

ΣΧΕΔΙΑΣΜΟΣ ΚΑΙ ΠΡΟΓΡΑΜ- ΜΑΤΙΣΜΟΣ ΠΑΡΑΓΩΓΗΣ

Έλεγχος Αποθεμάτων Υπό Αβέβαιη Ζήτηση

Γιώργος Λυμπερόπουλος

Πανεπιστήμιο Θεσσαλίας

Τμήμα Μηχανολόγων Μηχανικών

PRODUCTION PLANNING AND SCHEDULING

Inventory Control Under Uncertain Demand

George Liberopoulos

University of Thessaly

Department of Mechanical Engineering

The Newsvendor model

- **Assumptions/notation**

- Single-period horizon
- Uncertain demand in the period: D (parts) assume continuous random variable
Density function and cumulative distribution function of D : $f(x)$ and $F(x)$

$$F(a) = P(D \leq a) = \int_{x=0}^a f(x)dx \quad f(a) = \left. \frac{dF(x)}{dx} \right|_{x=a}$$

- Infinite production/replenishment rate (instantaneous replenishment)
- Zero lead time
- Overage cost rate: cost per unit of positive inventory remaining at the end of the period: c_o (€per left-over part)
- Underage cost rate: cost per unit of unsatisfied demand (negative ending inventory) : c_u (€per missing part or unsatisfied demand)
- No fixed setup production/order cost

- **Decision**

- Order quantity at the beginning of the period: Q (parts)

The Newsvendor model

- **Definitions**

- (Positive) inventory remaining at the end of the period: I^+
- Unsatisfied demand (negative inventory) at the at the end of the period: I^-

$$I^+ = (Q - D)^+ \equiv \max(Q - D, 0)$$

$$I^- = (D - Q)^+ \equiv \max(D - Q, 0)$$

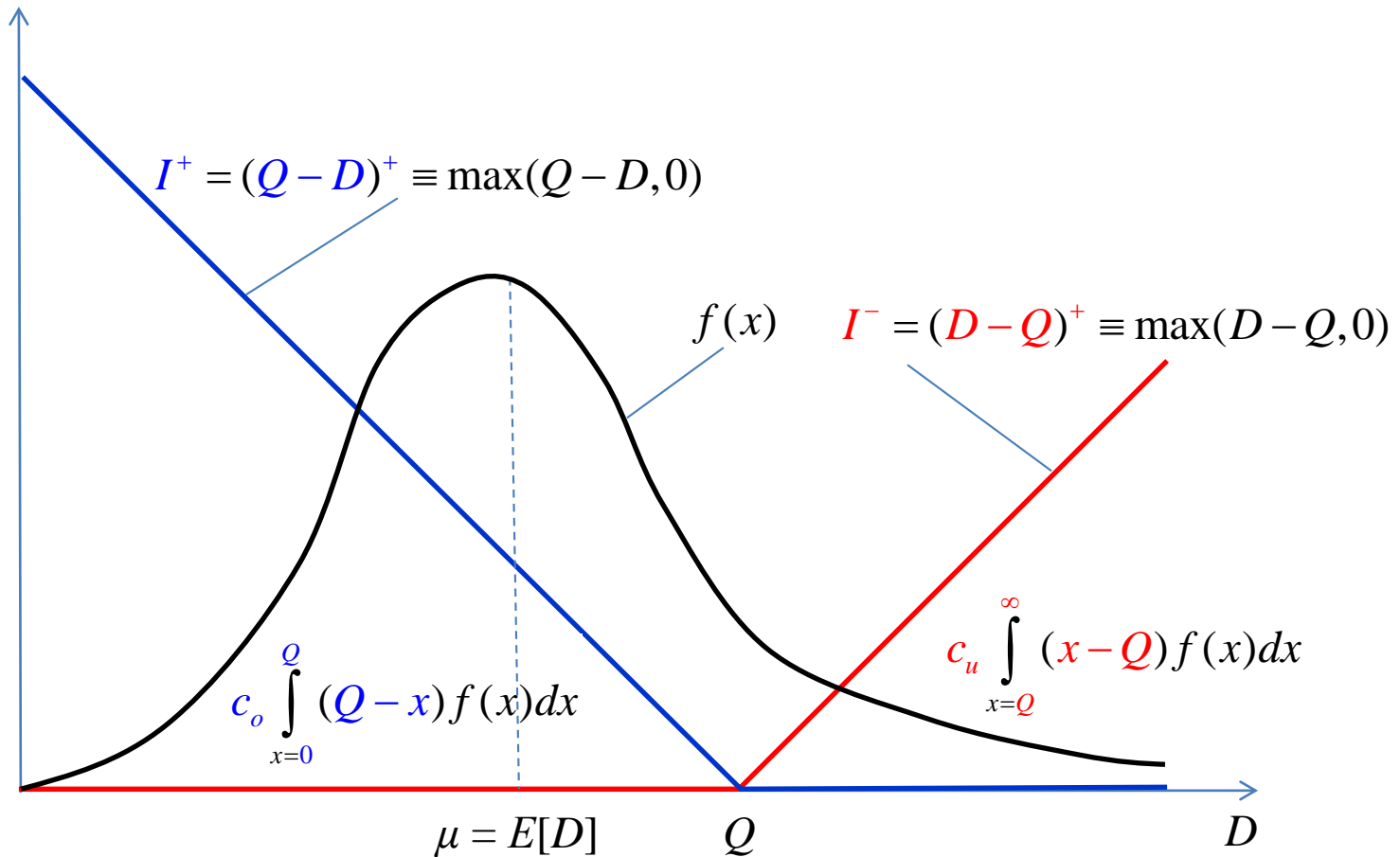
- Total overage and underage cost at the end of the period: $G(Q, D)$

