

# Discrete Event Simulation of Continuous Review (Q,R) Inventory Policy



**By**

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# Roadmap

- Problem Statement
- Objective
- Explain Simulation
- Steps
- Output
- Key Takeaways & Managerial Insights

# Problem

- The objective of inventory is to achieve satisfactory levels of customer service while keeping inventory costs within reasonable bounds.
- This is called **Inventory optimization** which is critical in supply chain management
- **But the complexity of real-world multi-echelon inventory systems under uncertainties results in a challenging optimization problem, too complicated to solve by conventional mathematical models.**

# Objective

1. We propose a novel **simulation-based optimization framework** for optimizing inventory systems that operate on **multi-period inventory management** with **continuous review (Q, R) inventory policy**.
2. **The objective is to minimize the inventory cost while maintaining acceptable service levels.**

# Multi Period Inventory Management ... 1

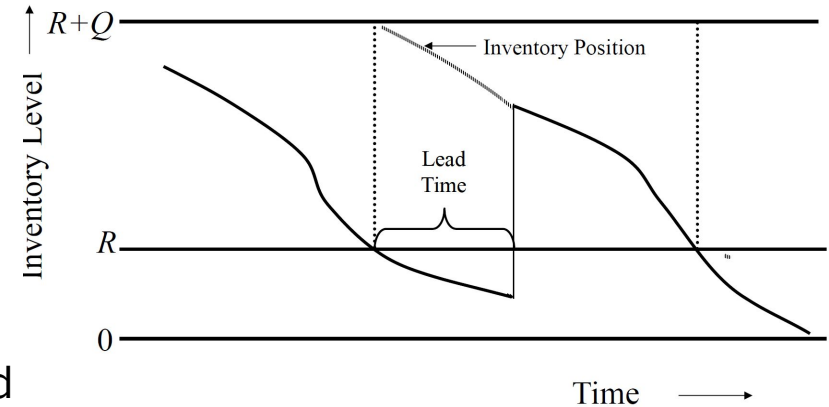
1. The inventory system operates continuously with many repeating periods or cycles.
2. The lead time ( $L$ ) for a new order is a random variable in general, reflecting the variation of delivery time.
3. The daily demand is  $D_i$  ( $i = 1, \dots, L$ ) within the lead time, where the  $D_0$  are also random variables, reflecting the variation of demand over time.
4. Inventory can be carried from one period to the next.

# Multi Period Inventory Management ... 2

1. New orders can be placed based on one of the following two basic classes of reorder policies:
  - a. **Event-driven:** These are reorder policies that are driven by reorder point (ROP) - **continuous review policy:** in which inventory is reviewed every day and a decision is made about whether and how much to order.
  - b. **Time-driven:** These are reorder policies that are driven by time - **periodic review policy:** in which inventory is reviewed at regular intervals and an appropriate quantity is ordered after each review.

# Continuous Review (Q, R) Inventory Policy

1. Whenever the inventory level reaches the reorder point  $R$ , place an order of  $Q$  to bring the inventory position to the order-up-to level  $R + Q$ .
2. Two decision to be made in the policy:
  - a. Decide the reorder level  $R$  - **when-to-order**
  - b. Decide the reorder quantity  $Q$  and hence the order-up-to level - **how much to-order**.
3. The two decision should be made based on the simulation results, to find the optimal  $Q$ , by minimizing the total inventory cost with the constraint of service level at 95%.



## 1. Assumptions

- a. Continuous Review (Q, R) inventory control policy
- b. There is an inventory holding cost (m.u./unit/day)
- c. There is fixed order cost (m.u./order)
- d. No backorders: when there is a stockout, the order is lost
- e. Random customer order demands
- f. Random order lead times

## 2. Given

- a. Unit holding cost and unit reorder cost
- b. Probability distributions for the random parameters

## 3. Determine

- a. Reorder point and optimal order quantity

## 4. Objective of the Simulation

- a. To minimize the inventory cost (= holding cost + reorder cost)
- b. To maintain service level no less than the minimum desired values



# Benefits of a simulation model

- Basic models such as the EOQ model assume that the demand for each period is constant and the lead time is not fluctuating. Further they fail to consider lost sales.
- Simulation has the capacity of drawing samples through the creation of artificial observations that behave like random events in the real system.
- Simulation can be used to investigate quickly the effects of a change in a real life situation that takes place over several years and thus can be used to study complex systems.

