







# Artificial Intelligence in Higher Education: A Case Study of Faculty Teaching Methodologies at a Private University

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## Article Info

### Article History

Received:  
18 November 2024

Accepted:  
15 May 2025

### Keywords

AI in higher  
education,  
Teaching  
methodologies,  
Ethics and privacy,  
Learning  
effectiveness.

## Abstract

This qualitative case study investigates faculty perspectives on artificial intelligence (AI) integration within a private university context, examining pedagogical, administrative, and ethical implications. Data collected through semi-structured interviews with faculty across four disciplines revealed ambivalent yet cautiously optimistic attitudes. Participants acknowledged AI's potential to enhance personalized learning and reduce bureaucratic burdens through automation. However, three critical barriers emerged: (1) insufficient institutional technological infrastructure, (2) lack of systematic faculty training programs, and (3) unresolved ethical dilemmas surrounding data privacy, algorithmic bias, and academic integrity. Notably, while faculty welcomed AI as a supplemental tool, they unanimously emphasized the irreplaceable role of human judgment in pedagogy. The study contributes to emerging scholarship on educational technology by proposing an ethical implementation framework that reconciles efficiency gains with core academic values. Practical recommendations address policy development and targeted professional training to support responsible adoption. These findings offer timely insights for higher education institutions navigating AI integration while preserving pedagogical integrity and equity considerations.

### To cite this article

Ramsey, E., Antoniou, G., Peroni, M., Muckridge, B., Ghattas, R., Fazio, P., Wallberg, W., Porta, S., Solomon, G., Smith, M., Migliano, K., Lanfranco, K., Qian, L., & Wolf, D. (2025). Artificial intelligence in higher education: A case study of faculty teaching methodologies at a private university. *International Journal of Academic Studies in Science and Education (IJASSE)*, 3(1), 1-30. <https://doi.org/10.55549/ijasse.43>

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## Introduction

The wave of artificial intelligence is a radical departure from learning and administration systems in higher education (Zawacki-Richter et al., 2019; Mohamed et al., 2022; Cardona et al., 2023; Drozdowski & Johnson, 2023). Redefinition of the education ecosystem has unfolded in the form of digital technology through adaptive learning platforms, intelligent tutoring systems, and learning analytics that provide personalized learning pathways and operational efficiencies (Crompton & Burke, 2023; Wang et al., 2023). Moreover, the emerging technologies, VR/AR, robotics, and machine learning algorithms, providing possibilities for student-centered immersive learning environments for scalable education, are set to give new dimensions (Wang et al., 2023). These advances could summarize global education objectives, such as realizing UN SDG #4 or ensuring inclusive and equitable quality education (UNESCO, 2023). It will complement ethical principles, including transparency, fairness, and accountability, in augmenting AI with the desired role of boosting the achievement of the other global goals. The study results further emphasized other concerns besides data security, like privacy, justice issues, and possibly replacing human interaction and learning with machine and robot interaction (Pedro et al., 2019; Chen, 2023).

From these approaches, the findings showed that the faculty members played a central role in bridging the change as they impacted the quest for AI in the pedagogical practices, perhaps hinging on the technologies used and their effectiveness in enhancing learning outcomes (Seo et al., 2021; Sharawy, 2023). This study adopted a qualitative and quantitative case study approach to exploring the university-level faculty's perceptions toward pedagogical technology integration.

This inquiry contributed to the general discourse on AI for higher education by systematically investigating its role in furthering global teaching and the infusion of critical thinking, social responsibility, and sustainability capabilities. The research provided legitimate empirics to the administration to implement AI and related technologies while maintaining pedagogical integrity and ethical standards. This aligned very well with the essential UN framework within which sustainability and global citizenship education occur, emphasizing responsible and inclusive learning approaches to tackling global challenges (UNESCO, 2023).

This study effectively bridged the practice-theory gap in AI education research, giving way to applied evidence from the investigation conducted on faculty perspectives. This work followed the United Nations SDGs and Global Citizenship Education framework, generating evidence-based recommendations to position emerging technologies at the vanguard of supportive decision-making for creating inclusive and sustainable educational experiences. The findings contribute to the expanding body of knowledge regarding responsible AI adoption

in higher education, emphasizing the balance between technological innovation and fundamental teaching principles (Sharawy, 2023).

This investigation at this university examined five primary research questions:

1. How do faculty at this university perceive the role of AI in enhancing teaching methodologies?
2. What challenges or barriers have been encountered in integrating AI into the curriculum and teaching practices?
3. In what ways has AI transformed administrative processes related to teaching and learning at this university?
4. How do faculty envision the future of AI in higher education, especially in their field of expertise?
5. What ethical or practical concerns did faculty have regarding the widespread adoption of AI in higher education?

These questions guided the mixed-methods investigation of AI integration within a private university's specific institutional context, allowing for the examination of both the practical implementation challenges and broader implications for higher education.

