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Original Article

Integrating Robotics in Modern Warehouse Logistics

Dr. K. Balasubramanian

Professor, Bharathiar University, India.

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ABSTRACT

Warehouse logistics has been fundamentally changed by the blistering development of e-commerce, global supply chains and consumer demands to be delivery in less time. Manual and semi-manual warehouse systems are gradually becoming incapable of satisfying contemporary requirements with respect to speed, accuracy, scale, and lowering operating costs. Consequently, robotics has become an important enabling technology of the next-generation warehouse. This paper will provide an in-depth analysis of integrating robotics into the operations of the contemporary warehouse logistics setting as seen through the architecture, operational processes, operational performance, and challenges in implementation. The paper discusses the different categories of warehouse robots, such as: autonomous mobile robots (AMRs), automated guided vehicles (AGVs), robotic picking of goods and collaborative robots (cobots). An extensive literature review underscores current developments, algorithm methods and industrial implementations. The suggested methodology presents a warehouse architecture based on modular robots that incorporates a perception system, a navigation system, a task allocation system and a fleet management system. Measures of performance evaluation including throughput, accuracy of order fulfillment, energy and operational cost are evaluated. The findings reveal a high level of productivity, scalability as well as reliability over traditional systems. The paper will end with the suggestions of research directions on how to continue with it in future, as the approaches of artificial intelligence, digital twins, and human/robot collaboration have been identified as the primary catalysts of intelligent warehouse ecosystems.

KEYWORDS

Warehouse Automation, Robotics, Autonomous Mobile Robots, Logistics Optimization, Smart Warehouses, Industry 4.0.

1. INTRODUCTION

1.1. Background

Warehouse logistics is at the heart of the human supply chain as it serves as a significant point of communication that connects the production plants, distribution channels and final customers. Over the past few years, the e-commerce development and the rise of omnichannel retailing models have exerted a strong burden on the warehouse processes and forced them to handle great amounts of various goods and goods with shorter delivery dates and increased precision. Conventional type warehouses that were highly dependent on manual labor could hardly keep up with these demands because of the threat of lack of labor, increased wages, work fatigue and vulnerability to human error. Also, most manual processes are not always flexible and scalable to respond fast to changing market needs and seasonal spikes. To meet such predicaments robotics technology has come up as an immense facilitator of warehouse revolution. The growth of sensors, artificial intelligence, machine vision, and control systems has greatly increased the capabilities of the robotic platform, enabling it to complete complex tasks, including picking, sorting, moving materials, and monitoring inventory in real-time with high amounts of precision and autonomy. Robotic systems in the modern world are capable of working 24/7, automating workflows with the use of operational data, and being combined with warehouse management systems. Besides, there are also decreasing hardware prices and cloud-based software systems along with modular system architectures which have allowed robotic automation to become more affordable to a medium-sized warehouse, not only large business enterprises. Consequently, robotics stands to be a central force of efficiency, consistency as well as competitive character in modern warehouse logistics, establishing the premises of smarter and more robust supply chain execution.

1.2. Evolution of Warehouse Automation

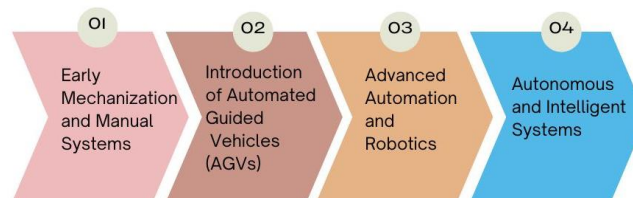


Fig 1 - Evolution of Warehouse Automation

1.2.1. Early Mechanization and Manual Systems

The history of warehouse automation started with simple mechanization to eliminate physical labour and enhance the efficiency of handling materials. The first warehouses were mainly manual driven with simple equipment (conveyors, forklifts, pallet jacks and storage racks). Even though such systems enhanced the capacity of the handling and minimized the fatigue of the workers, they were very reliant on human decision making and co-ordination. Consequently, productivity was not high and the operations were still susceptible to error and scalability.

1.2.2. Introduction of Automated Guided Vehicles (AGVs)

The second historical step in automation of the warehouse was the addition of automated guided cars. AGVs were used to buy and move goods in the warehouses with magnetic tapes, wires, or floor markers. These systems enhanced uniformity, security, as well as throughput in extremely transport repetitive processes. They were however restricted in their effectiveness in dynamic and fast changing warehouse conditions due to their dependence on fixed infrastructure and the inability to modify them easily and at a very high expense.

