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Breakthroughs in Back Orders

Matt Drake, Ph.D., CFPIM
Duquesne University, Pittsburgh, PA

Co-Authors: David Pentico and Carl Toews


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The slide features a large APICS 2013 logo at the top left. To the right of the text is a decorative graphic consisting of several overlapping, semi-transparent shapes: a blue square with white dots, a teal square with white circles, a red square with white circles, and a yellow square with white circles. The APICS logo at the bottom left includes the text 'The Association for Operations Management®'.

Presentation Overview

- EOQ with partial backordering
- Issues with non-linearly-changing backorder rate models
- Approximating the non-linearly-changing rates with constant and linear models
- Experimental results
- Conclusion

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The slide contains a presentation overview with five bullet points. At the bottom left is the APICS 2013 logo. At the bottom right is a small, stylized graphic of a gear or a cluster of shapes.

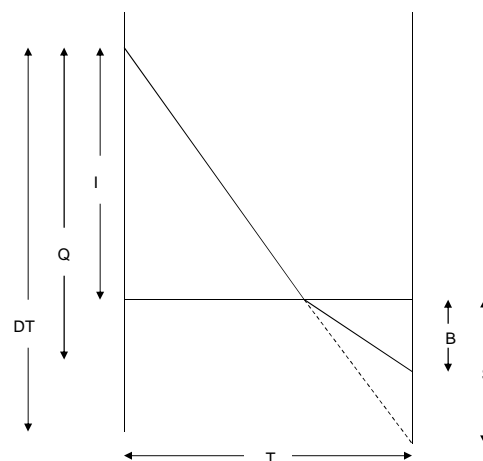
Rationale for Partial Backordering

- While some customers are willing to wait for delivery, others are not
 - Order cancellations
 - Supplier fulfills order using expensive methods of alternative supply
- Partial backordering EOQ model
 - Fraction (β) of demand that cannot be filled from stock is backordered
 - Remaining fraction ($1-\beta$) of demand is lost

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Basic EOQ with Partial Backordering



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Basic EOQ Results

- Objective: Minimize average cost per period

$$\Gamma(T,F) = \frac{C_o}{T} + \frac{C_h D T F^2}{2} + \frac{\beta C_b D T (1-F)^2}{2} + C_1 D (1-\beta)(1-F)$$

- First-order conditions yield optimal decisions

$$- T^* = \sqrt{\frac{2C_o [C_h + \beta C_b] - [(1-\beta)C_1]^2}{DC_h \beta C_b C_1}}$$

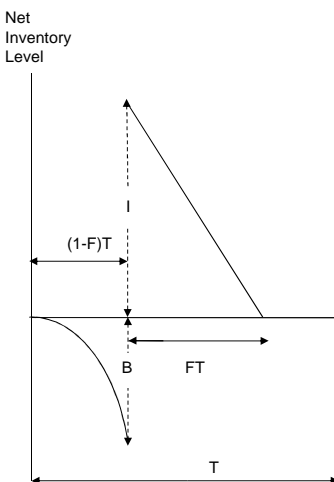
$$- F^*(T^*) = \frac{(1-\beta)C_1 + \beta C_b T^*}{T^*(C_h + \beta C_b)}$$

- Feasibility condition:

$$\beta \geq \beta^* = 1 - \frac{\sqrt{2C_o C_h D}}{C_1 D}$$

Otherwise, use either basic EOQ with no backordering or don't stock the item and lose all sales! (Choose the cheaper option.)

EOQ with Linearly-Increasing Backorder Rate



Time-Dependent Backorder Rates

- Linearly-increasing backorder rate (Toews et al. 2011)

$$- \beta(t) = \beta_0 + (1 - \beta_0) \left(\frac{t}{(1-F)T} \right) \text{ for } 0 \leq t \leq (1-F)T$$

- Exponential backorder rate (San Jose et al. 2006)

$$- B(\tau) = \rho \exp(-a\tau) \text{ for } \tau > 0$$

- Rational backorder rate (San Jose et al. 2005)

$$- B(\tau) = \rho / (1 + a\tau) \text{ for } \tau > 0$$

Issues with Non-Linearly-Changing Backorder Rates

- No closed-form solution like the constant and linearly-changing rate models have
- Solution procedures
 - Non-linear programming
 - Iterative process involving a search procedure such as Newton's Method
- Time consuming and harder to automate
- Difficult for many (perhaps *most*) managers to understand

Main Research Question

- Is it worth the hassle to use a non-linearly-changing backorder rate model to manage inventory, or will a constant or linearly changing backorder rate model perform well enough in practice?