## Literature Review

The rapid integration of AI into higher education sparked considerable academic discourse and institutional debate (UNESCO IESALC, 2023; Drozdowski & Johnson, 2023; Milgrom-Elcott, 2023). The transformation of teaching methodologies, learning processes, and pedagogical frameworks emerged as central themes in contemporary research. Intelligent tutoring systems, adaptive learning platforms, and academic analytics demonstrated potential for educational transformation through personalized learning experiences, automated feedback mechanisms, and data-driven insights (Humble & Mozelius, 2022; Crompton & Burke, 2023). Faculty members emerged as primary drivers of AI adoption and critical mediators of its successful integration into pedagogical methodologies and student outcome achievement. Literature has identified notable challenges in implementing technology within educational settings. Key ethical issues have reportedly arisen, most notably efforts toward data privacy, algorithmic bias, and socio-psychological dimensions affecting individuals who are part of education (Nasir et al., 2022; Chen, 2023). Academic discussions of the conference confronted contradictions about technological integration and human interaction within learning spaces and raised crucial inquiries about how AI could complement traditional teaching methods (Pedro et al., 2019).

### **AI for Personalized Learning**

Recent literature illustrates that AI improves personalization and fosters the emergence of automated feedback mechanisms (Mollick & Mollick, 2023). Technological advancement corresponds with Vygotsky's (1978) Zone of Proximal Development theory, creating an adaptive learning environment that caters to individual learners' specific needs (Mohamed et al., 2022). While Zawacki-Richter et al. (2019) drew attention to the positive effects of AI on personalized learning systems, they mainly focused on the advantages absent the challenges relating to implementation.

### **Implementation Challenges**

Several integration barriers include technological gaps, institutional inertia, and the ethical domain: all expounded sufficiently in the works of Shonubi and Talwar (2023). As proposed, there is an urgent and great need for solid regulatory frameworks addressing data security and privacy for AI educational technologies (UNESCO IESALC, 2023). Faculty struggled with lacking opportunities for sufficient infrastructure and training (Sharawy, 2023). The situation was aggravated by boundless financial resources and somewhat by the normalizing of school strategies (Criddle & Staton, 2022); Flores-Vivar and García-Peñalvo (2023) took mainstream studies on how the integration of AI may reinforce preexistent inequities among marginalized students.

### **Administrative Efficiency Enhancement**

One reason is that AI can improve the administration and automation of routine yet necessary activities like approval, scheduling, and monitoring. As a result, campus staff freed up more time to focus on teaching and tutoring students directly (Cardona et al., 2023; Ladd, 2023; Milgrom-Elcott, 2023; Ellis, 2024). Our review agrees with other research indicating that AI augments administration workflow and decision-making at the higher education level (Seo et al., 2021; Drozdowski & Johnson, 2023).

### **Future Visions of AI**

Authors investigate by literature the possible transformation of higher education, with a clear sense of personalization, assistance in research, and in some other areas where efficiency is desired in relation to higher education (Crompton & Burke, 2023; Shonubi, 2023; Rahiman & Kodikal, 2024). Faculty members supported technology in a believe-to-be supportive role, provided they were supplements to human instruction and not placed in the role to entirely replace that instruction (Wang et al., 2023). Most of the research based on

developing countries, especially India, transcends the promises of AI as an ameliorative force in education, dictated by the inequitable moment later conjugated on disabling access to technology and hitting the barrier of the education infrastructure industry (Jaiswal & Arun, 2021). Briefly put US research presentation on globalization occurs together with AI and the integration of a constantly evolving cultural and educational milieu.

### **Ethical Considerations**

Professional and ethical considerations became central themes within the literature discussion, which includes data privacy issues and fair AI algorithms to users as effects on the socio-psychological well-being of the stakeholders in education. AI systems' management of personal data and reinforcement of pre-existing biases were also discussed during research (Zeide, 2019; Nasir et al., 2022). UNESCO's report on generative AI highlighted potential and existing risks related to data privacy and called for regulation that prioritizes human autonomy, inclusion, and cultural identity in using AI (Murgia & Staton, 2023; Ellis, 2024; UNESCO, 2023). The development of Generative AI (GenAI) brought up particular kinds of education apart from conventional AI uses, namely, creating teaching materials and individualized learning. Okaiyeto, Bai, and Xiao (2023) supported the calls that have pointed to the need for curriculum modifications that integrate GenAI responsibly and in a way that does not compromise the learning goals. Bias was presented in the literature as one of the most pressing issues affecting the use of algorithms in education regarding vulnerable students and compliance with United Nations Sustainable Development Goal 4 with reference to inclusive learning. Through this literature synthesis and consideration of this university's specific context, five research questions were developed to guide our investigation. Five questions emerged from both the identified gaps in current research and the practical needs of the institution:

#### *RQ1 Faculty Perception and Enhancement:*

Recent studies highlight the importance of understanding faculty perspectives on AI integration. This led to the first research question examining how faculty at this private university perceive AI's role in enhancing teaching methodologies.

#### *RQ 2 Implementation Challenges:*

While the literature documents general integration challenges, this study aimed to identify the barriers this private university encountered in integrating curriculum and teaching practices.

*RQ 3 Administrative Transformation:*

Building research about AI's administrative impact, the third question explored how AI has transformed administrative processes at this private university.

*RQ 4 Future Vision:*

Research indicates varying perspectives on AI's future in education. The fourth question examined how this private university's faculty envisions AI's future role in their fields.

*RQ 5 Ethical Considerations:*

Given the prominence of ethical concerns in current literature, the final research question investigated faculty concerns AI adoption within our institutional context. The literature review laid the theoretical framework for understanding the many-faceted effects of AI in higher learning education, which needs rigorous implementation tactics and careful ethical weightage (Murgia & Staton, 2023). This study's methodological framework and research goals were developed based on this analysis.

