

# Research on Logistic Warehouse Scheduling Management With IoT and Human-Machine Interface

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## ABSTRACT

The automated deployment of the internet of things (IoT) and the human-machine interface provides the best advancement for dispersed warehouse scheduling management (WSM). In this paper, superior data systematic move toward warehouse scheduling management (WSM) has been suggested using the computational method to allow smart logistics. Furthermore, this paper introduces the human-machine interface framework (HMI) using IoT for collaborative warehouse order fulfillment. It consists of a layer of physical equipment, an ambient middleware network, a framework of multi-agents, and source planning. This approach is chosen to enhance the reaction capabilities of decentralized warehouse scheduling management in a dynamic environment. The simulation outcome has been performed, and the suggested method realizes a high product delivery ratio (96.5%), operational cost (94.9%), demand prediction ratio (96.5%), accuracy ratio (98.4%), and performance ratio (97.2%).

## KEYWORDS

Human-Machine Interface, IoT, Logistics Warehouse Management

## 1. INTRODUCTION

To maximize efficiency and respond accordingly to changing consumer demands, effective product design, manufacturing, packaging, and delivery is needed (Kumar et al, 2020). The potential logistics of physical items are delivered, handled, preserved, distributed, recognized, and used across the globe in the form of industry digital Internet should be transformed to make it feasible for better logistics performance, resilience, and replicated physical networks. The Internet of Things (IoT) links and interacts with industry-relevant objects such as components, devices, equipment, goods, and real Internet container (Bodkhe et al, 2020). It is possible to track and control all related products to allow producers to know the trends and results (Al-Turjman et al, 2020). The industry can be defined as the increasing digitalization and computerization of the developed process with decentralized intelligent decision-making or a modern supply chain that allows contact between goods, the industry, and business associates (Al-Turjman et al, 2019). By embracing IoT, information, and resources for the next development model, business reflects the emerging fourth industrial revolution (Al-Turjman et al, 2020). In the communication of the real and practical worlds, decentralized propensity assists in building smart networks and refine autonomous procedures, marking a significant new step in

DOI: 10.4018/IJSSCM.305846

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advancing the sector (Ponis et al, 2020). Industrial manufacturing equipment is no longer virtuously a development creation, and the item interacts with the machinery to express it exactly what to do. Industry uses a Bluetooth or sensor module among the latest manufacturing techniques (Mistry et al, 2020). Smart manufacturing methods pave the way to an innovative technology world, will drastically change commerce and manufacture cost restraints and marketing strategies (Dannapfel et al, 2020).

A Human-Machine Interface (HMI) is a dashboard or consumer interface for connecting an individual to software, system, or gadget (Younan et al, 2020). However, the expression can potentially be applied to any screen that enables a consumer to converse with a workstation (Rojas et al, 2019). HMI is widely used in the intelligence of a developed role in the industry (Molano et al, 2018). The key communication stage between the user and a computer or procedure is HMI (Bruzzzone et al, 2019 ; Sathishkumar et al, 2020). A well-designed HMI framework does more than present control features and data (Krdžalić et al, 2019). It supports an operator with active output functions, reviews the consequences of those activities, and performance statistics for the framework (Gaber et al, 2018; Majstorovic et al, 2018). Human Machine Interface (HMI) in industrial optimization's modern information technology field is an important concept (Elhoseny, 2020). Technological advancements have moved the limits of automation inside industrial warehouses and processing plants back to almost flawless operations (Mostafa et al, 2019 ; Younan et al, 2019). With the beginning of the Internet of Things (IoT) and the desire to fully automate processes, HMIs are becoming more significant (Pane et al, 2019). IoT, perhaps not a device, will perform in the way needed by an HMI (Ma et al,2017; Azgomi et al, 2019).

Logistics play a significant function in fulfilling customer orders to manufacturing (Fariva et al, 2019). Owing to the difficulty and wide range of customer requests, the order for concurrent information, and information accuracy, the warehouse's function has changed dramatically (Entchev et al, 2019; Abbasi et al, 2020). Thus, it raises the traditional or physical process that leads to low warehouse operations efficiency and is no longer receptive to client criteria (Lorenc et al, 2019; Bevilacqua et al, 2019). An advanced warehouse management system (WMS) should supplement the physical process because of its difficulties (Zhu et al, 2020). It recommends the features of the WMS, such as the process of order selection (Minashkina et al, 2020). An IoT computer can supply information on the products to the pickers' job position of the performance and the order picking development to increase (Lorenc et al, 2020; Chen et al, 2019). The heavy workload of staff is the central concern (Kutlu Gündoğdu et al, 2019).

