

# Research on the Pathway of Artificial Intelligence-Driven Management Accounting Transformation

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## ABSTRACT

The in-depth development of the digital economy has made artificial intelligence (AI) technology a key force driving transformation across various sectors of society. Management accounting, as a core system for enterprise decision support and value creation, is facing both transformational pressure and historical opportunities. This paper aims to explore how artificial intelligence drives a systematic transformation in the philosophy, model, and capabilities of management accounting, and to construct an evolutionary pathway. The research first analyzes the current application status and limitations of management accounting, elaborates on the theoretical foundation for transformation, then focuses on constructing a three-stage transformation path model: from tool empowerment to model reconstruction, and finally to ecosystem intelligence. It analyzes the characteristics and key implementation points of each stage, proposes countermeasures to address transformation challenges, and provides an outlook for the future, offering references for the practical upgrading of enterprise management accounting.

**Keywords:** Artificial intelligence; Management accounting; Digital transformation; Pathway research

## 1. INTRODUCTION

Currently, people are in a new era dominated by data-driven and intelligent technologies. Artificial intelligence technologies such as big data, machine learning, and natural language processing have been integrated into enterprise operations, becoming crucial for enhancing core competitiveness. In this context, the traditional working modes and value proposition of management accounting, as a vital component of enterprise management, are facing severe challenges brought by AI. Traditional management accounting relies on periodic reports and historical data, struggling to cope with rapidly changing market environments and uncertain operational risks, highlighting issues of information lag and superficial analysis.

From an industry perspective, the urgency for the intelligent transformation of management accounting is significant. Currently, management accounting in most Chinese enterprises remains at the stage of traditional accounting or primary

informatization, with only a few achieving deep integration with AI. The adoption rate among small and medium-sized enterprises (SMEs) is far lower than that of large enterprises. This situation constrains the improvement of enterprise management efficiency and makes it difficult for enterprises to fully leverage the decision-supporting role of management accounting.

Simultaneously, AI's capabilities in processing massive datasets, recognizing complex patterns, and performing self-optimizing predictions offer new possibilities for management accounting to overcome bottlenecks. AI is not merely a tool for improving efficiency but a disruptive force reshaping the logic and expanding the boundaries of management accounting. Although current intelligent technologies have covered basic accounting tasks, management accounting, due to its need to balance business complexity and decision comprehensiveness, exhibits a gap in practical AI applications, further underscoring the necessity for transformation.[1] Therefore, exploring the internal logic and specific pathways

for AI-driven management accounting transformation holds both theoretical value and practical significance. This paper aims to construct a clear transformation framework to provide references for both theory and practice.

## **2. RESEARCH STATUS AND THEORETICAL FOUNDATION**

### **2.1 Current Status of Management Accounting Application**

The application level of management accounting in Chinese enterprises is uneven, with most being in a transitional phase from traditional accounting to modern value-creation. Common problems in practice include: First, widespread data silos, where internal departmental data systems are independent, making it difficult to integrate financial and operational data, leading to a lack of comprehensive support for analysis. Second, traditional analysis methods relying primarily on static financial indicator analysis and historical data comparison, which struggle to adapt to dynamic markets and have insufficient capabilities for identifying risks and opportunities. Third, a static reporting system with long compilation cycles and standardized formats, failing to meet the personalized decision-making needs of managers. Fourth, low integration with business operations, where work is concentrated within the finance department and not embedded in business processes, making it difficult to provide targeted support for business decisions.

Considering enterprise size and industry distribution, large enterprises and capital-intensive, data-intensive industries show higher application levels. These enterprises have sufficient capital and technical reserves to introduce advanced systems promoting the informatization and intelligence of management accounting. In contrast, SMEs and traditional service industries, constrained by resources, still rely on management accounting primarily for basic budgeting, cost accounting, and performance evaluation, limiting its role in forward-looking decision support and hindering the full realization of its value-creation potential.

### **2.2 Theoretical Foundation for Transformation**

Traditional management accounting theory is built upon a relatively stable operating environment and limited internal data, with its core logic

centered on controlling and optimizing business activities through historical data analysis. However, with the increasing dynamism and complexity of markets, diversified consumer demands, and intense industry competition, the assumptions of traditional theory no longer align with reality. Therefore, new theories are needed to support the transformation.