$$G(Q, D) = c_o I^+ + c_u I^- = c_o (Q - D)^+ + c_u (D - Q)^+$$

- Expected cost: $G(Q)$

$$\begin{aligned} G(Q) &= E_D[G(Q, D)] = \int_{x=0}^{\infty} G(Q, x) f(x) dx \\ &= c_o \int_{x=0}^{\infty} (Q - x)^+ f(x) dx + c_u \int_{x=0}^{\infty} (x - Q)^+ f(x) dx \\ &= c_o \int_{x=0}^{\overset{Q}{Q}} (Q - x) f(x) dx + c_u \int_{x=Q}^{\infty} (x - Q) f(x) dx \end{aligned}$$

The Newsvendor model



The Newsvendor model

- **Problem**

Minimize $G(Q)$
 Q

- **First derivative of cost function**

$$\begin{aligned}
 \frac{dG(Q)}{dQ} &= \frac{d}{dQ} \left[c_o \int_{x=0}^Q (Q-x)f(x)dx + c_u \int_{x=Q}^{\infty} (x-Q)f(x)dx \right] \\
 &= c_o \int_{x=0}^Q \frac{d}{dQ} (Q-x)f(x)dx + 1(Q-x)f(x) \Big|_{x=Q} - 0(Q-x)f(x) \Big|_{x=0} + \\
 &\quad c_u \int_{x=Q}^{\infty} \frac{d}{dQ} (x-Q)f(x)dx + 0(x-Q)f(x) \Big|_{x=\infty} - 1(x-Q)f(x) \Big|_{x=Q} \\
 &= c_o \int_{x=0}^Q 1f(x)dx + c_u \int_{x=Q}^{\infty} (-1)f(x)dx \\
 &= c_o F(Q) - c_u [1 - F(Q)]
 \end{aligned}$$

The Newsvendor model

- **Second derivative**

$$\frac{dG^2(Q)}{dQ^2} = \frac{d}{dQ} [c_o F(Q) - c_u [1 - F(Q)]] = (c_o + c_u) f(Q) \geq 0$$

- **First-order condition for minimization**

$$\left. \frac{dG(Q)}{dQ} \right|_{Q=0} = 0 \Rightarrow c_o F(Q) - c_u [1 - F(Q)] = 0 \Rightarrow (c_o + c_u) F(Q) = c_u$$

$$\Rightarrow \boxed{Q^* : F(Q^*) = \frac{c_u}{c_o + c_u}} \Rightarrow \boxed{Q^* = F^{-1} \left(\frac{c_u}{c_o + c_u} \right)}$$

Note:

– $F(Q)$ is the fill rate, i.e. the probability that a demand will be satisfied!

– Recall: EOQ model with backorders: $F^* = \frac{b}{h + b}$

The Newsvendor model

- **Special case:** $D \sim \text{Normal}(\mu, \sigma)$

$$F(Q) = P(D \leq Q) = P\left(\underbrace{\frac{D - \mu}{\sigma}}_{\text{Normal}(0,1)} \leq \frac{Q - \mu}{\sigma}\right) = \underbrace{\Phi\left(\frac{Q - \mu}{\sigma}\right)}_{\text{standardized Normal cumulative distribution function}}$$

$$\Rightarrow \boxed{F(Q) = \Phi(z), \quad z = \frac{Q - \mu}{\sigma}}$$

$\Rightarrow \Phi(z)$ and hence $F(Q)$ can be evaluated from standardized Normal tables

z	$\Phi(z)$	$\Phi(z) - 0.5$
0	0.5000	0.0000
0.85	0.8023	0.3023
1.19	0.9015	0.4015
1.65	0.9505	0.4505
2.33	0.9901	0.4901
3.09	0.9990	0.4990

The Newsvendor model

- **Special case:** $D \sim \text{Normal}(\mu, \sigma)$ cont'd

$$Q^* : F(Q^*) = \frac{c_u}{c_o + c_u} \Rightarrow \Phi\left(\frac{Q^* - \mu}{\sigma}\right) = \frac{c_u}{c_o + c_u} \Rightarrow \frac{Q^* - \mu}{\sigma} = \underbrace{\Phi^{-1}\left(\frac{c_u}{c_o + c_u}\right)}_{z_{c_u/(c_o+c_u)}}$$

$$\Rightarrow \boxed{Q^* = \mu + \sigma z_{c_u/(c_o+c_u)}}$$

- **Example**

$D \sim \text{Normal}(120, 45)$

buying price $c = 30$

selling price $S = 110$

salvage price $s = 10$

$$\Rightarrow \begin{cases} c_u = S - c = 110 - 30 = 80 \\ c_o = c - s = 30 - 10 = 20 \end{cases}$$

$$\Rightarrow \frac{c_u}{c_o + c_u} = \frac{80}{20 + 80} = 0.80 \Rightarrow z_{0.80} = 0.85$$

$$\Rightarrow Q^* = \mu + \sigma z_{0.80} = 120 + 45 \cdot 0.85 = 120 + 38.25 = 158.85 \approx 159$$