# Discrete Event Simulation(DES) Model

Main reasons to use DES are:

1. Possibility to include dynamics
2. Simplicity of modelling

DES model follows a set of rules about the actual operations-

**Algorithmic:** Includes set of steps for algorithm

**Mathematical:** Includes a set of equations for variable relationships

**Logical:** Includes a set of conditions that define the control of the system dynamics

# Steps(1/3)

Step 1 - Create vector values for **Q** and **R**

Step 2 - Create a numeric variable : **CurrentInventory**

- Set it to initial estimated value

Step 3 - Create a logical variable : **Ordered**

- Set the initial value to False

Step 4 - Create a numerical value: **Lead**

- Set initial value to zero

Step 5 - Create a numerical variable: **Missed**

- Set initial value to Zero

## Steps(2/3)

For given value of Q and R,

Step 6 - Calculate Lead Time using condition:

- IF **Ordered** = TRUE & Lead > 0, Reduce Lead by 1

Step 7 - Receive Delivery using the condition

- IF **Ordered** = TRUE & Lead = 0,
  - Increment **CurrentInventory** by **Q** & Set **Ordered** = FALSE

Step8 - Use **Inventory** : Randomly generate **Demand** from demand data

8.1 IF **CurrentInventory** > **Demand**

- Reduce **CurrentInventory** by the **Demand** quantity

8.2 IF **CurrentInventory** < **Demand**

- **CurrentInventory** to zero
- Augment Missed - difference in **Demand** and **CurrentInventory**

# Steps(3/3)

Step 9: **Order Inventory:**

- IF **CurrentInventory**  $\leq$  R
  - Set **Ordered** = TRUE.
  - Randomly assign a sampled value lead-time distributions to Lead
  - Compute Total Ordering cost by adding variable cost

Step 10: Compute inventory holding cost

- Add to a variable having total Inventory holding cost

Step 11: Compute Total missed opportunities & Service level

- **ServiceLevel** =  $(\text{sum}(\mathbf{D}) - \text{Miss}) / \text{sum}(\mathbf{D})$

