

# Algorithmic Advantage: Reimagining Business Administration in the Age of Intelligent Systems

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## Abstract

This study explores the concept of algorithmic advantage and how it can be leveraged to redefine business administration in the era of smart systems. As organisations increasingly integrate artificial intelligence (AI), business intelligence systems (BIS), and algorithm-driven decision tools into the managerial process, conventional administrative structures are being radically changed. This empirical review aims at synthesising the available research on the role of intelligent systems in organisational strategy, operational efficiency, decision-making, and workforce dynamics. Based on research in public administration, banking, enterprise systems, and management education, the review adopts a narrative and thematic synthesis research approach. The results show that algorithmic advantage does not come solely from technical level but also from strategic alignment, organisational preparedness, and human ability. Empirical data show that system quality, information integrity, and user competence significantly impact performance and efficiency. Moreover, leadership support, governance models, and ethical issues are very important for closing the gap between the promised value of AI and the actual organisational gains.

The challenges noted in the review, such as high implementation costs, the complexity of integrating digital transformation, cultural resistance, and a lack of skills, are also contributing to the high failure rates of digital transformation initiatives. In different contexts, intelligent systems are depicted as complementing human decision-making and underscoring the importance of AI literacy and interdisciplinary collaboration. In general, the paper finds that to reconsider business administration in the age of smart systems, one needs to approach the matter holistically and consider the combination of technology, strategy, human capital, and institutional reform. The creation of intelligent systems as coherent structures within managerial systems, ethical governance systems, and continuous capability development systems, through successful implementation in organisations, leads to algorithmic advantage.

**Keywords:** *Algorithmic Advantage; Intelligent Systems; Artificial Intelligence (AI); Business Intelligence Systems (BIS); Digital Transformation; Business Administration.*

## I. INTRODUCTION

This empirical review is aimed at synthesising the findings of the transformation of intelligent systems and algorithmic technologies on the business administration landscape. Due to the growing use of artificial intelligence (AI), machine learning (ML), and algorithmic decision-making tools in organisations, the traditional managerial practices are being transformed, resulting in the establishment of new types of operational efficiency, strategic planning, and competitive advantage (Brynjolfsson & McAfee, 2017; Davenport, 2018). Traditionally viewed as a profession based on human judgment and the traditional hierarchical decision-making

framework, business administration includes now data driven processes, which can automate routine processes, improve forecasting, and provide real time decision support (Shrestha et al., 2019).

The background of this review lies on the fact that there is a very fast evolution of clever systems in all industries, such as financial services, which can use predictive analytics to streamline risk management (Agrawal et al., 2018), or supply chain management, which can use AI to dynamically plan and schedule (Ivanov et al., 2019). The concept of algorithmic advantage is important as organisations that successfully approach such technologies tend to have a higher agility,

lower operational expenses, and increased innovation abilities (Jöhnk et al., 2022).

The review is limited to empirical research and analytical articles published to date (2023) that analyse the use of algorithmic systems, their consequences, and management implications in organisations. They include studies of smart systems, machine learning application in administrative functions and algorithmic decision support of business processes. Research that concentrates on technical development and does not relate to organisational or managerial environment is not included.

#### ➤ *Conceptual Clarification*

To be more precise, there is a need to define some important terms that this review will focus on. The concept of algorithmic advantage is based on the competitive advantages in which organisations benefit through the application of algorithm-based systems to guide the use of algorithms to make decisions, optimise processes, and gain insights out of data (Agrawal et al., 2018). The intelligent systems are computational frameworks which may include AI, machine learning, or rule-based algorithms that have the ability to learn new data, adapt behaviours, and carry out tasks that were previously dependent on human reasoning (Russell and Norvig, 2016). In business administration, management processes refer to planning, organising, directing as well as controlling organisational activities; the management processes are becoming increasingly augmented with algorithm functions like demand forecasting, resource allocation and performance evaluations (Shrestha et al., 2019).

