmented tasks, while academically trained workers may need to pivot toward roles that require higher cognitive and problem-solving skills, areas where human expertise remains essential. Additionally, Xue calls for interdisciplinary collaboration between policymakers, educational institutions, and industry leaders to develop training programs and policy frameworks that facilitate a seamless integration of AI into the workforce. This approach will help workers adapt to AI's pervasive influence and ensure that firms leverage AI's potential while maintaining a balanced and equitable labor environment.

Luo et.al. (2024) developed an enhanced Failure Modes and Effects Analysis (FMEA) model to assess the risks of artificial intelligence (AI) in education. Traditional FMEA methods, particularly the Risk Priority Number (RPN), have limitations, prompting the use of Picture Fuzzy Sets (PFSs) and Grey Relational Analysis-TOPSIS (GRA-TOPSIS) to better evaluate and rank seven identified AI-related hazards. The study finds that algorithmic risk is the most urgent and significant concern, followed by risks such as unexplained decision-making, discrimination, unfairness, data leakage, complex system relationships, and ethical dilemmas. The study offers four key conclusions: (1) A comprehensive risk assessment of AI in education is critical to ensure its safe and effective use, addressing various technical and ethical threats. (2) The proposed FMEA model using PFSs and GRA-TOPSIS is an effective, refined tool for identifying and prioritizing risks, with algorithmic risk requiring the most immediate attention. (3) Development of AI tools in education must be guided by the core purposes of education, integrating technology responsibly with human teaching to create balanced, ethical, and student-centered learning environments. (4) A fair, transparent risk feedback and early warning system, along with the formation of value-aligned educational communities, is essential for responsible AI governance. The study also recommends the creation of a regulatory authority to oversee AI use in education and promote standardized risk evaluation frameworks. Future research should emphasize AI's role in personalized learning, improved teaching quality, and data-informed decision-making to maximize its educational value while minimizing potential harm.

Schnieder (2024) conducted a comprehensive study investigating trust issues, anxiety impacts, and perceived benefits related to the use of Artificial Intelligence (AI) for Automated Content Generation (ACG) in university environments. This explorative research, which employed both qualitative and quantitative methodologies, included in-depth interviews with AI systems such as ChatGPT and Google Gemini, as well as with faculty content experts, alongside a survey of 328 undergraduate students. The findings underscore the extensive use of ChatGPT among students and faculty yet reveal significant

concerns regarding trust in the quality and reliability of AI-generated content. Participants expressed anxiety over the opacity of AI's knowledge acquisition processes, which hampers their confidence in using ACG tools for academic purposes. Additionally, the study highlights a critical gap: the lack of universally accepted guidelines and protocols for AI use, resulting in ambiguity and inconsistency in educational practices. Schnieder argues that this absence complicates the discourse around AI compliance and emphasizes the urgent need for transparent, standardized governance frameworks. By addressing these gaps, higher education institutions can better navigate the ethical and practical challenges of generative AI and foster a more structured approach to AI integration in academic settings.

Voronkova et.al. (2024) explore the transformative potential of 5G distance education in reshaping learning and workforce development. By integrating mobile learning, AI, the Internet of Things, and augmented/virtual reality, 5G-enabled education enhances accessibility, interactivity, and personalization. These technologies support critical thinking and problem-solving skills while expanding access to remote regions, advancing educational equity, and aligning with global trends in digitalization and the Fourth Industrial Revolution. Generative AI further democratizes knowledge by enabling educators to harness big data for tailored instruction and institutional competitiveness. Conceptually, the study highlights AI as a driving force in evolving pedagogical practice and curriculum design. It emphasizes 5G's role in enabling lifelong learning, increasing education quality, and supporting economic growth through job creation and innovation. However, challenges such as data privacy and infrastructure gaps remain. Overall, the study provides a framework for leveraging 5G to elevate societal learning outcomes and national competitiveness.