The Newsvendor model

- **Extension: Discrete demand**

- Uncertain demand in the period: D (parts) assume discrete random variable

Probability mass function and cumulative distribution function of D : $p(x)$ and $F(x)$

$$F(a) = P(D \leq a) = \sum_{x \leq a} p(x) \quad p(a) = F(a) - F(a-1)$$

- Expected cost: $G(Q)$

$$G(Q) = E[G(Q, D)] = \sum_x G(Q, x) p(x) = c_o \sum_{x=0}^{Q-1} (Q-x) p(x) + c_u \sum_{x=Q}^{\infty} (x-Q) p(x)$$

- Problem: Minimize $G(Q)$

- First-order difference

$$G(Q+1) - G(Q) = c_o \sum_{x=0}^Q p(x) - c_u \sum_{x=Q+1}^{\infty} p(x) = c_o F(Q) - c_u [1 - F(Q)]$$

- First-order condition for minimization

$$Q^* : \text{smallest } Q \text{ such that } G(Q+1) - G(Q) \geq 0 \Leftrightarrow c_o F(Q) - c_u [1 - F(Q)] \geq 0$$

$$\Rightarrow \boxed{Q^* : \text{smallest } Q \text{ such that } F(Q^*) \geq \frac{c_u}{c_o + c_u}}$$

The Newsvendor model

- **Extension: Starting inventory $y > 0$**
 - Still want to be at Q^* after ordering, because Q^* is the minimizer of $G(Q)$
 - Order quantity: U
 - Optimal policy now depends on starting inventory:

$$U^*(y) = \begin{cases} Q^* - y, & \text{if } y < Q^* \\ 0, & \text{if } y \geq Q^* \end{cases}$$

Note:

- U^* \equiv optimal order quantity
- Q^* \equiv optimal “order-up-to” point \equiv inventory target level \equiv base stock level

The Newsvendor model

- **Interpretation of c_o and c_u for the single-period model**

- S = selling price (€per part)
- c = variable cost (€per part)
- h = holding cost (€per part per period)
- p = loss-of-goodwill cost (€per part short per period)

$$G(Q, D) = \underbrace{cQ}_{\text{order cost}} + \underbrace{h(Q-D)^+}_{\text{leftover inventory cost}} + \underbrace{p(D-Q)^+}_{\text{shortage cost}} - \underbrace{S \min(Q, D)}_{\substack{\text{sales} \\ \text{sales revenue}}}$$

$$G(Q) = E_D[G(Q, D)] = cQ + h \int_0^Q (Q-x)f(x)dx + p \int_Q^\infty (x-Q)f(x)dx - S \left[\underbrace{\int_0^Q xf(x)dx}_{\mu - \int_Q^\infty xf(x)dx} + \underbrace{\int_Q^\infty Qf(x)dx}_{\int_Q^\infty (x-Q)f(x)dx} \right]$$

$$= cQ + h \int_0^Q (Q-x)f(x)dx + (p+S) \int_Q^\infty (x-Q)f(x)dx - S \mu$$

$$\frac{dG(Q)}{dQ} = 0 \Rightarrow c + hF(Q) - (p+S)(1-F(Q)) = 0$$

$$\Rightarrow \boxed{Q^* : F(Q) = \frac{p+S-c}{p+S+h}} \Rightarrow \boxed{c_u = p+S-c}, \quad \boxed{c_o = h+c}$$

The Newsvendor model

- **Extension: infinite periods (infinite horizon) with backorders**

Same assumptions as single-period model except that:

- D_t = demand in period t ; D_1, D_2, D_3, \dots are i.i.d. with distribution $f(x), F(x)$
- U_t = amount ordered in period t
- Optimal policy in each period is “order-up-to” Q
- In the long-run, the inventory can never be higher than Q
⇒ In steady-state (long run): $U_t = D_{t-1}$



The Newsvendor model

- **Extension: infinite periods (infinite horizon) with backorders (cont'd)**

- Total cost in a period with demand D

$$G(Q, D) = \underbrace{(c - S)D}_{\text{order cost - sales revenue}} + \underbrace{h(Q - D)^+}_{\text{inventory holding cost}} + \underbrace{p(D - Q)^+}_{\text{backorder cost}}$$

- Expected average cost per period

$$G(Q) = E_D[G(Q, D)] = (c - S)\mu + hE[(Q - D)^+] + pE[(D - Q)^+]$$

- First-order condition for minimizing $G(Q)$

$$\frac{dG(Q)}{dQ} = 0 \Rightarrow hF(Q) - p(1 - F(Q)) = 0$$

$$\Rightarrow \boxed{Q^* : F(Q^*) = \frac{p}{p + h}} \Rightarrow \text{Newsvendor formula: } \boxed{c_u = p}, \quad \boxed{c_o = h}$$

Note:

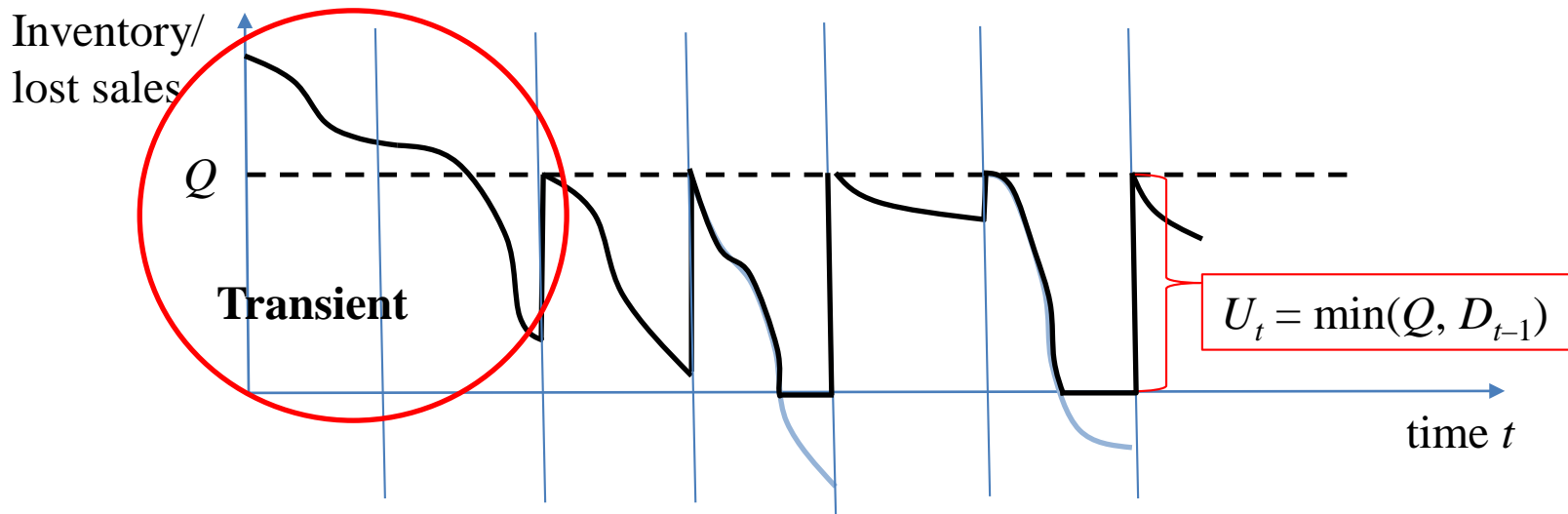
- c and S play no role in determining Q^* , because in the long run, all demands are satisfied regardless of Q ; therefore, the expected average ordering cost minus revenue per period is $(c - S)\mu$ regardless of Q .

The Newsvendor model

- **Extension: infinite periods (infinite horizon) with lost sales**

Same assumptions as infinite horizon with backorders except that:

- Unmet demand is not backordered but is lost
- In steady-state (long run): $U_t = \min(Q, D_{t-1})$



The Newsvendor model

- **Extension: infinite periods (infinite horizon) with lost sales (cont'd)**

- Total cost in a period with demand D

$$G(Q, D) = (c - S) \min(Q, D) + h(Q - D)^+ + p(D - Q)^+$$

- Expected average cost per period

$$G(Q) = E_D[G(Q, D)] = (c - S) [\mu - E[(D - Q)^+]] + hE[(Q - D)^+] + pE[(D - Q)^+]$$

- First-order condition for minimizing $G(Q)$

$$\frac{dG(Q)}{dQ} = 0 \Rightarrow hF(Q) - (p + S - c)(1 - F(Q)) = 0$$

$$\Rightarrow \boxed{Q^* : F(Q^*) = \frac{p + S - c}{p + S - c + h}} \Rightarrow \text{Newsvendor formula: } \boxed{c_u = p + S - c}, \quad \boxed{c_o = h}$$

Note:

- c and S now play a role in determining Q^* , because in the long run, the demand satisfied and the orders are $\min(Q, D)$, so they depend on Q ; therefore, the expected average ordering cost minus revenue per period is $(c - S)E[\min(Q, D)]$.

Lot size – Reorder point (Q, R) model

- **Assumptions**

- Infinite horizon
- Continuous review (as opposed to periodic review)
- D_t : random stationary demand per unit time (e.g., daily demand)
mean $\lambda \equiv E[D_t]$, variance $\sigma_t^2 = E[(D_t - \lambda)^2]$
- Unmet demand is either backordered or lost
- τ : Fixed replenishment order lead time
- Costs:
 - Variable unit production/order cost: c (€per part)
 - Fixed setup production/order cost: K (€per production run/order)
 - Inventory holding cost rate: h (€per part per unit time)
 - Stock-out (shortage/penalty) cost rate (2 cases / 4 situations: see next)