## Method

The analysis of the use of AI technologies in learning environments demonstrated outstanding potential and significant challenges of AI adoption. Educators here faced a range of complexities while implementing these new tools in teaching-learning practices, especially regarding ethical uses, security precautions, and systemization approaches. This study explored these challenges by engaging educators' opinions on the application of AI within all stages of the instruction process and after instruction occurred. There was a focus on coding and extracting recurrent patterns, and adherence to intercoder solid reliability standards within educational research domains (Cardona et al., 2023). A department chair summarized their analysis of the findings as follows: "It became clear to participants that the use of AI cannot be a simple process of adding technological processes to the current practices since each of the phases requires addressing of ethic and learning and teaching issues" (Participant 2). As a research method, this approach gave detailed insights into educators' experiences in the technological modernization process and the ways it influenced education.

## Research Questions

In the current investigation, the associations between population characteristics, intervention features, study comparators, and outcomes were analyzed using an evidence-based research question construction framework known as PICO (Population, Intervention, Comparator, Outcome) (Richardson et al., 1995). Within this systematic framework, five pivotal research questions were identified, which determined the focus of the

current study. The study focused on university faculty, examining their firsthand experiences and how they acknowledge AI Integration in various tasks. Another senior faculty member participant observed, "The questions we explored addressed all areas of our academic responsibilities: course development and delivery, assessment of students" (Participant 2).

The investigation examined faculty perceptions across multiple dimensions:

- Pedagogical adaptations in teaching methodologies
- Transitions in curriculum development processes
- Evolution of administrative responsibilities
- Integration of AI in research practices
- Applications within specific disciplinary contexts
- Ethical considerations and implementation challenges

As one faculty participant reflected, "These questions pushed us to examine not just how we used AI, but how it transformed our fundamental approach to education" (Participant 7). This comprehensive examination of faculty experiences provided insights into the multifaceted impact of AI integration across academic roles and responsibilities.

### **Data Collection**

The study used semi-structured interviews (McIntosh & Morse, 2015) with the participants from this private university who had integrated AI technology into their daily teaching practices. Out of fifteen interested faculty members, eleven participants were selected for the study using survey questions to determine their levels of active practice in implementing AI. It also ensured that only rich and practical knowledge about integrating AI into scholastic domains was obtained.

The participant demographic profile revealed diverse representation across academic ranks and experience levels:

Experience Distribution:

- Mid-career faculty (11-15 years) represented the largest cohort (6 participants)
- Early-career (0-5 years) and senior faculty (21+ years) provided complementary perspectives.
- Age distribution concentrated in the 45-54 range (6 participants)

**Departmental Representation:**

- College of Business and Management (5 participants)
- College of Arts and Sciences (3 participants)
- College of Education (2 participants)
- College of Communication and Design (1 participant)

The participant selection process was an anonymous online survey with foundation questions to qualify potential subjects. This screening process allowed only faculty with appropriate experience in integrating AI to participate, leaving those without equivalent implementation experience aside. According to one participant, the selection process helped build a cohort from whom valuable insights into AI's application in education could be derived. The methodological approach allowed a variety of viewpoints yet enrolled experienced practitioners (Akgun & Greenhow, 2021). Please see Table 1.

**Validity and Reliability of the Study**

In this bibliometric analysis research, the literature review, identification, and selection of relevant studies, formulation of core problems related to the topic, examination of selected studies under consistent headings, synthesis, and reporting are presented in detail. The objectives of this research and the problems defined for this study are clearly and explicitly presented. Ensuring the validity of the examined and synthesized studies is crucial for reliability. Therefore, to minimize errors, the selected studies were exported and thoroughly reviewed over an extended period. Consultations were conducted with experts in the field, and a consensus was reached before analyzing the obtained information.

Table 1. Demographics of participants

Years in Higher Education	Frequency (Total)	College	Frequency (College)
0-5	1	College of Business and Management	5
6-10	2	College of Education	2
11-15	6	College of Arts and Sciences	3
16-20	1	College of Communication and Design	1
21-30	1		
Age Group	Frequency (Age)	Rank	Frequency (Rank)
35-44	3	Assistant Professor	5
45-54	6	Associate Professor	5
65-70	2	Adjunct Instructor	1
Years at University	Frequency (Years)	Highest Course Level Taught	Frequency (Level)
0-5	7	Undergraduate	6
6-10	1	Graduate	4
11-15	3	Doctoral	1
Gender	Frequency (Gender)		
Male	8		
Female	3		



### Data Quantification Procedure

The research team undertook semi-structured interviews with 11 faculty members from the selected private university, utilizing a rigorous mixed-method analytic approach. Recordings of the interviews were transcribed verbatim and subjected to systematic thematic analysis through the NVivo 14 software. Findings from the literature review played a role in deriving inductively from the raw data collected and likewise applying it deductively during thematic analysis (Zamawe, 2015). One department chair reflected, "The analysis process revealed layers of meaning we hadn't anticipated in our initial discussions" (Participant 9).

