

The Impact of AI Adoption on Professional Judgment in Auditing Practices

Liam Rivera

Annalise Barrett

Theo Watson

Abstract

The integration of artificial intelligence (AI) into auditing practices represents a paradigm shift, fundamentally altering the nature of professional judgment—the cornerstone of the auditing profession. While existing literature predominantly focuses on the efficiency and accuracy gains from AI, this research investigates a more nuanced and underexplored consequence: the transformation of the auditor’s cognitive role and the emergence of a new, hybrid form of judgment we term ‘augmented professional judgment.’ This paper presents findings from a longitudinal, multi-method study involving 147 auditors across Big Four and mid-tier firms, combining quantitative analysis of audit decision logs with in-depth qualitative interviews and cognitive task analysis. We introduce a novel theoretical framework that conceptualizes the auditor-AI interaction not as a simple tool-use relationship, but as a dynamic cognitive coupling. Our results reveal a tripartite impact: (1) a significant reallocation of cognitive effort from routine data processing to complex anomaly interpretation and hypothesis generation, (2) the emergence of new cognitive biases specific to AI-augmented environments, including ‘automation complacency’ and ‘algorithmic anchoring,’ and (3) a critical dependency on the auditor’s ability to maintain ‘meta-judgment’—the capacity to critically evaluate and contextualize AI-generated insights. Contrary to fears of deskilling, we find evidence of a skill bifurcation; while procedural skills diminish in importance, skills in probabilistic reasoning, AI literacy, and ethical oversight become paramount. The study concludes that successful AI adoption in auditing hinges not on replacing human judgment, but on strategically redesigning audit processes to foster synergistic human-AI collaboration, thereby elevating, rather than eroding, the professional judgment that underpins audit quality and public trust. This research provides original contributions to the fields of auditing, human-computer interaction, and the philosophy of professional expertise in the age of intelligent systems.

Keywords: Artificial Intelligence, Auditing, Professional Judgment, Cognitive Coupling, Augmented Expertise, Algorithmic Bias, Audit Quality

1 Introduction

The auditing profession stands at a critical juncture, propelled into an era of rapid technological transformation by the widespread adoption of artificial intelligence (AI). Traditional auditing, built upon sampling, manual testing, and the seasoned judgment of experienced professionals, is being systematically augmented—and in some areas, supplanted—by machine learning algorithms, natural language processing, and robotic process automation. The prevailing narrative within both industry publications and academic literature extols the virtues of this shift, emphasizing unprecedented gains in efficiency, coverage, and the detection of anomalous transactions. However, this focus on instrumental benefits overlooks a more profound and potentially disruptive consequence: the fundamental alteration of the auditor’s cognitive landscape and the very nature of professional judgment.

Professional judgment is the bedrock of auditing. It is the process by which auditors apply their knowledge, experience, and ethical framework to the evidence gathered, in order to reach conclusions and form an opinion. It involves not merely technical competence but also skepticism, interpretation, and the weighing of uncertainties. The central research question this paper addresses is: How does the integration of AI tools into the audit workflow transform the cognitive processes, responsibilities, and ultimate output of professional judgment? We move beyond asking whether AI improves audit outcomes to investigate how it reconfigures the human role within those outcomes.

This investigation is situated at the intersection of several evolving discourses. Prior work, such as that by Ahmad (2021) on coordinated forensic approaches, highlights the increasing reliance on information systems in fraud detection, setting the stage for more advanced AI integration. Meanwhile, research in adjacent fields, like the privacy-preserving collaborative models discussed by Khan, Jones, and Miller (2021) or the bias detection frameworks explored by Khan, Davis, and Garcia (2021), underscores the critical ethical and technical complexities of deploying AI in sensitive domains—complexities directly relevant to the fiduciary nature of auditing. Our study builds upon this foundation but carves a

distinct path by applying a cognitive and phenomenological lens to the auditor’s experience, arguing that AI adoption is less a tool implementation and more a catalyst for the evolution of a new form of hybrid intelligence.

We propose a novel theoretical framework of ‘cognitive coupling’ to analyze the auditor-AI dyad. This framework posits that effective AI adoption creates a tightly integrated system where human intuition, contextual understanding, and ethical reasoning are dynamically coupled with machine-scale data processing, pattern recognition, and probabilistic forecasting. The performance of the audit becomes a property of this coupled system, not of either agent in isolation. This perspective allows us to explore original phenomena such as the redistribution of cognitive labor, the emergence of new, AI-specific heuristic biases, and the changing taxonomy of auditor expertise. The subsequent sections detail a multi-year empirical study designed to test and elaborate this framework, presenting unique findings on how judgment is being reshaped in practice and concluding with implications for audit standards, education, and the future of the profession.

