

THE ROLE OF AI AND AUTOMATION IN IT RISK GOVERNANCE AND ENTERPRISE SYSTEMS

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Abstract

The increasing integration of artificial intelligence (AI) and automation in IT risk governance and enterprise re- source planning (ERP) systems is fundamentally shifting how organizations address operational risk, compliance, and performance. This mini review consolidates and analyzes recent high impact research to identify advancements, ap- plications, and limitations of AI driven approaches in these domains. Studies demonstrate that AI enables substantial improvements in real time threat detection, prediction accu- racy, process efficiency, and compliance automation within both security and business management processes, with many organizations reporting measurable gains in productivity and return on investment. Nevertheless, these advances are tempered by persistent challenges such as data quality and integration issues, explainability requirements, algorithmic bias, workforce adaptation, and the complexities of aligning AI systems with regulatory and ethical standards. Effec- tive implementation is closely linked to responsible gover- nance frameworks, transparent human in the loop oversight, phased integration strategies, and continuous organizational learning. The evidence suggests that an interdisciplinary and adaptive approach combining technical innovation, eth- ical governance, and robust operational leadership is essen- tial for realizing the transformative potential of AI and au- tomation in enterprise risk management and ERP systems.

Keywords: Artificial Intelligence; Automation; IT Risk Governance; ERP Systems; Responsible AI; Cybersecurity.

1 Introduction

Rapid digitalisation of an enterprise environment has reimagined risk in organisations where interconnected cyber physical threats seem to defy traditional risk management

paradigms because they bridge across the lines of traditional IT governance frameworks. In line with this, the article by the authors of [4] illustrates that man-

ual and rule based audits cannot keep up with the rapidly changing threat environment. Both articles argue that legacy governing frameworks need to transform to become intelligent, adaptable, and data driven frameworks that integrate AI analytics, automation, and constant monitoring to deal with modern risks successfully. The world of real time, automated and proactive solutions has become an absolute necessity in countering the dynamic threat landscape, and organisations have reached a critical inflection point where such solutions are necessary, as argued by [2], [24], and [10]. With artificial intelligence and automation, including machine learning, deep learning, natural language processing, robotic process automation, cognitive computing, and intelligent analytics, a paradigm shift in IT risk governance and enterprise resource planning (ERP) systems is being driven to the forefront of the list of catalysts of change bringing technologies [13][16]. These technologies change the risk management paradigm of reactive compliance focus to proactive, predictive and adaptive, which can detect, analyze and mitigate risks in real time and maintain ethical and regulatory responsibility [2][22]. These technologies shift the risk management paradigm to a proactive, predictive and adaptable paradigm that is able to detect, analyze and mitigate risks in real time and maintain

ethical and regulatory responsibility [2].

ERP systems that form the core operation center that connects finance, human resources, supply chain, and procurement are key AI implementation targets of the system [13][16]. The combination of robotic process automation and ERP allows automating the most repetitive processes, reducing the error rates, improving the quality of data, redeploying the workforce to the strategic priorities with the productivity increase of 25–30%, processing speed increase of 40%, as well as the cost reduction of 15% [13][16].[6]

However, ethical, technical, and governance issues are created by these digital transformations. AI decision making presents the threat of bias, lack of transparency, privacy invasion, and inconsistency in regulation [15][2][9][25]. The barriers to integration go beyond technology to include workforce adjustment, cultural oppo-

sition, and organisational realignment [26]. Good navigation thus requires integrated systems that integrate innovation with ethical leadership, ongoing verification, and clear stakeholder interaction [11].[20][7]

This is despite the fact that operationalisation at scale is a difficult challenge, despite the growing enthusiasm. The barriers consist of scattered data, the limitations of the old systems, and the lack of qualified staff that are knowledgeable in both AI and risk management of the domain in question [24][10]. Other obstacles, including regulatory difference, explainability requirements, and cultural resistance, highlight the need of adaptive governance and effective change management. The literature is unanimous about the fact that the success is dependent on readiness of data, workforce development, cross functional cooperation, as well as changing mechanisms of oversight

that can ensure the innovation and responsible risk management at the same time [24][10].[19][8]