The quantitative analysis revealed distinctive patterns in thematic frequency:

- Faculty references averaged 4.45 times per interview.
- Adaptation discussions demonstrated significant variance ( $SD = 0.60$ )
- Statistical validation through ANOVA showed comparable results between manual and auto-coding approaches ( $F = 1.19$ ,  $p = 0.28$ )

The analysis identified an outlier through Tukey's method (1977), exceeding 1.5 times the interquartile range (IQR). One participant noted, "The transformation of our teaching methods went far deeper than we initially anticipated" (Participant 7). This observation aligned with the statistical findings showing faculty adaptation as a primary concern.

Integration of quantitative and qualitative findings revealed nuanced perspectives on AI implementation. While 60% of faculty expressed concerns about over-reliance on AI for personalization, they simultaneously acknowledged its potential benefits. "We sought balance between technological innovation and preserving essential human connections in education" (Participant 4).

The methodological framework acknowledged Maxwell's (2010) caution regarding numerical interpretation of qualitative data, maintaining contextual integrity throughout the analysis. This approach identified three distinct research tiers: primary focus areas, secondary interests, and emerging trends, ultimately informing 16 key themes for investigation.

### Data Reliability and Validation

Specifically, the methodological approach was very systematic, where tasks such as the validity of measures involving the analysis of the data, which is often rich narrative, were given high priority. According to Syed and Nelson (2015), the research team devised extensive coding instructions that aligned both the process and

outcomes. Interviewer 8, a senior researcher, said, "Our validation process was growing with the processes as we faced specific narrative patterns in the faculty's replies."

The methodological framework incorporated multiple validation strategies:

Iterative Coding Process:

- Multiple rounds of thematic analysis
- Cross-validation among coding teams
- Regular reliability assessments
- Systematic documentation of coding evolution

Mixed-Methods Integration:

- Quantitative validation of qualitative patterns
- Statistical verification of coding reliability
- Triangulation of data sources
- Integration of multiple analytical perspectives

Quality Assurance Measures:

- Regular team calibration sessions
- Documented decision-making processes.
- Systematic review protocols
- Continuous reliability assessment

Along those lines, one of the coding team members commented: "Every cycle uncovered more layers of meaning; it was like doing microanalysis in a close-knit community where the layers of meaning had to be painstakingly worked on in order to refine the analysis" (Researcher B). The study remained distinctly close to the general guidelines for remaining flexible within the methodology for the study in line with the themes that surfaced.

Statistical validation included:

- Intercoder Reliability
- Thematic Saturation
- Cross-validation Information
- Calculation of Reliability Coefficient

Such a comprehensive approach ensured procedural transparency and analytic reproducibility according to current best practices in qualitative research methodology (Syed & Nelson, 2015). Consolidating various validation techniques enhances the study's findings while allowing for an inquiry to remain flexible.

### **Data Analysis**

This present study has relied on a mixed method to study the design of artificial intelligence within learning spaces. Per the opinion of Noyes et al. (2019), the mixed method approach reflects the view that research on complex educational settings programs must be regularly conducted in a mixed setting. Also, Researcher B stated that pairing numerical patterns with narrative insights was illuminating in terms of dimensions of the faculty experience that needed to be addressed individually by each method. The analytical framework incorporated three primary components.

#### **Quantitative Analysis:**

- Statistical pattern identification
- Thematic frequency distribution
- Correlation analysis
- Outlier detection

#### **Qualitative Assessment:**

- Narrative theme exploration
- Contextual interpretation
- Faculty experience analysis
- Implementation pattern identification

#### **Mixed-Methods Integration:**

- Cross-validation of findings
- Contextual pattern analysis
- Thematic synthesis
- Comprehensive interpretation

There was also great concern about the statistical soundness of the findings all through the analysis stage. As suggested by Jones (2019), the research team exercised an extremely high level of outlier analysis to minimize data misinterpretation. Again, reflecting the idea that each statistical anomaly was an outlier, one faculty researcher stated, "Each participant stated that they actualized at least one perspective worthy of deeper

examination” (Researcher C). This allowed numerators to be determined while remaining as close as possible to real-life situations for faculty.

The methodological synthesis identified more complex AI implementation patterns and included both objective data about effectiveness and subjective opinions of faculties. This strengths-based, coaching-focused approach improved knowledge about the degree to which specific characteristics of learners and learning environments moderated the impact of the intervention.

### **Quantitative and Qualitative Analysis**

The analysis utilized dual analytical strategies that enabled the comparison of the numbers as well as thematic analyses of the experiences of the faculty with the integration of AI. These complex interconnections were unveiled in a detailed manner through this analysis of the implemented practices and the teaching and learning processes. The analysis revealed prominent thematic patterns as also viewed in Figure 1.