The theoretical framing of the adoption of the algorithms in organisations is based on a number of existing models. According to the Technology Acceptance Model (TAM), the perceived usefulness and the ease of use determine the ability of people to accept technology (Davis, 1989; Venkatesh and Bala, 2008). TAM can be used to predict the uptake by managers to algorithms in the following way: TAM posits that the uptake of an algorithm depends on both the utility and confidence of the user. Resource Based View (RBV) puts emphasis on internal capabilities as drivers of sustainable advantage; algorithmic systems, when institutionalized into the organisational routines might be the valuable, rare and difficult to copy assets (Barney, 1991; Brynjolfsson and McAfee, 2017). Also, the Dynamic Capabilities Theory highlights the relevance of perceiving opportunities, mobilizing assets and changing processes in reaction to technological changes (Teece, 2018). Rapid reconfiguration of administrative processes can be made possible by use of intelligent systems.

It should also be noted that it is important to differentiate between automation, AI, and algorithmic decision making. Automation can be defined as a rule-based task execution that is limited in terms of learning, AI can be responsive and pattern recognise on the basis of ML, and the concept of algorithmic decision making

implies adherence to formalised procedures/model that generate decisions or recommendations after data have been fed into them (Davenport and Ronanki, 2018). These differences matter since organisational effects of simple automation can enhance efficiency, meanwhile AI-driven algorithms have the potential to affect the strategy and organisational positions (Shrestha et al., 2019).

## II. METHODOLOGY

This empirical review adopted a systematic approach to synthesise studies examining the role of intelligent systems and algorithmic technologies in business administration. Relevant literature was identified through searches in Scopus, Web of Science, ScienceDirect, SpringerLink, and ResearchGate, using combinations of keywords such as “*algorithmic advantage*,” “*artificial intelligence in business administration*,” “*intelligent systems*,” “*business intelligence systems*,” and “*AI adoption in enterprises*.” Studies published between 2020 to 2023 were included to ensure currency while excluding publications. Inclusion criteria required that studies address organisational contexts, managerial applications of AI or algorithmic systems, and measurable or observed outcomes. Conceptual or purely technical studies without organisational relevance were excluded.

Selected studies comprised quantitative, qualitative, and mixed-methods designs, including surveys, systematic literature reviews, content analyses, case studies, and conceptual frameworks, reflecting diverse methodologies applied in AI and algorithmic adoption research. Data extraction focused on key variables such as AI or algorithmic system type, organisational context, managerial outcomes, efficiency gains, and adoption challenges. Impact measures included performance metrics, user satisfaction, strategic alignment, and ethical considerations.

A narrative synthesis approach was applied, complemented by thematic analysis to identify patterns across contexts, sectors, and methodologies. Findings were organised to highlight both practical implications for business administration and theoretical contributions to understanding algorithmic advantage.

