

Next, let us consider stochastic models for perishable products, also called "newsvendor problem".

We consider/assume:

- A single perishable product, e.g., newspaper, food, flowers, seasonal goods such as clothing (but also, e.g., airline reservations)
- single time period
- at the end of period, product has salvage value (e.g., selling clothes out of season at a discount)
- no initial inventory
- decision variable y = # of items to stock
- the demand D is a random variable (we will need to make reasonable assumptions for its probability distribution)
- k = setup cost, irrelevant here (exactly one order is placed)
- c = unit cost of purchasing/producing
- h = holding cost per item = cost of storage - salvage value
- p = shortage cost (penalty) per item, e.g., lost revenue or lost customer goodwill

The amount sold is $\min\{D, y\} = \begin{cases} D & \text{if } D < y, \\ y & \text{if } D \geq y. \end{cases}$

The cost is $C(D, y) = c y + p \max\{0, D - y\} + h \max\{0, y - D\}$

$\underbrace{\hspace{10em}}$
order cost
 $\underbrace{\hspace{10em}}$
penalty
 $= \begin{cases} 0 & \text{if } D < y \\ p(D - y) & \text{if } D > y \end{cases}$
 $\underbrace{\hspace{10em}}$
holding cost
 $= \begin{cases} 0 & \text{if } D > y \text{ (all sold)} \\ h(y - D) & \text{if } D < y \text{ (leftovers)} \end{cases}$

Goal: minimize expected cost, given some probability distribution $P_D(d)$ for the demand.
 $\underbrace{\hspace{10em}}$
probability that demand = d

Expected cost $\mathbb{E}[C](y) = \sum_{d=0}^{\infty} C(d, y) P_D(d)$.

How do we model $P_D(d)$?

Possibility (1): brute-force using empirical data, i.e., $P_D =$ empirical probability distribution

E.g., suppose in the last 20 days, we sold the following # of products:

9 15 14 10 7 9 8 3 12 18 5 20 16 17 7 10 12 8 9 12

Then: # sold:	3	5	7	8	9	10	12	14	15	16	17	18	20
frequency: (how many days?)	1	1	2	2	3	2	3	1	1	1	1	1	1
empirical P_D $= \frac{\text{frequency}}{\text{total \#}}$	$\frac{1}{20}$	$\frac{1}{20}$	$\frac{2}{20}$	$\frac{2}{20}$	$\frac{3}{20}$	$\frac{2}{20}$	$\frac{3}{20}$	$\frac{1}{20}$	$\frac{1}{20}$	$\frac{1}{20}$	$\frac{1}{20}$	$\frac{1}{20}$	$\frac{1}{20}$

Then $\mathbb{E}[C] = C(3) \frac{1}{20} + C(5) \frac{1}{20} + C(7) \frac{2}{20} + \dots$

- Problems:
- might not have enough data (e.g., 1, 2, 4, 6 items sold never occurred above)
 - historical data not always good

Possibility (2): use a theoretic $P_D(d)$ (maybe using additionally mean or spread of historical data)

If d ranges over large number of values, it makes sense to approximate it by a continuous probability distribution $q(d)$.

$$\begin{aligned}
 \text{Then } \mathbb{E}[C](y) &= \int_0^{\infty} C(x, y) q(x) dx \\
 &= \int_0^{\infty} \left[c y + p \max\{0, x-y\} + h \max\{0, y-x\} \right] q(x) dx \\
 &= c y \underbrace{\int_0^{\infty} q(x) dx}_{=1, \text{ since } q \text{ is a probability distribution}} + p \underbrace{\int_0^{\infty} \max\{0, x-y\} q(x) dx}_{= \int_y^{\infty} (x-y) q(x) dx} + h \underbrace{\int_0^{\infty} \max\{0, y-x\} q(x) dx}_{= \int_0^y (y-x) q(x) dx} \\
 &= c y + p \int_y^{\infty} (x-y) q(x) dx + h \int_0^y (y-x) q(x) dx
 \end{aligned}$$

Goal: minimize $\mathbb{E}[C](y)$. Thus we compute:

$$\frac{d \mathbb{E}[C](y)}{dy} = c + p \underbrace{\int_y^{\infty} (-q(x)) dx - p (x-y) q(x) \Big|_{x=y}}_{=} + h \int_0^y q(x) dx + h (y-x) q(x) \Big|_{x=y}$$

Note: $\frac{d}{dy} \int_y^{\infty} f(x, y) dx = \frac{d}{dy} [F(\infty, y) - \underbrace{F(y, y)}_{= F(x_1, x_2) \Big|_{x_1=x_2=y}}]$ (F anti-derivative in first variable, i.e., $\frac{\partial}{\partial x} F(x, y) = f(x, y)$)

$$\begin{aligned}
 &= \frac{\partial}{\partial y} F(\infty, y) - \underbrace{\frac{\partial}{\partial x_1} F(x_1, y) \Big|_{x_1=y}}_{= f(y, y)} - \frac{\partial}{\partial x_2} F(y, x_2) \Big|_{x_2=y} \\
 &= \int_y^{\infty} \frac{\partial}{\partial y} f(x, y) dx - f(y, y)
 \end{aligned}$$

$$\Rightarrow \frac{d \mathbb{E}[C](y)}{dy} = c - p \int_y^{\infty} q(x) dx + h \int_0^y q(x) dx$$

Let us introduce the cumulative distribution function $\Phi(y) := \int_0^y \varphi(x) dx$.