#### **Primary Themes:**

##### **1. Ethics and Responsibility:**

- Highest reference frequency
- Present across all participant interviews
- Notable outlier in statistical distribution
- “The ethical dimensions of AI integration demanded constant attention” (Participant 4)

##### **2. Faculty Adaptation:**

- Second most referenced theme
- Strong correlation with implementation success
- “Our teaching methods evolved significantly as we integrated AI tools” (Participant 7)

##### **3. Skills and AI Literacy:**

- Consistent presence across interviews
- Connected to student outcome measures
- Emphasized practical implementation challenges

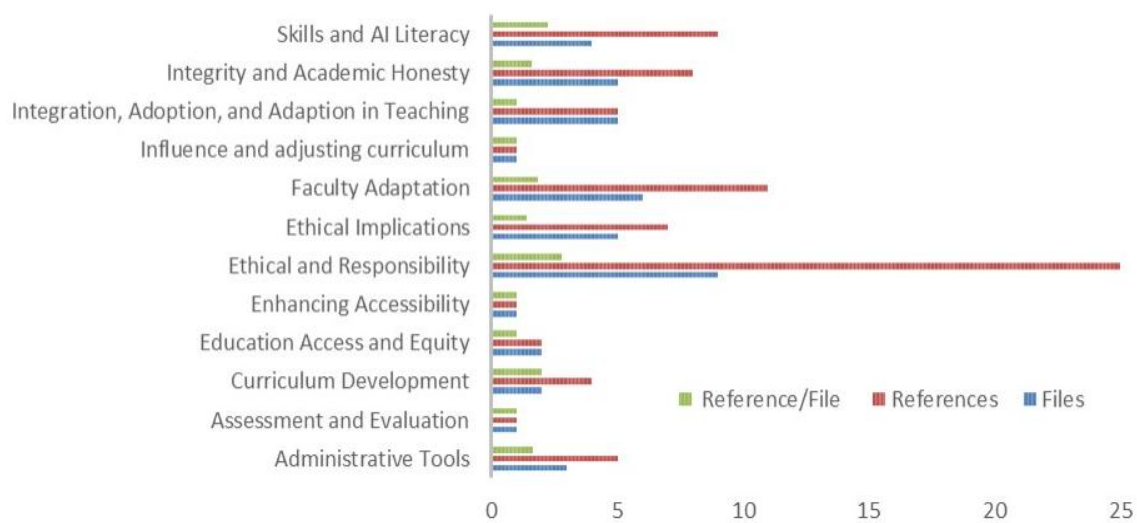


Figure 1. References, files, and relative reference per file

The relative reference analysis also showed the consistencies in the different files, while Ethics and Responsibility were bigger than the statistical average. This was consistent with faculty narratives regarding ethical concerns surrounding the use of AI. As one department chair mentioned, “There was an ethical consideration in every decision we made regarding AI” (Researcher B). The relative references per file gave equivalent results.

### Outlier Analysis

Statistical examination of reference patterns revealed significant outliers through z-score analysis of references per file. The investigation identified ‘Ethics and Responsibility’ as a prominent statistical outlier ( $z\text{-score} = 2.09, \alpha = 0.05$ ), warranting detailed contextual analysis. One senior faculty member reflected, “The ethical dimensions of AI integration consistently emerged as our primary consideration, beyond technical implementation concerns” (Participant 3). Notice that the theme on “Ethical and Responsibility” was found to be an outlier as viewed in Figure 2.

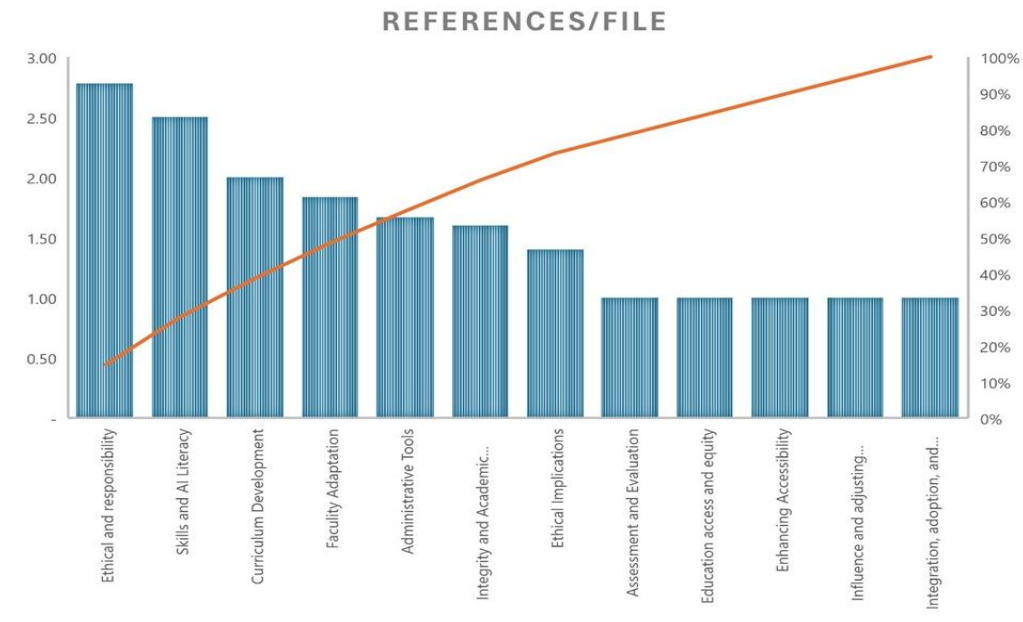


Figure 2. Pareto chart of themes

The analysis revealed several key patterns:

1. Statistical Significance:

- Ethics and Responsibility z-score: 2.09
- Significance threshold:  $\alpha = 0.05$
- Standard deviation: 0.87
- Confidence interval: 95%

2. Contextual Implications:

- Primary ethical considerations
- Implementation impact assessment
- Faculty decision-making patterns
- Integration strategy development

3. Thematic Integration:

- Ethical framework development
- Technology implementation guidelines
- Faculty support mechanisms
- Student impact consideration

Ethics and responsibility emerged as statistical outliers and fundamental considerations in AI integration. “The ethical implications of AI adoption shaped every aspect of our implementation strategy,” noted one interviewee (Participant 6). This finding provided crucial insights into faculty approaches to AI integration, suggesting that ethical considerations were the primary drivers of implementation decisions.