Approximating Solutions for the Non-Linearly-Changing Backordering Rates

- Estimating the constant backorder rate or the starting point for the linear rate requires an estimate of the stockout interval for the non-linear rate model
- Three possibilities based on iterations discussed in paper:
 1. Use only the first estimate, skipping re-estimation
 2. Use alternative with lowest estimated cost
 3. Use alternative with lowest actual cost (*DIFFICULT!*)

Experimental design

- Five parameters of interest
 - Demand per period [20, 200]
 - Fixed ordering cost [0.5, 5, 50]
 - Backorder cost per period [0.5, 5.0]
 - Ratio of cost of lost sale to backorder cost per period [2, 5]
 - Backorder resistance [0.10, 0.25, 0.50, 1.00]
- $2 * 3 * 2 * 2 * 4 = 96$ problem instances

Results for Exponential Model

		Constant-Exponential		Linear-Exponential	
		Avg	Max	Avg	Max
a = .10	Alt 1	1.00391	1.0120	1.00370	1.0121
	Alt 2	1.00405	1.0120	1.00384	1.0121
	Alt 3	1.00372	1.0113	1.00354	1.0114
a = .25	Alt 1	1.01389	1.0421	1.01342	1.0429
	Alt 2	1.01522	1.0421	1.01500	1.0429
	Alt 3	1.01168	1.0347	1.01152	1.0352
a = .50	Alt 1	1.03062	1.0859	1.03011	1.0970
	Alt 2	1.04037	1.0859	1.03966	1.0970
	Alt 3	1.01838	1.0473	1.01948	1.0575
a = 1.0	Alt 1	1.06562	1.1587	1.06683	1.1938
	Alt 2	1.12622	1.2846	1.16452	1.4592
	Alt 3	1.01608	1.0505	1.02126	1.0885

Results for Rational Model

		Constant-Rational		Linear-Rational	
		Avg	Max	Avg	Max
a = .10	Alt 1	1.00329	1.0119	1.00354	1.0120
	Alt 2	1.00345	1.0119	1.00368	1.0120
	Alt 3	1.00327	1.0116	1.00339	1.0114
a = .25	Alt 1	1.01129	1.0416	1.01195	1.0423
	Alt 2	1.01291	1.0416	1.01345	1.0423
	Alt 3	1.01062	1.0361	1.01029	1.0349
a = .50	Alt 1	1.02442	1.0847	1.02730	1.0893
	Alt 2	1.03385	1.0847	1.03612	1.0893
	Alt 3	1.01831	1.0623	1.01788	1.0558
a = 1.0	Alt 1	1.05307	1.1489	1.05187	1.1672
	Alt 2	1.07626	1.2812	1.13427	1.3107
	Alt 3	1.01893	1.0467	1.01909	1.0525

Summary of the Results

- Approximation performance is best for small values of a
- Alternative 3 (selecting estimate with lowest actual cost) performs significantly better for high values of a
- Cost ratios were slightly lower for:
 - Demand of 20 compared to 200
 - Higher ordering costs (50 compared to 5 and 0.5)
 - Backorder cost of 0.5 compared to 5.0
 - Lost sales to backorder cost ratio of 2 compared to 5

Conclusions

- The functional form of the backorder rate does not make much of a difference
- Exotic functional forms often do little more than muddy the analytical waters
- Researchers examining other factors such as deterioration or inflation can reasonably assume a constant backorder rate
- Identifying simpler forms for modeling the additional considerations should be a goal because this increases the practical applicability of the model

Survey



www.tinyurl.com/lc3s3fm