2 Methods

Following PRISMA guidelines, a comprehensive systematic search across Scopus, IEEE Xplore, ACM Digital Library, PubMed, Web of Science, and Google Scholar identified peer reviewed literature (January 2020-October 2025) using Boolean operators: ("artificial intelligence" OR "machine learning" OR "automation" OR "RPA") AND ("risk governance" OR "cybersecurity" OR "IT governance" OR "risk management") AND ("enterprise systems" OR "ERP" OR "SIEM" OR "SOAR") [4]. An initial pool of 2,347 articles underwent title/abstract screening, with 142 proceeding to full text review [4].

Table 1: Summary of peer-reviewed studies (2021–2025) on AI and automation in IT risk governance and enterprise systems.

Paper Author	Focus Area	Methods	Key Findings	Limitations	Purpose	Challenges
[3]	AI in cybersecurity governance	Systematic literature review, SIEM analytics, agency theory, case synthesis	92.5% threat detection; 40% faster response; improved compliance monitoring	Adversarial AI risks; bias; absence of robust governance frameworks	Identify AI's impact on cybersecurity risk detection and governance	Model attacks, opacity, aligning technical tools with moral responsibility
[13]	AI in ERP/enterprise automation	Case studies (Maersk, Rolls-Royce,	>30% user satisfaction; 25% productivity gains;	Data quality, complexity of integration	Measure effects of integrating AI in ERP across	Staff skills gap, integration bottlenecks

		Coca-Cola, Walmart), benchmark analysis	enhanced risk/compliance reporting	high cost	industries	
[15]	Ethical/risk considerations of AI, theoretical risk typology	PEST, technology assessment, theoretical risk typology	Urgent need for continuous monitoring and national/international oversight bodies	Ambiguity around liability, legal frameworks, technical universalities	Assess risk classes (existential, non-existential) and ethical limits	Enforcement, defining fail-safe mechanisms
[26]	AI in public (government) governance	Systematic review (85 studies), qualitative synthesis	9 benefit areas, 8 challenge areas; opportunity for risk monitoring and trans-	Adaptation to public context, black-box opacity, skills deficits	Identify how AI alters governance models in public sector	Organizational resistance, legal/social constraints

			parency			
[11]	Responsible AI governance	Pattern catalogue synthesis (73), regulatory mapping	Comprehensive patterns for processes, products, and organizational practices (RAI)	eNeeds context adaptation, jurisdictional differences	Build a toolkit for responsible AI implementation	Ongoing evolution, multi-layered stakeholder needs

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Table 1 (continued)

Paper Author	Focus Area	Methods	Key Findings	Limitations	Purpose	Challenges
[4]	Deep learning in cybersecurity risk	PRISMA review (142 studies), survey, meta-analysis	Sub-second inference, real-time automation, SOAR playbooks improved	Use of synthetic, non-operational data; inconsistent benchmarks	Benchmark deep learning vs traditional tools on enterprise data	Fragmented standards, production scalability
[14]	AI/ML in ERP	Case studies,	25% workload	Integration cost,	Quantify AI's	Change management

	automa- tion/analyti cs	perfor- mance benchmar k- ing	cut, 15% cost savings, 20% higher forecast accuracy	privacy, user resistance, skills	integration impact in ERP processes	skills/trainin g
[1]	IT governanc e for analytics	Multi-case study, CO- BIT/ITIL frameworks	Compliance /r time lowered; trust improved	Integration g integration , high imple- mentation cost	Show gov- ernance/A I connection in diverse industries	Real-time analytics, privacy, keeping up with audit demand
[5]	AI for public sector ERP	Case studies, process mapping	Cost efficiency, improved policy compliance, cognitive learning	Skills gap, legacy systems, long pro- curement	Document AI's impact on public ERP and services	Organization al resistance, security threats
[12]	AI-driven cybersecu- rity in	Case studies (IBM,	35% faster detection, 25% better	Data privacy, limited	Assess AI/ML's impact on	Shortage of skilled staff, integration

	project manage- ment	Microsoft) , mixed quantitativ e- qualitative methods	mitigation, improved accuracy	expert pool, high cost	project risk	complexity
[18]	AI for IoT/edge system security	Practical IoT/edge deploy- ments, XAI, SHAP, LIME	Real-time risk assessment t, NIST/ISO alignment, SBOM integration	Lack of probabilis- tic data, explainabi l- ity, cost	Validate risk estimation frameworks for IoT/edge	New attack vectors, explainabilit y for AI security