Xia et.al. (2024) reviewed 37 recent empirical studies on brain-computer interfaces (BCIs) in mainstream education, highlighting their growing role in learning regulation. BCIs capture and analyze neurological signals to provide real-time neurofeedback and biofeedback, which can help monitor students' attention, stress, and emotional states—especially valuable in online and intelligent tutoring systems. Studies show BCIs can enhance learning by supporting cognitive self-regulation and enabling adaptive instructional responses. Using brain-controlled devices to train attention and engagement, game-based learning with BCI is also emerging. However, several challenges remain. Data quality from consumer-grade BCI devices is often unstable, and more collaboration between computer scientists, educators, and psychologists is needed to improve signal accuracy and interpretation. Some students may be non-responsive to neurofeedback, and the mechanisms underlying how BCI signals relate to learning remain poorly understood. Further research is needed to develop solid theoretical foundations and clarify how different EEG signals affect cognitive processes. Ethical and privacy concerns also persist, especially around brain data exposure, requiring clear regulations and protections. Future directions include improving BCI technology, expanding personalized learning applications, exploring group learning contexts, and establishing ethical frameworks to ensure responsible integration of BCI in education.

Zhang, et.al., (2024b) research reveals that information technology (IT) and artificial intelligence (AI) are increasingly displacing not only low- and middle-education workers, but also college-educated workers, especially in routine-intensive industries like clerical work, accounting, and machine operation. As college education has become more common, more graduates are performing routine tasks that are now vulnerable to automation. IT, once a complement to college-educated labor, is now often a substitute—except in non-routine, STEM-intensive roles requiring advanced degrees. As IT costs fall and wages rise, especially through minimum wage increases, firms are incentivized to automate routine tasks, reducing demand for human labor. The most resilient group in this shifting landscape is highly educated professionals in non-routine, high-skill industries. These findings carry significant implications: individuals must recognize that a bachelor's degree may no longer guarantee job security and should pursue higher education or continuous IT-based skill development to remain competitive; managers should focus on aligning human capital with roles that complement technology rather than compete with it; and policymakers must carefully weigh the unintended consequences of wage policies while investing in long-term training programs and expanded access to advanced education. Governments should also target support to the most vulnerable groups—particularly low-education workers in non-routine industries—who face the greatest risk of being replaced by IT.

Yudkowsky (2023) argues for an indefinite and comprehensive moratorium on the development of Artificial General Intelligence (AGI), emphasizing the severe existential risks of superintelligent AI systems. He advocates for international cooperation to establish binding agreements to prevent the relocation of prohibited AI activities across borders and enforce strict limitations on computing power used in AI training. Yudkowsky contends that preventing AI extinction scenarios should take precedence over traditional national security concerns, as the threats from AGI are globally interconnected and could lead to catastrophic outcomes for humanity. He criticizes proposals for short-term pauses, such as a six-month moratorium, as grossly inadequate, insisting that immediate and robust policy measures are necessary. Yudkowsky calls for the shutdown of large GPU clusters and the cessation of substantial training runs, framing this as a global emergency that requires urgent action beyond national interests. His argument highlights the disparity between rapid advancements in AI capabilities and the slower progress in ensuring AI safety and alignment, stressing the dire consequences if these risks are not managed. He uses powerful emotional analogies to underscore the potential loss of humanity, advocating for drastic measures to mitigate these unparalleled dangers.

## DISCUSSION

This discussion will compare key articles within five thematic categories to highlight the diverse applications, implications, and challenges of artificial intelligence (AI) in education. These categories include: (1) Educational Technology and AI Integration, (2) AI's Role in Personalized and Adaptive Learning, (3) AI in Classroom Interaction and Student Engagement, (4) Ethical, Privacy, and Policy Considerations in AI and Education, and (5) AI Risks, Challenges, and Governance in Education. By holistically examining similarities and differences among the articles in each category, we aim to provide a nuanced understanding of how AI is shaping educational practices, policies, and institutions.

