

# Strategic Analysis of Coordinated Optimization in Warehousing and Distribution of Perishable Agricultural Products

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**Abstract:** The logistics of perishable agricultural products face multifaceted challenges in maintaining quality, controlling costs, and ensuring timeliness due to the perishable nature of the goods. Traditional segmented management often leads to internal inefficiencies within the system. To address this, this paper focuses on the coordinated optimization of warehousing and distribution. By analyzing the system's requirements for coordination, interaction mechanisms, and key bottlenecks, the research constructs a spatiotemporally coupled integrated modeling framework. This framework dynamically quantifies quality decay and incorporates it into a multi-objective optimization model, followed by an exploration of algorithm design for dynamic environments. Based on this, a strategic system comprising information-integration synchronization, resilient network linkage, and vertical quality coordination is proposed. This study provides a systematic analytical framework and strategic support for breaking down segmental silos and achieving holistic resource and value coordination across the full chain.

**Keywords:** Perishable Agricultural Products; Warehousing and Distribution; Coordinated Optimization; Quality Decay; Spatio-Temporal Network; Multi-Objective Optimization

## Introduction

The realization of value for perishable agricultural products is highly dependent on the efficiency of their logistics and the level of quality preservation from post-harvest to the end consumer. Due to the significant biological activity and time-sensitive fragility of these products, inherent tension exists among the goals of cost, efficiency, and quality within their logistics system. Under traditional management models, warehousing and distribution, as core operational stages, are often planned and optimized independently. This fragmentation leads to systemic issues such as mismatches between inventory strategies and transportation plans, where local cost savings trigger increased global losses, severely constraining the improvement of overall supply chain performance. Consequently, breaking down the barriers between stages and achieving coordinated optimization of warehousing and distribution has become a critical pathway for enhancing the efficiency of perishable agricultural product logistics and reducing value loss. Conducting an in-depth analysis of the conceptual underpinnings, models, and methodologies of coordinated optimization, and constructing a practical and actionable strategic framework, holds significant theoretical importance and presents an urgent practical necessity for guiding the refined operational management of fresh produce supply chains. This study aims to systematically address this need, providing reference for theoretical development and practical application in related fields.

## 1. The Connotation and Key Challenges of Coordinated Optimization for Warehousing and Distribution of Perishable Agricultural Products

### 1.1 The Uniqueness of the Perishable Agricultural Product Logistics System and the Need for Coordination

The core characteristic of the perishable agricultural product logistics system lies in the biological activity and time-sensitivity of the managed objects. Their value loss exhibits a strong correlation with time and environmental parameters during the logistics process. This inherent property dictates that its optimization objective is fundamentally a multi-objective balancing problem: simultaneously

suppressing quality decay while controlling logistics costs and meeting service requirements. Traditional segmented management separates warehousing and distribution for isolated optimization. Warehousing focuses on inventory costs and in-warehouse loss control, while distribution concentrates on routing costs and delivery timeliness. However, this isolated optimization model is prone to causing systemic inefficiencies. For instance, batch strategies at the warehousing end may reduce operational costs but can extend inventory turnover time, thereby accelerating quality deterioration and constraining the flexible operational space in the distribution stage. Consequently, the need for coordination is rooted in the biological principles of the products themselves. It necessitates the construction of an integrated decision-making framework that jointly optimizes inventory policies, sorting operations, transportation routing, and delivery scheduling. This aims to achieve holistic resource allocation and value preservation across the entire chain from post-harvest handling to end-consumer delivery.