An IoT computer can supply information on the products to the pickers' job position to increase the order picking method's performance (Azadeh et al, 2019). It is very prevalent that the employee randomly places the goods in the manual process, and the picking method relies on the worker's recollection and familiarity (YILMAZ et al,2020). Consequently, the worker's productivity would be decreased, leading to a high earnings rate (Arena et al,2020). An IoT-based WSM is recommended to mitigate the warehouse service process because of the above issues. By implementing the IoT-based technologies in the delivery of the process, instead of physical paper evidence of the material, the elimination of redundant procedures will minimize the picker's workload. It will help enhance the factory's operation's quality and increase employees' contentment at work.

A recent research analysis on warehouse management dealing with low-volume, a high item for consumption combine with IoT technology is described in this paper in Section 2. The proposed system's entire structure is described in Section 3, and case analysis is provided to support our proposed system in Section 4. Section 5 presents conclusions and future work.

## 2. RELATED WORKS

(Vatumalae et al, 2020) suggested the Warehouse Management System of a Third Party Logistics Provider for Important advantages resulting from the Factory implementation 3rd Party Logistics Service (3PLs) Supplier Control Scheme. Through performing an in-depth case study in one of the core

third-party logistics (3PLs) firms, providing much of the logistics facilities, including transport and warehousing, a qualitative research approach was introduced. These research results have important consequences for better discussing the Warehouse Scheduling Management Method's ideas, methods, and application in the warehouse processes, academicians, and 3PLs organizations. Future research can lead to an examination of the partnership between the warehouse Schedulingmanagement system and 3PL market competition within the organization of 3PL.

(Dharmasiri et al, 2020) recommended implementing Multiple Automated Ground Vehicles Traffic Real-Time Control algorithms for Warehouse Operations; in the transport of products, parts, and raw materials, various industrial warehouses have gained from automated guided vehicles (AGV) systems. Despite the great stability provided by the AGV systems, the AGV systems' introduction raises significant challenges in managing traffic between several ground vehicles within the warehouse. In combination with an anti-collision route preparation algorithm, an AGV traffic management system provides efficiencies for the movement of warehouse materials. This paper's findings offer timely and useful insights into the smart warehouses and logistics phenomena as a possible method for improving material movement to be more effective and collision-free through the use of new technology AGV systems in warehouse management.

(Alsahfi et al,2020) planned a Survey on Trajectory Data Warehouse for Advanced location acquisition technologies to track moving objects in geographical space (people, aircraft, cars, livestock, ships). Such technologies provide a large amount of trajectory data warehouse (TDW). Trajectory data can be managed exclusively with either Moving Object Databases (MOD) or Warehouse of Trajectory Data. This paper will review the current research in this paper to organize, handle, and review TD using data warehouse technology. This paper suggests a system that helps to include the criteria for constructing the TDW. In addition, using the TDW, this study analyzes multiple applications and how the TDW is used for these applications. This paper fixes some problems with current TDWs and explore potential work in this context.

(Zimon, 2019) suggested the impact of quality management on inventories in commercial ventures operating within-cluster purchasing organizations. Inventory storage costs are always high, and appropriate approaches need to be implemented to optimize them. Different types of techniques supporting logistic processes (SLP) are applied for this purpose. Often common are joint events within community buying organizations. Companies are given an incentive to reduce prices. The study focused on chosen financial metrics to infer that quality control schemes' implementation increases supply management effectiveness. The adoption of such programs allowed large corporations to coordinate their warehouse management and minimize stockpile stocks, which affected their management expenditure.

(Afifudin et al, 2020) recommended executing a Structured Object-Oriented Formal Language for Warehouse Management systems for scheme architecture is inseparable from applications' development. Many issues face the design process in all software development phases, such as inappropriate and vague requirements. By applying formal engineering approaches, these challenges can be solved. Structured Object-oriented Formal Language is one of which (SOFL). The study and development of the plan and execution of SOFL are carried out to solve the difficulty. Using a systematic analysis and manageability index, architecture, and execution are assessed and tested. This investigation utilizes a method of warehouse organization, a security scheme, as a case study.