#### **2.2.1 Resource Orchestration Theory**

This theory posits that a firm's competitive advantage depends not only on the resources it possesses but also on its ability to bundle, configure, and leverage these resources. In the digital era, data is a core strategic resource, and AI technology is a key tool for efficiently orchestrating data resources. Through AI, enterprises can integrate multi-source internal and external data, cleanse, analyze, and apply it, transforming disordered data into orderly decision-making information, optimizing resource allocation, and enhancing management efficiency and decision quality to support building competitive advantages.

#### **2.2.2 Technology Empowerment Perspective**

Technology empowerment emphasizes the reshaping of organizational processes, capabilities, and value creation models by technology. The empowerment of management accounting by AI is reflected at multiple levels: At the data processing level, it breaks through the limitations of traditional technologies, rapidly handling massive, multi-dimensional, unstructured data. At the analytical capability level, leveraging machine learning and deep learning algorithms to mine hidden patterns and correlations within data, enhancing prediction and decision support capabilities. At the process optimization level, it automates part of the repetitive work, freeing up human resources and promoting the transition of management accounting personnel towards higher-value decision-support work.

It is worth noting that the characteristics of blockchain technology—distributed ledger, consensus mechanisms, and immutability—highly align with the core requirements of management accounting for value management and risk management. Blockchain can ensure the authenticity and traceability of data generation, transmission, and storage, address data trust issues, enable collaborative data sharing among multiple

parties, break down information barriers, strengthen data resource reliability and synergy, and provide a solid technical foundation for AI-driven management accounting transformation. [2]

### **3. THE INTERNAL LOGIC OF AI-DRIVEN MANAGEMENT ACCOUNTING TRANSFORMATION**

AI's drive for management accounting transformation is not a simple technological addition or tool update. It represents a profound structural change involving philosophy, methods, processes, and value positioning. The internal logic is reflected in fundamental shifts across three core dimensions: decision timeliness, analytical methods, and value creation.

#### **3.1 Decision Timeliness**

The core function of traditional management accounting focuses on ex-post accounting, organizing, summarizing, and analyzing financial data of past business activities to answer "what happened" and "why it happened." Its essence is to provide feedback on historical operating results. Under this model, management decisions rely on lagging information, making it difficult to intervene in operational issues promptly, let alone predict future risks and opportunities, leaving enterprises passive when responding to market changes.

The introduction of AI technology fundamentally changes the temporal positioning of management accounting. On one hand, embedding AI tools into business processes allows for real-time collection and instant analysis of operational dynamic data. Management accounting can then answer "what is happening" and "what potential impact it may have," shifting from ex-post accounting to in-process control. Managers can identify operational deviations based on real-time data, adjust strategies, and prevent problems from escalating. On the other hand, using algorithms like machine learning and time-series analysis, AI deeply mines multi-source information such as historical data, market data, and industry data to build predictive models, answering "what will happen," extending from in-process control to ex-ante prediction. This forward-looking decision support capability enables enterprises to proactively position themselves for market opportunities, avoid potential risks, and enhance management precision and proactivity.

#### **3.2 Analytical Methods**

Limited by the computing power and data processing capabilities of traditional technologies, traditional management accounting data analysis often selects partial sample data, inferring the overall situation from the sample. This has obvious limitations: subjective sample selection may lead to biased analysis results, failing to reflect the true operational situation; limited sample data makes it difficult to identify complex correlations and niche patterns, showing insufficient sensitivity to low-probability, high-risk anomalous events; sample analysis primarily focuses on structured data, with weak utilization capabilities for unstructured data like text, images, and audio, resulting in a single analytical dimension.

The breakthroughs in AI technology make it possible for management accounting to achieve a leap from sample analysis to full-scale analysis. Leveraging big data processing technologies and high-performance computing capabilities, AI can process the full volume of internal and external enterprise data, including structured and unstructured data, historical and real-time data, internal and external data. The full-scale analysis model has three main advantages: Covering all data avoids sample bias, ensuring objective and comprehensive analysis results, providing a reliable basis for decisions. Deeply mining full-scale data reveals hidden complex correlations and potential patterns, accurately identifying anomalies and risk points. Breaking data type limitations, transforming unstructured data into valuable information, enriching the analytical dimensions of management accounting, making operational analysis more comprehensive and multi-dimensional.

