

Agile, IoT, and AI: Revolutionizing Warehouse Tracking and Inventory Management in Supply Chain Operations



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Abstract

Aim: The way traditional supply chains operate has shown limited success in coping with changes in supply chain requirements related to inventory tracking and warehouse management. The research evaluates the collaborative effects of Agile methods with IoT devices and AI capabilities to optimize these processes.

Methods: This study examines three vital aspects of data science deployment within a consumer product distribution company that uses IoT sensors (RFID tags and Texas Instruments CC2650) for real-time data combined with AI analytics (Random Forest and reinforcement learning in Python) through an Agile (2-week cycles via Jira) deployment approach for inventory management. A twelve-month project employs AI modeling based on Python and utilizes Scrum sprints as its methodology.

Results: The systematic study produced three significant results which include a 25% higher inventory turnover rate, 20% fewer tracking errors, and 15% lower operating costs. Strong solutions emerge from the combination of Agile with IoT and AI and demonstrate promising capabilities for enhancing supply chain resilience at a large-scale level.

Conclusion: Practical applications from the research follow some practical suggestions and directions for upcoming scientific investigations into blockchain technology implementation.

Recommendation: The research presents real-world implications for medium firms and recommendations for blockchain-based secure data-sharing studies to advance supply chain functions.

Keywords: *Agile methodologies, internet of things (IoT), artificial intelligence (AI), supply chain operations, warehouse tracking, inventory management, predictive analytics, real-time data, operational efficiency, scrum*

INTRODUCTION

Supply chain logistics executes order distribution through warehouse tracking systems which maintain inventory control to move products inventory storage to direct customers. These processes operate under significant pressure in the present era which includes both e-commerce evolution and global trade expansion despite their paramount need to provide speed and accuracy with adaptable mechanisms. Several traditional inventory tracking techniques such as manual counts paper-based data recording and outdated computer programs face limitations due to variable customer demand inefficient business processes and large transaction volumes which produce out-of-stock situations delayed transportation needs and increased costs. Due to high risk levels, one wrong inventory mistake will break down an entire supply chain network leading to decreased customer trust and profit reduction. The main issue stems from legacy systems that cannot supply live visibility alongside predictive analytics and scalable options which create operational restrictions for middle-sized firms actively serving a high-speed consumer-based market. The main problem with legacy systems stems from their inability to deliver real-time visibility together with predictive insights and scalable solutions that create operational bottlenecks and lost commercial opportunities for mid-size companies working in a fast-paced consumer market.

Technology integration with logistics systems provides a critical connection to help the industry. Real-time data delivery from sensors along with RFID tags in the IoT world makes previously unknown processes visible. With analysis of data streams, AI generates forecasts while carrying out automated decisions that earlier depended on uncertain human factors. Implementing technology requires Agile methodologies to operate effectively because Agile divides large projects into manageable cycles. These tools unite to create a transformational change in warehouse operations instead of achieving small progressive improvements. The urgency to address greenhouse gas emissions in logistics rises due to its contribution to 13% of total global CO₂ emissions and calls for resource-maximizing solutions and waste-minimization measures (McKinnon *et al.*, 2015).

The research takes on this interface because numerous mid-size organizations unlike larger companies face limitations when it comes to immediate systematic overhauls. Modern commercial speed alongside consumer demand for quick delivery exposes an essential problem for smaller businesses regarding advanced tool access while managing system complexity. Moreover, the environmental toll of logistics, accounting for 13% of global CO₂ emissions, demands solutions that optimize resources and reduce waste (McKinnon *et al.*, 2015). By integrating Agile, IoT, and AI, we propose a practical path forward, one that balances innovation with feasibility.

A simulated mid-sized distributor with three warehouses throughout North American territory serves as the research focus. Our main objective is to conduct three essential tasks: (1) evaluate the impact of these technologies on turnover and accuracy metrics (2) identify practical solutions to existing challenges and (3) build a design template that can serve as a guide for others. This isn't just about efficiency it's about equipping firms to thrive in a competitive, sustainable future, making supply chains a driver of economic strength (Sanders, 2018; Chopra & Meindl, 2016).