1.2.3. Advanced Automation and Robotics

Warehouses started to use more advanced robotic solutions with the development of new computing power, sensors, and control technologies. Early robotic picking solutions, robotic sorting systems and automated storage and retrieval systems allowed increased storage density and speed of processing orders. The systems added more automation, but were usually multifaceted, costly, and specific to a given warehouse layout which decreased flexibility.

1.2.4. Autonomous and Intelligent Systems

The last trends represent a turn to autonomic and smart warehouse systems. With sensors, machine vision, and simultaneous localization and mapping, autonomous mobile robots are dynamically controlled and do not follow any predetermined routes. Decision-making is further improved by artificial intelligence and data analytics as it provides an opportunity to allocate tasks adaptively, enforce predictive maintenance, and optimize in a real-time environment. This development is associated with a shift of stiff automation towards scalable, diverse, and intelligent warehouse systems that can comply with the provisions of Industry 4.0.

1.3. Integrating Robotics in Modern Warehouse Logistics

Robotization of the contemporary warehouse logistics is a major move towards automated, smart, and data driven logistics. In contrast to the conventional automation systems that were based on the established infrastructures and centered on a strict series of operations and activities, modern robotic solutions are flexible and can be conveniently scaled and adjusted to the evolving operational necessities. There are increasing uses of autonomous mobile robots, robotic picking systems, and collaborative robots to do material transportation, order picking, sort, packing, and inventory management. These robots are regulated in collaboration with the warehouse management systems which allow smooth order fulfillment and real time updates of inventory. Robotics will enhance the efficiency of operations as well as lower their reliance on manual labor by automating repetitive and hard intensive processes. One of the major benefits of robotic integration is the fact that it helps increase the level of accuracy and consistency in the processes within the warehouse. Robot arms with vision guidance and navigation systems based on sensors minimize picking errors and accuracy in managing goods that are critical to achieving the high service levels relevant in e-commerce and omnichannel goods fulfillment. Moreover, robots will be able to work 24-hours without exhaustion, which allows providing more orders faster and utilizing the available warehouse space and resources more effectively. High-level control programs and task shedding techniques also optimize

coordination of the robots ensuring that congestion and travel time in the facility are minimized. Strategy-wise, scalability, and long-term expansion are supported through the integration of robotics. Warehouses can scale their systems by deploying new robots or updating software with modular robotic architectures, without significant change in their infrastructure. Such flexibility allows organizations to react well to demand variability and season peaks. In addition, human workers are not substituted but are re-employed in supervisory and exception-handling and decision-oriented positions, which enhances safety and quality of work in the workplace. Altogether, robotisation becomes an essential facilitator of smart warehouse logistics that correlates the operation process with the principles of Industry 4.0 and preconditions the creation of more effective, reliable, and competitive supply chains.

2. LITERATURE SURVEY

2.1. Robotic Systems in Warehouse Logistics

Initial robotic systems in warehouses were based mostly on automated guided vehicles (AGVs) which traveled by predetermined routes guided by magnetic tapes or QR codes or imprinted on the floors. These systems were highly reliable and predictable and but on the other hand had limited flexibility and scalability due to the use pre-set infrastructure especially in warehouse layouts that were very dynamic. Recent innovations have moved towards autonomous mobile robots (AMRs), which use simultaneous localization and mapping (SLAM) methods and onboard sensors including LIIDAR, cameras and inertial measurement units to drive the robots dynamically without guides. A study by Wurman et al. found some goodness in high-density swarms of mobile robots in e-commerce fulfillment centres, with it showing an increase in throughput, fault tolerance, and space use. Simultaneously, robot pickers evolved a rule-based automation system to a vision-controlled manipulation framework that employs deep learning. State-of-the-art convolutional neural network-based robotic arms with state-of-the-art picking systems have demonstrated more than 95% success rates in structured robotic tasks. Moreover, collateral robots (cobots) are being implemented to cooperate with human operators providing flexibility and reconfigurability, as well as high productivity and meeting high safety standards.