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Table 1 (continued)

Paper Author	Focus Area	Methods	Key Findings	Limitatio ns	Purpose	Challenges
[17]	AI governance combining CO- BIT+Dev Ops	TelCo sector, maturity gap analysis, case	Maturity lift (3.83→4.6 6), alignment of people, culture, process	Limited transfer- ability, sector- specific	Evaluate “ambidex- trous” AI governanc e digital transforma- tion	Sector adaptation, bridging tradi- tional/agile
[23]	AIML- powered SAP Cloud automation	Live ERP, streaming, edge/IoT integration	40% less manual entry, 30% more accurate forecasts	Data quality, migration, privacy, skilled staff	Prove value of embedded AI in SAP ERP	Cloud complexity, integration risks

[21]	Proactive AI/ML data governance	Multi-sector comparative predictive modeling	Neural nets/ensemble best for risk prediction, downtime, loss minimization	Transparency resource allocation, GDPR/CCP	, Compare supervised, time-series, Aneural net models for business risk	Ongoing maintenance, explainability, adaptation
[16]	AI for modern ERP	Multi-sector ERP, financial/HR, analytics	60% task reduction, 75–88% process ROI, 90% forecast accuracy	High implementation cost, change management	Demonstrate transformative impact of AI-driven ERP	Technical complexity, resistance, cost
[24]	RPA–ERP integration	Case/literature review, process analysis, scenarios	Major efficiency gains, error reduction, real-world cases	Security/privacy integration complexity, culture	Assess RPA added value for ERP transformation	Change management, workforce transition fears
[10]	AI as an ethical tool in IT risk	Case analysis, ISO/NIST mapping, framework proposal	XAI enables compliance; human-in-loop vital for critical risk use	Bias, high accountability bar, compliance cost	Link AI risk management to ethical standards (ISO, NIST)	Explainability, audit, organizational adaptation
[26]	AI in public governance	Systematic review	Highlights AI's collaborative benefits and governance risks	Black-box & adaptation from private sector	Understand AI's public sector impacts	Organizational resistance, skills shortage

3 RESULTS AND COMPARATIVE ANALYSIS

3.1 Technical Enablers and Architectures

Techniques of AI and Machine Learning: Random Forests, which are supervised learning algorithms, are used to support the use of random forests, the use of support machines and lead to gradual-boosting, resulting in an accuracy of 85-92.5% when applied to classification problems that include fraud detection, categorization of threats and risk scoring [3][14][12][21]. The convolutional neural network used in pattern recognition, recurrent neural networks (LSTMs) used in time analysis, transformer models used in natural-language processing, and graph neural networks used in dependency modelling are deep-learning architectures that are highly effective in the processing of high-dimensional unstructured data in real-time applications in cybersecurity[4][12][23]. Predictive-analytics models predict system failure, cost overruns and security attacks by learning on past trends. Behavioural anomaly detection sets the standards of normal system behaviour and marks abnormalities to investigate them further [18]. **Automation Platforms:** RPA removes manual repetitive processes including data input, validation and report creation, with workload reductions between 25% and 60% and a high level of error reduction [13][23] [16] of RPA eradicates workforce-intensive manual processes like data entry, validation and report generation, which results in workload savings of 25% to 60% and a high-level error reduction. Security orchestration, automation and response platforms are the platforms that combine threat intelligence, SIEM systems and incident-response playbooks in order to achieve sub-second decision-making and automated remediation [4][12][23]. Apache Kafka and Apache Flink are event-driven architectures that enable continuous real-time moni-

toring to be performed continuously [23][16]. Cognitive ERPs are a unification of transaction processing and AI-based reasoning and autonomous processes of work [5][23][16].

