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Certificate in Automated Storage and Retrieval System for Warehouses

## Performance Monitoring and Data Analytics

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### Alarm Management

Related terms: event notification, threshold setting, incident response

Definition: A systematic process for configuring, prioritizing, and handling alerts generated by AS/RS equipment and monitoring software. Effective alarm management reduces noise, ensures critical issues are addressed promptly, and supports root-cause analysis.

Example: An AS/RS controller triggers a high-temperature alarm on a storage rack; the monitoring system escalates the alert to the maintenance team while suppressing non-critical vibration warnings.

Practical application: Integrating alarm management with a centralized dashboard enables operators to view real-time alerts, acknowledge them, and assign corrective actions, improving mean-time-to-respond (MTTR).

Challenges: Over-alerting can desensitize staff; setting appropriate thresholds requires historical data analysis and periodic review.

### Automated Guided Vehicle (AGV) Metrics

Related terms: fleet utilization, travel time, energy consumption

Definition: Quantitative indicators that assess the performance of AGVs used for material transport within a warehouse, including uptime, load factor, and route efficiency.

Example: A fleet of AGVs reports a 92% utilization rate, meaning each vehicle spends 92% of operational time carrying loads versus idle.

Practical application: Monitoring AGV metrics helps planners adjust dispatch algorithms, balance workloads, and schedule charging cycles to avoid bottlenecks.

Challenges: Data granularity may be limited by legacy controllers; integrating disparate AGV brands into a single analytics platform can be complex.

### Batch Picking Efficiency

Related terms: order consolidation, pick density, travel distance

Definition: A measure of how effectively the system groups multiple orders into a single pick cycle, reducing travel and handling time.

Example: In a zone with high SKU variety, batch picking reduces average pick travel from 120m to 78m per order.

Practical application: By analyzing pick-list data, managers can determine optimal batch sizes that maximize picker productivity without causing excessive queuing.

Challenges: Larger batches may increase order hold time, affecting delivery promises; balancing batch size against service level agreements (SLAs) requires careful simulation.

### Capacity Utilization Rate

Related terms: storage density, space allocation, throughput

Definition: The proportion of available storage volume or rack slots that are actively used, expressed as a

percentage.

Example: An AS/RS with a 1,200 m<sup>3</sup> capacity stores 960 m<sup>3</sup> of inventory, achieving an 80% utilization rate.

Practical application: Tracking capacity utilization supports decisions on adding mezzanine levels, re-configuring aisle widths, or implementing dynamic slotting algorithms.

Challenges: Over-utilization can impede equipment movement and increase wear; under-utilization wastes capital and may raise per-unit handling costs.

### Cycle Time

Related terms: process latency, order fulfillment time, throughput time

Definition: The elapsed time from the initiation of a request (e.g., a retrieval command) to the completion of the associated action (e.g., item delivery to the pick station).

Example: A retrieval cycle averages 45 seconds, while a storage cycle averages 38 seconds in a high-rise AS/RS.

Practical application: Reducing cycle time directly improves order turnaround, enabling higher order volumes without additional equipment.

Challenges: Cycle time variability may stem from fluctuating load weights, equipment aging, or network latency; isolating the dominant factor often requires multivariate analysis.

### Data Acquisition System (DAS)

Related terms: sensor network, PLC, edge gateway

Definition: The hardware and software infrastructure that collects raw signals from equipment (temperature, motor current, position) and forwards them to analytics platforms.

Example: A DAS aggregates vibration data from 150 conveyor motors and streams it to a cloud-based analytics engine for anomaly detection.

Practical application: DAS enables real-time condition monitoring, feeding predictive-maintenance models with high-frequency data.

Challenges: Ensuring time-synchronization across heterogeneous sensors and handling bandwidth constraints in legacy facilities can be problematic.

### Data Lake

Related terms: raw data repository, schema-on-read, big data

Definition: A centralized storage architecture that holds large volumes of structured and unstructured data in its native format, allowing flexible analytics.

Example: All AS/RS event logs, video feeds, and IoT sensor streams are ingested into a Hadoop-based data lake for later querying.

Practical application: Data scientists can explore historical performance trends, train machine-learning models, and extract insights without predefined schemas.

Challenges: Without proper governance, data lakes can become “data swamps” where low-quality or duplicate data hampers analysis.