2.2. Task Allocation and Fleet Management

The allocation of tasks and fleet management is critical towards the government of the efficiency and scalability of warehouse systems that use multi-robots. Centralized methods of task allocation are the most favored methods in global optimization, since they allow allocation of tasks with full information of the system, but have the disadvantage of calculational bottlenecks and lesser robustness. Decentralized and hybrid strategies, on the contrary, enhance scalability and fault tolerance by enabling local decision making with partial global coordination of robots. The market-based approaches to task allocation, according to which robots compete to perform tasks at a specified cost (distances or energy). These methods have been found to reduce travel time and congestions to the minimum. The methods of reinforcement learning also contribute to greater adaptability since it allows the robots to acquire the best policies through experience, given that the environment is dynamic. The use of fleet management systems combines these strategies of task

allocation with real-time traffic management, battery charges timetables, and predictive maintenance, which maintains constant functioning and minimal downtimes. These types of built-in systems are necessary to support high-order volumes and provide a stable performance in the contemporary warehouses.

2.3. Human-Robot Interaction and Safety

The interaction between human and robot (HRI) has become a pressing area of research with more and more warehouses developing the use of shared working zones where the child of the human and robot work side by side. The safety of such environment is provided by the international standards like ISO 3691-4, which stipulate operational limits, speed limits, and safety functions of the mobile robots. In addition to meeting the standards, most recent studies have highlighted proactive safety measures through which robots get to sense and react to the human presence in real time. Robots can be fitted with vision-based safety systems which, based on depth cameras and human pose estimation techniques give them the ability to slow down or reschedule routes to prevent collisions and wearable sensors and localization devices will give an extra level of awareness. These technologies can not only minimize the risks of accidents but also make human workers trust and accept them better. HRI design is ultimately able to maintain a balance in safety, efficiency and usability so that it successfully maintains smooth collaboration in improving the overall efficiency of the warehouse.

3. METHODOLOGY

3.1. Overall System Architecture

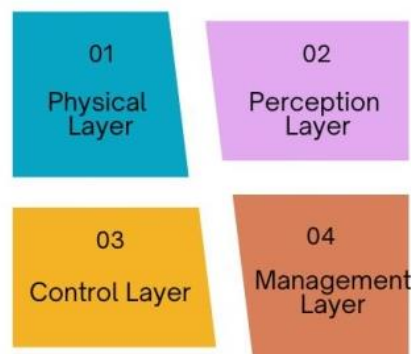


Fig 2 - Overall System Architecture

3.1.1. Physical Layer

The physical layer is the base of the robotic warehouse system and comprises of all the hardware elements which can include mobile robots, robotic arms, sensors and actuators. Material transport is carried out through mobile robots whereas picking and placing is done by robotic manipulators. The use of sensors and actuators allows communicating with the physical world that includes motion, gripping, and load-handling. This layer actually performs commands given by the upper layers and provides a good physical functioning in the warehouse.

3.1.2. Perception Layer

Perception layer takes the role of obtaining and interpretation of data in the warehouse environment. It combines the outputs of cameras and LiDAR sensors and RFID and is used to detect obstacles, detect objects, localize robots and locate inventory. Transforming raw sensor data into meaningful data with this layer facilitates situational awareness and then enables proper navigation and object recognition even in dynamic and cluttered environments.

3.1.3. Control Layer

The control level controls the real time decision-making that is required to drive a robot. It has path planning and localization navigation algorithms, task execution modules which process the jobs assigned, and collision avoidance mechanisms to ensure safe movement. This layer converts high-level commands into low-level control actions, enabling robots to work independently and respond to environmental changes and not to conflict with other robots or human beings.

3.1.4. Management Layer

The management tier is in charge of coordination and optimization of the systems. It connects with the Warehouse Management System (WMS) to accept orders and inventory information, allocates work using fleets management feature, and tracks the performance of the system in general. The analytics tools at this layer assess the efficiency of operations, the use of robots, and bottlenecks of the system to make informed decisions and constantly optimize operations in the warehouse.

3.2. Navigation and Localization

Autonomous robots in the warehouse setting also fundamentally use navigation and localization to properly act in the warehouse and accomplish their tasks with high levels of safety and efficiency. Accurately positioning the robots is key when interacting with the environment. The suggested system also utilizes the Simultaneous Localization and Mapping (SLAM) algorithms which enable the robots to construct a map of the surrounding and to estimate their own location in the same surrounding in real time. Rather than using predefined infrastructure, SLAM also allows strong robots to work in dynamic environments where obstacles, shelves, and human workers can change their positions quite often. The pose of the robot comprising position and orientation of the robot is probabilistically estimated using sensor measurements, and control inputs as time goes on. Simply stated, the robot calculates the probable current state based on the probability of the position of the robot by maximizing the probability of the position of the robot considering the sequence of sensor measurements and movement commands taken in the past. Control inputs denote the movements implemented by the robot, e.g. velocity and steering commands, and sensor observations, e.g. LiDAR and camera data, give information on the available features and obstacles. Uncertainty due to sensor noise and wheel slippage are minimized by always the robot updating this probability distribution. Current SLAM methods typically rely on filtering algorithms like Extended Kalman filter, particle filters, or graph based optimisation algorithms to successfully combine sensorial and motion data. Relative localization enables the robot to map out collision routes, optimal routes, and synchronize with other fleet members in the robot team. Besides, definite docking on