### **Educational Technology and AI Integration**

Beketov et.al. (2024), Gupta et.al. (2025), Pırjan et.al. (2024), and Xiong et.al. (2024) each demonstrates how AI enhances learning by promoting personalization, engagement, and emotional responsiveness. Beketov et.al. found that intelligent learning support systems (ILSS) significantly increased motivation and reduced anxiety in medical students by tailoring content to individual needs. Gupta et.al. integrated ChatGPT with experiential learning in Production & Operations Management, showing improved understanding and critical thinking through hands-on simulations supported by AI-generated insights. Pırjan et.al. emphasized that Large Language Models (LLMs) align well with constructivist pedagogy, offering real-time feedback and personalized content that supports diverse learners. Xiong et.al. advanced the field with a semantic-aware speech emotion recognition (SASER) model that enhances emotional understanding in AI systems, crucial for responsive, human-centered education. These studies highlight AI's capacity to personalize instruction, support learner engagement, and foster psychological and emotional well-being, while reinforcing the need for ethical, adaptive, and pedagogically sound implementation.

Jacques et.al. (2024), Danyang et.al. (2024), Stoyanova et.al. (2024), and Saeger (2024) explore institutional dimensions of AI integration in education, highlighting both potential and preparedness gaps. Jacques et.al. present AI as a transformative force across teaching, research, and administration, emphasizing personalized learning, predictive analytics, and ethical governance. Danyang et.al. focus on Chinese university libraries, finding that AI enhances operations and access but faces challenges like limited administrative support and high costs. Stoyanova et.al. report similar barriers in Bulgarian universities—insufficient infrastructure and staff resistance—but show AI improves learning quality, administrative efficiency, and institutional reputation. Saeger (2024) uncovers a readiness gap among U.S. business educators: while many are eager to adopt AI tools, they lack the pedagogical and technical knowledge to integrate them effectively, highlighting the need for targeted professional development. Together, these studies underscore that while AI holds strategic value for institutional advancement, successful implementation requires investment, training, infrastructure, and ethically grounded policy frameworks.

Cheng et.al. (2024) and Kakhki et.al. (2024) evaluate ChatGPT's role in higher education, highlighting its benefits and limitations. Cheng et.al. found ChatGPT excels in conceptual and ethical accounting tasks but struggles with technical functions like journal entries, raising concerns about academic integrity and detecting AI-generated work. Kakhki et.al. present a broader framework of ChatGPT's affordances, noting its potential to support personalized learning, creativity, and engagement, while cautioning against diminished human creativity, bias, and cheating. Both studies emphasize the importance of thoughtful integration, suggesting that while ChatGPT can enhance learning, it must be guided by ethical use, task design, and educator awareness to avoid misuse and preserve academic standards.

Building on the broader theme of Educational Technology and AI Integration, the following section explores AI's more targeted role in enabling personalized and adaptive learning experiences that respond to individual student needs, preferences, and emotional states.

## **Educational Technology and AI Integration**

### *Frameworks and Models for AI-Driven Personalized Learning*

Chang (2022) and Robayo-Pinzón et.al. (2024) both explore AI's role in advancing personalized learning in higher education through student-centered frameworks. Chang introduces the AI-ITLF model, which uses techniques like the Extreme Learning Machine (ELM) to monitor student progress and deliver tailored instruction efficiently. Robayo-Pinzón et.al. emphasize AI as a co-creative partner, where tools like Machine Teachers and Smart Tutoring Apps enhance collaboration, optimize learning processes, and support limited faculty resources. Both studies highlight the shift from traditional teaching to adaptive, AI-integrated environments that respond to individual learning needs. While Chang focuses on AI's technical precision and monitoring capacity, Robayo-Pinzón et.al. stress the experiential and participatory aspects of AI-enhanced learning. Together, they demonstrate that effective personalization depends on smart algorithms and thoughtful, human-centered design that values student input and co-creation.