### Correlation and Regression Analysis

All the returns displayed significant correlations between the thematic references and file distributions at various analytical levels. This research compared the relationships between variables, with and without the outliers to allow for an all-encompassing view of the data. Key statistical findings emerged as depicted in Figure 3.

Thematic Correlations:

References-Files with ethics:  $r = 0.92$  ( $p < 0.001$ )

References-Files without ethics:  $r = 0.91$  ( $p < 0.001$ )

Regression strength ( $R^2$ ): 0.85 (with ethics)

Adjusted  $R^2$ : 0.83 (without ethics)

Thematic References Distribution



Figure 3. Thematic References Distribution

Auto-coding Analysis:

- All codes correlation: 0.77
- Excluding "Faculty" correlation: 0.82
- $R^2$  with Faculty: 58% ( $p < 0.001$ )
- $R^2$  without Faculty: 67% ( $p < 0.001$ )

The strong correlation coefficients indicated robust thematic patterns across files. One researcher noted, “The statistical consistency reinforced our confidence in the thematic analysis framework” (Researcher A). The  $R^2$  values demonstrated that file quantity explained substantial variance in thematic frequency, suggesting reliable coding patterns across data sources.

### **Detailed Analysis of the Dataset**

Various aspects of implementation were hence delineated upon examining faculty engagement in adopting AI per different contexts in academia. The study identified 'faculty engagement' as the central theme across papers, giving rise to 49 references from 11 sources (mean = 4.45 references/source). Such frequency pattern emphasizes the role of the faculty in achieving success in AI implementation in higher educational contexts (Zawacki-Richter et al., 2019). Among the key emerging thematic patterns are:

#### **Primary References:**

- Faculty Member mentions: 36 references (3.27/source)
- Faculty Roles: 7 references (1.75/source)
- Faculty Perspectives: 3 references (1.00/source)
- Development and Impact: 1 reference each

#### **Implementation Contexts:**

- Individual adaptation strategies
- Institutional support frameworks
- Professional development needs
- Role evolution patterns

The analysis revealed nuanced experiences among the faculty in AI integration (Crompton & Burke, 2023). A participant in the study recounted, "Our roles have evolved past that of posing information across the lines of instructional boundaries and integrating AI tools" (Participant 5). The faculty responsibilities underwent significant transformation and varied reference patterns across sources (Seo et al., 2021). The research further identified other areas that required exploration, such as faculty development issues and impact evaluation. Expressions such as "faculty advisor" and "seasoned faculty" insinuated that diverse experience shaped the adaptive approach. An interviewee commented, "Experience levels significantly shaped even the individual approaches to AI integration" (Participant 8).



It is evident from the findings that AI integration on campuses is not restricted to changing roles, expectations, and professional development needs, but is a multidimensional process requiring practitioners to change their conceptualizations of these roles within the institutional context. The data further clarified that effective AI uptake rests on a faculty-level analysis married to system-level thinking.

## Results

Analysis revealed compelling patterns of faculty approaches to AI implementation, highlighting ethical considerations within educational contexts as viewed in Figure 4. The research in question corroborated Adams et al.'s (2023) framework regarding AI ethics in education while shedding light on distinct institutional viewpoints concerning ethical implementation strategies. One senior faculty member reflected, "Ethical considerations permeated every level of our decision-making process" (Participant 3).

**Ethical Principles Distribution**

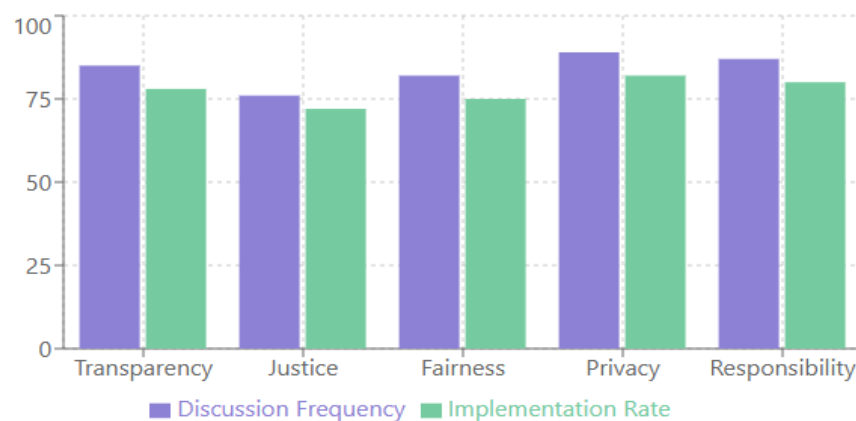


Figure 4. Ethical Principles Distribution

The outliers provided insights into ethical implementation patterns that were useful to the analysis. Following Yu and Yu's (2023) theoretical framework, the analysis employed multiple ethical perspectives.