#### **3.3 Value Creation**

In the initial stage of transformation, AI's contribution to management accounting value primarily manifests at the level of process automation. Through technologies like Robotic Process Automation (RPA), it replaces manual operations with clear rules and high repetition, such as data entry, invoice verification, reconciliation, and report generation, reducing human errors, improving efficiency, and lowering operational costs. Value creation at this stage remains at the basic level of cost reduction and efficiency improvement, without changing the core decision logic of management accounting.

As AI technology is applied more deeply, the value creation model of management accounting gradually upgrades towards decision intelligence. AI systems can simulate human cognitive processes, analyze, optimize, and judge complex decision problems, providing or even automatically executing optimal decision solutions in specific domains. For example, in budget management, AI dynamically adjusts budget allocations based on strategic goals and changes in internal/external environments; in cost control, it automatically identifies cost drivers and optimizes cost structures; in performance evaluation, it builds multi-dimensional evaluation models, achieving dynamic assessment and early warning of operational performance. This shift towards decision intelligence liberates management accounting personnel from procedural, repetitive decision tasks, allowing them to focus on creative work such as strategic analysis, business negotiation, and cross-departmental collaboration, maximizing human resource value.

#### **4. PATHWAY MODEL FOR AI-DRIVEN MANAGEMENT ACCOUNTING TRANSFORMATION**

Based on the aforementioned internal logic, enterprises advancing the intelligent transformation of management accounting is a gradual and deepening process. Combining practical experience and theoretical analysis, this paper constructs a three-stage transformation pathway model: from tool empowerment to model reconstruction, and finally to ecosystem intelligence. Each stage is independent yet closely linked, forming a complete transformation system.

##### **4.1 Tool Empowerment**

Tool empowerment is the foundation of transformation. Its core goal is to introduce AI tools to solve the problems of low efficiency and high consumption in traditional management accounting, improving basic work efficiency and accuracy, while simultaneously accumulating high-quality data resources.

The first is to introduce Robotic Process Automation (RPA) to automate repetitive, rule-based tasks in management accounting by simulating manual operational processes. The second is to introduce Optical Character Recognition (OCR) to process unstructured

documents (invoices, contracts, receipts, etc.), automatically recognizing key information and converting it into structured data, avoiding human entry errors and inefficiency, and improving data collection efficiency.

Tool empowerment can liberate the productivity of management accounting personnel, freeing them from repetitive tasks to focus on high-value work. It also reduces operational risks, minimizes errors in manual entry and calculation steps, and accumulates structured data to provide data support for subsequent transformation stages.

##### **4.2 Model Reconstruction**

This stage is the key to transformation. Its core goal is to deeply apply complex AI technologies, break the boundaries of traditional management accounting processes and analytical limitations, improve the quality and timeliness of decision support, and achieve fundamental changes in the management model.

Model reconstruction primarily involves applying advanced AI technologies. First, introduce advanced machine learning models to build predictive models, cost analysis models, and risk assessment models, enabling deep analysis and prediction of operational data. Second, introduce Natural Language Processing (NLP) to handle textual data, automatically extracting, analyzing, and interpreting textual information, transforming unstructured text into structured information. It also supports querying financial data and generating analysis reports through natural language, enhancing decision efficiency and providing intuitive, efficient decision support for management.

Model reconstruction can break through the limitations of traditional processes, improving the quality and timeliness of decision support, providing management with more accurate and comprehensive operational insights. It also promotes the shift of the management accounting reporting system from static documents to dynamic decision support platforms, meeting the personalized needs of managers at different levels.