- **Order policy**

- (Q, R) policy: order Q when *inventory position* falls below R

- **Decision variables**

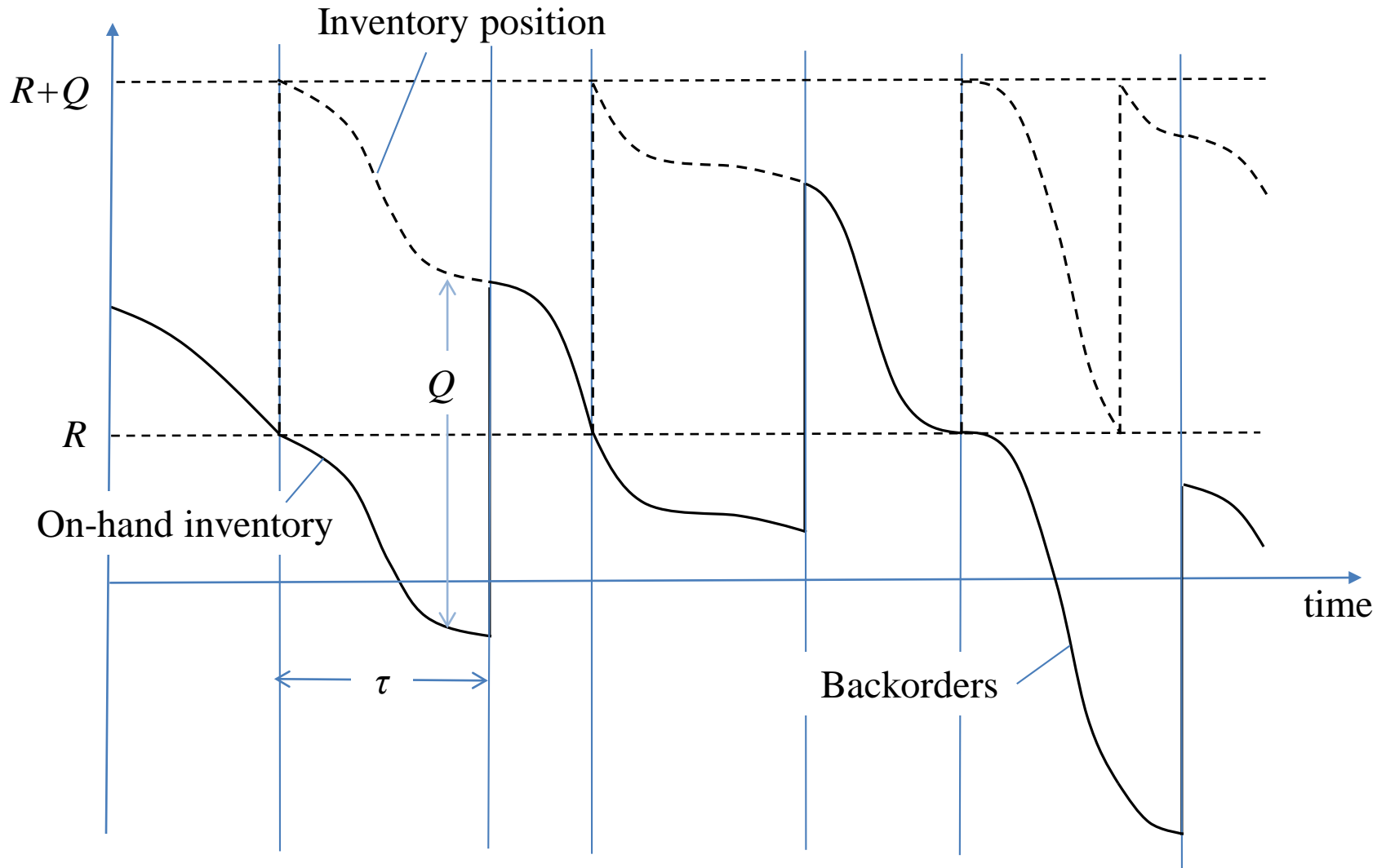
- Q : lot size (reorder quantity)
- R : reorder point

(Q, R) model

- **Assumptions on stock-outs and stock-out cost rate**
 - **Case 1: Backordered demand**
 - p_1 (€per stock-out occasion)
 - p_2 (€per part short)
 - p_3 (€per part short per unit time)
 - **Case 2: Lost sales**
 - p_L (€per lost sale)