### Data Mining

Related terms: pattern discovery, clustering, association rules

Definition: The process of extracting useful patterns, correlations, or anomalies from large datasets using

statistical and machine-learning techniques.

Example: Mining retrieval logs reveals that items stored on the top tier of a particular rack experience 15% longer pick times due to elevator latency.

Practical application: Insights from data mining guide layout optimization, slotting strategies, and preventive maintenance schedules.

Challenges: High dimensionality and noisy sensor data can produce spurious patterns; domain expertise is essential to validate findings.

#### Data Visualization

Related terms: dashboards, heat maps, KPIs

Definition: The graphical representation of performance metrics and trends, enabling stakeholders to quickly interpret complex data.

Example: A heat map shows aisle congestion levels during peak hours, highlighting a bottleneck near the loading dock.

Practical application: Interactive dashboards allow supervisors to drill down from overall OEE to individual equipment MTBF, fostering data-driven decision making.

Challenges: Over-crowded visualizations can obscure key insights; selecting appropriate chart types and ensuring real-time refresh rates require careful design.

#### Downtime Analysis

Related terms: unplanned outage, availability, root-cause analysis

Definition: The systematic examination of periods when equipment is not operational, distinguishing between scheduled maintenance and unexpected failures.

Example: An AS/RS crane experiences 3 hours of unplanned downtime per month, primarily due to motor bearing wear.

Practical application: By categorizing downtime events, managers can prioritize corrective actions, negotiate better service contracts, and improve overall equipment effectiveness.

Challenges: Capturing accurate start/stop timestamps in legacy systems can lead to under-reporting; distinguishing between true downtime and idle periods for strategic reasons may be ambiguous.

#### Key Performance Indicator (KPI)

Related terms: metric, benchmark, performance target

Definition: A quantifiable measure used to evaluate the success of an organization, department, or process against defined objectives.

Example: An AS/RS KPI of 95% on-time order fulfillment reflects the proportion of orders shipped within the promised window.

Practical application: KPI dashboards provide real-time visibility of critical factors such as OEE, cycle time, and inventory accuracy, guiding corrective actions.

Challenges: Selecting irrelevant or redundant KPIs can dilute focus; continuous alignment with evolving business goals is necessary.

#### Mean Time Between Failures (MTBF)

Related terms: reliability, failure rate, maintenance interval

**Definition:** The average elapsed time between consecutive equipment failures, calculated by dividing total operational time by the number of failures.

**Example:** A shuttle robot with an MTBF of 2,400 hours indicates high reliability, allowing longer maintenance cycles.

**Practical application:** MTBF informs spare-parts inventory levels, warranty negotiations, and predictive-maintenance model calibration.

**Challenges:** Accurate MTBF calculation requires comprehensive failure logging; intermittent faults that do not trigger alarms may be omitted, skewing results.

#### Mean Time To Repair (MTTR)

**Related terms:** repair duration, downtime, service efficiency

**Definition:** The average time required to restore equipment to operational status after a failure, including diagnosis, part replacement, and testing.

**Example:** An MTTR of 45 minutes for a conveyor belt motor indicates efficient maintenance procedures.

**Practical application:** Reducing MTTR improves overall equipment availability and supports tighter production schedules.

**Challenges:** Variability in technician skill levels and parts availability can cause MTTR fluctuations; capturing precise start and end times is essential for reliable measurement.

#### Overall Equipment Effectiveness (OEE)

**Related terms:** availability, performance, quality

**Definition:** A composite metric that multiplies availability, performance efficiency, and quality rate to assess how effectively equipment is used.

**Example:** An AS/RS with 90% availability, 95% performance, and 99% quality yields an OEE of 84.5%.

**Practical application:** OEE serves as a benchmark for continuous improvement initiatives, highlighting areas where losses occur (e.g., speed loss vs. quality defects).

**Challenges:** Calculating OEE for multi-modal systems (shuttle, crane, conveyor) requires consistent data definitions; misaligned data sources can produce misleading scores.

#### Predictive Maintenance

**Related terms:** machine learning, condition monitoring, failure prediction

**Definition:** A proactive approach that uses sensor data, statistical models, and algorithms to forecast equipment failures before they occur.

**Example:** A predictive model flags a rise in motor vibration amplitude as a precursor to bearing failure, prompting a scheduled replacement during off-peak hours.