Step 12: Repeat steps 6 through the previous step for all values of Q and R

# Advantages & Challenges of DES Model

## Major Advantages

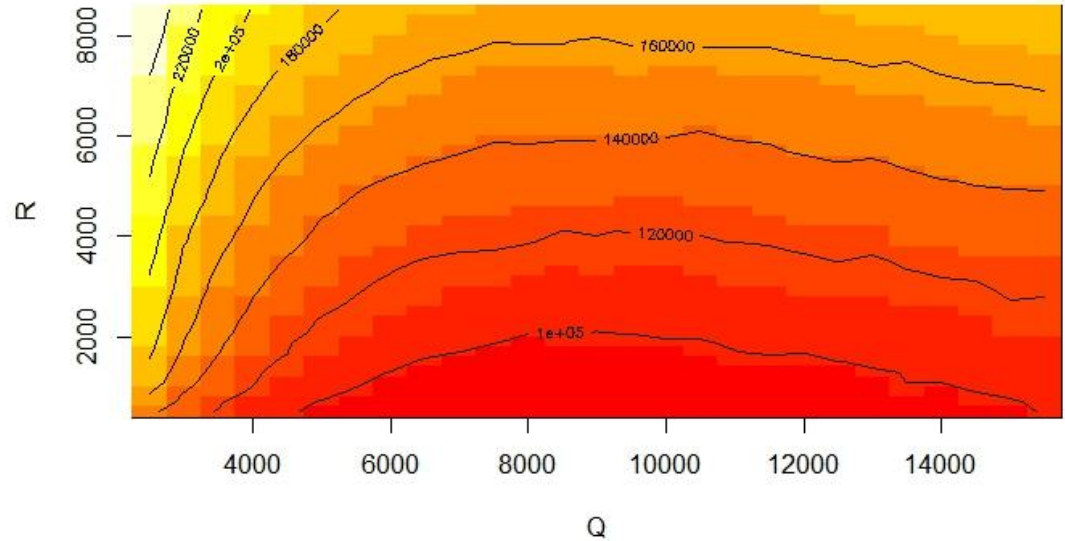
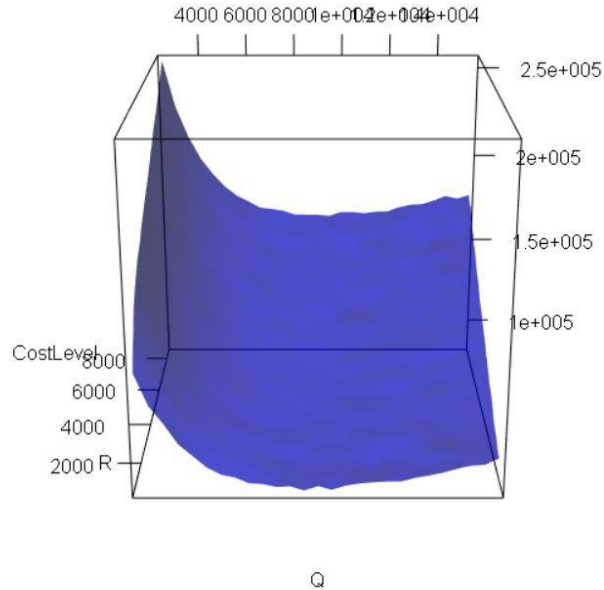
- Study new designs without interrupting real systems & understand existing
- Provides decision frameworks that are realistic and robust
- Visualize and understand system dynamics
- Uses actual system level logical or mathematical models and minimizes artificial assumptions or usage of stylized models
- Perform What-if kind of analysis for decision framework

## Challenges:

- Validation of Supply Chain could be difficult with lack of data
- Missing support for logistics process in simulation