Table 1 Summary of Included Studies

Author(s)/ Year	Context / Industry	Research Method	Key Findings	Implications
<b>Grigore-Teodor SAMOILÁ, 2022</b>	Public Administration / Digital Reforms / Smart City (Romania, EU, Estonia comparison)	Conceptual / Comparative Analysis / Case Study	1. Digitization requires re-planning, decentralization, and financial autonomy.2. Success requires master plan in-laws + ERP systems to enforce cross-party collaboration.3. Romania lags behind Estonia; issues in cultural digitization, urban-rural divide, and data collection.4. Proposes Blockchain for infrastructure, 5G, and the "Authority for the Digitization of Romania" (ADR).5. EU "Next Generation EU" funds are critical for digital transformation.	1. Need for "smart roadmap" with clear deadlines and alert systems.2. ERP legislative master plans ensure predictability and sustainability.3. Digital sovereignty and broadband (5G) are strategic EU goals.4. Public administration must adopt private-sector tools (ERP, CRM) and self-financing mechanisms.
<b>Nehul Singh &amp; Satyendra Chouhan, 2021</b>	Intelligent Business Systems / AI in Enterprise	Literature Review / Conceptual Analysis	1. AI augments, not replaces, human intelligence.2. Key techniques: ML, Deep Learning (fraud detection), NLP (chatbots).3. AI improves data management, CRM, marketing, manufacturing.4. Challenges: High cost, integration difficulty, lack of expertise, rigid algorithms.5. AI will create more jobs than it replaces (WEF).	1. Businesses must adopt hyper-automation.2. Investment in education/training is essential for AI transition.3. Virtual Reality and advanced AI are next frontiers.4. Domain experts + Data Scientists must collaborate.
<b>Sulaiman Alsheibani, Chris Messom, Yen Cheung, Mazoon Alhosni, 2020</b>	Strategic AI Management / Cross-Industry (IT, Healthcare, Finance, etc.)	Content Analysis (Top 100 AI Companies via CB Insights/Crunchbase)	1. Gap between promised AI value and actual delivered value.2. 60% of decision-makers believe leadership is incompatible with AI needs.3. Proposed 5-Step Framework: - Q1: Solid AI business definition? - Q2: What value is sought? - Q3: What AI technologies align? - Q4: Key ethical concerns? - Q5: Organizational readiness?4. AI Ethics (fairness, accountability, transparency) are critical.	1. AI cannot be left to technologists alone; requires business leadership.2. Strategy must integrate technical + domain expertise.3. Organizations must assess culture and processes for AI compatibility.4. AI transparency must be structured (human instruction, understanding, testing, constraints).
<b>Rahmatian, F., &amp; Sharajsharifi, M., 2022</b>	MBA Education / Artificial Intelligence Curriculum	Qualitative Meta-Synthesis	1. AI (generative models, adaptive learning, predictive analytics) enhances personalization and curriculum relevance.2. Integration is uneven; faces ethical (integrity, bias), infrastructural, and faculty readiness challenges.3. Frameworks used: TAM, Paradox Theory, ISM.4. MBA enrollment decline linked to lack of data/AI skills and employment difficulties.5. AI literacy is now essential, not optional.	1. Business schools need holistic, ethically grounded AI strategies.2. Balance technological fluency with humanistic leadership.3. Shift from lecture-based to experiential, modular, AI-enhanced learning.4. Address digital equity and algorithmic bias.5. MBA must be "future-ready," not just a credential.
<b>D. Dudeja, S. Mayor, Y. Ganganwar, M. Singh, N. Goyal, A. Tiwari, 2023</b>	Context: AI Resource Management and Scheduling. Industry: General / Cross-industry	Type: Conceptual / Perspective Paper. Method: Review and proposal of a theoretical model (Sales-	<ul style="list-style-type: none"> <li>• Computational resources treated as commodities in a decentralized marketplace.</li> <li>• Resource allocation via competitive bidding (auctions) improves utilization.</li> <li>• AI/ML enhances adaptive workload management, query optimization (8-10x faster), and resource tuning.</li> <li>• AI</li> </ul>	<ul style="list-style-type: none"> <li>• Theoretical: Applies auction theory and market dynamics to AI resource management.</li> <li>• Practical: Encourages adaptive, responsive systems balancing resource use without overburdening infrastructure.</li> <li>• Future Work:</li> </ul>