**Practical application:** Implementing predictive maintenance reduces unplanned downtime, extends component life, and optimizes maintenance workforce allocation.

**Challenges:** Model accuracy depends on data quality and volume; false positives can lead to unnecessary interventions, while false negatives may miss critical failures.

#### Real-Time Monitoring

**Related terms:** stream processing, live dashboard, event handling

**Definition:** Continuous observation of system parameters as they occur, enabling immediate detection of

deviations and rapid response.

Example: A live dashboard shows current load factor of each storage lane, updating every second to reflect new insertions.

Practical application: Real-time alerts allow operators to intervene before a minor issue escalates into a full-scale outage, supporting high-availability service levels.

Challenges: High-frequency data streams demand robust network bandwidth and scalable processing infrastructure; latency must be minimized to retain relevance.

### Slotting Optimization

Related terms: dynamic storage, pick frequency, travel distance

Definition: The strategic assignment of SKUs to storage locations based on demand patterns, dimensions, and handling characteristics to minimize retrieval effort.

Example: Fast-moving items are placed on the nearest levels of the AS/RS, while slow-moving bulk goods occupy deeper, lower-density slots.

Practical application: Regular slotting analysis, powered by analytics on order history, reduces average travel distance and improves throughput.

Challenges: Constantly changing demand can render static slotting plans suboptimal; implementing dynamic slotting requires real-time system integration and may increase repositioning workload.

### Statistical Process Control (SPC)

Related terms: control charts, process variation, Six Sigma

Definition: A methodology that uses statistical techniques to monitor and control a process, ensuring it operates within predefined limits.

Example: Control charts for conveyor speed show that most measurements stay within  $\pm 2\%$  of the target, indicating stable operation.

Practical application: SPC helps detect early signs of drift in equipment performance, prompting corrective actions before quality or productivity suffers.

Challenges: Setting appropriate control limits requires historical data; insufficient sample sizes can lead to over-reactive adjustments.

### Throughput

Related terms: units per hour, capacity, processing rate

Definition: The volume of material that passes through the AS/RS system within a given time frame, typically expressed as items or pallets per hour.

Example: An AS/RS achieves a throughput of 1,200 pallets per hour during peak shift.

Practical application: Monitoring throughput helps determine whether the system can meet seasonal demand spikes and guides capacity-expansion decisions.

Challenges: Throughput is influenced by multiple variables—cycle time, equipment availability, and order mix—making root-cause attribution complex.

### Utilization Ratio

Related terms: equipment usage, idle time, productivity

Definition: The proportion of time that a piece of equipment is actively engaged in work versus being idle or

waiting.

Example: A shuttle crane reports a utilization ratio of 68%, indicating that 32% of its scheduled time is idle.

Practical application: Low utilization may signal suboptimal scheduling, excess capacity, or bottlenecks elsewhere in the workflow.

Challenges: High utilization can increase wear and reduce MTBF; balancing utilization with maintenance windows is essential.

#### Warehouse Management System (WMS) Integration

Related terms: ERP, API, data synchronization

Definition: The process of linking the AS/RS control software with the broader WMS to exchange inventory, order, and location data seamlessly.

Example: When a sales order is entered in the ERP, the WMS forwards a retrieval request to the AS/RS, which then updates the inventory count upon completion.

Practical application: Tight integration eliminates manual data entry, reduces errors, and enables end-to-end order visibility.

Challenges: Differing data models and communication protocols can cause mismatches; robust error-handling and regular reconciliation are required.

#### Workload Balancing

Related terms: load distribution, resource allocation, queue management

Definition: The technique of evenly distributing tasks among multiple AS/RS resources (cranes, shuttles, conveyors) to avoid over-loading any single asset.

Example: An algorithm reassigns retrieval jobs from a heavily loaded crane to a less-utilized shuttle during peak periods.

Practical application: Balanced workloads improve overall system throughput and extend equipment lifespan by preventing localized stress.

Challenges: Real-time balancing demands fast decision-making algorithms and accurate status feeds; latency in data can lead to suboptimal assignments.

#### Yield Loss

Related terms: defect rate, quality loss, scrap

Definition: The proportion of inventory that becomes unusable or requires rework due to handling errors, mis-placement, or equipment malfunction.