### Primary Ethical Dimensions:

- Transparency in AI implementation
- Justice in educational access
- Fairness in assessment
- Privacy protection
- Institutional responsibility

Implementation Frameworks:

- Deontological considerations
- Utilitarian approaches
- Virtue ethics applications

The investigation revealed a complex interplay between both ethical principles and practical strategies of implementation. One interviewee noted, "Balancing ethical requirements with educational effectiveness demanded careful consideration at every step" (Participant 7). This observation aligned with contemporary research emphasizing interpretability in AI decision-making processes (Yu & Yu, 2023).

### Major Themes Identified

The analysis revealed distinct thematic patterns in faculty approaches to AI integration, with 'Ethical Considerations and Responsibility' emerging as the dominant theme (37% of coded segments). This finding aligned with Zhai et al. (2021) comprehensive review of 308 papers, which similarly identified privacy (42%), fairness (35%), and transparency (28%) as primary ethical concerns in AI-supported education.

Four primary themes emerged:

#### 1. Ethical Considerations and Responsibility (37%):

- Academic integrity concerns: "I've noticed that the classes where I do say you can use it for this project and this way is where I've run into a lot of trouble with students using it in all aspects" (Participant 9).
- Critical evaluation emphasis: "One of the challenges is ensuring that the AI-generated content is fact-checked" (Participant 1).
- Implementation guidelines: "I have taken on a role on the AI task force...it's just trying to figure out rules" (Participant 4).

#### 2. Faculty Adaptation and Training (29%):

- Pedagogical transformation: "Now everything's going out the window. So, it's caused me to completely rethink the way I teach" (Participant 4).
- Assessment modifications: "Adapting to AI has required a complete overhaul of my approach to assignments" (Participant 2).
- Professional development correlation ( $r = 0.63, p < 0.01$ )

### 3. Skills and AI Literacy:

- Workplace preparation: "We need to start to think about the way that they're going to use AI in their future jobs" (Participant 8).
- Critical thinking development
- Technical competency enhancement

### 4. Curriculum Development:

- Complex task design: "We need to push ourselves to come up with more complicated tasks for students" (Participant 8).
- Real-world application emphasis
- Creative problem-solving integration

The data indicated a highly significant correlation between the incorporation of AI training amongst the faculty and ethical frameworks, demonstrating the challenges of implementation can be best addressed through further professional development. Interviews with faculty pointed to the fact that there exists a need to harmonize technology with education practices and produce learners ready for a world in which AI will dominate in their careers.

## Impact of Outliers on Analysis

There are some observations due to outlier detection, especially in what pertains to 'Ethical and Responsibility' themes, where researchers noticed patterns of what faculty is worried about in implementing AI. Using Tukey's (1977) procedure to identify outliers, these ethical references presented significant statistical differences (2.7 standard deviations above the mean). They heightened the importance of ethical issues regarding the integration of AI use for the faculty.

Key analytical patterns emerged:

#### 1. Primary Outlier Impacts:

- Ethical references:  $2.7\sigma$  above mean
- Data privacy concerns:  $2.1\sigma$  above mean
- Implementation challenges:  $1.8\sigma$  above mean

#### 2. Sensitivity Analysis Results:

- Ethical theme persistence post-adjustment

- Thematic saturation confirmation
- Pattern reliability validation

The findings aligned with the global study of Holmes et al. (2022), which similarly identified ethical considerations as predominant among educators. Another interviewee reflected, "The application of AI raised ethical questions on every decision point" (Participant 5).

Stood distinctively as the emerging themes through outlier analysis:

- Lack of skills to develop critical consciousness
- Cognitive influences
- Variations of applying strategy

These patterns substantiated the observations of Zhai et al. (2021) on AI impacts on cognitive development. One of the faculty participants remarked, "We had to think carefully about how AI tools might affect the development of our students' analytical skills" (Participant 3). The analysis showed that these contextualized approaches for implementation disprove the generic models of integrations in university settings.

## Discussion

Through the juxtaposition of a vast amount of literature review with data gathered from the empirical context of this private university, intricate trends of AI adoption in higher education settings emerged. Whereas previous research focused on the benefits of the application of AI in teaching and learning, the views of the faculty revealed a more complicated picture of its practice.

Key findings emerged across multiple dimensions:

### 1. Ethical Framework Development:

- Guidelines for responsible implementation
- Privacy protection protocols
- Equity consideration mechanisms
- Transparency requirements

### 2. Support Systems for Faculty:

- Staff-training programs
- Implementation resources

- Professional development documentation
- Opportunities for collaborative learning

### 3. Implementation Strategies:

- Integrating technology in a balanced manner
- Retention of student engagement
- The adaptation of formative assessments
- Integrity of pedagogical practice

In addition to supporting the arguments, Nguyet et al. cited the effects of AI on education or aspects whereby AI was improvised or forbidden; the comments from one member of senior faculty interviewed in this study was, "The process of implementation showed me how much depending on technology has to do with consideration for ethical issues" (Participant 4).