Logistic warehouse scheduling management with IoT and human-machine interface has been used to overcome the existing model issues, 3PLs, AGV systems, TDW, SLP, and SOFL. In this study, WSM has been suggested to increases product delivery ratio, operational cost, demand prediction ratio, accuracy ratio, and performance ratio. The following section discusses the proposed model briefly.

### 3. WAREHOUSE SCHEDULING MANAGEMENT (WSM)

A logistic warehouse management system (WMS) consists of software and systems that allow companies to control and handle warehouse transactions as soon as goods or machinery reach the warehouse before it moves out. As rapid, accurate, and effective as warehouse operations helps the supply chain to operate. A WMS plays an important role in handling the supply chain operations, from raw materials to finished product delivery. In the processing of new orders, the sharing and modification of information is a key concern, though order alters often occur. The future IoT-based warehouse scheduling management (WSM) completely uses RFID technologies and wireless sensors to the path and trace raw parts, semi-finished products, and completed goods. The raw components and semi-finished products involved are in limited quantity with a great variety in the highly personalized and versatile low-volume high-product mix industry. The main component of a Warehouse Control System solution is the Human Machine Interface (HMI). The HMI is a graphical management tool used to provide operators and maintenance staff with insight into comprehensive automation system status information. To track the critical statistics of the material handling equipment system based on HMIs have real-time visibility.

Figure 1. Warehouse operation

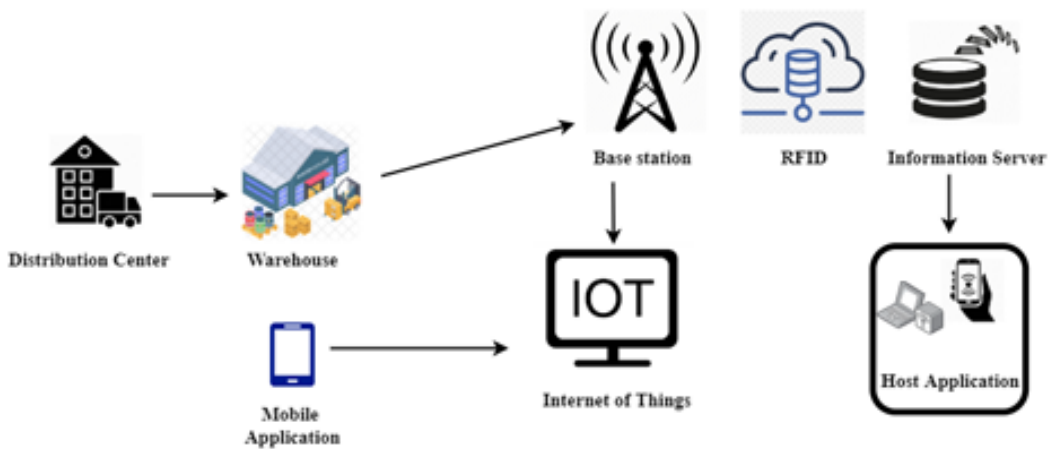


Fig 1 shows the Warehouse operation. Intelligent supply organization/production scheduling may decrease the supply levels and, thereby, the operational costs for storage/recovery and order picking. A supply chain is a network for a firm to produce and distribute a specific product or service between its suppliers. Product development, marketing, operations, distribution, finance and customer services are all functions of the provider chain. Stock reductions may be established by having smaller ordering quantities delivered more repeatedly. They determine ordering policies for all products, which reduce the long-run supply holding and ordering expenditure per unit time by resolve the following difficulty. The management of the warehouse means the supervision of warehouse operations. This includes receipt, tracking, stockpiling, training personnel, shipping management, workload planning and goods movements monitoring.