In this course, we only deal with p_2

(Q, R) model: Backordered demand

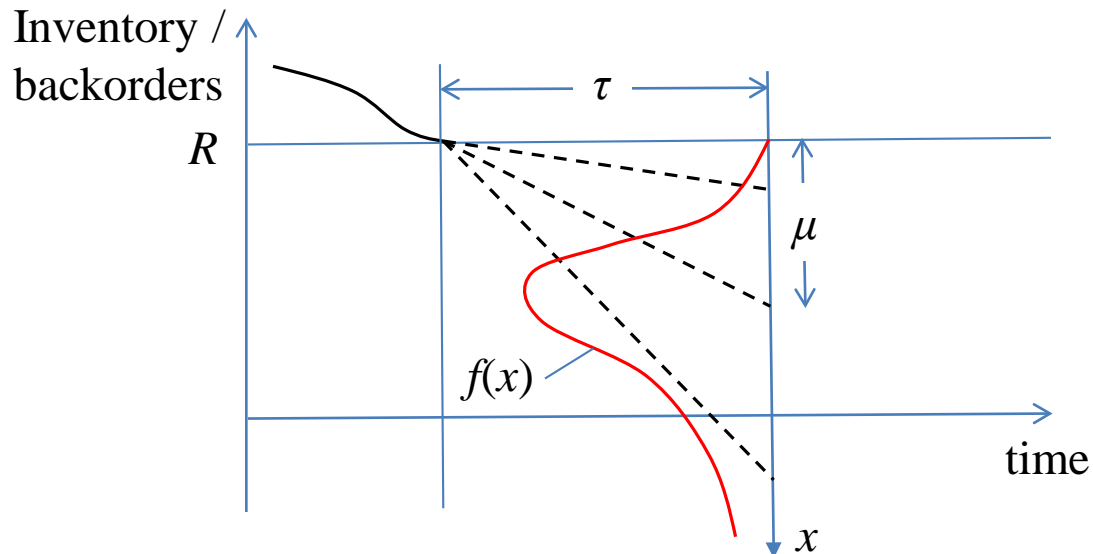


(Q, R) model: Backordered demand

- **Analysis**

- D : demand during lead time τ

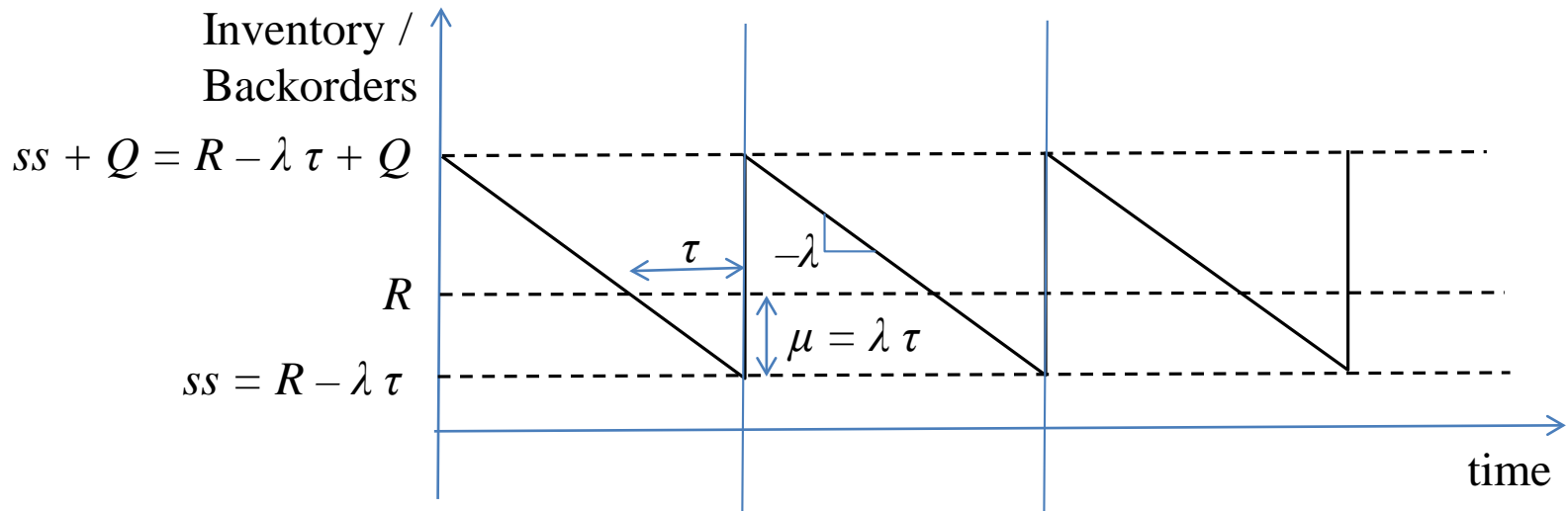
- Density function and cumulative distribution function of D : $f(x)$ and $F(x)$
 - $D = D_1 + D_2 + \dots + D_\tau$
 - Mean: $\mu \equiv E[D] = E[D_1 + D_2 + \dots + D_\tau] = \tau E[D_t] = \tau \lambda$
 - Variance: $\sigma^2 \equiv Var[D] = E[(D - \mu)^2] = Var[D_1 + D_2 + \dots + D_\tau] = \tau Var[D_t] = \tau \sigma_t^2$



(Q, R) model: Backordered demand

- **Inventory holding cost**

- Safety stock $ss \equiv R - \mu = R - \lambda \tau$
- Expected average inventory approximation $\bar{I} \approx ss + Q/2 = R - \lambda \tau + Q/2$ (underestimates true value)
- Expected average inventory hold cost = $h \bar{I} = h(R - \lambda \tau + Q/2)$



(Q, R) model: Backordered demand

- **Setup cost**
 - Expected average order frequency = λ/Q
 - Expected average setup cost per unit time = $K \lambda/Q$
- **Stock-out (penalty) cost**
 - Assumption: $\tau \ll Q/\lambda \Rightarrow$ stock-out per cycle depends only on R
 - $B(R)$: Expected stock-out cost per cycle (depends on definition of stock-out cost rate)
 - Expected average stock-out cost per unit time = $B(R) \lambda/Q$
- **Total expected average cost per unit time**

$$G(Q, R) = h \left(\frac{Q}{2} + R - \lambda \tau \right) + K \frac{\lambda}{Q} + B(R) \frac{\lambda}{Q}$$

(Q, R) model: Backordered demand

- **Optimization problem**

$$\text{Minimize}_{Q,R} G(Q, R) = h \left(\frac{Q}{2} + R - \lambda \tau \right) + K \frac{\lambda}{Q} + \frac{\lambda}{Q} B(R)$$

