
| RESEARCH ARTICLE

Digital Transformation of Supply Chain Quality Management: Integrating AI, IoT, Blockchain, and Big Data

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| ABSTRACT

The digital era really changed the face of supply chain quality management. This article explores the transformative power of some emergent technologies, including the Internet of Things, Artificial Intelligence, Machine Learning, and Blockchain, on a global supply chain's practice. In this context, while the challenges of managing and maintaining globalization are at an all-time high for companies, this set of technologies offers unprecedented opportunities for visibility, accuracy, and efficiency imperatives in a supply chain. Digital innovations have become vital to mitigate risks, optimize processes, and maintain high standards of quality by facilitating real-time data collection, predictive analytics, and decentralized ways of information sharing. The article examines the integration of these technologies within traditional supply chain management frameworks, identifying their potential to drive sustainable competitive advantage. However, it also looks at the challenges created through digital transformation, including data security concerns, system interoperability, and growing pressure to innovate more.

| KEYWORDS

Digital Transformation; Supply Chain Quality Management; AI, IoT, Blockchain, Big Data

| ARTICLE INFORMATION

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1. Introduction

SCM has been developing for a few decades from its traditional form into more developed and technology-based operations. The fact of such evolution is a very important context to be given while assessing the impact of digitalization on the quality management of supply chains.

1.1 Contextual Background

SCM involves coordinating all activities; right from the preparation of raw materials to finished products reaching the consumers in the production and distribution of goods and services. In other words, SCM focuses on ensuring efficiency by cost reduction with the aim of maximum value delivery in the supply chain. Earlier, SCM practices were not so complex as they focused on local and regional logistics. As globalization grew, this supply chain did as well and necessitated similarly intelligent logistics and inventory management due to the complexity of several tiers of suppliers and distributors.

Technological advancement and shifting priorities of business have brought in significant changes in the evolution of SCM. The early advances included the introduction of Enterprise Resource Planning (ERP) systems and better forecasting tools that made available real-time data and better decision-making. Recent developments focus on increased integration and collaboration within the supply chain, with strategies aligned to broader business objectives. The modern landscape highlights sustainability, risk management, and the use of digital technologies such as the IoT, AI, and blockchain. These real-time monitoring and

predictive analytics are adding transparency to major elements of the supply-chain [1]. With extended and more complicated supply chains, in general, rising quality management has emerged as an important factor in ensuring consistency, reliability, and customer satisfaction. This increases the demand for robust quality management approaches that may help address the challenges and opportunities in maintaining high standards across diverse and independent supply chains.

The need for continuing product and customer satisfaction within various complex networks highlights the increasing importance of quality management within global supply chains. Quality management has become so vital since supply chains have continued over recent years to increasingly develop with multi-tiered suppliers and production sites globally. Nowadays, quality management practices include processes like Total Quality Management (TQM) and Six Sigma to help overcome challenges in variability in supplier performance and lack of standardization. Presently, such practices are often combined with big data analytics and IoT to further improve quality control and ensure high standards across the supply chain. Above all, depending on the complicated developments of globalization, a well-developed and strong quality management system is necessary for modern global product markets regarding maintaining quality and competitiveness [2]

1.2 The Digital Transformation

The digital era has completely changed the way business processes are conducted and has thereby influenced the performance of every organization. Digitalization has been at the core of the redefinition of access to supply chain activities, including better technologies that allow efficiency and complete transparency. Due to technological advancements, digital tools like the Internet of Things and big data analytics can now be employed to monitor events in real-time and make focused data-driven decisions and quicker responses toward changes in situations and optimization of supply chain activities. It also allows for more precise forecasting and predictive analytics due to the integration of AI, further smoothing operations. Added to this value of digital transformation in improving process efficiency is better collaboration among supply chain partners, which translates to more resiliency and competitive advantage in the global marketplace [3].

2. The Evolution of Supply Chain Quality Management

Traditional quality management methodologies in supply chains have focused on systematic toolkits that are meant to reduce the level of inefficiency and defects while maintaining uniformity in production and distribution. Some important methodologies include Statistical Process Control (SPC), Quality Function Deployment (QFD) to cater a system in translating the needs of the consumers to actionable engineering requirements. Failure Modes and Effects Analysis (FMEA) has been utilized to identify potential failure points and mitigate risks before they impact the final product. Additionally, Root Cause Analysis (RCA) has been essential for diagnosing the underlying causes of quality issues, enabling organizations to implement effective corrective actions. The use of Lean Management and just in time (JIT) principles have been critical in minimizing waste and synchronizing production programs in accordance with the customer's needs, thereby enhancing the performance of the supply chain in general [4]. These have been strongly helping to optimize supply chain processes and ensure high-quality outputs. This has, however taken a new meaning with the increasing complexity and globalization in supply chains, where the need for the integration of the various digital technologies has shifted the priority towards better quality management and increased operability efficiency.