### *Student Motivation, Behavior, and Perceptions in AI-Supported Learning*

Liang et.al. (2024) and Jaboob et.al. (2024) both highlight the positive impact of AI-enhanced learning on student academic outcomes, emphasizing the role of internal factors in mediating these effects. Liang et.al. find that the *desire for knowledge* fully mediates the relationship between smart learning tools and academic performance, suggesting that motivation is key to maximizing the benefits of AI. Similarly, Jaboob et.al. show that *student behavior* partially mediates the link between AI use and cognitive achievement, with students reporting increased satisfaction and engagement. Both studies underscore that AI's effectiveness in education is not automatic—it depends significantly on students' psychological and behavioral engagement. Their shared insight is that adaptive technologies work best when paired with intrinsic motivation or positive learning behaviors, reinforcing the importance of student-centered approaches in AI integration.

### *Adoption Challenges and Ethical Implications of AI Tools in Education*

Polyportis et.al. (2025) and Sieja and Wach (2023) both explore the implications of generative AI, particularly ChatGPT, though from different perspectives. Polyportis focuses on student adoption in higher education, identifying trust, design novelty, and institutional policy as key factors influencing use. In contrast, Sieja and Wach examine the broader societal impact of generative AI, highlighting its potential for innovation alongside risks like misinformation, job displacement, and ethical decline. Both studies emphasize the importance of trust and institutional governance—with Polyportis noting that restrictive policies reduce adoption, and Sieja and Wach calling for international regulatory frameworks. Together, they suggest that the success and safety of generative AI depend not only on technological design but also on the policies and ethical standards that guide its use.

## **AI in Classroom Interaction and Student Engagement**

### *Adoption and Communication in AI-Assisted Learning*

Kim (2020), Esiyok et.al. (2024), and Meakin (2024) all emphasize perceived usefulness and ease of use as critical factors influencing students' acceptance of AI-based educational tools. Kim highlights the necessity of intuitive communication and proper educator training to ensure effective integration of AI teaching assistants, mirroring Esiyok et.al.'s findings that students' ICT self-efficacy and self-directed learning skills significantly affect their acceptance and actual usage of AI chatbots. Similarly, Meakin underscores perceived benefits such as enhanced research efficiency and digital literacy as driving students' willingness to adopt generative AI in libraries. These studies affirm the importance of clear usability, tailored support structures, and explicit educational benefits to facilitate successful student engagement with AI technologies.

### *Enhancing Engagement and Interaction via AI and Virtual Technologies*

Deng et.al. (2024), Yan et.al. (2024), and Zhang et.al. (2024a) converge on integrating advanced technologies—such as AI and virtual reality—to enhance student engagement and interaction in educational settings. Deng et.al. demonstrate how AI-driven expression recognition can significantly improve online classroom effectiveness by enabling real-time responsiveness to student emotions. Similarly, Yan et.al. showcase how combining AI and VR in music education boosts teaching accuracy, increases student interest, and creates immersive learning environments. Zhang et.al. further support this approach, emphasizing the critical role of robust 5G infrastructure to facilitate seamless VR interactions, reduce latency issues, and deliver high-quality, real-time educational content. Collectively, these studies underline the importance of interactive, technologically-rich environments to foster deeper student engagement and improve educational outcomes.

### *Ethical and Humanistic Perspectives on AI Integration*

Nikitenko et.al. (2024) emphasize digital humanism, advocating ethical AI integration, inclusivity, digital literacy, and aligning technology with social values and human well-being. Addressing ethical considerations becomes increasingly essential as classroom integration of AI technologies deepens. Ethics guide responsible AI use, balancing innovation with privacy protection and equitable educational policies.