- **Optimality conditions**

$$\frac{\partial G(Q, R)}{\partial Q} = \frac{h}{2} - \frac{K\lambda}{Q^2} - \frac{\lambda B(R)}{Q^2} = 0 \Rightarrow Q^2 = \frac{2\lambda[K + B(R)]}{h}$$

$$\Rightarrow \boxed{Q = \sqrt{\frac{2\lambda[K + B(R)]}{h}}} \quad (1)$$

$$\frac{\partial G(Q, R)}{\partial R} = h + \frac{\lambda}{Q} \frac{dB(R)}{dR} = 0$$

$$\Rightarrow \boxed{\frac{dB(R)}{dR} = -\frac{hQ}{\lambda}} \quad (2)$$

(Q, R) model: Backordered demand

- **Optimality condition (2): Case 2:** Stock-out cost p_2 € per part short

$$B(R) = p_2 \underbrace{E[(D - R)^+]}_{n(R) \equiv \text{expected number of stock-outs per cycle}} = p_2 \underbrace{\int_{x=R}^{\infty} (x - R) f(x) dx}_{n(R)} \Rightarrow \frac{dB(R)}{dR} = -p_2[1 - F(R)]$$

$$\text{Condition (2): } -p_2[1 - F(R)] = -\frac{hQ}{\lambda} \Rightarrow \boxed{F(R) = 1 - \frac{hQ}{p_2\lambda}}$$

(Q, R) model: Backordered demand

- **Simultaneous solution of conditions (1) and (2)**

Solve by fixed-point iteration:

1. Given R , solve (1) to find Q
2. Given Q , solve (2) to find R
3. Repeat until convergence

(Q, R) model: Backordered demand

Illustration for case 2: Stock-out cost p_2 € per part short

– **Optimality conditions for case 2**

$$\boxed{Q = \sqrt{\frac{2\lambda[K + p_2 n(R)]}{h}} \quad (1)} \quad \boxed{F(R) = 1 - \frac{hQ}{p_2 \lambda} \quad (2)}$$

– **Assumption:** $D \sim \text{Normal}(\mu, \sigma)$

Use standardized cumulative distribution function (cdf) $\Phi(z)$ to compute $F(R)$

$$F(R) = P(D \leq R) = P\left(\underbrace{\frac{D - \mu}{\sigma}}_{\text{Normal}(0,1)} \leq \frac{R - \mu}{\sigma}\right) = \Phi\left(\frac{R - \mu}{\sigma}\right)$$

$$\Rightarrow \boxed{F(R) = \Phi(z), \quad z = \frac{R - \mu}{\sigma}}$$

– $\Phi(z)$ and hence $F(R)$ can be evaluated from standardized Normal cdf tables

(Q, R) model: Backordered demand

Use standardized loss function $L(z)$ to compute $n(R)$

$$Y \sim \text{Normal}(0,1) \Rightarrow L(z) \equiv E[(Y - z)^+] = \int_{y=z}^{\infty} (y - z) \underbrace{\varphi(y)}_{\substack{\text{Normal}(0,1) \\ \text{density function}}} dy$$

$$n(R) = E[(D - R)^+] = E \left[\sigma \left(\underbrace{\frac{D - \mu}{\sigma}}_{\text{Normal}(0,1)} - \frac{R - \mu}{\sigma} \right)^+ \right] = \sigma L \left(\frac{R - \mu}{\sigma} \right)$$

$$\Rightarrow \boxed{n(R) = \sigma L(z), \quad z = \frac{R - \mu}{\sigma}}$$

$L(z)$ and hence $n(R)$ can be evaluated from standardized loss function tables

It can be shown that

$$L(z) = \varphi(z) - z[1 - \Phi(z)]$$

$$\Rightarrow n(R) = \sigma L(z) = \sigma \varphi(z) + (\mu - R)[1 - \Phi(z)], \quad z = \frac{R - \mu}{\sigma}$$

(Q, R) model: Backordered demand

Under the assumption $D \sim \text{Normal}(\mu, \sigma)$, the optimality conditions become

$$Q = \sqrt{\frac{2\lambda[K + p_2\sigma L(z)]}{h}} \quad (1)$$

$$\Phi(z) = 1 - \frac{Qh}{p_2\lambda} \quad (2)$$

$$z = \frac{R - \mu}{\sigma} \quad (3)$$

(Q, R) model: Backordered demand

Fixed point iteration algorithm for case 2 under the assumption $D \sim \text{Normal}(\mu, \sigma)$

$$Q_0 = \sqrt{\frac{2\lambda K}{h}}, \quad z_0 = \Phi^{-1}\left(1 - \frac{Q_0 h}{p\lambda}\right), \quad R_0 = \mu + \sigma z_0, \quad n = 1$$

Step 1: $Q_n = \sqrt{\frac{2\lambda[K + p_2\sigma L(z_{n-1})]}{h}}$

Step 2: $z_n = \Phi^{-1}\left(1 - \frac{Q_n h}{p_2\lambda}\right)$

Step 3: $R_n = \mu + \sigma z_n$

Step 4: $|Q_n - Q_{n-1}| \geq \varepsilon$ OR $|R_n - R_{n-1}| \geq \varepsilon \Rightarrow n \leftarrow n + 1$, GOTO Step 1

(Q, R) model: Service Levels

- **Service levels in (Q, R) systems**

- **Type 1 Service** (replaces stock-out cost p_1 €/per stock-out occasion)