A shift toward digital supply chains represents a paradigm shift in managing quality and enabling the smooth functionality of organizations. Advanced technologies, such as AI, IOT, and blockchain, are influential in the integration of more connectivity with driven data into the core of the supply chains [3]. Supply chains are becoming increasingly interconnected and data-driven with the advent of advanced technologies such as Artificial Intelligence, Internet of Things (IoT), and blockchain [3]. But with the power of AI and machine learning algorithms, predictive analytics can foresee possible quality issues even before they occur, while IoT devices monitor production and supply chain activities in real-time. Blockchain enhances transparency and traceability along the supply chain, ensuring accurate and immutable records of transactions and product origin within the reach of all parties concerned [3]. This nudges quality management toward being more proactive and responsive, hence letting organizations adapt to changed circumstances and challenges in the global marketplace. Integration of digital tools, would represent a new frontier of quality management from traditional practices with agiler and more resilient supply chains.

3. Digital Tools and Technologies for Supply Chain Quality Management

Innovative digital tools and technologies have been introduced in recent times in supply chain quality management, hence improving the practice of maintaining and enhancing product and service quality. Generally, these include innovations like IoT, AI, and blockchain. Such technologies introduce real-time monitoring, data analysis, and predictive insight. Consequently, the application of such technologies enhances the level of accuracy, efficiency, and transparency of quality management practices while sustaining continuous improvement and strengthening the resilience of supply chains in a globalized market.

3.1 Internet of Things (IoT)

IoT stands for the network of interconnected devices, sensors, and other items that collect and share data over the internet. IoT is a quantum leap in supply chain quality management, providing it with real-time visibility on various aspects concerning the operation of supply chains. The embedded sensors and smart devices will be able to track many quality-related key metrics such as temperature, humidity, or the state of a product throughout the supply chain. It detects deviations from quality standards automatically and in real time, thus enabling timely intervention and ensuring high quality throughout the supply chain. IoT also acts as an important enabler in agile and responsive supply chain management, which seems to be certainly the future of logistics in the context of growing emphasis on real-time data and process optimization in the digital era [1].

The Internet of things, on the other hand, works based on connected sensors and other data-collecting device chains located at different points of the supply chain. These transmit the recoded data into centralized systems that, after processes, turn it with actionable insights. For example, sensors in the cargo container will indicate weather conditions, temperatures, and humidity conditions under which products need to be transported. This real-time data collected through those sensors allows organizations to address issues proactively, reduce risks related to quality degradation, and cut down possible waste. IoT allows quality management to be transparent and traceable, enhances the accuracy of process monitoring, and permits the customer to access real-time information in making better decisions [5]. With the integration of IoT, better efficiency in quality management across the organization will lead to a decrease in operational costs and improve general supply chain resilience.

A notable example of IoT in supply chain quality management is detailed in the case study by Ramanathan et al. [6]. This study explores how IoT technology is employed to ensure the quality and safety of human milk during its collection, storage, and distribution in milk banks. IoT sensors monitor the storage conditions of human milk, which is very sensitive to even slight changes in temperature. It gives an account of how these embedded sensors in the storage units and transport containers track and record temperature and humidity levels incessantly. This is a way of real-time monitoring that allows immediate alerts in case of deviation from the required standards, thus allowing immediate corrective actions to prevent spoilage. For example, if a sensor detects that the temperature in a storage unit has risen or dropped beyond the bounds of the safe zone, an alert is immediately sent to both the milk bank and the logistics team for prompt action, such as altering the refrigeration settings or relocating the milk to a different unit. That resulted in enhanced quality control of the human milk supply chain, reduced spoilage rates, and improved safety. This study shows how IoT technology works in concert to ensure high standards of quality and safety relative to sensitive biological products [6].