$$\text{Min} \sum D_i E_i + B_i E_i / R_i + s D_i R_i / 2 \tag{1}$$

$$\sum G_i R_i = G \tag{2}$$

where  $DH$  is the demand rate in units per year for

As shown in equations (1) and (2) where  $E_i$  is the demand speed in units per year for an item for consumption  $i$ ,  $B_i$  the fixed ordering costs for product  $i$ ,  $D_i$  the unit variable purchase cost for product  $i$ ,  $s$  the yearly supply transport cost rate,  $R_i$  the order quantity for product  $i$ ,  $G_i$  the amount of space occupied by one unit-load of product  $i$ , and  $G$  the available stock compartment space. The paper suggests a condition found in many services where device loads are resupplied one at a time, e.g., pallet storage). They use the following notation:  $T$  product package allocated to the forward-facing area,  $Q_{j\pi}$  is a random variable representing the number of items for consumption  $j$  during the selection process  $j = 1, \dots, O$ ,  $S_{j\pi}$  is a random variable representing the number of concurrent replacement for product  $j$ , if the forward area contains  $\bar{I}$  unit-loads of product  $j$  at the beginning of the selection stage,  $j = 1, \dots, O$ ,  $i = 1, \dots, n_j$ .  $V_j$  is the random vector reflecting the number of commodity unit-loads  $j$  that is needed to fulfill requirements during the picking era.

Expressions (3) and (4) include the estimated number of picks from the backward and forward fields.

$$\sum_{j=T} F(Q_j), \tag{3}$$

$$\sum_{j \neq T} F(Q_j), \tag{4}$$

Let  $X_j$  denote the number of commodity unit loads  $j$  at the beginning of the picking era, and it is deposited in the forward zone. The estimated number of simultaneous replenishment is, therefore, given by the expression (5).

$$\sum_{j=T} F(S_{jx}). \tag{5}$$

A mathematical expression has been indicated, and this paper extracts for  $F(S_{jx})$ .

$$F(S_{jx}) = \sum_{l=x+1}^{\infty} (m-x) \cdot Q(V_j = m) = \sum_{l=x+1}^{\infty} Q(V_j \geq l) = F(V_j) - \sum_{l=1}^x Q(V_j \geq l) \tag{6}$$

The future IoT-based WSM completely uses RFID technologies and wireless antenna to track and trace raw division, semi-finished products, and completed goods. The input device helps to gather all modifications and inform the warehouse operations records. The incoming bits, with IoT technology, both tasks and activities are monitored, and the discrepancy can be managed and corrected immediately by the suggested framework due to order alteration or upgrading. The raw materials, semi-finished products, and completed products are stock up in the warehouse, in the sub-store, or the fulfillment center before delivery. All the pieces are marked with an RFID tag in the IoT environment.

The person defines the components or items who read the material, and information is then sent to the radio recognition and eventually to the information server via the RFID middleware. Since this proposed device is intended for a slightly lower scenario, which is the industry's standard situation in the Industry 4.0 era, the high versatility of updating details is needed for order customization. RFID information gathered can be mortified or removed anytime at any time by approved personnel through mobile apps.