$S_1 \equiv$ Probability of not stocking out during the lead time

$$S_1 = P(D \leq R) = F(R)$$

- **Optimization problem**

$$\text{Minimize}_{Q,R} G(Q, R) = h \left(\frac{Q}{2} + R - \lambda \tau \right) + K \frac{\lambda}{Q}$$

subject to $F(R) \geq \alpha$ (i.e., subject to $S_1 \geq \alpha$)

- **Solution**

$$Q^* = \sqrt{\frac{2K\lambda}{h}} = \text{EOQ}$$

R^* = minimum R such that $F(R) \geq \alpha$

$$D \text{ continuous r.v.} \Rightarrow R^* = F^{-1}(\alpha)$$

(Q, R) model: Service Levels

- **Type 2 Service** (replaces stock-out cost p_2 €per part short)

$S_2 \equiv$ Proportion of demands met from stock

$$S_2 = 1 - n(R)/Q$$

- **Optimization problem**

$$\text{Minimize}_{Q,R} G(Q, R) = h \left(\frac{Q}{2} + R - \lambda \tau \right) + K \frac{\lambda}{Q}$$

subject to $1 - \frac{n(R)}{Q} \geq \beta$ (i.e., subject to $S_2 \geq \beta$)

- **Note:** Now the constraint depends on both R and Q

(Q, R) model: Service Levels

– Type 2 Service (cont'd)

Approximate solution

$$Q^* \approx \sqrt{\frac{2K\lambda}{h}} = \text{EOQ}$$

R^* = minimum R such that $n(R) \leq Q^*(1 - \beta)$

$$D \text{ continuous r.v.} \Rightarrow n(R^*) = Q^*(1 - \beta)$$

$D \sim \text{Normal}(\mu, \sigma) \Rightarrow n(R^*) \equiv \sigma L(z^*) = Q^*(1 - \beta)$

$$\Rightarrow R^* = \mu + \sigma z^*, \quad z^* = L^{-1}\left(\frac{Q^*(1 - \beta)}{\sigma}\right)$$

(Q, R) model: Service Levels

– Type 2 Service (cont'd)

More accurate solution

Consider first-order conditions (1) and (2) for case 2

$$Q = \sqrt{\frac{2\lambda[K + p_2 n(R)]}{h}} \quad (1), \quad F(R) = 1 - \frac{Qh}{p_2 \lambda} \quad (2)$$

$$(2) \Rightarrow p_2 = \frac{Qh}{[1 - F(R)]\lambda} \equiv \text{imputed stock-out cost}$$

$$(1) \Rightarrow Q = \sqrt{\frac{2\lambda\{K + Qhn(R)/[1 - F(R)]\lambda\}}{h}} \equiv \text{quadratic function in } Q$$

positive root:
$$Q = \frac{n(R)}{1 - F(R)} + \sqrt{\frac{2K\lambda}{h} + \left(\frac{n(R)}{1 - F(R)}\right)^2} \quad (3)$$

$$\boxed{n(R) = (1 - \beta)Q} \quad (4) \Rightarrow \quad L(z) = \frac{(1 - \beta)Q}{\sigma}$$

(Q, R) model: Random Lead Time

- **Extension: Random lead-time**

- L : random lead time

- Mean: $\tau \equiv E[L]$, variance $\sigma_L^2 \equiv E[(L - \tau)^2]$

- D : demand during lead time L

- Density function and cumulative distribution function of D : $f(x)$ and $F(x)$

- $D = D_1 + D_2 + \dots + D_L$, where L is a random variable

- It can be shown (see next page) that:

- Mean: $\mu \equiv E[D] = \tau \lambda$

- Variance: $\sigma^2 \equiv \text{Var}[D] = E[(D - \mu)^2] = \tau \sigma_t^2 + \lambda^2 \sigma_L^2$

Everything else holds!!

(Q, R) model: Random Lead Time

- Derivation of μ and σ^2

$$\text{Mean: } \mu \equiv E[D] = E_{L, D|L}[E[D | L]] = E_L[L\lambda] = \tau\lambda$$

$$\begin{aligned}\text{Variance: } \sigma^2 &\equiv \text{Var}[D] = E[(D - \mu)^2] = E[D^2 - 2\mu D + \mu^2] = E[D^2] - 2\mu E[D] + E[\mu^2] \\ &= E_{L, D|L}[E[D^2 | L]] - 2\mu^2 + \mu^2 = \tau\sigma_i^2 + \lambda^2\sigma_L^2 + \lambda^2\tau^2 - \mu^2 = \tau\sigma_i^2 + \lambda^2\sigma_L^2 + \lambda^2\tau^2 - \tau^2\lambda^2 \\ &= \tau\sigma_i^2 + \lambda^2\sigma_L^2\end{aligned}$$

where we used:

$$E_{D|L}[D^2 | L] = E_{D|L}[\text{Var}[D | L] + E[D | L]^2] = L\sigma_i^2 + L^2\lambda^2$$

$$E_{L, D|L}[E[D^2 | L]] = E_L[L\sigma_i^2 + L^2\lambda^2] = \tau\sigma_i^2 + \lambda^2(\text{Var}[L] + \tau^2) = \tau\sigma_i^2 + \lambda^2(\sigma_L^2 + \tau^2) = \tau\sigma_i^2 + \lambda^2\sigma_L^2 + \lambda^2\tau^2$$