### ***1.2 The Interactive Mechanism and Conflict Analysis between Warehousing and Distribution Processes***

The warehousing and distribution processes are tightly coupled through material flow and information flow, constituting a nonlinear interactive system. The output rate and order completeness from warehousing directly constrain the initiation time and loading efficiency of distribution. Conversely, the structure of the distribution network, along with the time windows and distribution density of customer demands, inversely defines the order wave planning and inventory deployment strategy in warehousing<sup>[1]</sup>. Inherent conflicts exist between the two processes across multiple dimensions. From a cost perspective, the economic lot-size principle of warehousing and the principle of frequency economy in distribution mutually constrain each other, manifesting as a trade-off between inventory holding costs and transportation costs. In the time dimension, the duration of warehousing operations erodes the available in-transit time for distribution, while lenient delivery schedules increase the static storage period of inventory, both exerting pressure on quality preservation. Regarding resources, the order-picking capacity and temporary storage area capacity of warehousing, along with the vehicle capacity and loading/unloading resources of distribution, form interconnected capacity bottlenecks, which become particularly prominent during peak demand periods. Analyzing this bidirectional constraint and multi-objective conflict forms the basis for identifying key leverage points for coordination and constructing an integrated optimization model.

### ***1.3 Major Bottlenecks and Constraints in Achieving System-wide Coordinated Optimization***

System coordination faces multiple intrinsic bottlenecks and rigid constraints. Information fragmentation prevents the synchronized sharing of inventory status, quality data, and dynamic routing information, leading to decisions based on localized or outdated information and making precise synchronous planning difficult to achieve. At the organizational and management level, fragmented performance evaluation systems hinder the identification and equitable distribution of overall benefits, thereby dampening the motivation for coordination. On the technical front, the non-linear nature of product quality decay processes presents a core modeling challenge, making it difficult to accurately model and embed within optimization algorithms. Physical constraints include the fixed layout of cold chain network nodes, the processing capacity and temperature zone limitations of warehousing facilities, and the refrigeration performance and load capacity of transportation equipment; these collectively delineate the boundaries of the feasible solution space for coordination strategies. Furthermore, market demand uncertainty poses a continuous test to the dynamic adjustment capability of the coordinated system, requiring the optimization strategy to possess inherent flexibility to cope with fluctuations.

## **2. Theoretical Models and Analytical Methods for Coordinated Optimization of Warehousing and Distribution of Perishable Agricultural Products**

### ***2.1 An Integrated Spatio-Temporal Modeling Framework for Warehousing and Distribution***

#### ***2.1.1 Representation and Extension of the Spatio-Temporal Network***

The fundamental model extends the logistics network into a spatio-temporal network. The spatial dimension retains the conventional topology of nodes (e.g., warehouses, distribution centers, customer points) and arcs (i.e., transportation routes). The temporal dimension is incorporated by discretizing the

planning horizon into multiple time periods, thereby replicating each physical node across each time point to construct an expanded network containing spatio-temporal nodes. Material flow, information flow, and decision flow propagate within this three-dimensional network. Warehousing operations (such as storage and sorting) correspond to the temporal dwell of goods at specific spatial nodes, while distribution and transportation correspond to the movement of goods between spatio-temporal nodes. This representational method can explicitly capture the cumulative impact of operation durations, waiting times, and in-transit time consumption on subsequent processes<sup>[2]</sup>.

### **2.1.2 Embedding and Quantifying the Dynamics of Quality Decay**

The core innovation of the modeling framework lies in embedding the dynamic process of quality decay for perishable agricultural products into the flow of the spatio-temporal network. The quality state of a product (e.g., freshness, intactness rate) is defined as one or more state variables that evolve over time and under environmental conditions, primarily temperature. At warehousing nodes, these state variables are updated based on the temperature control level of the storage environment and the storage duration. On transportation arcs, they are updated according to the transportation time and the average temperature within the vehicle. Ultimately, quality loss can be translated into economic costs through predefined functional relationships (such as exponential decay models), which are then incorporated into the objective function alongside logistics operational costs. This enables the complete quantification of the "time-temperature-quality-cost" transmission chain within the model.