3.2 Artificial Intelligence and Machine Learning

Artificial Intelligence (AI) and Machine Learning (ML) are changing the way supply chain management operates by offering refined methods of data analysis and decision-making. AI refers to those abilities of machines that usually require human intelligence, such as problem-solving and pattern recognition. Machine Learning is a subset of AI which involves algorithms that allow systems to learn from and improve upon experience without explicit programming. AI and ML are reshaping quality management in the digital era through sophisticated tools for predictive analytics, anomaly detection, and process optimization within supply chains. In addition, AI and ML boost predictability and proactive problem-solving before any quality problems can affect supply chain operations [7].

Also, AI and ML use humongous data and sophisticated algorithms to tailor the quality management system for the supply chain. They analyze the data from production lines, supplier networks, and customer feedback to find patterns and trends. Machine Learning is, thus, a part of AI in which specific algorithms are developed to enable a system to recognize intricate models and enhance predictive abilities based on new data. For example, ML algorithms can identify an anomaly in product quality if real-time sensor reading, in collaboration with the historical data of the manufacturing process, are used to predict potential flaws in the process before they occur. Subsequently, AI-enabled systems churn out actionable insights and recommendations, like revisions in manufacturing parameters or reassessments of supplier quality, to reduce the associated risks and ensure constant quality. A data-driven approach ensures accurate forecasting, timely intervention, and optimization of the process in quality management [8].

A very notable use case would be the food production plant, which deals in various consumer-packaged foods and combines both manual and automated production lines. For years, quality control at the plant has depended on visual inspections at the stock-keeping unit level. The plant recently extended this process by adding AI-based camera stream analysis to monitor and analyze products running on the production line. It utilizes Deep Neural Networks (DNNs) for product fault detection and quality evaluation of packaging in real time. The entire process involves an AI system taking feeds from cameras at the production lines, then running the video streams through DNNs that are trained on huge amounts of image data. These networks can locate a wide range of product defects and packaging errors, accommodating variants created by product rotation and positioning. The key objectives of the system are to replace manual, sampled quality control with 100% inspection without adding personnel;

initiate a systematic learning loop from quality assurance results; and reduce process waste. The AI continuously improves its detection capabilities by storing the quality criteria in a structured training format, thus making the quality control process more effective and dependable. The resulting AI-driven inspection will not only raise the bar for defect detection accuracy but also further optimize the whole quality management process at plants [7].

3.4 Blockchain Technology

Whereas blockchain technologies originally developed from cryptocurrency, they've grown into a game-changing methodology that transforms industries and supply chains across the world. In layman's terms, it could be said that blockchain is merely a decentralized, immutable ledger that records transactions across a network of computers. It links a transaction block with a previous one, creating an immutable and transparent chain. This technology was just starting to change quality management in supply chains globally by bringing more transparency, traceability, and security within the flow of goods and knowledge. Blockchain applied in supply chains allows for the real-time tracking of products, limits fraud, and ensures quality compliance checks to ensure integrity in a supply chain [9].

Blockchain technology allows several nodes on a network to validate and log transactions onto a shared ledger. It works through the maintenance of a copy of the blockchain on each of the nodes in the network, updating the Gets in the blockchain in real time when new transactions arise. In this case, supply chain quality management involves the use of blockchain to obtain tamper-proof records of every transaction, from the production of goods down to delivery. For example, each stage of the supply chain can be recorded on this blockchain: raw material sourcing, manufacturing, and distribution are reduced to blocks. This ensures that all parties will have the same indelible view of the history of products. All these features tend to give way to better traceability, quicker sorting out of disputes, and adherence to better quality standards, thus realizing more reliability in supply chain operations [10].

The wood construction industry is one interesting context in which blockchain technology can change supply chain management [11]. It allows players at every level in the industry—from the forest owner and preliminary processors to architects and users access to a secure, transparent ledger system containing all relevant information. Forest owners, for instance, outline with blockchain information the state of the entire forest, including the types of wood and where the wood is positioned. This information enables optimized forest management as well as provides full traceability within the supply chain. Logistics companies also record transportation and quality information coming from the IoT sensors in the blockchain to extend material tracking and verification. Architects and construction companies obtain immutable building design records along with processing data through blockchain. This is a holistic approach that not only boosts transparency and gains the trust of all stakeholders but also opens the door to new business models and innovations, such as offering detailed historical data on the end user's maintenance and renovation needs. Consequently, applying blockchain to the wood construction industry opens up one source of dependability concerning the handling of such a convoluted supply chain where material integrity and traceability are preserved from the forest to the finished structure.