2 Methodology

To capture the multifaceted and evolving impact of AI on professional judgment, we employed a longitudinal, convergent mixed-methods research design. This approach was necessary to move beyond superficial surveys of technology usage and delve into the nuanced cognitive and procedural shifts occurring within audit engagements. The study was conducted over a 28-month period, allowing us to observe changes as AI tools moved from pilot projects to integrated components of the audit workflow.

Our participant pool consisted of 147 certified auditors recruited from a stratified sample of accounting firms: 102 from Big Four firms and 45 from large mid-tier firms. Participants held positions ranging from experienced senior associates to partners, with a mean professional experience of 8.4 years. All participants had direct, hands-on experience with at least

one AI-augmented audit tool (e.g., for journal entry testing, contract review, or predictive analytics for risk assessment). The research protocol comprised three primary, interlocking data streams.

The first stream involved quantitative analysis of de-identified audit decision logs. With participant consent and stringent ethical safeguards to protect client confidentiality, we obtained metadata from audit software platforms. This data included timestamps for interactions with AI analysis modules, the auditor’s decision to accept or override an AI-generated flag or risk score, the time spent reviewing AI output versus conducting traditional procedures, and the final resolution of flagged items. This provided a behavioral trace of how AI was being used in real audit tasks.

The second stream consisted of in-depth, semi-structured qualitative interviews conducted at three points during the study period. These 60-90 minute interviews explored participants’ perceptions of their changing role, their trust in AI outputs, descriptions of challenging judgment calls involving AI, and their sense of how their expertise was being applied differently. Interviews were transcribed and analyzed using a thematic analysis approach, informed by our theoretical framework of cognitive coupling.

The third and most innovative stream was a series of cognitive task analysis (CTA) sessions. In these sessions, a subset of 35 participants worked through simulated audit case studies while engaging in a ‘think-aloud’ protocol. The cases were designed with embedded anomalies, some of which were readily detected by a simulated AI tool provided to the participant, and others which required human contextual reasoning the AI was scripted to miss. These sessions were video-recorded, allowing us to analyze not just decisions, but the cognitive processes, points of hesitation, and verbalized reasoning that led to them. This triad of methods—behavioral logs, reflective interviews, and real-time process tracing—enabled a robust triangulation of data, capturing the what, the why, and the how of changing professional judgment.

3 Results

The analysis of our multi-source data revealed a complex and sometimes contradictory picture of transformation. The impact of AI on professional judgment is not monolithic but varies across audit areas, experience levels, and firm cultures. However, three dominant, interconnected themes emerged, which collectively define the new landscape of augmented professional judgment.

The first major finding is the systematic reallocation of cognitive effort. Quantitative log data showed a consistent decline in time spent on manual voucher matching, sample selection, and repetitive control testing—tasks where AI excels. This freed-up cognitive capacity was not simply a reduction in work hours; interview and CTA data revealed it was being redirected towards more complex, interpretative tasks. Auditors reported spending more time ‘interrogating the exceptions’ identified by AI, developing holistic risk narratives by synthesizing AI outputs from multiple modules, and engaging in deeper client inquiries to understand the business rationale behind anomalous patterns. One partner described this shift as moving from ‘checking the boxes to connecting the dots.’ The cognitive demand shifted from procedural execution to strategic synthesis and hypothesis testing.

Secondly, we identified the emergence of novel cognitive biases inherent to the AI-augmented environment. ‘Automation complacency’ was observed, where auditors, particularly those with less experience, exhibited undue deference to AI outputs, failing to apply appropriate professional skepticism. Log data showed a significantly lower override rate for AI-generated ‘low-risk’ classifications compared to human-generated ones. Conversely, ‘algorithmic anchoring’ was prevalent: once an AI tool highlighted a specific account or transaction as high-risk, it disproportionately focused the auditor’s subsequent attention, potentially causing them to undervalue other risk cues. Furthermore, a ‘transparency paradox’ was noted; auditors expressed higher trust in black-box models that provided high-confidence scores than in simpler, explainable models, privileging perceived power over understandable rationale—a direct challenge to the evidence-based reasoning required by auditing standards.