### **2.1.3 Construction of a Unified Decision Space**

Within the spatio-temporal network framework, the previously separate warehousing and distribution decisions are integrated into a unified decision space. Key decision variables include: the inventory holding quantity at each spatio-temporal node; from which spatio-temporal node an order is picked and released; and which spatio-temporal path (i.e., when, by which vehicle, and along which route) is selected to transport the goods to their destination. All these decisions collectively determine the specific trajectory of the material flow within the spatio-temporal network. This construction fundamentally avoids the local optimality and systemic conflicts caused by segmented optimization, providing a rigorous mathematical model foundation for implementing genuine coordinated optimization.

## **2.2 Construction of Decision Variables and Objective Functions under Multi-Objective Coordinated Optimization**

### **2.2.1 Integrated System of Decision Variables**

The model involves a multi-level system of decision variables. At the strategic and tactical level, variables include the inventory positioning strategy and safety stock levels for each facility within the cold chain network, as well as the vehicle resource allocation plan for distribution centers. At the operational level, core variables encompass: the dynamic inventory routing for each product (specifying when and where it is stored); the order consolidation and wave division rules; the picking start and completion times for each batch of orders; the sequence of customer points assigned to each delivery vehicle; the precise service time window for each customer point; and the temperature control level settings for in-transit transportation. These variables are interrelated; for instance, inventory routing influences order fulfillment feasibility, which in turn constrains the generation of vehicle routes.

### **2.2.2 Multi-Objective Conflicts and Function Composition**

The optimization objectives for perishable agricultural product logistics inherently involve conflicts, primarily manifested as a triangular trade-off among cost, efficiency, and quality<sup>[3]</sup>. The objective function typically takes the form of minimizing total system costs or maximizing total net revenue, and its composition includes:

Operational costs, encompassing fixed and variable costs such as warehouse leasing/depreciation, labor, and energy consumption, as well as vehicle fixed usage costs and variable transportation costs (fuel, toll fees);

Quality loss costs, calculated through the aforementioned quality decay model, directly reflecting the product value depreciation caused by time delays or temperature control failures;

Timeliness penalty costs, imposing economic penalties for deliveries made earlier or later than the promised time windows to quantify service level;

Stockout loss, measuring the lost sales opportunities due to insufficient inventory or untimely allocation.

### **2.2.3 Integration of Constraints**

The model's constraint system integrates multiple limitations: flow balance constraints ensure the continuity of material flow within the spatio-temporal network; capacity constraints include warehouse storage capacity, order-picking processing rates, vehicle loading weight and volume, and driver working hours; time window constraints originate from customer requirements or store operating hours; logical constraints guarantee the continuity and rationality of delivery routes. Additionally, quality safety constraints are included, such as requiring that the remaining shelf life of a product upon reaching the customer must be above a certain threshold. These constraints collectively define the feasible region within the high-dimensional decision space.

## **2.3 Dynamic Optimization Algorithm Design for Freshness Preservation and Cost Control**

### **2.3.1 Algorithm Strategy Selection and Integration**

Given the large-scale, multi-objective, and dynamic nature of the problem, pure exact algorithms (such as branch and bound) are often only applicable to small-scale instances. In practice, meta-heuristic algorithms are predominantly adopted as the core solution framework. For example, multi-objective genetic algorithms (such as NSGA-II) or particle swarm optimization algorithms, due to their strong global search capabilities, are suitable for exploring the vast solution space and generating Pareto front solution sets. Simultaneously, problem-specific heuristic rules (such as the savings algorithm or the nearest neighbor method) can be embedded to rapidly generate high-quality initial solutions, or targeted local search operators (such as intra-route/inter-route customer point exchange, inventory point reallocation) can be designed to enhance neighborhood search efficiency.