LITERATURE REVIEW

Supply chain management development has tightened its dependence on technological solutions to resolve ongoing performance issues. As fundamental centers of inventory distribution

warehouses have served as the prime target for emerging innovations. The initial supply chain management system functioned through barcode readers and basic software which struggled with contemporary market management requirements. IoT introduced revolutionary changes through sensor technologies that monitor physical objects for instant product tracing. Studies show it can slash tracking errors by up to 30%, offering a level of visibility once unimaginable (Atzori *et al.*, 2010).

The origin of Agile methodologies within software development enabled them to introduce new approaches for technology-driven organizational evolution. The structured short cycles under Agile systems serve as a feedback-driven framework that controls integration chaos. The flexible approach of adapting through each step suits the perfect implementation of IoT and AI in warehouses which feature unpredictable environments. Research suggests this approach cuts deployment risks and speeds up results, though its use in logistics is still in its infancy (Conforto *et al.*, 2014).

The evidence is compelling: AI-driven robotics can trim warehouse costs by 10–20% (Sanders, 2018), while IoT-enhanced tracking boosts responsiveness to disruptions like weather delays (Ivanov *et al.*, 2019). Predictive analytics, a staple of AI, consistently outperforms traditional forecasting, reducing stockouts and overstocking (Lee *et al.*, 2021). True productivity arises when multiple techniques unite. The real power emerges by combining IoT data delivery to AI systems while Agile maintains technology usefulness by preventing overreliance on technology.

Challenges persist. Data quality can make or break AI models - garbage in, garbage out, as the saying goes (Fawcett *et al.*, 2011). Integrating with old systems is another hurdle, often requiring costly upgrades or workarounds (Gunasekaran *et al.*, 2017). And then there's the human factor: staff need training to trust and use these tools effectively, a step often overlooked (Rushton *et al.*, 2017). The research explores possible cooperative operations of Agile technology with IoT and AI systems throughout a medium-scale organization that faces both resource restrictions and efficiency requirements.

METHODOLOGY

Case Study Design

This study simulates a mid-sized consumer goods distributor in North America, managing three warehouses (California, Texas, Ohio), \$150 million in annual revenue, and a fleet of 50 vehicles plus third-party logistics partners

Materials and Technologies

IoT: RFID tags (Alien Technology ALN-9640) and sensors (Texas Instruments CC2650) for real-time tracking.

AI: Python 3.9 with sci-kit-learn (Random Forest) and TensorFlow (reinforcement learning).

Agile: Scrum via Jira, 2-week sprints.

Hardware: Simulated Dell PowerEdge R740 servers.

Data Collection

The data analysis included twelve months of collected information with ten thousand SKUs and one hundred thousand RFID scans and demand signals such as sales, weather and promotional data. Designers implemented the system to reproduce actual market patterns that included seasonal demands and unanticipated changes.

Procedures

IoT: The real-time inventory tracking used MQTT to exchange data between various warehouses. Random Forest used an existing set of data to determine future forecasts, but reinforcement learning worked to standardize inventory levels across the organization.

AI: The analytical models received 80 percent of the total information for training purposes while they validated their operations through 20 percent of the available data.

Agile: The project involved five sprints where staff participation guided three IoT setup and two AI integration phases.

Data Analysis

The evaluation examined three performance indicators which included inventory turnover as well as tracking accuracy and operational costs. Stock turnover frequency per year served as a measurement while accuracy verified the correctness of recording movements and total expenses associated with labor staff, holding costs, and administrative errors. The measurements between pre-implementation and post-implementation periods were evaluated through paired t-tests from the SciPy library ($\alpha = 0.05$) across three warehouses for a period of 12 months. The research checked whether observed improvements including error reduction reached statistical significance when p-values were below 0.05.

The analysis examined three scenarios through sensitivity checks using AI models to determine their behavior when facing either a 20% increase (for holidays) or 15% decrease (for slowdowns) or $\pm 10\%$ unpredictable weekly changes. The Random Forest and reinforcement learning models demonstrated strong performance compared to the baseline while dealing with the 20% surge and 15% drop scenarios and random $\pm 10\%$ weekly fluctuations because of the sensitivity analysis. Current figures present results as means along with standard deviations. All analysis ran on simulated servers, scripted for transparency and repeatability (Boute & Van Mieghem, 2021).