3.5 Big Data and Advanced Analytics

Big Data has become a key driver in advanced analytics concerning supply chain quality management, totally changing how organizations manage and improve their supply chains. Big Data is usually characterized by five Vs: Volume, Variety, Velocity, Veracity, and Value. These dimensions draw into focus the volume or magnitude of data, variety from different sources, the velocity or speed of generation, veracity, which depicts the accuracy or reliability of data, and value, which pertains to the possible insight that could be derived. All these elements put together enable supply chain management to visualize all aspects in greater detail and in real time. Accordingly, advanced analytics can help organizations bring insight from vast volumes of data into view with the help of sophisticated methodologies, thus enabling data-driven decision-making. Companies can make their supply chains work much more effectively, and respond to changing markets with increased agility at lower costs, and better customer satisfaction using this approach [12].

Big Data Analytics in quality management of supply chain works through use of combined technologies such as sensors, RFID, IoT, and advanced machine learning algorithms. Then advanced analytics models are put into use against this data, gathered on regular intervals at various stages of the supply chains, and will extract patterns, predict the obstacles that are likely to come, and identify the processes. For example, predictive analytics will forecast demand fluctuations, while machine learning algorithms identify defects in the products on the production line. The approach of analyzing data systematically and in real time would ensure not only higher quality control but also the perpetuation of a learning loop that constantly improves efficiency and effectiveness within the supply chain over time. While this is the case, challenges occur with regard to the intricacies involved in setting up such systems, issues relating to the security of data, and skilled professionals who are needed to manage and interpret these data effectively.

For example, Big Data and Predictive Analytics can be used in quality management across the complex supply chain in the automobile industry. The company installs sensors at all production lines and supplier sites to collect big amounts of real-time data related to variables like temperature and vibration. The data will then be populated into the central platform with prior history for analysis along with predictive models on conditions that would likely lead to defects. Further, such insight would enable the firm to focus quality checks on high-risk areas by leveraging an optimized process to reduce defective parts in final assembly. Moreover, sharing this data with the suppliers will directly enable collaboration in maintaining quality standards all along. While this approach can improve product quality and reduce warranty costs, it also streamlines production efficiency.

4. Challenges and Opportunities in the Digital Era

Modern supply chains, which now inarguably go hand in hand with digitization, also reveal a door to very modern concerns in quality management: cybersecurity. A digital supply chain has different technologies integrated amongst them, such as IoT sensors, cloud computing, and blockchain. Increased connectivity sends data flying across these platforms; in a digital supply chain, the whole value chain will be prone to cyber threats such as theft of data, manipulation of systems, and operational disruption. A cyberattack on the supply chain systems will expose data integrity to product quality and safety, flawed products reaching consumers, financial losses, and reputational damage. Strong cybersecurity measures in place, such as encryption, access controls, and continuous monitoring, contribute to the protection of sensitive quality related data and ensure the resilience of digital supply chains against cyber threats [13].

4.1 Data Privacy and Compliance Issues

Big data acts as the basis for organizations to further refine their services. With this comes a web of complexities pertaining to data privacy and compliance issues, as wide varieties of data will be collected and analyzed by these companies, ranging from transaction records to social networking trends. Ensuring the accuracy and security of such data to avoid misuse and retain the trust of the users is an important task. The integration of big data analytics into different fields requires the implementation of stern measures concerning personal data protection and access to unauthorized quarters. In addition, due to the fact that different regions have different levels of frameworks regarding data protection, added proper awareness and knowledge among consumers makes the task even more complex and requires strong internal data governance practices within companies [14].

4.2 The Digital Divide and Adoption Challenges in Different Regions

Because of the digital divide, the adoption of the same digital technologies of supply chain quality management does not take place uniformly in different regions. Advanced AI, Big Data, and blockchain are integrated into the supply chains in developed regions, while in some developing regions, their usage is at a very initial stage. However, in developing regions, limited access to technology, inadequate infrastructure, and lower levels of digital literacy can stifle this momentum. This leads to an inconsistency in the quality management of supply chains worldwide, since locations with a proportionately modern environment can commence advanced digital disclosure, while other locations might be more traditional [15]. The digital gap can be overcome only with rightful investment in technology and infrastructure, combined with proper training and support, so that all the regions can contribute to and benefit from the modern quality management of supply chains.