### **2.3.2 Rolling Optimization Mechanism Under Dynamic Uncertainty**

To address real-time uncertainties such as demand fluctuations, traffic congestion, and equipment failures, the static optimization model must be upgraded to a dynamic optimization framework. Rolling Horizon Control is an effective method. It divides the entire operational cycle into consecutive decision windows. At each decision point, the system re-solves the integrated model for a limited future period (the prediction horizon), starting from the current moment, based on the latest state information (real-time inventory, newly arrived orders, traffic predictions). After solving, only the optimized plan for the immediate or near-term period (the execution horizon), such as immediate picking instructions or vehicle dispatch orders, is implemented. As time progresses, the decision window continuously rolls forward, achieving a closed-loop dynamic optimization of "plan-execute-feedback-replan"<sup>[4]</sup>.

### **2.3.3 Balancing Solution Quality and Computational Efficiency**

Ensuring the algorithm returns a high-quality solution within an acceptable time frame is a core challenge. Hybrid intelligent algorithms represent the mainstream direction, combining the strengths of different approaches. For instance, simulation-based optimization methods can be employed, where complex constraint and objective evaluations (particularly quality decay calculations) are handled within a simulation module, while the optimization algorithm is responsible for iteratively searching the decision variables. Additionally, the decomposition-coordination philosophy can be utilized, which appropriately decomposes the original problem into sub-problems based on time or space. These sub-problems then iteratively update parameters through a coordination mechanism (such as Lagrangian relaxation or goal programming) to approximate a global optimum. Algorithm parameters (e.g., population size, iteration count) also require adaptive adjustment according to the problem scale to balance solution depth and computational time.

## **3. Construction of a Strategic Framework for Coordinated Optimization of Warehousing and Distribution of Perishable Agricultural Products**

### **3.1 Information Fusion-Based Synchronization Strategy for Inventory and Transportation Planning**

A prerequisite for achieving coordinated optimization between warehousing and distribution is breaking down information silos and establishing an information fusion environment that supports real-time decision-making. This requires the construction of an integrated data platform capable of continuously collecting and processing multi-source, heterogeneous data from warehouse management

systems (e.g., real-time inventory levels, storage locations, and environmental parameters), order management systems (e.g., customer demand, delivery time windows), transportation management systems (e.g., vehicle location, status, estimated time of arrival), and external systems (e.g., traffic conditions, weather forecasts). The key to information fusion lies in the alignment and integration of data across semantic, temporal, and spatial dimensions to form a unified, time-varying "supply chain state image." This image not only reflects the current state of physical resources but can also project their short-term evolution through predictive models, thereby providing a dynamic and consistent decision-making basis for synchronized planning.

Based on this panoramic information view, inventory planning and transportation planning can transition from traditional serial or loosely-coupled modes to a deeply synchronized optimization mode. The core strategy involves adopting a joint optimization engine that simultaneously considers the real-time capacity constraints, cost structure, and routing efficiency of the distribution network when formulating replenishment plans, safety stock strategies, and intra-warehouse operation schedules. For instance, the decision-making engine can dynamically adjust target inventory levels at different storage nodes based on forecasts of future demand orders and analyses of vehicle availability, and generate vehicle loading and departure schedules that are tightly dovetailed with outbound plans. This synchronization strategy can effectively reduce inventory overstock, vehicle idle time, or emergency transfers caused by information delays or planning conflicts, thereby systematically lowering inventory holding costs and emergency transportation costs while meeting service commitments<sup>[5]</sup>.

### ***3.2 Resilient Warehousing Network and Coordinated Distribution Routing Strategy Adapting to Demand Fluctuations***

The high volatility of market demand for perishable agricultural products necessitates that the coordinated system possesses structural resilience. At the warehousing network level, resilience strategies are reflected in the flexible design of facility functions and the dynamic distribution of inventory. A hybrid network structure of "central warehouses + forward micro-warehouses" can be adopted. Central warehouses handle large-volume, low-frequency storage and cross-docking functions, while widely distributed forward micro-warehouses focus on rapidly responding to end-point demand. By establishing a dynamic inventory allocation model, inventory deployment within the network can be intelligently adjusted based on real-time demand hotspot predictions and the inventory consumption rates at each node. This pre-positions the right products in nodes closer to potential demand, thereby shortening emergency response distances and times.