Table 1. Comparison of traditional management accounting model vs. AI-driven model characteristics

Management Aspect	Traditional Model	AI-Driven Model	Technological Support
Budgeting	Annual static budgeting, reliant on historical data, lagging adjustments	Rolling dynamic budgeting, integrates multi-source data, real-time adjustments	Machine Learning, Time Series Analysis
Cost Accounting	Single-dimension accounting, manual allocation of indirect costs, low accuracy	Multi-dimension accounting, automatic allocation, high accuracy	Big Data Processing, Intelligent Allocation Algorithms
Performance	Evaluation Ex-post evaluation, single indicators, reliant on manual statistics	Real-time monitoring, multi-dimensional indicator system, automatic alerts	Data Visualization, Alert Algorithms

### 4.3 Ecosystem Intelligence

This stage represents the highest level of transformation. Its core characteristic is achieving deep integration of management accounting with the enterprise's business ecosystem, breaking internal boundaries, and achieving data synergy and value co-creation with external entities such as supply chain partners, customers, and collaborators. The core goal is to drive business model innovation and maximize organizational value.

The realization of ecosystem intelligence relies on the application of cutting-edge AI technologies. First is Deep Learning, processing more complex and massive datasets to enhance analysis and decision-making capabilities for complex business scenarios. Second is Knowledge Graphs, constructing internal and external enterprise knowledge networks, integrating knowledge about corporate strategy, business processes, customer relationships, supply chain relationships, etc., enabling structured storage and relational analysis of knowledge, providing knowledge support for cross-domain decision-making.

Ecosystem intelligence aims to achieve the deep integration of management accounting capabilities with business processes. It embeds management accounting tools and methods into the entire value chain—R&D, production, supply chain, sales, customer service—achieving complete synergy between finance and operations. Simultaneously, it breaks down enterprise data boundaries, establishing data-sharing mechanisms with upstream/downstream supply chain partners, customers, financial institutions, and other external entities. Utilizing blockchain technology ensures data security and trust. Through multi-party data collaboration, it achieves goals such as supply chain optimization, upgraded customer relationship management, and joint risk prevention and control.

Ecosystem intelligence will drive the functional transformation of management accounting, shifting

from a traditional finance department to a value center, actively participating in high-level decision-making such as corporate strategy formulation and business model innovation, leading the direction of enterprise value creation.

## 5. TRANSFORMATION CHALLENGES AND COUNTERMEASURES

While AI provides powerful momentum for management accounting transformation, enterprises still face a series of challenges during implementation. Targeted strategies are needed to ensure smooth progress.

### 5.1 Data Governance and Talent Shortage Challenges and Countermeasures

#### 5.1.1 Challenges

Data is the core foundation for AI-driven management accounting transformation, but currently, most enterprises have significant shortcomings in data governance: lack of unified data standards, inconsistent data definitions across departments; uneven data quality with issues like missing, erroneous, or redundant data; incomplete data security and privacy protection mechanisms, posing leakage risks. These problems prevent AI models from obtaining high-quality data inputs, affecting the accuracy and reliability of analysis results.

Simultaneously, talent shortage is a key factor constraining transformation. The transformation requires compound talents possessing expertise in management accounting, data science skills, and business acumen. Currently, the supply of such talents in the market is insufficient. Existing internal staff often lack relevant skills in data processing and algorithm application, making it difficult to meet transformation needs.

### 5.1.2 Countermeasures

First, it is to improve the data governance system, formulate data governance plans from a strategic level, clarifying goals, responsible entities, and implementation paths, establish unified data standards and definitions, standardizing the entire process of data collection, storage, transmission, and application, strengthen data quality control by establishing evaluation indicators and assessment mechanisms, and conduct regular data cleansing and optimization, as well as building a data security and privacy protection system using encryption technologies, access controls, data anonymization, and other means to ensure data security and compliance.

Second, it is to build a compound talent pipeline. Combine external recruitment with internal cultivation. Externally recruit professionals with a compound background in management accounting and data science. Internally, enhance the data skills and intelligent tool application capabilities of existing personnel through training, job rotation, and project practice, establish cross-departmental collaboration mechanisms within the enterprise, forming project teams composed of management accountants, data analysts, and technical personnel to promote knowledge sharing and complementary skills, and improve incentive mechanisms, rewarding outstanding performers in the transformation to stimulate enthusiasm and creativity.