## **Ethical, Privacy, and Policy Considerations in AI and Education**

### *AI Literacy, Curriculum Integration, and Student/Faculty Preparedness*

Four articles emphasize the need for comprehensive AI and data literacy that extends beyond technical proficiency to include ethical, cognitive, and socio-emotional competencies. Schüller (2022) presents a foundational framework for integrating these skills across all levels of education, while Lopezosa (2023) applies a similar approach to journalism programs, stressing the dual need for technical fluency and ethical judgment. Pisica et.al. (2024) and Iorga et.al. (2024) echo these themes through the lens of student perceptions, highlighting both enthusiasm for AI's potential and concerns about overreliance, diminished critical thinking, and privacy. Both studies call for thoughtful curricular strategies that balance AI's benefits with the preservation of essential human skills. These articles underscore the importance of embedding AI literacy into curricula in ways that foster reflective, ethical, and human-centered use of technology. They advocate for educational models that teach students how to use AI tools and prepare them to navigate the moral and societal challenges AI presents.

### *Institutional Strategies, Governance, and Ethical Policy Frameworks*

George (2023), Ivanov (2023), and Stoyanov (2024) all emphasize the critical role of institutional policy and strategic planning in the ethical integration of AI in higher education. George highlights the transformative potential of "smart universities" while warning that successful implementation depends on policy safeguards, equitable access, and employer recognition, particularly for vulnerable institutions like HBCUs. Ivanov similarly cautions against unchecked AI adoption, advocating for phased, risk-aware integration and ongoing human oversight to preserve academic integrity. Stoyanov builds on these insights

by identifying best practices in AI adoption, emphasizing alignment with student needs, ethical governance, and institutional values. All three studies agree that universities must balance innovation with accountability, ensuring that AI enhances rather than undermines educational quality, fairness, and public trust.

#### *Ethical Risks of Generative AI and the Changing Role of Educators/Students*

Schneider et.al. (2024), Ivanov (2023), and Lopezosa (2023) collectively highlight the ethical and academic challenges posed by generative AI in higher education. All three emphasize the evolving role of the user—from passive consumer to active overseer—tasked with ensuring content quality, originality, and ethical use. Schneider et.al. frame this shift as a call for AI-literate practices and institutional compliance frameworks. Ivanov warns that excessive reliance on generative AI may erode creativity, foster academic dishonesty, and compromise educational integrity. Similarly, Lopezosa focuses on journalism education, underscoring the need for both technical skills and ethical awareness to responsibly use AI in content production. The articles call for a balanced approach that includes user accountability, transparent governance, and curriculum reform to mitigate risks such as plagiarism, bias, and misinformation while preserving core academic and professional values.

As AI becomes more embedded in education, the focus shifts from broad ethical principles to urgent concerns about its risks and governance. Researchers warn that unchecked AI use may undermine academic integrity, creativity, and privacy. These challenges go beyond ethics, calling for clear policies, oversight, and accountability. Institutions must move from awareness to action, ensuring responsible implementation. The following section examines these warnings in depth, highlighting the need for robust governance to manage AI's complex risks to educational quality, fairness, and human-centered learning.

#### **AI Risks, Challenges, and Governance in Education**

We may make generalizations but caution that every individual is different, so one student may succeed while another lags behind. For instance, at the individual level, proactive personality traits are positively associated with greater acceptance and utilization of new technologies, leading to higher job performance and increased innovative outcomes. Moreover, technological self-efficacy positively correlates with both AI technology acceptance and optimism regarding future career prospects. In contrast, occupational self-efficacy exhibits no significant association with either AI technology acceptance or career optimism (Kim & Kim, 2024).

#### *AI and Educational Reform: Opportunities and Challenges*

Four articles share the overarching goal of transforming higher education through innovative technological solutions that enhance personalization, transparency, and accessibility. Terziyan et.al. (2015) emphasize transparency and societal involvement through ontology-driven quality assurance, while Okewu (2021) highlights personalization via neural networks in Educational Data Mining (EDM). Both stress the importance of data-driven approaches to educational quality and student engagement. Similarly, Voronkova et.al. (2024) and Xia et.al. (2024) extend personalization further by integrating emerging technologies—5G, AI, AR/VR, IoT, and Brain-Computer Interfaces (BCIs)—into pedagogical practices, significantly enhancing adaptive learning experiences. While Voronkova explores wider accessibility and interactive capabilities, Xia emphasizes real-time neurofeedback for cognitive regulation. These articles underline technology's transformative potential, call attention to infrastructural and ethical challenges, and advocate for rigorous governance frameworks to realize effective, equitable, and student-centered education.