This resilient warehousing network must be matched with a coordinated adjustment strategy for distribution routes. When demand fluctuates, the coordinated system should not merely adjust a single segment but must initiate a joint re-optimization mechanism for the network and routes. This includes: dynamically assigning delivery zones or adjusting responsible vehicles based on the inventory status of forward micro-warehouses and the geographical clustering of demand orders; consolidating delivery routes during low-demand periods to improve load rates, and having standby vehicles and flexible route contingency plans ready for peak periods. Developing the coordinated strategy relies on in-depth analysis of demand fluctuation patterns and rapid simulation capabilities. By simulating the coordinated performance of the network and routes under different fluctuation scenarios, a tiered response strategy library can be pre-established. This ensures that when actual fluctuations occur, a near-optimal warehousing-distribution coordinated plan can be swiftly deployed to maintain the overall efficiency and stability of the system.

### ***3.3 Vertical Coordination Management Strategy Incorporating Quality Decay and Loss Control***

Quality control represents the value core of logistics coordination for perishable agricultural products, necessitating the establishment of a vertical coordination management strategy that permeates both warehousing and distribution stages<sup>[6]</sup>. This strategy fundamentally relies on applying a unified quality decay model to the development of operational standards for the entire chain. Based on the product-specific time-temperature sensitivity, maximum permissible time thresholds and temperature control standards are set for different product categories across all stages: from inbound receiving, storage, and sorting to transportation and final delivery. These standards are then translated into specific operational parameters and embedded within the task scheduling logic of warehouse management systems and the in-transit monitoring commands of transportation management systems. This ensures that operations at each stage are conducted within the predefined constraints for quality preservation.

The deeper challenge of vertical coordination lies in resolving the issues of cost attribution and responsibility determination caused by quality decay, which requires designing mechanisms to achieve incentive compatibility across stages. One strategy is to introduce a settlement or evaluation mechanism based on delivered quality, linking the final product quality grade upon delivery to the performance assessment of each stage, including warehousing and transportation. For example, a transparent quality traceability and data recording system can be established. When a product's quality upon reaching the customer falls below expectations, the system can trace its time consumption and environmental records at each stage, thereby objectively assessing the responsibility proportion of each node. This incentivizes warehousing operators to optimize workflows to reduce storage time, and distribution parties to optimize routing and temperature control to minimize in-transit losses. Both sides shift from adversarial gaming to cooperating in enhancing end-point quality, ultimately forming a quality preservation-oriented closed-loop for vertical coordination management.

## Conclusion

This study systematically examines the core issues of coordinated optimization for the warehousing and distribution of perishable agricultural products, revealing the nature of the system's multi-objective conflicts and the complex interaction mechanisms between different stages. By constructing a spatio-temporally coupled integrated modeling framework, the quality decay process is quantified and embedded into the multi-objective optimization function, thereby establishing a mathematical model foundation for coordinated decision-making. In response to the model's complexity and environmental dynamism, an optimization approach integrating meta-heuristic algorithms with rolling horizon control is proposed. Building on this, a three-tier strategic framework is developed, encompassing information fusion-driven plan synchronization, a resilient network and route linkage responsive to demand fluctuations, and quality-oriented vertical coordination. This framework aims to enhance the overall resilience and economic efficiency of the logistics system through systematic integration. Future research could delve deeper into areas such as multi-agent coordination mechanisms, complex cross-chain scenarios, the application of artificial intelligence in real-time prediction and adaptive optimization, and the more precise characterization of quality decay models.

## Fund Projects

2025 Hainan Provincial Philosophy and Social Sciences Planning Project, Project Title: Research on the Collaborative Optimization of Storage and Distribution for Fresh Agricultural Products in Hainan, Project Approval Number: HNSK (YB) 25-11

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