

# Operations-Driven Analytics in E-Commerce

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First Approach

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# Today's Journey

- 1 Set the opening scene and objectives
- 2 Link analytics to operations leverage
- 3 Surface review paper takeaways
- 4 Demystify the Newsvendor logic
- 5 Map relevance, gaps, and risks
- 6 Align on datasets, next moves, and discussion



Source: Unsplash

# Responsive Inventory in Motion

Mišić & Perakis (2020) document how retailers fuse POS, clickstream, and supplier telemetry to shrink the latency between demand sensing and replenishment. Their case analyses show safety-stock reductions of 15–25% once inventory targets refresh with real-time signals across stores and DCs.

## Signals Highlighted in the Review

- SKU-level velocity forecasts updated from streaming data
- Reverse logistics information feeding net-demand estimates
- Supplier reliability scores guiding multi-echelon positioning



## Control Tower Actions

- Predict route saturation by lane and time window
- Re-sequence stops when telemetry flags bottlenecks
- Sync promised slots with capacity forecasts

## Analytics Payoff

The review cites same-day pilots in North America and Asia where integrating vehicle tracking, order forecasts, and workforce data trimmed late deliveries by 12–18% while containing surge labor. These examples underscore the value of feedback loops that tighten the promise-to-fulfillment cycle.

*Mišić & Perakis (2020), Section 3.4*

# Revenue Management Playbook

Mišić & Perakis (2020) argue that revenue analytics succeeds when pricing, assortment, and inventory share a common demand backbone. They highlight e-commerce cases where Bayesian or reinforcement learning models update price ladders daily while honoring inventory guardrails.

- ① **Dynamic Pricing:** Learning-based engines respond to elasticity signals while preserving minimum inventory buffers.
- ② **Assortment Tuning:** Stochastic programs allocate digital shelf space across substitutes using margin and supply risk inputs.
- ③ **Targeted Promotions:** Uplift models prioritize segments that marketing can serve without overwhelming fulfillment capacity.

*Mišić & Perakis (2020), Sections 4.1–4.3*

## Newsvendor Intuition

*“The single-period inventory problem is a balancing act between disappointing customers today and holding excess stock tomorrow. The Newsvendor model quantifies that trade-off so planners can align service levels with risk tolerance.”*

- Underage cost ( $C_u$ ) = contribution margin + customer impact.
- Overage cost ( $C_o$ ) = holding, markdown, or salvage gap.
- Critical fractile balances the two and sets the service target.

## Critical Fractile Rule

$$q^* = F^{-1}\left(\frac{C_u}{C_u + C_o}\right)$$

- $F(q)$ : forecast distribution for the selling window.
- $C_u$ : cost of an unmet unit of demand.
- $C_o$ : cost of a leftover unit.
- Service level target =  $\frac{C_u}{C_u + C_o}$ .

## Flash Sale Example

Limited-edition gadget drop, priced at €120 with a supplier cost of €70 and liquidation value of €40.

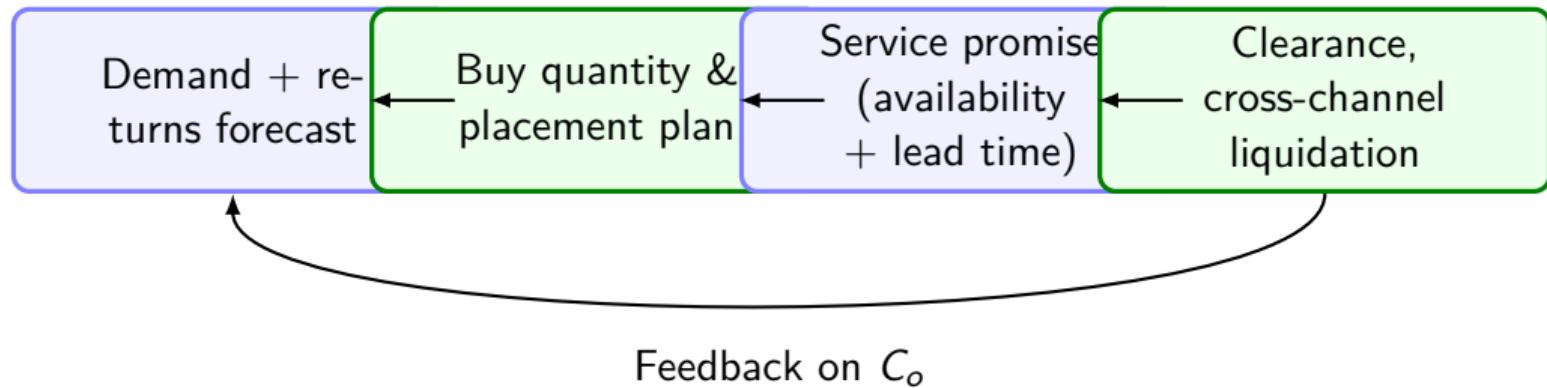
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Underage cost $C_u$ (lost margin)	€50
Overage cost $C_o$ (liquidation gap)	€30
Demand forecast	Normal(900, 180)
Critical fractile	$50/(50 + 30) = 0.625$
Optimal order $q^*$	$\approx 958$ units

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- Decision: order 958 units to balance stockout risk and overage cost.
- Interpretation: target a 62.5% service level for this promotion.

# From Model to Decisions



# Out of Scope Today

## Topics Parked for Later Deep Dives

- Feature-rich demand forecasting and machine learning pipelines
- Customer-level uplift modeling for personalization
- Automated parameter learning for  $C_u$  and  $C_o$

## Preview

These advanced methods will enrich the inputs to our classical models once we establish the baseline decisions and data flows.

## Where the Model Strains

Demand shocks Seasonality and trend breaks can render  $F(q)$  obsolete overnight.

Cost opacity Loyalty erosion and brand damage are tough to price into  $C_u$ .

Portfolio coupling Shared capacity across SKUs violates the single-period assumption.

Data hygiene Returns data and lead times require ongoing cleansing to stay credible.

## Datasets We Are Scoping

Dataset	Use Case
JD.com flash sale records	Demand and fulfillment volatility during high-pressure events
Amazon last-mile routing benchmarks	Link delivery promise to actual route constraints
Retail returns and refurbishment logs	Quantify net demand after returns and salvage value
Promotion calendar + fulfillment metrics	Calibrate $C_u/C_o$ under different campaign types

# Closing & Discussion

Operations analytics bridges insight and action.

Continued collaboration will shape our next iteration.

Thank you for your partnership.

Source: Unsplash.com

## References

- Velibor V. Mišić & Georgia Perakis, “Data Analytics in Operations Management: A Review,” *Manufacturing & Service Operations Management*.
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