Example: A yield loss of 0.8% was recorded when a storage robot mis-aligned pallets, causing damage to the top layer.

Practical application: Tracking yield loss helps identify problematic processes, such as improper loading procedures, and drives corrective training.

Challenges: Small percentages can be difficult to detect without high-resolution data; attributing loss to a specific cause may require detailed incident logs.

#### Zero-Defect Initiative

Related terms: continuous improvement, Six Sigma, quality management

Definition: A strategic program aimed at eliminating errors in storage and retrieval operations, striving for

perfect accuracy.

Example: Implementing barcode verification at each pick point reduced mis-picks from 0.3% to 0.02%, approaching zero-defect goals.

Practical application: The initiative leverages performance monitoring data to pinpoint error hotspots and deploys targeted interventions such as operator training and automated checks.

Challenges: Achieving absolute zero defects is often unrealistic; maintaining momentum requires sustained leadership support and measurable incentives.

### Advanced Analytics

Related terms: machine learning, predictive modeling, prescriptive optimization

Definition: The application of sophisticated statistical and algorithmic techniques to extract deeper insights, forecast future states, and recommend optimal actions.

Example: An advanced analytics model predicts the optimal time to batch store incoming pallets to minimize future retrieval travel distance.

Practical application: Prescriptive analytics can automatically adjust slotting rules, schedule maintenance, and re-route AGVs based on predicted demand patterns.

Challenges: Model interpretability, data privacy, and integration with existing control systems can hinder adoption; ongoing model validation is essential.

### Automated Storage and Retrieval System (AS/RS) KPI Dashboard

Related terms: visual analytics, real-time reporting, performance scorecard

Definition: A consolidated visual interface that presents key metrics—such as OEE, cycle time, utilization, and alarm frequency—in a single view for rapid assessment.

Example: The dashboard displays a traffic-light indicator for each crane: green for acceptable MTBF, yellow for approaching maintenance, red for critical failure risk.

Practical application: Managers can spot trends, drill into specific equipment, and trigger corrective workflows directly from the dashboard.

Challenges: Data latency, inconsistent metric definitions across equipment vendors, and information overload can reduce the dashboard's effectiveness.

### Batch Consolidation

Related terms: order grouping, pick sequencing, workload reduction

Definition: The process of merging multiple customer orders into a single retrieval operation to reduce travel and handling steps.

Example: Consolidating ten small orders into one batch cut the average pick travel distance by 30%.

Practical application: Batch consolidation improves picker efficiency and reduces energy consumption of moving equipment.

Challenges: Larger batches may increase order lead time; careful planning is required to balance efficiency gains against SLA commitments.

### Cold-Start Problem

Related terms: model training, data scarcity, initialization

Definition: The difficulty faced when predictive models lack sufficient historical data to generate reliable

forecasts, common in newly commissioned AS/RS installations.

Example: A newly deployed predictive-maintenance algorithm cannot accurately predict bearing wear until six months of vibration data are collected.

Practical application: Bootstrapping techniques, such as transfer learning from similar facilities, can mitigate the cold-start issue.

Challenges: Inaccurate early predictions may erode stakeholder confidence; ongoing model refinement is required as data accumulates.

### Data Governance

Related terms: metadata management, data quality, access control

Definition: The set of policies, procedures, and standards that ensure data is accurate, consistent, secure, and usable across the organization.

Example: A governance policy mandates that all sensor timestamps be synchronized to UTC and that any data anomalies be logged for review.

Practical application: Strong data governance underpins reliable analytics, regulatory compliance, and effective decision making.

Challenges: Implementing governance in a fragmented IoT environment involves aligning multiple stakeholders and legacy systems.

### Dynamic Slotting

Related terms: real-time allocation, inventory turnover, algorithmic placement

Definition: The continuous reassignment of inventory locations based on real-time demand signals, rather than static, pre-defined slots.

Example: As a seasonal product's demand spikes, the system automatically moves its pallets to the most accessible lanes.

Practical application: Dynamic slotting maximizes picking efficiency during demand fluctuations and reduces unnecessary repositioning.

Challenges: Frequent moves increase handling costs; balancing the benefits of accessibility against the labor cost of relocation requires cost-benefit analysis.

### Event Correlation

Related terms: root-cause analysis, incident management, log aggregation

Definition: The technique of linking multiple system events (alarms, sensor readings, operator actions) to identify common underlying causes.