Figure 2. Warehouse management in supply chain

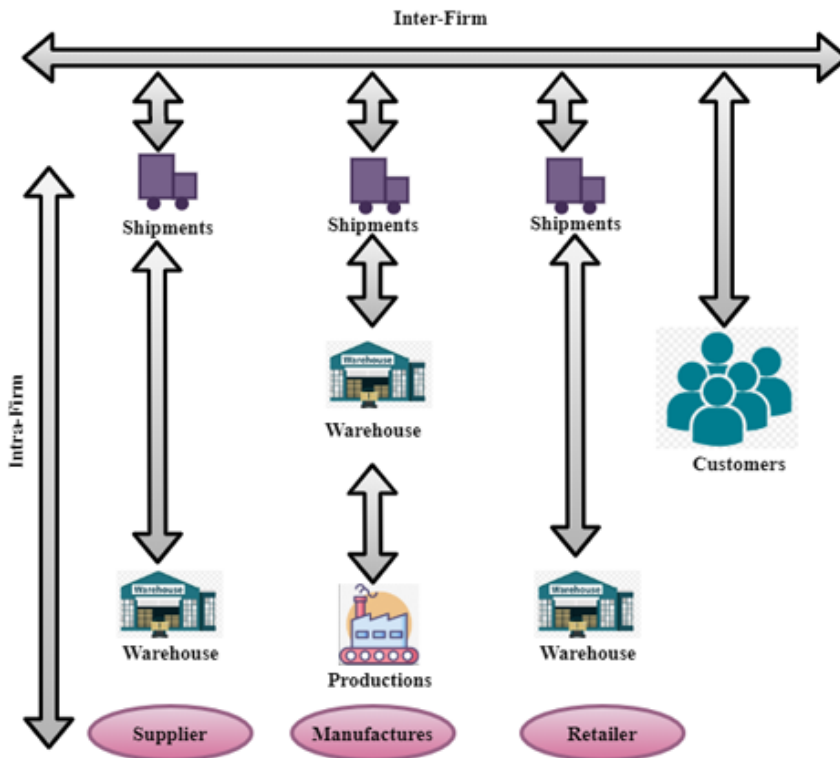


Fig 2 shows Warehouse management in the supply chain. A Warehouse Management System (WMS) is a core component of the supply chain and is mainly targeted at managing the transport and handling of warehouse goods and handling the payment process, including loading, collecting, put away, and picking. Warehousing and delivery operations are part of the framework of allocation management, which is part of the supply chain itself. A warehouse is a large firm work area used for products to be stored or build up. Time value is formed by stock up items during the year and discharge them when they are desirable. The most significant aspect of the supply chain and the distribution sector is the warehousing operation. It allows an organization to manage the inventory, and data collection tends to minimize shipping delays. Tracking the shipping times with the warehousing management system will become much simpler. Intra-firm logistics (warehousing) equipment, technology. Inter-firm logistics (transportation) comprehensive network planning, fleet, and driver assignment.

Figure 3. Order fulfillment processes

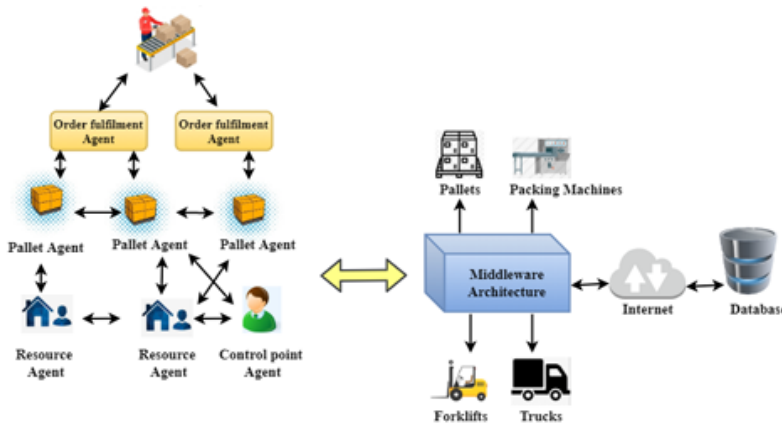


Fig 3 shows Order fulfillment processes. The method of receiving goods, then sorting and shipping orders to clients, is order fulfillment. The procedure begins with the placement of an order by a client and finishes after they collect it. If the customer needs to refund a product, order fulfillment still performs the return transaction. The steps involved in order fulfillment: • Accepting merchandise packages. • Handling of inventory. • Sorting of orders. • Packaging, processing of returns.

The method of receiving goods, then sorting and shipping orders to clients, is order fulfillment. The procedure begins with the placement of an order by a client and finishes after they collect it. However, if the customer wishes a package to be returned, order management still handles the refund transaction. The order fulfillment process addresses all the supply chain tasks involved in receiving, sorting, and shipping customer orders to their final destinations. The selection of increasingly advanced and digitized technological solutions for logistics is continually increasing: driverless transport vehicles, robot swarms, artificial intelligence. The breakup includes warehouse operations from reception, organization, performance, and distribution processes in a number of important fields. The following areas are: receipt of goods, Commodity cross-docking, Inventory organization and storage.