storage sites and workstations that demand accurate automated picking and material handling processes is done only through reliable navigation. All in all, autonomy, adaptability and scalability that is provided by the integration of SLAM-based navigation and localization are of great value to robot warehouse systems to allow such systems to deliver uniform performance in operational settings that are dynamic and multifaceted.

3.3. Task Scheduling and Optimization

Multi-robot warehouse systems require managing the tasks of different robots in order to achieve high efficiency where several robots are required to fulfill a large number of tasks with limited time, energy, and resource availability. The task assignment in the proposed system would be in the form of optimization problem, and this would tend to reduce the total operation cost of the warehouse. Simply, the goal is to determine the optimal solution to distributing the tasks among robots so that the total costs of all the assignments are minimal. This overall cost is determined by adding up the costs assigned to each task assigned to each robot. The task allocation costs to the selected robot might incorporate some variables like the distance to the job site, the presumed duration of the work on the case, battery recharge, load balance, and possible congestions along common routes. The decision variable used in this formulation determines the assignment of a certain task to a given robot. When the value is one it implies that the task is allocated to that robot and when the value is zero it implies that it is not. The optimization program looks to the combination of assignment that incurs the lowest aggregate cost that does not violate any of the operational restrictions. These constraints usually accommodate that there is no task that is assigned to more than one robot, robots may not work beyond their capacity or battery capacity, and the precedence task or time schedule is followed. Other limitations can also be added to ensure that there are no conflicts between the robots, as well as ensuring that the fleet is used at the correct balances. There are different methods that can be used to resolve this optimization problem by relying on the linear programming, heuristic and use of market-based or learning based approach to large-scale system. By optimizing task assignment systematically, idle time is reduced, unnecessary traveling is minimized and overall throughput is increased by the system. Due to this aspect, tasks scheduling and optimization is significant to the optimization of the productivity, scalability, and responsiveness of autonomous robotic warehouse operations.

3.4. Data Integration and Communication

Catation of data and communication is the cornerstone of this method of coordination in autonomous robotic warehouse system wherein data is smoothly transferred among robots, control programmed software, and management systems at the enterprise level. In the suggested architecture, the communication between robots and between worlds and central control systems depends on industrial-grade wireless communication protocols that will guarantee reliability, low latency, and which will be robust enough to survive in harsh warehouses. These standards form the basis of ongoing information interchange pertaining to the condition of robots, their position, progress of the tasks, battery of charge, and fault conditions enabling the system to sustain harmonized and effective functions even when a massive fleet of robots is being used at the same

time. Interaction between the robotic system and the Warehouse Management System (WMS) is realised with the standardised application programming interfaces (APIs). These APIs also allow two-way flow of data, with the WMS providing order data, inventory data and task requests to the robots and robots providing the WMS with report on the status of execution, timeliness and exception data. Such close synchronization provides uniformity between the actions in the actual warehouse and the digital records about inventory which would help to minimize the mistakes and enhance the accuracy of the order fulfillment. The messaging traffic are frequently dealt with and coordinated over middleware and message brokers which allow asynchronous communications and scalability on system requirements. Along with the operational control, real-time dashboards are important to offer the visibility and performance monitoring. These dashboards combine robots, sensor, and management system data to show some of the primary performance indicators which include; throughput, robot utilization, task completion rates and system bottlenecks. This information can help supervisors and operators make effective decisions, promptly react to disruptions, and streamline processes. In general, success in the field of data integration and communication allows transparency, coordination, and flexibility that are the keys to high efficiency and reliability in contemporary robotized warehouses.