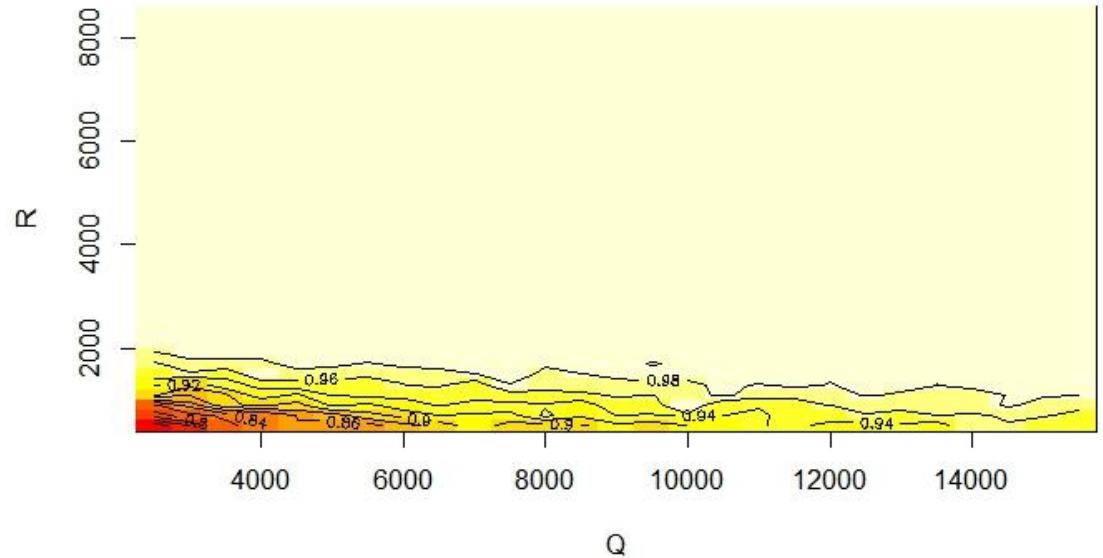
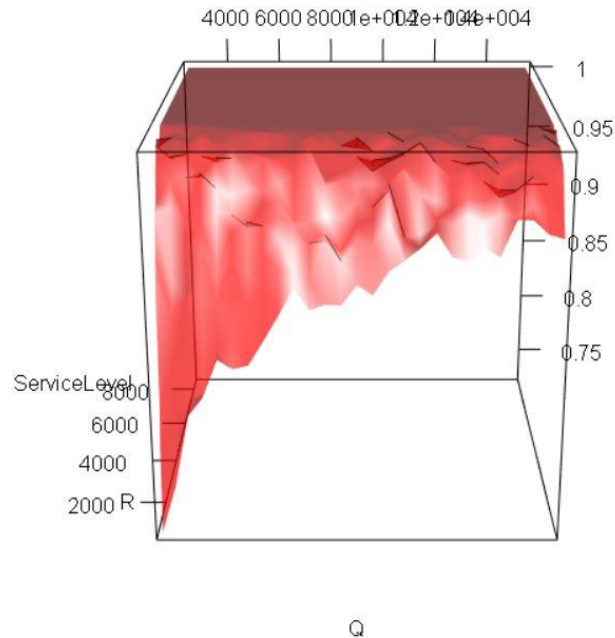
# Output

## Different Views of Cost Profiles for Different Q and R Combinations



# Output

Different Views of Service Levels for Different Q and R Combinations





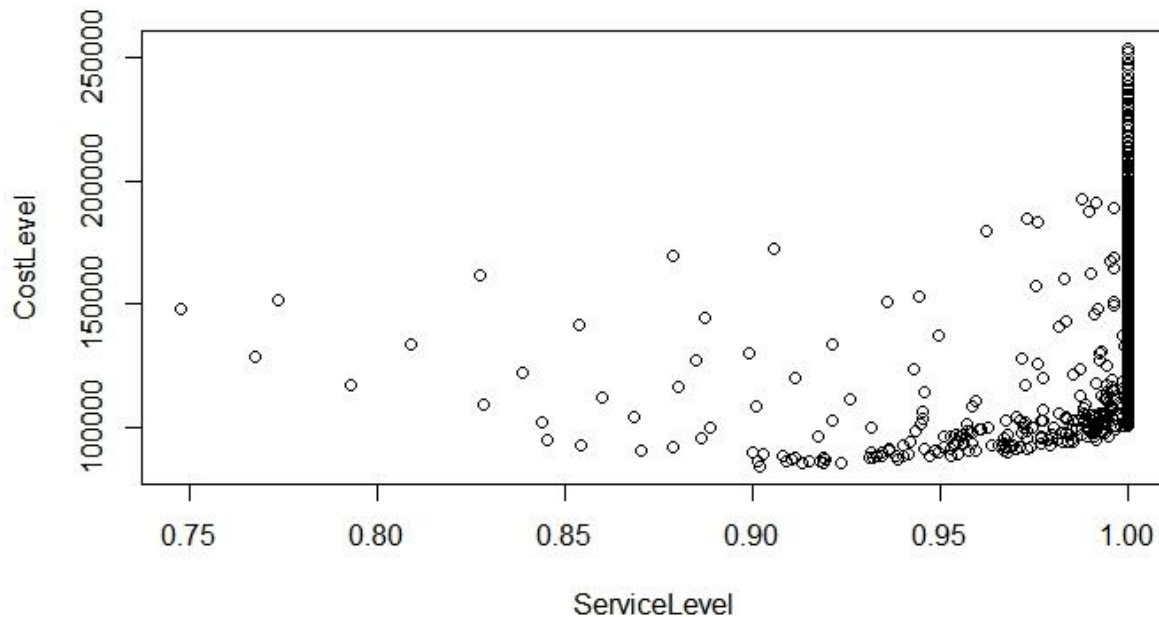
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Optimal Q & R for given service level and minimum cost level