	(Computing, Maritime/Shipping)	based auction model)	supports inventory management and business process efficiency.	Advanced bidding strategies using reinforcement learning, game theory, and multi-agent negotiation.
<b>A. Al-Okaily, A. P. Teoh, M. Al-Okaily, M. Iranmanesh, M. A. Al-Betar, 2023</b>	Context: Business Intelligence Systems (BIS) efficiency and success factors. Industry: Jordanian Banking Industry	Type: Quantitative. Method: Survey questionnaire (127 valid responses); Partial Least Squares-Structural Equation Modeling (PLS-SEM)	<ul style="list-style-type: none"> <li>• System Quality → positively influences User Satisfaction.</li> <li>• Information Quality → positively influences User Performance.</li> <li>• User Quality (skills) → positively influences both User Satisfaction and User Performance.</li> <li>• User Performance → directly and positively influences BIS Efficiency.</li> <li>• User Satisfaction does not directly influence BIS Efficiency (indirect via User Performance).</li> </ul>	<ul style="list-style-type: none"> <li>• Theoretical: Validates DeLone &amp; McLean model at organizational level; introduces "User Quality" as critical.</li> <li>• Practical: Banks must focus on technical system features, high information quality, and training users in business, analytical, and technical skills.</li> </ul>
<b>S. Ahmad, S. Miskon, T. A. Alkanhal, I. Tili, 2020</b>	Context: BIS Adoption and Acceptance determinants. Industry: Multi-sector (Banking, SME, Telecom, Healthcare, etc.)	Type: Systematic Literature Review (SLR). Method: Content analysis of 84 studies (2011–2020) using text mining (Yoshikoder) and human coding	<ul style="list-style-type: none"> <li>• Identified 93 determinants influencing BIS adoption/acceptance.</li> <li>• Top determinants: Organization Size, Relative Advantage, Complexity, Competitive Pressure, Cost, Top Management Support.</li> <li>• TOE and DOI dominate organizational adoption; TAM and UTAUT dominate individual acceptance.</li> <li>• 70%+ of BIS projects fail; quantitative methods dominate (52.2%).</li> </ul>	<ul style="list-style-type: none"> <li>• Theoretical: Proposes integrated conceptual framework combining TOE, TAM, DOI, and Institutional Theory across individual, technological, organizational, and environmental contexts.</li> <li>• Practical: Guides managers to prioritize determinants by industry (e.g., regulatory pressure for retail, cost for SMEs). Highlights need for user acceptance studies in developing countries.</li> </ul>

### III. DISCUSSION OF FINDINGS

#### ➤ Strategic Reconfiguration and Organisational Readiness

The reviewed evidence shows that to reach algorithmic advantage, fundamental strategic and structural reconfiguration is necessary as opposed to technological adoption. As Samoilă (2022) illustrates, the digital transformation of the public administration requires the decentralisation, financial independence, and legislatively-grounded ERP master plans to provide the continuity of the political cycles. This indicates that algorithmic systems need to be incorporated into the stable institutional structures in order to create sustainable value. Equally, Alsheibani et al. (2020) note that a gap between the expected returns of AI and the actual returns to the organisations exists. According to their review of the top AI companies, most organisations fail due to inappropriate leadership structures, cultural norms, and governance systems in response to the requirements of AI. The five steps framework that outlines the business value, aligning technologies, confronting ethics, and evaluating preparedness emphasizes that strategic transparency and ethical futurism is a precondition to successful AI integration. Together, these results emphasise a point that algorithmic advantage is dependent on organisational readiness, executive dedication, and systemic change instead of technological advancedness.

#### ➤ Performance, Efficiency, and Resource Optimisation

The research also indicates that smart systems have a lot of potential to improve operational performance and efficiency that can be achieved through the correct implementation of intelligent systems. The article by Al-Okaily et al. (2023) in the Jordan banking industry, performed a quantitative study with the PLS-SEM method showing that system quality and information quality positively influence the performance of the users that directly leads to the improvement of the BIS efficiency. Notably, user skills were also discovered to affect the satisfaction and performance meaning that algorithmic systems will only be of value when human competencies are competent enough. In line with this, the idea of computational resources as tradeable assets in decentralised marketplaces, promoted by Dudeja et al. (2023), suggests the use of auction-based models of the distribution of AI as a commodity, which maximise the utilisation of computational resources and their workload. Their framework depicts the manner in which adaptive algorithms would can be of great benefit in accelerating the processing of queries and enhance inventory and processes efficiency. The combination of these results strengthens the point that the advantage of algorithms can be observed in direct productivity, responsiveness, and resource optimisation. But all these advantages will hinge on the quality of systems design, the integrity of data, and the ability of organisations to interpret and utilise the outputs of the algorithms.