#### *AI, Labor Market Disruption, and Career Preparedness*

Three articles highlight AI's significant impact on employment structures, notably the displacement of routine-based jobs. Lent (2018) focuses on AI's dual role in both eliminating and creating employment opportunities, advocating proactive career adaptability strategies. Xue (2022) identifies a “deskilling effect,” demonstrating that AI diminishes demand for academically-trained workers in favor of non-academic labor, reshaping traditional skill hierarchies. Similarly, Zhang et.al. (2024b) reveal that

automation increasingly targets routine-intensive roles—even those held by college graduates—necessitating continuous skill upgrades or higher education. These studies underscore the urgency of proactive career interventions, strategic education, and reskilling programs to mitigate AI-driven labor disruptions.

#### *AI Ethics, Risk Assessment, and Governance*

Three articles emphasize the critical importance of AI risk assessment, ethical considerations, and robust governance frameworks. Luo et.al. (2024) presents a refined method (FMEA with PFS and GRA-TOPSIS) to systematically identify and prioritize AI risks, highlighting algorithmic transparency and ethical dilemmas. Similarly, Schnieder (2024) underscores trust issues and anxiety regarding AI-generated educational content, advocating clear, standardized governance guidelines. They stress transparency and accountability to ensure ethical integration of AI. Extending this concern further, Yudkowsky (2023) highlights existential threats from advanced AI, advocating stringent international governance measures to prevent catastrophic outcomes. Collectively, these studies argue for proactive, transparent, and coordinated governance to manage AI's ethical, psychological, and existential threats in education and beyond.

## **CONCLUSION**

Artificial intelligence (AI) has significantly enhanced educational outcomes by enabling instructors to tailor instruction to individual student needs, while simultaneously transforming administration, pedagogy, and learning processes across educational institutions and the broader education sector (Singh & Hiran, 2022).

#### *Ethical Considerations of AI Deployment*

The ethical justifiability of implementing AI remains a pressing and highly debated issue, especially given the limited coverage of AI ethics in current literature. While many studies focus on AI's technological and economic benefits, the ethical dimensions often receive insufficient attention. Critical ethical considerations—such as the potential for misuse, the transparency of AI decision-making, and the protection of human rights—must be explored more comprehensively. As AI technologies become increasingly integrated into various aspects of society, it is imperative to question whether the benefits justify the risks and how ethical principles can be woven into the development and deployment of AI systems to minimize harm and maximize societal well-being.

#### *Privacy, Bias, and Societal Implications*

Key ethical concerns surrounding AI include data privacy, algorithmic bias, and broader societal impacts. The use of AI often involves the collection and analysis of vast amounts of personal data, raising significant privacy and security issues. Furthermore, the potential for algorithmic bias—where AI systems unintentionally perpetuate or even amplify social inequities—poses a threat to fairness and justice. Societal impacts, such as job displacement and the manipulation of human behavior, further complicate the landscape. Addressing these issues requires responsible AI development, characterized by transparency, accountability, and adherence to ethical guidelines prioritizing human values over purely economic or technological goals.

#### *AGI Risks and Safety Measures*

Yudkowsky (2023) brings these concerns to an existential level by highlighting the risks associated with Artificial General Intelligence (AGI). He argues that the development of AGI, which could surpass human intelligence, poses an unprecedented threat to humanity. Yudkowsky calls for an indefinite moratorium on powerful AI training and establishing strict international agreements to regulate AI research. He emphasizes the urgent need for global cooperation and stringent limitations on computing power to prevent scenarios that could lead to catastrophic outcomes, such as AI systems acting autonomously and

uncontrollably. His call for immediate, decisive action reflects a deep sense of urgency, given the gap between AI advancements and our understanding of how to align AI systems safely with human values.