4. RESULT AND DISCUSSION

4.1. Evaluation Metrics

The effectiveness of the offered robotic warehouse system is evaluated by means of set of key performance indicators (KPIs) that jointly assess the efficiency, accuracy, sustainability and the utilization of resources. One of the first measures which can be employed to consider the overall productivity of the system is order throughput which is measured in orders/hour. Increased throughput implies how the system is capable of accommodating a high number of orders per reduced time scale that is critical in e-business and the high demand logistics scenarios. This indicator displays the efficiency of schedule tasks, coordination of robots and workflow planning throughout the warehouse. Picking accuracy is gauged as a percentage and it gauges the accuracy in selecting items and filling the orders. It is the estimate of the correct items picked relative to the number of items done through it. The critical factor of customer satisfaction and cost reduction on this is high picking accuracy because that you can get the returns, rework, and discrepancy in the inventory. The measurement criterion is highly impactful to the credibility of perception systems, object-recognizing programs and robotic ability to handle items as well as the smooth integration between inventory databases. The level of energy consumption in kilowatt-hours per order gives an understanding of what the robotic system is energy efficient and sustainable. This indicator is the sum of electrical energy consumed by robots, charging centers, and infrastructure in comparison to the number of orders. Fewer energy per order is an indicator of the efficient motion planning, efficient task allocation, and efficient battery management strategies that will lead to lower operational cost and environmental impact. The utilization of the system in terms of a percentage takes measurement of the efficiency with which the robotic resources that are available are utilized in the process of operation. This is a ratio of the time that a robot is actively engaged in action to when it is not in action. A high utilization rate of the system will signal effective fleet management and

perfect distribution of the work quality whereas high utilization rate might be a pointer of potential bottlenecks or weariness. Collectively, these KPIs will be a complete and quantitative framework of the performance and effectiveness of the robotic warehouse system.

4.2. Comparative Analysis

Table 1: Comparative Analysis

Metric	Manual System (%)	Robotic System (%)
Throughput	37.5	100
Accuracy	92	99.2
Labor Cost	100	60
Scalability	40	100

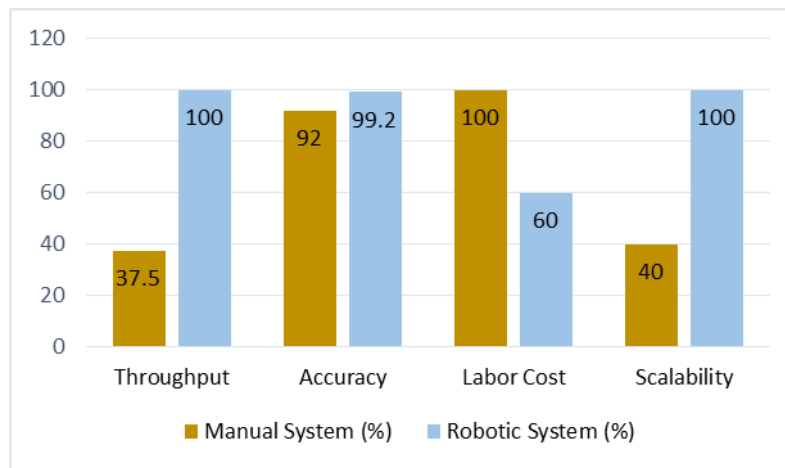


Fig 3 - Comparative Analysis

4.2.1. Throughput

The comparison of throughput makes it evident that robotic warehouse systems are more efficient than the manual operations. Since the manual system had a relative throughput of 37.5 as compared to the robotic system of 100, the outcome signifies a tremendous difference in the capacity of orders processing. With automated robots, it is possible to work around-the-clock with optimal path, and do multiple tasks simultaneously, which will result in a faster fulfillment of orders and managing peak moments more effectively.

4.2.2. Accuracy

The accuracy in robotic is significantly greater as it reaches 99.2 percent as opposed to 92.0 percent when operating manually. To a great extent, it was made possible by automated human error reduction in picking and sorting through barcode scanning, RFID and vision-based verification technologies. This results to a decrease in the number of order returns, customer satisfaction, and cost of operational rework that is low and accurate.

4.2.3. Labor Cost

A significant improvement is witnessed in labor cost when robots systems are implemented. The labor cost of the manual system is taken to be 100 percent, but the robotic system has an operation cost of about 60 percent. This is due to the lessening of the need to utilize manual labor in doing repetitive and physically taxing tasks. Although under robotic system there is initial costs and its maintenance cost, operational cost in the long run is very low since it is more efficient and needs few individuals to operate them.