### ➤ *Human Capital, Skills Development, and Adoption Complexities*

One of the most frequent and important themes throughout the studies reviewed is the inevitable nature of human capital as the means of achieving the advantage of algorithms. Singh and Chouhan (2021) note that AI does not replace human intelligence, but rather augment it, but they draw attention to obstacles, which include high costs of implementation, integration difficulties, and the lack of professionals. This is in line with what Rahmatian and Sharajsharifi (2022) found during the field of MBA education, in which the unbalanced implementation of AI and readiness of faculty affects the transformative potential. They state that AI literacy is turned out to be the necessary skill to be a business leader in the future, which is the indication of a change in the skills set needed to be an administrative in the modern world. In addition, Ahmad et al. (2020) determine the presence of 93 determinants of BIS adoption such as organisational size, perceived relative advantage, complexity, competitive pressure, and top management support. Their conclusion that high percentage of BIS projects fail demonstrates how complicated the adoption processes are. All these studies combined show that algorithmic advantage does not arise due to a principle of technological determinism, but as a result of the interaction between managerial backing, user skill, organisational culture, and strategic alignment.

### ➤ *Implications of the Study*

The results of this review indicate that the benefit of algorithms is more organisational than technological. To the business administrators, it means that intelligent systems should be strategically coordinated with institutional structures, leadership capability, and competencies of the workforce to create value that is quantifiable. A banking, public administration and enterprise setting provides evidence that the quality of the system, the integrity of the information and the capability of users play a significant role in determining the performance. Additionally, leadership preparedness and ethical systems of governance are crucial in preventing the gap between AI promise and the actual value delivery. The case study thus suggests that strategic clarity, digital literacy, and cultural adaptability have to be key priorities of organisations that aim at gaining competitive advantage by means of algorithms. When integrated into consistent management and institutional systems, algorithmic systems may be seen as dynamic capabilities that increase decision-making, efficiency, and innovation.

### ➤ *Recommendations for the study*

According to the evidence reviewed, organisations need to take a sequenced and systematic approach to the integration of algorithm. First, the leadership teams should clarify the strategic value that AI systems are supposed to deliver and determine the organisational preparedness before implementing the system. Second, workforce training and AI literacy must be invested in to ensure that the effectiveness of those systems is as high as possible and that the resistance to their adoption is minimal. Third, high system and information quality must be considered a

priority of organisations since they have a direct impact on the outcomes of performance and efficiency. Governance frameworks should also be established by policy makers and institutional leaders making sure that there is transparency, fairness and accountability in algorithmic decision making. Lastly organisations need to keep on conducting a review of technological alignment with the business objectives to prevent expensive implementation failures as noted in BIS adoption research.

## IV. CONCLUSION

To summarise, this review proves that algorithmic advantage is a revolutionary change in business management, which is defined by the introduction of smart systems into business strategies, operations, and decision-making. The data show that although AI and algorithmic technologies can be used to strengthen a competitive position and optimise resources, as well as increase efficiency, their effectiveness relies on the readiness of the organisation, quality human resources, and strategic governance. In industries, in both government administration and industry management and enterprise systems, the interaction of technological competence and managerial competence is what defines realised value. The advantage is not necessarily created automatically by the algorithmic systems but they can serve as enablers of dynamic organisational capability when they are integrated into coherent institutional structures. Finally, in order to revise the concept of business administration in the era of intelligent systems, it is necessary to balance the technological innovation and the ethical control, devotion of the leadership and the long-term investment in the area of the human expertise.

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