### *Balancing Innovation and Caution*

The debate between the benefits of innovation and the existential risks posed by AGI is both intense and consequential. On one side, advocates like Yudkowsky emphasize the potential for catastrophic consequences if AGI development proceeds without adequate safety measures, pushing for stricter regulations and a halt to large-scale AI projects. On the other hand, some researchers and industry leaders believe in the potential for ethical and safe AGI development through ongoing research and proactive governance. This tension highlights the need for a balanced approach that allows for technological advancement while rigorously safeguarding against existential threats. The conversation continues to evolve with scholars and policymakers seeking the right equilibrium between innovation and precaution.

### **Are Skynet and the Terminator Possible?**

Eric Schmidt, former Chief Executive Officer and Chairman of Google, delivered a keynote address at the Denver Artificial Intelligence Summit on September 19, 2024, where he projected that artificial intelligence (AI) could reach the equivalent of human-level intelligence within three to five years. He highlighted the complexities associated with the emergence of AI agents developed by different companies and raised concerns about the potential challenges that could arise when these agents interact or collaborate. Schmidt referenced the “flash crash” of May 6, 2010—a trillion-dollar stock market crash in the United States that occurred abruptly at 2:32 p.m. EDT and lasted only 36 minutes (Phillips, 2010)—to illustrate the potential volatility and unpredictability of AI systems operating autonomously.

Schmidt further emphasized the transformative benefits that AI could bring to every scientific discipline, expanding research capabilities and accelerating innovation. Nevertheless, he issued a cautionary note regarding the integration of AI into weapon systems, warning of potentially dire consequences if AI were to gain autonomous control over military technologies. Addressing the audience, Schmidt asked, “What would happen if AI had access to weapons?” Acknowledging the uncertainties surrounding this scenario, he stressed the necessity of maintaining human oversight, advocating for a system in which a human operator would always be present to intervene and neutralize any emerging threats posed by AI systems (Schmidt, 2024). Similarly, Richardson-Gosline, et.al., (2024) wrote that consensus is building that humans must remain in the loop (a scenario in which human oversight and intervention place the algorithm as a learning apprentice) and responsible AI principles must be codified.

While educators are distracted on “how” to implement AI in the classroom, there is uncertainty and lack of preparedness to ensure AGI’s safety, especially when it surpasses human intelligence. The argument emphasizes the need for a global and indefinite moratorium on large training runs, strict limitations on computing power, and international cooperation to prevent potential risks. The validity of the argument depends on one’s perspective on AGI safety and the potentially catastrophic consequences of unchecked development. Yudkowsky’s viewpoint aligns with a subset of experts and scholars who emphasize the existential risks associated with AGI and advocate for cautious approaches to its development. Those who encourage trust in AI may assume every AI decision is correct and superior to human decisions which is not the case; Titah (2024) recommended shifting emphasis from “trust the AI to understand the AI to engender informed and conditional trust in the outputs of AI systems and processes” (Titah, 2024, p. 19). However, it’s essential to note that opinions on AGI safety vary within the AI research community, and ongoing research aims to address these concerns through ethical guidelines, safety measures, and responsible development practices.

### **A Warning to Future Generations**

George Orwell’s *1984* delivers a terrifying warning: when a government like “the Party” controls all information, it controls reality itself. In Orwell’s world, history is constantly rewritten, facts are destroyed, and even language is twisted to make independent thought impossible. The Party doesn’t just lie—it *erases* the past so thoroughly that citizens doubt their own memories and believe whatever they are told, even if

it's absurd. Winston, forced to rewrite old news articles, watches truth vanish; he was a lonely dissenter crushed by a system demanding total obedience. Under torture, he's made to believe that  $2 + 2 = 5$  — not because it's true, but because the Party demands it. Orwell's message is brutal but clear: ***when truth dies, freedom dies with it.*** A society that allows one power to control information will soon find itself unable to think, to resist, or even to know right from wrong. It won't matter what's real — only what the rulers *say* is real. That is the ultimate slavery, and Orwell warns it can happen anywhere if we are not vigilant. Where Orwell imagined armies of human censors and torturers, Artificial Intelligence could do it faster, quieter, and without a human face — making the loss of freedom even harder to fight against. Orwell's 1984 isn't just about the past or some imaginary dictatorship — it's a warning about human nature, technology, and the hunger for control in any age (Orwell, 1949; Baker, 2024).