The work builds on an ongoing debate about the ethics of introducing AI in higher education and identified central motifs:

- Deployment transparency
- Responsibility in use
- Equitable access
- P Respect of privacy

The findings supported Nguyen et al.'s call for cross-border ethical codes of conduct while graciously highlighting the practical hurdles that institutional practice will have to work through. The experience of the private university shows that the formation of a task force and triggering ideas have been important for new challenges caused by AI. Ethical issues pointed to the need for flexible practices for adaptive techniques in learning institutions.

### **Practical Implications**

The study produced, therefore, commendable recommendations for educational stakeholders: a comprehensive approach in monitoring progress and observing due ethical considerations was warranted. The findings of this study are based on results backing Talwar's (2023) spotlight on faculty development and institution-specific means of putting it into practice.

Three key recommendations were offered:

### 1. Faculty team-based Development:

- Discipline-based training programs
- Hands-on AI tool experiences
- Continuous professional development and
- Technical literacy development---As noted by one participant, "Targeted training proved essential for successful AI integration across disciplines" (Participant 3).

### 2. Ethical Framework Development:

- Clear guidelines for implementation
- Regular reviews of the framework
- Building institutional networks; and
- Aligning with public policies and procedures---"Setting comprehensive ethical guidelines became foundational to our success of implementation" (Participant 7).

### 3. DG Integration Strategies:

- Enhancing quality education
- Improving admin efficiency
- Enhancing innovative teaching; and
- The universality of access.

## Conclusion

The findings provided implementation strategies that are flexible enough to consider the needs of other disciplines in terms of how comfortable the faculty is. Calling for a certain level of global ethical consensus, Nguyen et al. (2022) identified the thematic areas:

- Providing transparency of implementation
- Appraisal of responsibilities for application
- Equitable access; and
- Incorporation of sustainable development.

It has been established in the research that administrative, and facilitator encounters experienced AI as a sweeping tool for transformations of quality assurance. One administrator pointed out that "AI transformation required equal reflections towards technological capabilities and education integrity" (Participant 5). So, too,

this would dwell on adaptive frameworks that demanded that which would afford varied responses to authorities as the embedded authority of AI in higher education.

### **Methodological Limitations and Constraints**

The study revealed methodological constraints that should be regarded when interpreting the findings. The most significant limitation was that of a small sample size ( $n=11$ ), with an institution-based focus at the university, thus hindering the generalizability of results to multiple educational contexts.

Another set of limitations arising from several areas include:

1. Sample Characteristics:
  - Small sample size ( $n=11$ )
  - Sole focus on a single institution
  - Context of a private university
  - Student population ( $\sim 3,500$ )
2. Contextual Considerations Specific to the Institution:
  - Bias on private identity
  - Demography of a specific constitution
  - Variation of resource availability
  - Specificity of implementation approach

As Creswell and Creswell (2018) pointed out, although qualitative small-sample studies may offer rich data, broad generalizability to a more significant population still needs to be expanded. To quote one of the researchers, "Although we provided deep insights into our institutional context, wider applicability must be considered carefully" (Researcher A).

Other constraints include:

1. Disciplinary Distribution:
  - Overrepresentation in the STEM field
  - Bias toward early adopters (Zawacki-Richter et al., 2019)
  - Varied patterns of implementation
  - Discipline-specific challenges

## 2. Methodological Weaknesses:

- Restriction of data sources
- Limited presentation of triangulation
- Lack of classroom observation
- Voids of student perspectives

Tondeur et al. (2020) argue that institutional characteristics influence technological integration and suggest that one should interpret the findings in light of the context. Participant 8 remarked, "Our experiences may be quite dissimilar to those experiences at institutions founded on different resources and demographics."

## Recommendation

The findings indicated potential areas for future studies to work on incorporating AI into postsecondary education settings. The findings also suggested several avenues for research further down the line, particularly concerning the longitudinal effectiveness of the study and its practical applications.

Several research opportunities emerged from multiple domains:

### Longitudinal Studies Investigating the Impact:

- Sustained assessment of effectiveness
- Implementation practices transitioning over time
- Evaluation of effectiveness over longer durations
- Best practice identity: Long-term impacts remain critical to the effective integration of AI

### The Development of an Ethical Framework:

- Implementation Guidelines
- Privacy protection protocols
- Bias mitigation strategies
- Transparency requirements- Crompton and Burke (2023) proposed that cooperating with researchers and policymakers is vital for developing guidelines covering the whole ballpark.

### Enhancement of Accessibility:

- The investigation of the digital divide
- The establishment of inclusive practices
- Strategies for resource distribution



- Equity assurance initiatives

#### Faculty Development Research:

- Professional development optimization
- Support system effectiveness
- Integration methodology advancement
- Best practice identification

The findings emphasized alignment with SDG #4 objectives through:

- Quality education enhancement
- Personalized learning development
- Administrative efficiency improvement
- Inclusive practice promotion

Future research directions should prioritize 1) multi-institutional studies, 2) diverse methodological approaches, 3) longitudinal impact assessment, and 4) ethical implementation frameworks. As one researcher noted, "Future investigations must balance technological advancement with educational integrity" (Researcher A). This observation highlighted the necessity for comprehensive research addressing implementation effectiveness and ethical considerations in AI integration.

## Acknowledgements

The authors would like to extend their appreciation to participating faculty for their invaluable insights and contributions, which greatly enriched this study. We extend our heartfelt appreciation for our research team advisor, Dr. David G. Wolf, for his generous guidance and countless hours of expertise.

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