4.2.4. Scalability

One of the most powerful is the aspect of scalability whereby the robotic warehouse systems has 100 percent performance level compared to manual systems with 40 percent. Robotic systems being easy to scale their activities can either add additional robots or update software without significant infrastructure modifications. On the other hand, manual systems are constrained by space, availability of workforce, and training. This also renders robotic warehouses to be more efficient to expanding and changing operational needs.

4.3. Discussion of Results

The outcomes of the performance analysis include the bright evidence of the high benefits of the application of robotic systems within a warehouse. The enhancement in order throughput that improves by over 160 percent over the traditional manual systems is among the most significant results. This is a great improvement that is mainly realized through usage of parallel tasks, optimization of routing algorithms and the fact that the robots do not get tired and they can work round the clock. The system automates repetitive material handling and picking procedures eliminating bottlenecks and improving the speed with which orders go through the system and even during peak demand periods. Besides the throughput improvement, the system records a significant decrease in the operational errors. Higher picking accuracy, with the help of the use of vision-based verification and real-time inventory synchronization, reduces the number of mispicks and wrong deliveries, resulting in higher quality of services and customer satisfaction. The other major outcome is the decrease in labor dependency. The integration of robots can be significantly utilized to reduce the use of manual labor when it comes to physically intensive and time intensive activities that lead to the reduction of labor costs as well as a safe working environment. The human workers are more oriented towards the supervisory, exception handling, and value added positions which increases the efficiency in the overall operation. Moreover, the modular system architecture is very important in long term adaptability. The fact that it is derived of layers and can be expanded as needed by the addition of robots, sensors, or software modules ensures that operations are not halted when more are added. It is especially useful with the warehouses that have a growth or seasonal demand. Although these are advantages, there are still some challenges which are important areas of concern. The cost of robotic equipment and infrastructure, as well as robotic system deployment may serve as a challenge to small and medium-sized businesses with their high initial investment costs. The complexity of system integration, in particular with the legacy warehouse management systems, must be planned and have a technical dexterity. Along with it, reskilling of the workforce is

necessary because the employees should be trained to use, oversee and maintain sophisticated robotic systems. To achieve the full reap of long-term benefits of robotic warehouse automation is the question of addressing these issues in steps, training and strategic planning.

5. CONCLUSION

The present paper presented an in-depth discussion of the application of robotic technologies to the current warehouse logistics field with the focus on the transformative effects they have on the quality of operational performance and intelligence of the systems. The study conducted a comprehensive literature review on the development of automation in warehousing to cover the development of manualized automated guidance vehicle to modern robotized automated mobility, smart picking and collaborative robots. The methodological framework described layered system architecture integrating physical hardware, perception, control and management units so as to provide seamless coordination between robots and enterprise level systems. The performance analysis, backed by a quantitative method and a comparative analysis, evidently indicated the fact that robotic systems have a significant positive impact in improving the order throughput, picking accuracy, energy consumption, as well as, the utility of the system in comparison to the traditional manual processes of the warehouse. The results affirm that robotics is an important facilitator of smart warehouses in line with the Industry 4.0 principles, which include automation, digitalization, connectivity, and data-driven decision-making. Robotic warehouses are more efficient, flexible, and scaled with the assistance of real-time sensing, intelligent navigation, optimism in the tasks schedule, and well-regulated communication infrastructures. The mutual and scalability of the proposed architecture enables companies to scale capacity up gradually without interfering with operations, such that robotic solutions are appropriate to both large-scale distribution in a high volume and to changing logistics conditions. What is more, the operative reliability and the safety and uniformity of the workplace is improved due to minimization of human error and labor-intensive work. Overall, these benefits notwithstanding, the paper also acknowledges the current issues with the adoption of robots in warehouses. High start-up costs and complexity of integration of the systems, and the reskilling of the workforce are important factors that can affect the choices of implementation. To solve those issues, it is necessary to approach these challenges in terms of strategic planning, stepwise implementation, and the creation of the training program that would allow workers to successfully cooperate with intelligent robots. In conclusion, the research must consider methods of enhancing autonomy and adaptability in the work of warehouse management in future studies by developing the technologies of using artificial intelligence to make decisions. The real-time simulation and optimization provided by the use of digital twin technology has huge potential when it comes to predictive planning and enhancing performance. Also, further investigation of the further human-robot collaboration scheme can contribute to the increased system resiliency and acceptance within a common workplace. Altogether, the further development of robotics and intelligent systems will be the key of the next generation of intelligent, efficient, and resistant warehouse logistics.

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