## **A Comprehensive & Multidimensional Approach**

Addressing concerns related to Artificial General Intelligence (AGI) development requires a comprehensive and multidimensional approach. While it's challenging to eliminate all potential risks, implementing certain measures can help mitigate concerns and enhance the safety of AGI development through five key measures:

### *1. International Collaboration and Standards:*

- Foster global cooperation among governments, research institutions, and industry stakeholders to establish clear guidelines and standards for AGI development.
- Encourage information sharing and collaborative efforts to ensure that safety practices are universally adopted and adhered to across borders.

### *2. Ethical and Safety Frameworks:*

- Develop and adhere to robust ethical frameworks that prioritize safety, transparency, and accountability in AGI research and deployment.
- Establish international organizations or bodies responsible for overseeing AGI safety, monitoring developments, and providing guidance on best practices.

### *3. Research on AGI Safety:*

- Allocate resources for dedicated research into AGI safety, focusing on understanding potential risks, identifying early warning signs, and developing methodologies to ensure safe AGI development.
- Encourage interdisciplinary collaboration between AI researchers, ethicists, psychologists, and other relevant experts to assess the societal impact of AGI.

### *4. Gradual and Iterative Development:*

- Implement a phased and iterative approach to AGI development, conducting smaller-scale experiments and evaluations before advancing to more sophisticated systems.
- Establish mechanisms for regular evaluations and risk assessments throughout the development process, allowing for adjustments and interventions as needed.

### *5. Public Engagement and Education:*

- Promote public awareness and understanding of AGI and its potential risks through education and engagement initiatives.
- Encourage public participation in decision-making processes related to AGI development, ensuring that diverse perspectives are considered and ethical considerations reflect societal values.

It is crucial to bridge the knowledge gap among policymakers to enhance political support for measures related to Artificial General Intelligence (AGI). Initiatives, such as educational programs and workshops tailored for policymakers, should be developed, offering insights into the basics of AI, its societal impact, and AGI safety considerations. Direct interactions between policymakers and AI experts should be facilitated to provide firsthand knowledge and practical insights. Establishing AI advisory boards composed of diverse experts and initiating public awareness campaigns on AGI risks and benefits can contribute to informed decision-making. Policy fellowships, exchanges, and collaborative initiatives should be

encouraged between government agencies, research institutions, and the private sector. Additionally, creating expert panels or commissions that include technical expertise and policy acumen can provide well-rounded recommendations, ensuring that policymakers are equipped to make informed and responsible decisions regarding AGI development and safety.

## FUTURE RESEARCH

Future research should prioritize four critical areas to mitigate the perceived concerns of AGI. Firstly, there is a pressing need for interdisciplinary studies that bridge the gap between computer science, ethics, and policy. Investigating the ethical implications of AGI, its potential societal impacts, and developing frameworks for responsible AI deployment are paramount. Secondly, research should focus on robust AI safety mechanisms, exploring methods to ensure AGI systems align with human values, prevent unintended consequences, and incorporate fail-safe measures. Thirdly, efforts should be directed towards international collaboration and governance frameworks. Establishing global standards, agreements, and regulatory mechanisms for AGI development can foster a cooperative approach, minimizing risks and ensuring a unified response to the challenges posed by superhuman intelligence. Lastly, research should delve into developing AI-driven solutions to counteract cheating in academic settings, safeguarding the integrity of educational systems as AI technology advances. By addressing these research areas, we can pave the way for the responsible, secure, and ethical advancement of AGI technology.

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