Figure 4. Order picking processes

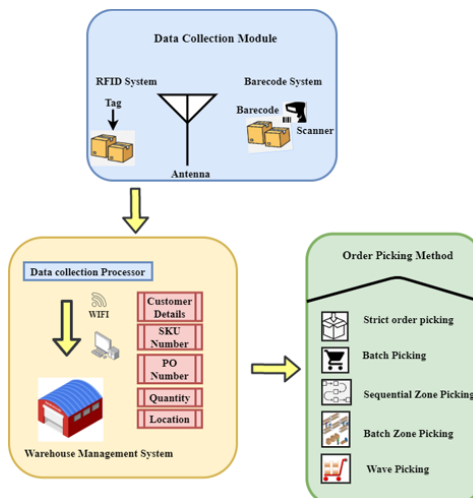


Fig 4 shows order picking processes. To fulfill many independent customer requests, order picking is the process of deciding products from a warehouse. This is an important aspect of the delivery chain process, and with any warehouse, it has been considering the manual and expensive operation. Warehouse delivery and optimized distribution allow for increased labor productivity and customer satisfaction. In the order fulfillment process instead it helps reduce errors and damages. Moreover, during handling it will prevent everyone's goods from being lost or stolen. Order picking is essentially the act of taking out the correct items for an order from a warehouse, regardless of the system selection. The order fulfillment process is the first step because if a warehouse gets the order picking process right, then one step towards achieving customer loyalty. In the product picking process, a single order proceeds the packing slip. Users take the order to the packing station to be packaged until. A negotiated number of packets (consumer packs) are collected and made ready for shipment in the shop. Space is often needed for warehousing, order picking, and handling for order picking in general. Their efficiency and utility can largely be determined by how the relationship between human and machine is constructed ergonomically between the warehouse employee and digitized picking aids or a logistics manager and the resource planning IT systems. The cost associated with taking a piece of product out of a warehouse and preparing it for shipping is a pick and carry fee. The suggested WSM model enhances product delivery ratio, operational cost, demand prediction ratio, accuracy ratio, and performance ratio.

#### 4. RESULT AND DISCUSSION

The proposed WSH model's experimental results have been executed. This paper analyzed the performance metrics such as product delivery ratio, operational cost, demand prediction ratio, accuracy ratio, and overall performance ratio.

Figure 5. Product delivery ratio

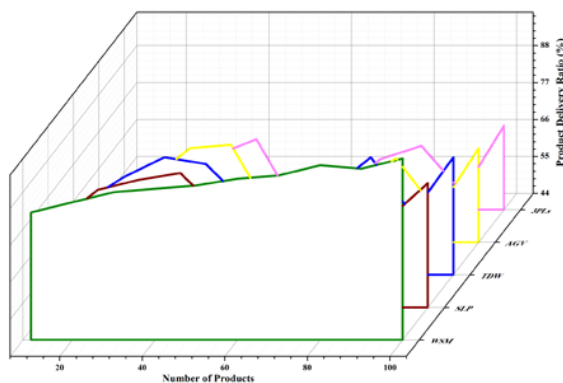


Fig 5 shows the product delivery ratio. When it comes to the supply chain performance, the Ideal Order Rate is another critically significant logistics parameter. It calculates the number of orders



collected, transported, and delivered without any accidents route. Both the shipping time and arrival time are appreciated, the order is not faulty, and the items are not broken. The Estimated Time Arrival is calculated from when the order is placed to be sent to the instant it is delivered to the client/post office. After work surface spot and understanding the normal distribution time from the factory to anyplace, the aim is to reduce it where potential providing unique release packages. It is easier to say that an order will arrive in 4-5 commerce days than to say it will reach 1-5 commerce days. Furthermore, it is much easier if it can decide the shipping hours. This way, the customer knows when he should be house to pick up the pack, increasing the order's correctness rate and preventing returns.