Example: A surge in temperature alarms coincides with a drop in motor current, revealing a cooling-system failure.

Practical application: Correlating events accelerates troubleshooting, reduces mean-time-to-repair, and helps prevent cascading failures.

Challenges: High-volume log streams can overwhelm correlation engines; filtering noise without losing critical signals is essential.

### Forecast Accuracy

Related terms: demand planning, prediction error, bias

**Definition:** The degree to which predicted demand or system performance matches actual outcomes, typically expressed as a percentage error.

**Example:** A demand forecast for a SKU achieved 92% accuracy, with a mean absolute percentage error (MAPE) of 8%.

**Practical application:** Accurate forecasts enable optimal slotting, inventory levels, and labor planning, reducing stockouts and excess holding costs.

**Challenges:** Sudden market shifts, promotional events, and supply-chain disruptions can degrade forecast accuracy; adaptive models are needed.

### Heat Map Visualization

**Related terms:** spatial analytics, density mapping, performance hotspots

**Definition:** A graphical representation that uses color gradients to display the intensity of a metric across a physical layout.

**Example:** A heat map of the AS/RS floor shows high congestion (red) near the inbound dock during morning shifts.

**Practical application:** Managers can identify and alleviate bottlenecks by re-routing traffic or expanding capacity in highlighted zones.

**Challenges:** Generating accurate spatial data requires precise location tracking; outdated maps can mislead decision makers.

### Inventory Accuracy

**Related terms:** cycle count, stock discrepancy, reconciliation

**Definition:** The proportion of inventory records that correctly reflect the physical quantity and location of items in the warehouse.

**Example:** An inventory accuracy of 99.4% indicates that only 0.6% of items are mismatched between system and shelf.

**Practical application:** High accuracy reduces order errors, improves customer satisfaction, and minimizes costly re-stocking activities.

**Challenges:** Mis-reads from RFID, human error during manual moves, and delayed data updates can lower accuracy; continuous monitoring and corrective cycles are required.

### Key Alert Threshold

**Related terms:** limit setting, sensitivity, alarm configuration

**Definition:** Predetermined values for a metric that, when exceeded, trigger a notification to operators or maintenance personnel.

**Example:** A temperature threshold of 75°C on a motor triggers a high-priority alarm when surpassed.

**Practical application:** Properly calibrated thresholds help focus attention on truly critical events while suppressing trivial fluctuations.

**Challenges:** Static thresholds may become obsolete as equipment ages; periodic review based on trend analysis is essential.

### Latency

**Related terms:** response time, network delay, processing lag

**Definition:** The time delay between the generation of a data point (e.g., a sensor reading) and its availability for analysis or action.

**Example:** A latency of 250 ms between a conveyor speed sensor and the control system is acceptable for real-time adjustments.

**Practical application:** Low latency enables closed-loop control, allowing the system to react instantly to deviations.

**Challenges:** Bandwidth constraints, edge-to-cloud data transfer, and computational overhead can increase latency, degrading control fidelity.

### Machine Learning Model Drift

**Related terms:** concept drift, model retraining, performance degradation

**Definition:** The gradual loss of predictive accuracy as the underlying data distribution changes over time.

**Example:** A model trained on historic pick-frequency data begins under-predicting demand for a new product line, leading to suboptimal slotting.

**Practical application:** Monitoring model drift prompts timely retraining, ensuring analytics remain relevant and reliable.

**Challenges:** Detecting drift requires continuous validation metrics; frequent retraining may strain computational resources.

### Predictive Analytics Dashboard

**Related terms:** forecast visualization, scenario modeling, what-if analysis

**Definition:** An interactive interface that displays forward-looking metrics, such as projected equipment failures or anticipated throughput, based on statistical models.

**Example:** The dashboard shows a 20% increase in expected crane downtime next month, prompting pre-emptive maintenance scheduling.

**Practical application:** Decision makers can evaluate the impact of potential interventions before implementation, optimizing resource allocation.

**Challenges:** Dashboard usability depends on clear visual design; overly complex predictions without context can confuse users.

### Quality Control Loop

**Related terms:** feedback mechanism, process improvement, defect detection

**Definition:** A cyclical process where performance data is collected, analyzed, and used to adjust operational parameters to maintain or improve quality.