Note that $\Phi(\infty) = \int_0^{\infty} \varphi(x) dx = 1$ (all probabilities integrate up to 1).

$\Phi(y)$ tells us the probability that the demand is satisfied if we order y items.

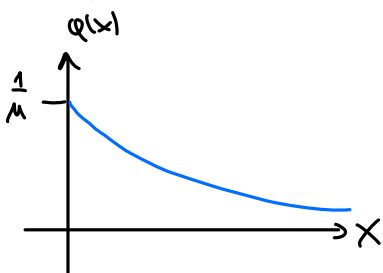
$$\begin{aligned} \text{Then } \frac{d\mathbb{E}[C](y)}{dy} &= c - p \left(\underbrace{\int_0^{\infty} \varphi(x) dx}_{=1} - \underbrace{\int_0^y \varphi(x) dx}_{=\Phi(y)} \right) + h \underbrace{\int_0^y \varphi(x) dx}_{=\Phi(y)} \\ &= c - p + (p+h)\Phi(y) \stackrel{!}{=} 0 \end{aligned}$$

$$\Rightarrow \boxed{\Phi(y^*) = \frac{p-c}{p+h}} \quad \text{i.e., we should choose } y^* \text{ s.t. this equation is satisfied.}$$

$\Phi(y^*)$ is called "optimal service level".

Solutions can be found algebraically or numerically/graphically

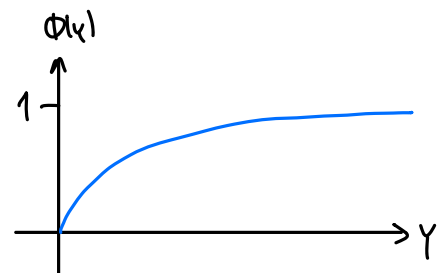
Example: Assume exponential distribution $\varphi(x) = \frac{1}{\mu} e^{-\frac{x}{\mu}}$, with $\mu > 0$ the mean value.



Note: μ is indeed the mean, since $\mathbb{E}(X) = \int_0^{\infty} x \varphi(x) dx = \int_0^{\infty} \frac{x}{\mu} e^{-\frac{x}{\mu}} dx = \mu \int_0^{\infty} \gamma e^{-\gamma} d\gamma$ (change of variables: $\frac{x}{\mu} = \gamma$)

$$\begin{aligned} \mathbb{E}(X) &= \int_0^{\infty} x \varphi(x) dx = \int_0^{\infty} \frac{x}{\mu} e^{-\frac{x}{\mu}} dx = \mu \int_0^{\infty} \gamma e^{-\gamma} d\gamma \\ &\stackrel{\text{integration by parts}}{=} \underbrace{\mu [-\gamma e^{-\gamma}]_0^{\infty}}_{=0} + \mu \int_0^{\infty} e^{-\gamma} d\gamma = -\mu e^{-\gamma} \Big|_0^{\infty} = 0 - (-\mu) = \mu \end{aligned}$$

$$\text{Then } \Phi(y) = \int_0^y \frac{1}{\mu} e^{-\frac{x}{\mu}} dx = -e^{-\frac{x}{\mu}} \Big|_0^y = -e^{-\frac{y}{\mu}} + 1$$



$$\Phi(y) = \frac{p-c}{p+h} \Leftrightarrow 1 - e^{-\frac{y}{\mu}} = \frac{p-c}{p+h} \Leftrightarrow e^{-\frac{y}{\mu}} = 1 - \frac{(p-c)}{p+h} = \frac{p+h}{p+h} - \frac{(p-c)}{p+h} = \frac{c+h}{p+h}$$

$$\Rightarrow -\frac{y}{\mu} = \overset{\substack{\text{natural logarithm} \\ \downarrow}}{\ln} \frac{c+h}{p+h} \Rightarrow y = -\mu \ln \frac{c+h}{p+h} \overset{\substack{\ln \frac{a}{b} = -\ln \frac{b}{a} \\ \downarrow}}{=} \mu \ln \frac{p+h}{c+h}$$

\Rightarrow For exponential distribution with mean μ , the optimal order quantity is $y^* = \mu \ln \frac{p+h}{c+h}$.

Numerical example: For $\mu = 10\,000$, $c = 200$, $p = 450$, $h = -90$, we find
a large salvage value can make h negative.

$y^* \approx 11\,856$ (so 1856 items more than the average should be stocked).

Note that $\Phi(y^*) = \frac{450-200}{450-90} = 0.694$ i.e., the demand is satisfied with 69.4% probability here.
optimal service level