Figure 6. Operational cost

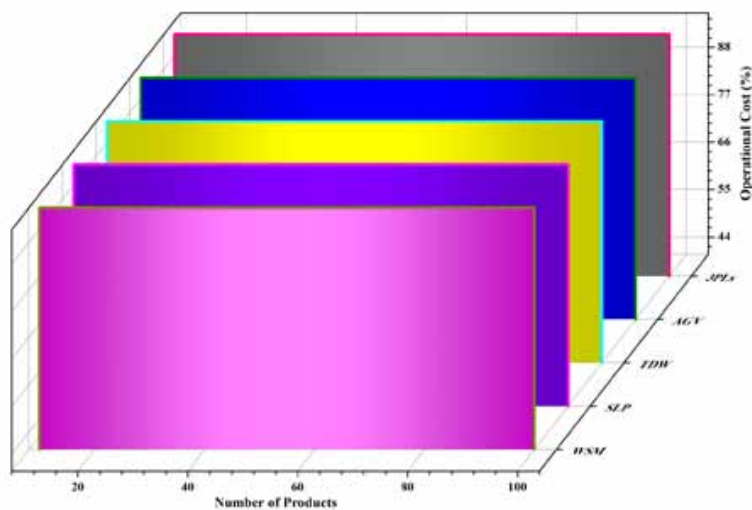


Fig 6 shows operational cost. Administration of operations costs is borne to sustain the service of the fulfillment center. Line supervision charges, clerical effort, information technology, equipment, premiums, and taxes are included. Typical warehouse processes engage packing missing items, transporting items inside or from warehouses, and receiving congregation, processing, or delivering items. Warehouse operations can be known to organize items for sale or supply, and these are protected elsewhere. The separate revenue expenses come under the general sub-categories of direct labor, direct supplies, and overhead and can be assumed to include the sales-related commissions' expense. The cost of revenue is measured as inventory starting + transactions - inventory finishing. The proposed WSM model reduces the operational cost when compared to other existing methods.

Figure 7. Demand prediction ratio

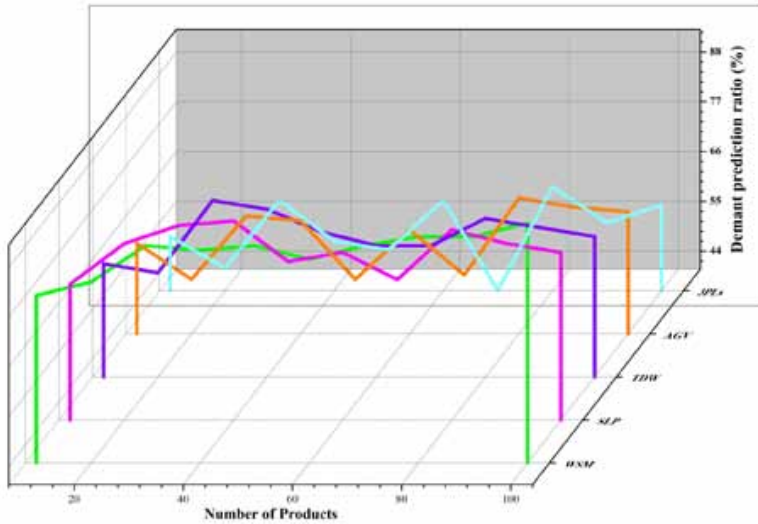


Fig 7 shows the demand prediction ratio. In assessing the overstock, the proposed logistics WSM metric is better. It compares the ratio of the product available for sale to the total amount that is consumed. This is a perfect predictor of success and can inform whether the business can face unforeseen circumstances. If calculating and use it with inventory turnover, or the carrying cost of inventory, it is an even greater metric. Demand prediction is an area of business intelligence that helps understand and anticipate consumer demand through the corporate supply chain and market management to improve supply decisions. The demand analysis ratio requires objective approaches such as data usage, particular historical revenue data, and test-market statistical methods.

Figure 8. Accuracy ratio

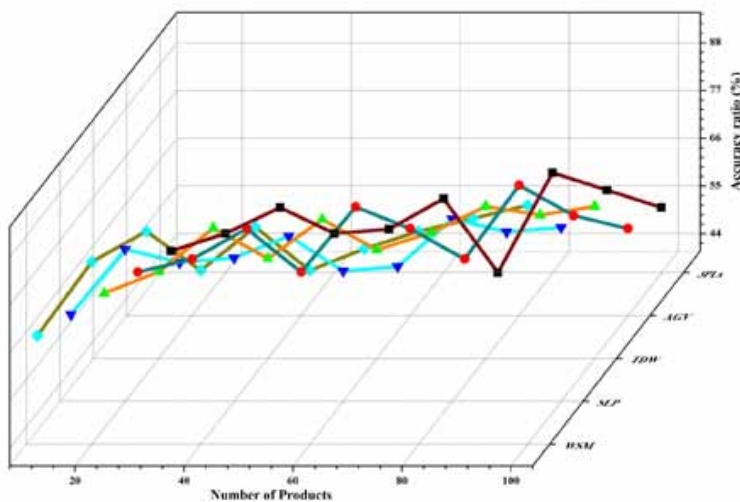


Fig 8 shows the accuracy ratio. Tracking and tracing will become much quicker when using IoT and HMI to stock-taking precision will become much higher in warehouses with associated pallets. In the transportation business, vehicle optimization and maintenance preparation can be carried out easily and effectively. To translate and relay critical information (product location, inventory details), sensors are mounted in the storage area and on inventory items and monitor inconsistencies, such as missing goods, directly to the mobile devices/dashboards of warehouse managers. Corrective steps take place in real-time, dramatically enhancing the picking process precision and performance.