**Example:** After detecting a rise in mis-picks, the system automatically tightens the alignment tolerance on the storage robot and schedules a calibration check.

**Practical application:** Continuous loops enable rapid correction of deviations, sustaining high inventory accuracy and order fulfillment rates.

**Challenges:** Over-reliance on automated adjustments may mask underlying systemic issues; human oversight remains critical.

### Resource Utilization Forecast

**Related terms:** capacity planning, predictive scheduling, workforce optimization

**Definition:** An estimate of future equipment and labor usage derived from historical performance data and upcoming demand projections.

**Example:** Forecasts indicate that crane utilization will peak at 85% during the holiday season, suggesting the need for supplemental temporary staff.

**Practical application:** Accurate forecasts support proactive staffing, maintenance planning, and investment decisions.

**Challenges:** Forecasts can be disrupted by unforeseen events such as supply-chain delays or equipment failures; contingency buffers are advisable.

### Scalable Architecture

**Related terms:** modular design, cloud integration, horizontal scaling

**Definition:** A system design that can accommodate growth in data volume, processing load, and functional scope without major redesign.

**Example:** Adding new sensor clusters to the DAS is achieved by simply registering them in the existing data pipeline, preserving performance.

**Practical application:** Scalability ensures that analytics platforms remain viable as the warehouse expands or adopts new technologies.

**Challenges:** Legacy components may limit scaling; careful planning of data ingestion, storage, and processing layers is required.

### Signal-to-Noise Ratio (SNR)

**Related terms:** data quality, sensor calibration, measurement fidelity

**Definition:** A metric that compares the level of useful signal (e.g., genuine vibration data) to background noise, indicating the reliability of sensor readings.

**Example:** An SNR of 30 dB for a motor temperature sensor suggests clear data, whereas 10 dB indicates significant interference.

**Practical application:** High SNR values improve the accuracy of predictive-maintenance models and reduce false-positive alarms.

**Challenges:** Environmental factors, electromagnetic interference, and sensor aging can degrade SNR; regular calibration and shielding may be necessary.

### Statistical Forecasting

**Related terms:** time-series analysis, exponential smoothing, ARIMA

**Definition:** The use of statistical methods to predict future values of a metric based on historical patterns.

**Example:** Applying exponential smoothing to weekly throughput data yields a forecast that matches actual performance within  $\pm 5\%$ .

**Practical application:** Statistical forecasts guide inventory replenishment, staffing, and equipment procurement decisions.

**Challenges:** Assumptions of stationarity may not hold in volatile demand environments; combining statistical with machine-learning approaches can improve robustness.

### Throughput Capacity Planning

**Related terms:** load forecasting, system sizing, peak analysis

**Definition:** The process of determining the maximum volume of items the AS/RS can handle while maintaining service levels, based on current and projected demand.

**Example:** Capacity planning shows that the current system can sustain 1,500 pallets per hour, but a planned promotion will require 2,000 pallets per hour.

**Practical application:** Insights inform decisions to upgrade hardware, add parallel lanes, or implement load-balancing software.

**Challenges:** Under-estimating peak demand leads to missed orders; over-provisioning results in unnecessary capital expenditure.

#### Virtual Sensor Fusion

**Related terms:** data synthesis, model-based estimation, digital twin

**Definition:** The creation of derived metrics by combining multiple physical sensor inputs through algorithms, providing insight that no single sensor can deliver.

**Example:** Combining motor current, temperature, and vibration data yields a virtual health index for a crane, indicating early wear before a failure occurs.

**Practical application:** Virtual sensors enhance condition monitoring, enabling more precise predictive-maintenance triggers.

**Challenges:** Fusion algorithms require careful tuning; inaccurate weighting can produce misleading health scores.

#### Work-In-Progress (WIP) Monitoring

**Related terms:** queue length, process flow, lead time

**Definition:** Tracking the quantity and status of items currently being processed within the AS/RS, from receipt to storage or retrieval.

**Example:** The WIP dashboard shows 250 pallets awaiting storage, with an average wait time of 12 minutes.

**Practical application:** Real-time WIP visibility helps balance inbound and outbound operations, preventing bottlenecks and ensuring smooth flow.

**Challenges:** Inconsistent tagging or delayed status updates can cause inaccurate WIP counts; integrating all touchpoints into a single data stream is essential.