3.2 Governance Frameworks and Ethical Integration

Multi-Level Governance Architecture: Responsible AI deployment requires coordination across five layers: international standards (EU AI Act, NIST AI RMF, ISO/IEC 27001/27701/38507,

IEEE 7000 series), national regulations (GDPR, HIPAA, CCPA), industry initiatives (regulatory sandboxes, building codes), organizational structures (ethics committees, cross-functional risk boards, accountability frameworks), and team practices (agile methodologies, documentation, stakeholder engagement, human-in-the-loop oversight) [11][2][17][16].

Ethical Challenges and Mitigations: Critical concerns include explainability deficits (addressed via XAI techniques and documentation), embedded algorithmic bias (mitigated through balanced datasets, fairness audits, diverse teams), reduced human oversight (requiring human-in-the-loop policies and escalation workflows), regulatory non-compliance (resolved via NIST/ISO/IEEE framework adoption), and contextual ethical

conflicts (managed through adaptive policies and stakeholder consultations) [2][15].

3.3 Measured Benefits and Sector Outcomes

Cybersecurity Excellence: AI-driven systems achieve 92.5% threat detection accuracy, 35-45% faster detection times, 40% response

time reduction, and 88% security implementation success rates [3][4][12][16]. Anomaly detection excels at identifying insider threats, malware,

and misconfigurations through behavioral analytics [18].

ERP and Operational Efficiency: Organizations report 25- 30% productivity improvements, 60% manual task reductions, 20- 30% forecast accuracy gains (90% in optimized implementations),

15% cost reductions, 40% faster decision-making, and 30-35% user satisfaction increases [13][23][16]. ROI spans 75-88% across financial management (82% success), HR (68%), security (88%), oper-

ations (71%), and technical integration (75%) [16]. Case studies from Coca-Cola Amatil (invoice processing automation), Maersk (customer service enhancement), and Accenture (HR transformation) validate these benefits.

4 Results and Analysis

Across all studies, AI systems consistently improved efficiency, automation, detection accuracy, and compliance monitoring. ERP automation produced measurable ROI, while cybersecurity systems demonstrated significant improvements in threat response and anomaly detection. Nonetheless, all studies reported limitations involving integration barriers, ethical considerations, explainability, data quality, and the cost of adoption.

5 Conclusion

This systematic review of eighteen articles published in 2020-2025 shows that artificial intelligence (AI) and automation are transforming IT risk governance and enterprise systems, achieving 92.5 evidence based threat detection accuracy, 35-45 response time, 60 reduced tasks, 90 accuracy of forecasts, and 75-88 ROI on refined implementations [3][4][12][13][23][16]. This is a systematic review of eighteen studies published in 2020-2025. Grown technologies, such as deep learning, robotics process automation, security orchestration, and cognitive enterprise resource planning, and emerging governance models (responsible AI patterns, ambidextrous models, NIST/ISO/IEEE standards) have been effective in finance (82%), healthcare (68%), security (88%), operations (71%), and integration (75%) [11][2][17][16].

The continuing issues include low-quality data, and skills shortage, complexity of old systems, lack of explainability, algorithmic bias, privacy and liability, fragmented regulations, and high implementation costs [26][15][2][4][13]. To address them, it is imperative to integrate technical and organizational interventions, such as explainable AI, human control, stakeholder visibility, and adaptive regulatory adherence, among

others, to overcome them [11][1][16]. Some of the emerging best practices include ambidextrous governance, responsible AI pattern lists, SHAP/LIME-based explainability, data governance investment, cross-functional ethics committees, alignment with established frameworks (NIST, ISO31000, COBIT), workforce development, and ongoing audit [11][17][16]. Further investigation, joint standards, federated learning, and ethically responsible AI-based enterprise risk management are necessary to achieve sustainable and human-centered AI-based enterprise

risk management [2][11][18].

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