| Service Level | Minimum CostLevel | Optimal Q | Optimal R |
|---------------|-------------------|-----------|-----------|
| 1             | \$ 98,212         | 11000     | 1500      |
| 0.99          | \$ 93,202         | 10500     | 1100      |
| 0.98          | \$ 90,224         | 10000     | 900       |
| 0.97          | \$ 90,893         | 10000     | 1100      |
| 0.96          | \$ 89,109         | 9000      | 1100      |
| 0.95          | \$ 88,441         | 9000      | 900       |

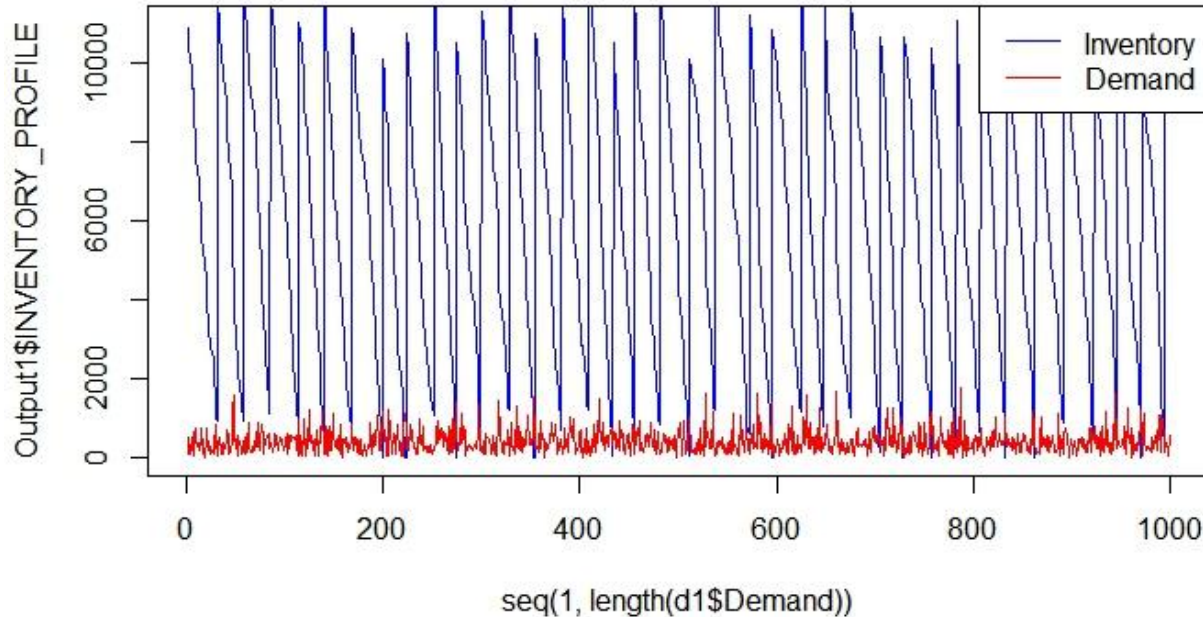
# Output

Plot of Cost and Service Level



# Output

Inventory Profile for  $Q = 11000$ ,  $R=1500$



# Key Takeaways

- Inventory Decision Questions - How Much? When? - are answered
- EOQ under this model acts as good approximation
- Inventory optimization model can be used to find optimal inventory levels on safety stock targets
- Implementing simulation in inventory optimization, retailers and supermarkets can understand shelf-life, lead time, and replenishment rules



**THANK YOU**