5. Best Practices and Strategic Recommendations

5.1 Guidelines for Integrating Digital Tools into Supply Chain Quality Processes

The integration of digital tools for supply chain quality management should be done in a strategic way, taking the tools as an enhancement of the quality processes. Organizations should identify those very specific challenges and objectives they want to achieve with the use of digital transformation. This will include the identification of digital tools supporting such objectives, including IoT sensors for real-time monitoring, advanced analytics platforms for predictive insight, and artificial intelligence for process automation and error detection. A full assessment of the present process must be done to find out where digital tools can create maximum value. It would, therefore, be relevant to ensure that any chosen tool can work with existing systems and scale up when and if necessary. In that regard, the organization may consider integrated data capabilities, where different digital tools and processes can converse well [15].

With the integration of such tools, for instance, comprehensive training that includes new technologies and workflows should be imparted to all employees. Since change management strategies are supposed to facilitate smooth transitions, it is reassuring to think that the technical competencies in the use of digital tools, together with the benefits these tools derive in the performance of daily tasks, are imparted to the staff. More so, clear plans in the areas of data management and quality assurance would go a long way to ascertaining that standards are maintained right through the value chain. Additionally, regular evaluation and refreshment of digital tools and processes should be done to meet the challenges emerging through emerging technologies and changing market expectations [16]. If an organization follows this recommendation, then it can improve the quality management process to attain better efficiency and accuracy of supply chain operation.

5.2 Frameworks for Successful Digital Transformation

Digital transformation in supply chain quality management can be effective by implementing structured frameworks that guide an organization through this rather complex process. A classic framework is one of the step-by-step approaches: pilot projects test and refine new technologies, followed by full-scale deployment. This will let organizations test effectiveness in a real environment using various digital tools, and make whatever adjustments are needed. The clear establishment of milestones and performance metrics is quite instrumental in monitoring the progress of the steps involved and the impacts on quality management brought about by these digital tools. They should align with quality objectives pertaining to the organization, besides offering appropriate insights for improvements continuously [17].

Another way digital transformation can be measured is through the formation of a cross-functional committee comprising IT experts, supply chain managers, and quality assurance professionals. It ensures that the integration of all areas of business regarding digital transformation goes on properly and smoothly, and communication between departments does not get affected. Apart from this, strong data governance policies are needed to manage and ensure safety for large volumes of information generated through newer digital tools. The ability to maintain the quality standard and compliance across the supply chain could be further enhanced by integrating real-time monitoring capabilities. Such a framework will, hence, enable organizations to manage their digital transformation efforts effectively and, in turn, accomplish better outcomes related to quality management [[15][16].

6. Integrated model for supply chain quality management

Supply chain quality management in the context of an automotive assembly line integrates in a rather structured manner, which is quite evident from the advanced technologies at different spectrums of the process. The integrated model comprises three stages: Assembly Parts Management, Material Movement from Storage to Assembly Sequence, and Predictive Maintenance of Assembly Machines. Each stage consists of certain digital tools and methodologies being deployed for optimization regarding quality and efficiency within the supply chain.

6.1 Assembly Parts Management

Basically, at this stage, it's all about inventory control of the parts released for assembly. Big Data and Advanced Analytics will be applied to analyze the historical trends and forecasts of future demands so that data-driven decisions can be made on the level of inventory. The aim is to ensure the cost of carrying inventory is optimized with the risk of stockouts so that the right pieces can be available at the right time without accruing unnecessary costs.

The economic order quantity (EOQ) is the mathematical means that shows the level of inventory that minimizes the total inventory costs, comprising the costs of ordering and the costs of holding inventory. The EOQ formula is given by [18].

$$EOQ = \sqrt{\frac{2DS}{H}}$$

(1)

Where;

D represents the annual demand for a specific part.

S is the cost of placing an order.

H denotes the holding cost per unit per year

EOQ will help the company calculate the number of optimal parts that can be ordered. EOQ minimizes costs yet keeps the stock volume adequate to cover production demands. This approach will make sure that inventories are properly managed to reduce waste and enhance the overall quality of the supply chain.