Figure 9. Performance ratio

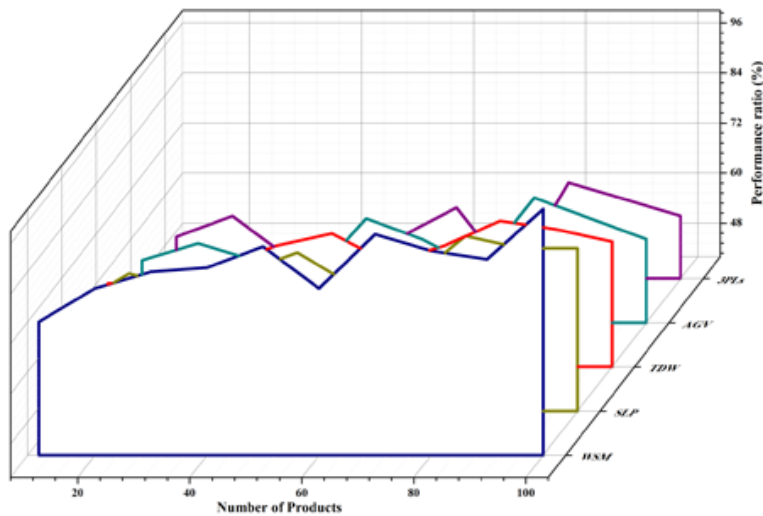


Fig 9 shows the performance ratio. Quality metrics are numerical qualities that are usually used to illustrate a storage or delivery center’s successes and limitations. Typically benchmarks concentrate on external outputs such as on-time shipments and concentrate on receiving, transportation, picking/packing orders, and shipping. Each operation should adapt Key Performance Indicators (KPIs) to their condition, and a small number of KPIs that should be worn by all warehouses to work more effectively and efficiently: inventory accuracy, receiving efficiency, cost of picking and packing, inventory proceeds, and order occasion for the customer cycle.

The proposed WSH models enhance the product delivery ratio, operational cost, demand prediction ratio, accuracy ratio, and performance ratio when compared to other existing 3rd Party Logistics Service (3PLs), Automated Ground Vehicles (AGV), Trajectory Data Warehouse (TDW), Supporting Logistic Processes (SLP), and Structured Object-oriented Formal Language (SOFL) methods.

## 5. CONCLUSION

Owing to the complexity and wide variance of consumer requests and the need for real-time details, a warehouse’s activities are expected to adapt accordingly. Therefore the conventional service of a manual warehouse is no longer appropriate for business employees. To increase performance and allow customized order fulfillment, an advanced WSM is therefore very important. The proposed

WSM is a technique in this paper to choose the most effective form of order picking, thereby improving the order picking process's efficiency. It was inferred from the case study findings in this study that the WSM assist in supplying and enhance depending on the warehouse process in terms of both social and economic benefits. For the tangible gains, order fulfillment such as order fill rate and order consistency may be improved. In addition, order picking will increase the time of receipt, inventory accuracy, and warehouse efficiency. It will boost the packaging process for the intrinsic gains; furthermore, the stock can be noticeable with RFID. The preparation method considers the complex essence of consumer orders' demand, picking field layout, and immersive human-machine interface. Furthermore, it is possible to boost the confidence of the employees. As this study's focal point primarily on realistic request and order selection, there has been no detailed discussion of the routing and storage policies. The integration of fake attitude will be one of the potential paths to allow smart logistics and knowledge to be more streamlined to streamline warehouse operations with greater productivity, presentation and fewer expensive in the extended word, and more investigate on smart robotics be supposed to carry out as the warehouse pick and pack operations are shifted from picker-to-goods to goods-to-picked. The simulation results have been executed, and the suggested method achieves a high in product delivery ratio (96.5%), operational cost (94.9%), demand prediction ratio (96.5%), accuracy ratio (98.4%), and performance ratio (97.2%).

## **FUNDING AGENCY**

This research received no specific grant from any funding body in the public, commercial, or not-for-profit sectors.

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