## 5.2 Organizational Change and Ethical Risk Challenges and Countermeasures

### 5.2.1 Challenges

AI-driven management accounting transformation is not only a technological change but also an organizational one, profoundly altering enterprise workflows, power structures, and cultural atmosphere. The process may trigger employee resistance: some employees fear job replacement by intelligent tools, leading to job changes or unemployment; others are accustomed to traditional work modes and find it difficult to adapt to new technologies and processes. Additionally, insufficient attention and support from top leadership can lead to a lack of overall planning and resource allocation, hindering progress.

Furthermore, the "black box" nature of AI also brings ethical risks. The decision-making process of models can be difficult to explain, potentially

leading to opaque decisions. If models contain biases or errors, they may cause decision-making mistakes, harming enterprise interests or social fairness. The process of data collection and application may infringe on customer privacy or leak business secrets, affecting corporate reputation.

### 5.2.2 Countermeasures

The first is to promote organizational change and cultural cultivation. Leadership must establish transformation awareness, incorporate the intelligent transformation of management accounting into corporate strategic planning, clarify goals and pathways, coordinate resources across departments, and provide solid support. At the same time, it is necessary to strengthen communication and publicity, conveying the necessity and value of transformation to employees, dispelling concerns, and gaining understanding and support, establish a trial-and-error mechanism, allowing for certain mistakes during transformation, and encourage employees to try new technologies and work modes, and cultivate a data-driven, collaborative, and innovative organizational culture, promoting a work philosophy among employees that is based on data and oriented towards innovation.

The second is to establish an ethical governance framework. Firstly, there is a must to formulate ethical guidelines for AI application, clarifying requirements for decision transparency, fairness, and accountability, ensuring the decision-making process is traceable and explainable. Secondly, it is necessary to establish an AI model auditing and evaluation mechanism, regularly testing model accuracy and fairness, and promptly correcting biases. Thirdly, there is also a must to strengthen compliance management, strictly adhere to data privacy protection laws and regulations, and standardize data collection and application processes. Finally, it is also necessary to establish an ethical oversight mechanism, setting up a dedicated oversight team to supervise the application of AI in management accounting and prevent ethical risks.

## 6. CONCLUSION AND OUTLOOK

### 6.1 Research Conclusion

Through in-depth research, this paper draws the following conclusions: First, current management accounting in Chinese enterprises faces issues like data silos, traditional analysis methods, and low

integration with business operations. Traditional theoretical foundations struggle to adapt to the new era's environment. Introducing Resource Orchestration Theory, the Technology Empowerment perspective, and blockchain technology can provide theoretical and technical support for transformation. Second, the internal logic of AI-driven management accounting transformation is reflected in the extension of decision timeliness from ex-post accounting to in-process control and ex-ante prediction, the leap of analytical methods from sample analysis to full-scale analysis, and the upgrade of value creation from process automation to decision intelligence. Third, transformation can follow the three-stage pathway from tool empowerment to model reconstruction and then to ecosystem intelligence. Each stage progresses gradually, achieving systemic change in management accounting. Fourth, transformation faces challenges such as data governance, talent shortage, organizational change, and ethical risks. These require countermeasures like improving the data governance system, building a compound talent pipeline, promoting organizational change and cultural cultivation, and establishing an ethical governance framework.

## 6.2 Future Outlook

With the continuous development of AI technology, the intelligent transformation of management accounting will usher in broader prospects. At the technological level, breakthroughs in Generative AI will further optimize human-computer interaction, enabling more natural and intelligent conversational analysis and report generation, promoting the embedding of management accounting capabilities into various business aspects of the enterprise in a service-oriented manner.

AI-driven management accounting transformation is an inevitable trend in the era of the digital economy. Only by proactively embracing change, rationally planning the transformation pathway, and actively addressing challenges can enterprises fully leverage the value of AI, propel management accounting to achieve qualitative leaps, and provide solid support for sustainable development.

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## REFERENCES

- [1] Li Minjie, Wu Huaping. Research on the Architecture of Management Accounting System Based on Intelligent Technology[J]. Finance and Accounting, 2021, (15):44-48.
- [2] Liu Guangqiang, Wei Jingjing, Qi Miao. Research on Intelligent Management Accounting Based on "Blockchain+" Digital Skills[J]. Commercial Accounting, 2022, (16):36-46.