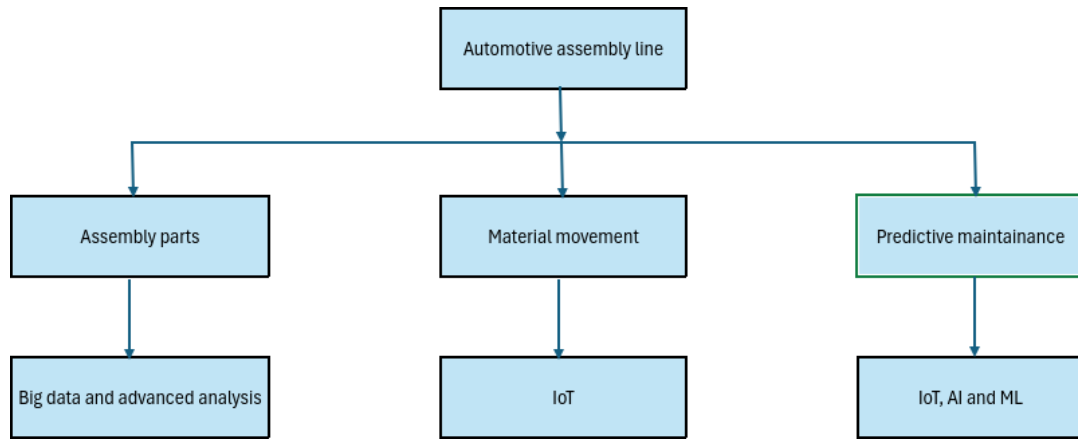


Fig. 1 Integrated model of the automotive assembly line

6.2 Material Movement from Storage to Assembly Sequence

Once parts are in stock, they need to be transferred onto the assembly line in the quickest and most effective manner possible. At this stage, which involves utilizing the IoT to deliver real-time notifications regarding when and where the parts need to be moved based on needs determined by the sequence of assembly. The Just-In- Time approach assures that the parts are fed onto the assembly line right on time, minimizing storage time and reducing inventory costs.

The process can be modeled using the Reorder Point (ROP) formula, which calculates the point at which an order should be placed to replenish inventory before it runs out. The ROP is determined by [19]:

$$ROP = d \times L + SS \quad (2)$$

Where:

d is the average demand per period. L is the lead time.
 SS represents the safety stock.

This equation helps in decision-making on material movements within an assembly sequence with the goal of minimizing downtimes, hence improving production flows. The integration of IoT devices means real-time tracking and alerting will enhance the responsiveness and flexibility of the supply chain.

Predictive Maintenance of Assembly Machines

In this integrated model, the last step refers to maintaining the machinery used in assembling. Predictive maintenance identifies equipment failures in advance before they really occur, which minimizes losses due to wasted time and ensures flow in production. At this stage, IoT, AI, and ML come together to provide real-time monitoring of the condition of assembly machines and allow for predictive analytics.

Reliability modeling in machines can be conducted with the use of a hazard rate function, which denotes the probability of failure at every instant of time. A typical hazard rate function is expressed as [20]

$$h(t) = \frac{f(t)}{1-F(t)} \quad (3)$$

Where:

$h(t)$ is the hazard rate at time t .
 $f(t)$ is the probability density function of the failure time.
 $F(t)$ is the cumulative distribution function of the failure time

In predictive maintenance, machine learning models could take a set of input parameters (for instance, temperature and vibration) and generate some kind of probability of failure P (failure). Converting this to customized functionality may be represented as [20]:

$$P(\text{failure}) = f(x_1, x_2, \dots, x_n) \quad (4)$$

Where, x_1, x_2, \dots, x_n are the variables being monitored, such as sensor data collected from the machines. With the help of historical data, we can train these models to predict the possible failure of a machine, allowing us to schedule maintenance proactively.

Eventually, integration of the three stages, namely, Assembly Parts Management, Material Movement, and Predictive Maintenance, will yield an optimized quality management in the supply chain system for automotive assembly lines. This comprehensive approach not only enhances the efficiency and reliability of the production process but also ensures that quality is maintained throughout the supply chain, ultimately leading to higher customer satisfaction and lower operational costs.

7. Conclusion

In the digital era, quality management in supply chains has increased tremendously owing to the rapid development taking place in technologies such as IoT, AI, and Blockchain. The development was able to let organizations improve their supply chain by leveraging better supply chain visibility, real-time monitoring, and predictive analytics. The integration of these technologies is so deep that conventional supply chains have turned into resilient, lean, and agile systems.

However, all these technological gains come with a lot of challenges: data security concerns, system integration issues, and continuous innovation. While companies can harness these digital enablers, they must combine this effort with the need to invest in higher standards of quality in every supply-chain process that defines them in global markets today.

Eventually, the future of quality management in supply chains would integrate seamlessly with the digital world. Companies that emerge successfully through this integration will not only enhance operational efficiency but also ensure superior value delivery to customers for leadership in the respective industry.

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