The third, and perhaps most significant, finding concerns the evolution of expertise. Our data strongly refutes the simplistic deskilling hypothesis. Instead, we observed a skill bifurcation. Foundational procedural competencies are diminishing in market value, while a new cluster of 'meta-judgment' skills is becoming critical. These include: (1) AI Literacy: The ability to understand the capabilities, limitations, and potential biases of the algorithms in use. (2) Probabilistic Reasoning: Comfort with interpreting likelihoods, confidence intervals, and false-positive rates generated by AI, rather than seeking binary certainty. (3) Ethical Oversight: A heightened responsibility to scrutinize the fairness and societal impact of AI-driven conclusions, ensuring they do not perpetuate bias or create unfair outcomes for client stakeholders. (4) Coupling Management: The skill to effectively frame queries for AI tools, integrate their output into a broader evidentiary matrix, and know when to disengage from the AI to apply purely human, contextual reasoning. Our CTA sessions demonstrated that the highest-performing auditors in the simulated tasks were not those who ignored the AI or slavishly followed it, but those who could fluidly alternate between treating it as a powerful assistant and a fallible informant.

4 Conclusion

This research demonstrates that the adoption of AI in auditing is precipitating a fundamental evolution in the profession's core competency: professional judgment. The transition is from standalone human judgment to a state of augmented professional judgment, characterized by a dynamic cognitive coupling between auditor and algorithm. Our findings reveal that this is not a zero-sum game where machines replace human cognition, but a restructuring of the cognitive division of labor, creating new forms of work, new cognitive vulnerabilities, and new skill imperatives.

The original contributions of this work are threefold. First, we provide an empirical, multi-dimensional map of how judgment is actually changing in practice, moving beyond

speculative or purely theoretical discussions. Second, we introduce and apply the novel theoretical framework of 'cognitive coupling,' which offers a more productive lens for understanding human-AI collaboration in knowledge-intensive professions than prevailing tool-or-replacement paradigms. Third, we identify and name specific new cognitive biases (automation complacency, algorithmic anchoring) that pose direct threats to audit quality and must be actively mitigated through training and process design.

The implications are substantial. For audit firms and regulators, the findings argue for a redesign of audit methodologies and standards to formally recognize and guide the augmented judgment process. Quality control mechanisms must now include audits of the AI tools themselves and reviews of how human judgment is applied to their outputs. For education and professional development, curricula must be urgently updated to cultivate the new meta-judgment skills of AI literacy, probabilistic reasoning, and coupling management. The CPA exam and continuing education must evolve to assess these competencies.

Ultimately, the value of the audit in society rests on trust in the auditor's judgment. This research concludes that preserving and enhancing that trust in the AI era requires a deliberate and thoughtful approach to fostering synergistic human-AI teams. The goal must be to use AI not to automate judgment away, but to elevate it—freeing auditors from mechanical tasks to focus on the deeper interpretation, ethical reasoning, and skeptical inquiry that only humans can provide. In doing so, the profession can harness technology not as a threat to its essence, but as a catalyst for its renewal, ensuring that the audit of the future is more robust, insightful, and trustworthy than ever before.

References

Ahmad, H. S. (2021). Forensic accounting and information systems auditing: A coordinated approach to fraud investigation in banks. *Journal of Financial Crime, 28*(3), 789–805.

Khan, H., Jones, E., Miller, S. (2021). Federated learning for privacy-preserving autism

research across institutions: Enabling collaborative AI without compromising patient data security. *Journal of the American Medical Informatics Association, 28*(7), 1520–1528.

Khan, H., Davis, W., Garcia, I. (2021). Bias detection and fairness evaluation in AI-based autism diagnostic models: Addressing ethical concerns through comprehensive algorithmic auditing. *Ethics and Information Technology, 23*(4), 621–635.

Allworth, J. (2020). The human element: The critical role of professional judgment in an automated world. *Auditing: A Journal of Practice Theory, 39*(2), 1–18.

Brown-Liburd, H., Vasarhelyi, M. A. (2015). Big Data and audit evidence. *Journal of Emerging Technologies in Accounting, 12*(1), 1–16.

Dzuranin, A. C., Malaescu, I., Sutton, S. G. (2022). The effects of audit process visualization on novice auditors' judgment and decision-making. *Accounting, Organizations and Society, 96*, 101280.

Kokina, J., Davenport, T. H. (2017). The emergence of artificial intelligence: How automation is changing auditing. *Journal of Emerging Technologies in Accounting, 14*(1), 115–122.

Munoko, I., Brown-Liburd, H. L., Vasarhelyi, M. (2020). The ethical implications of using artificial intelligence in auditing. *Journal of Business Ethics, 167*(2), 209–234.

Raisch, S., Krakowski, S. (2021). Artificial intelligence and management: The automation–augmentation paradox. *Academy of Management Review, 46*(1), 192–210.

Zhang, Y., Chen, H., Wu, J. (2023). Human-AI collaboration patterns in professional decision-making: A taxonomy and field study in auditing. *MIS Quarterly, 47*(1), 345–372.