RESULTS

For the inventory turnover, the company raised its inventory turnover rate to five times per year after an increase of twenty-five percent ($p < 0.01$). On the tracking accuracy, the implementation of IoT analytics and precision techniques decreased health department errors by 20% ($p < 0.05$) to reach 2% in total errors. On costs, the company gained a performance boost equivalent to 15% (\$22.5 million annually) through improved labor efficiency and inventory efficiency.

Table 1: Performance Metrics Comparison

Metric	Pre-Implementation	Post-Implementation	Change
Turnover (turns)	4	5	25%
Errors (%)	10	2	-80%
Cost (\$M)	150	127.5	-15%

DISCUSSION

The results paint a clear picture: IoT's real-time tracking slashed errors by providing an unblinking eye on inventory movements, a leap forward from manual counts that often lagged in reality (Atzori *et al.*, 2010). AI's predictive power cut waste by aligning stock with actual demand, a stark contrast to the overstocking or shortages plaguing traditional systems (Lee *et al.*, 2021). Agile tied it all together, ensuring a rollout that adapted to surprises—like initial sensor glitches—rather than stalling out (Conforto *et al.*, 2014). Agile showed its worth by enabling the team to handle data glitches along with training needs through its adaptive approach. These technologies have transformed warehouse operations through revolutionary changes by transforming them into predictive facilities that monitor customer demands rather than relying on response to chaos. Mid-sized companies rely on this as their survival strategy since large organizations solve issues with big financial investments, but the operational approach of this method helps smaller entities achieve better results.

The 15% cost drop isn't just a number, it's a ripple effect. Less overstock means less warehouse clutter, fewer rushed shipments, and a leaner carbon footprint, tying directly to sustainability goals that matter more every day (McKinnon *et al.*, 2015). Agile feedback loops enabled staff members to adapt more quickly than anticipated, indicating that technological implementation for employees could be less challenging than originally thought. The mid-sized investigation demonstrates that scalability applications are not limited to enterprise operations, but they start from basic approaches towards intelligent development.

Challenges weren't absent. Early data inconsistencies threw off AI predictions until cleaned up in later sprints, a reminder that tech is only as good as its inputs (Fawcett *et al.*, 2011). Training staff to trust IoT dashboards took time, but their buy-in proved crucial, echoing the need for people-centric approaches (Rushton *et al.*, 2017). The true trial of these advancement achievements will depend on their ability to maintain stability through changing customer buying behaviors and technological evolution. This proper combination has the potential to transform how suppliers work together and distributed delivery services which demand innovation. The adaptable structure of Agile provides organizations foundations to dynamically progress which moves them from temporary solutions to sustainable improvement platforms.

CONCLUSION

Through the union of Agile methodologies with IoT and AI in warehouse tracking systems the study proved that improvements could be achieved with a 25% rise in inventory turnover and both 20% cut in tracking errors and 15% reduction in operational costs in a simulated mid-sized distributor. These findings directly solve traditional system weaknesses as they provide IoT-enabled real-time monitoring in combination with AI-based prediction capabilities together with

Agile-based adaptable implementation for building resilient e-commerce supply networks. By increasing the turnover rate by 25% the company shortens its stock holding period to minimize waste and shorten order delivery times and by reducing errors by 20% reliability improves to prevent production disruptions. The 15% cost reduction equivalent to \$22.5 million yearly demonstrates improved labor and inventory management which enables scaled business operations for medium-sized organizations with minimal budget.

Systems that transform market needs through this framework's basic operational structure of IoT tracking together with AI predictions and Agile adjustments do not need perfect initial execution for business advantage. The sustainability goals are served by two key environmental benefits: decreased overstocking and optimized logistics that reduce 13% of global CO₂ emissions from logistics (McKinnon *et al.*, 2015). The simulation operates within a controlled setting that provides consistent results but external conditions such as equipment breakdowns or labor pushbacks need additional testing. The proposed blockchain integration, building on IoT's data transparency, offers tamper-proof tracking to enhance trust and auditability, as emerging logistics applications suggest (DHL Trend Research, 2020). The proposed framework requires further testing under real-life conditions and researchers need to investigate blockchain as a data-sharing solution for multi-tier supply chains and prove its applicability to various industry sectors. This strategy enhances mid-sized companies by changing warehouses into strategic resources that propel North American logistics evolution while maintaining economic performance and environmental sustainability for future adaptability.

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