

# INDR 450/550

Spring 2022

Lecture 20: Prescriptive analytics 2

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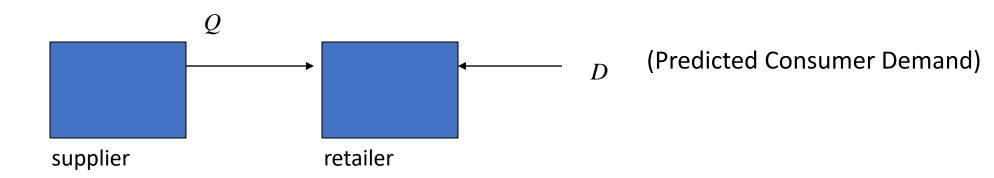
## Announcements

- Class Exercise at the end of lecture today. If you are participating online, please upload your document under Course Contents/Class Exercises
- Lab 8 material (on trees, bagging, random forests, boosting) and a short video are available
- Exam scheduled for May 7 is postponed to May 13.
  - Review exercises are available
  - Make sure that you also review the class exercises and the homeworks
- HW 3 postponed to April 29

# The Assumptions

- Short selling season
- Decision made in advance of the season
- No replenishments or capacity additions during the season (purchasing in advance is required)
- Demand forecasts have considerable uncertainty
- Items lose value significantly after the season

## The Newsvendor Problem



D: demand (random variable)

Q: quantity ordered from supplier

w : wholesale price (of supplier)

r: retail price (r>w)

s: salvage value (s < w)

m: unit manufacturing cost of supplier (m < w)

### The Newsvendor Problem

#### The profit as a function of Q:

$$\prod_{R}(Q) = r \min(Q, D) + s(Q - D)^{+} - wQ$$

Because D is a r.v., we choose to maximize:

$$E[\prod_{R}(Q)] = E[r\min(Q,D) + s(Q-D)^{+} - wQ]$$

#### The Newsvendor Problem: the result

Solving for the optimal Q:

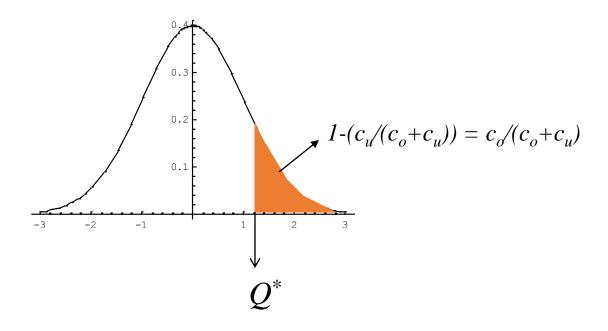
Solving for the optimal Q: 
$$F_D(Q^*) = \frac{c_u}{c_u + c_o} \qquad \Rightarrow \qquad C = F_D(Q^*) = \frac{c_u}{a^* c_o}$$

 $Q^*$  is such that the probability of satisfying all the demand  $P(D \le Q^*)$  is equal to the critical fraction :  $c_u/(c_u+c_o)$ 

fo (a)

$$area = \frac{cu}{at}$$
 $area = \frac{cu}{at}$ 
 $area = \frac{cu$ 

# The Newsvendor Problem: the optimal order quantity



A visual interpretation

$$Ex: Uniform (0,a), fo(x) = \begin{cases} \frac{1}{a} & o(x) < a \\ o & otherwise \end{cases}$$

$$F_D(x) = \int_0^x \frac{1}{a} dy = \frac{\pi}{a} \quad o(x) < a$$

$$\Rightarrow F_D(Q^*) = \frac{\omega}{\omega + \omega} \Rightarrow Q^* = \frac{C_n}{\omega + c_0} a$$

EX 2: 
$$D_{\text{NeXPO}}(\lambda)$$
 folk? =  $\begin{cases} \lambda e^{-\lambda x} \times 50 \\ 0 \end{cases}$  otherwise

$$F_{D}(x) = \int_{0}^{x} \lambda e^{-\frac{1}{2}} dy = 1 - e^{-\frac{1}{2}x}$$

$$= 1 - e^{-\frac{1}{2}} = \frac{\alpha e^{-\frac{1}{2}}}{\alpha + \epsilon_{0}}$$

$$1 - e^{-\lambda Q^{*}} = \frac{\alpha}{\alpha + \alpha}$$

$$e^{-\lambda Q^{*}} = 1 - \frac{\alpha}{\alpha + \alpha} = \frac{c_{0}}{\alpha + c_{0}}$$

$$= \lambda Q^{*} = \ln \left(\frac{c_{0}}{\alpha + \alpha}\right) \Rightarrow Q^{*} = \ln \left(\frac{c_{0}}{\alpha + \alpha}\right)$$

$$= \lambda Q^{*} = \ln \left(\frac{c_{0}}{\alpha + \alpha}\right) \Rightarrow Q^{*} = \frac{\ln \left(\frac{c_{0}}{\alpha + \alpha}\right)}{-\lambda}$$

$$= \lambda Q^* = \ln \left( \frac{\omega}{\omega + \omega} \right) \Rightarrow Q^* = \frac{\ln \left( \frac{\omega}{\omega + \omega} \right)}{-\lambda}$$

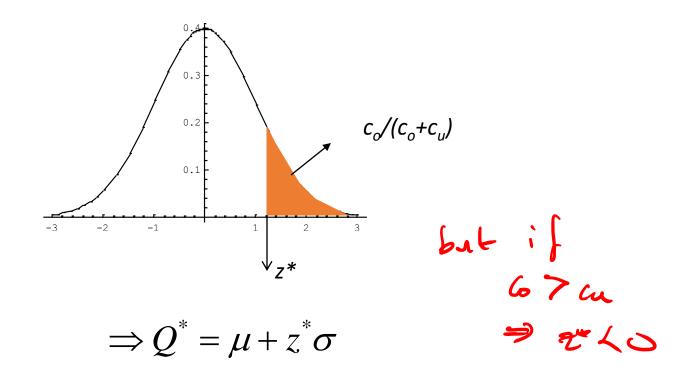
but  $f_0(x)$  may not be in aboved form or may not lead to an easy solution for  $f_0(x) = \frac{u}{at}$  $\int_{a}^{b} (x) = \frac{1}{\sqrt{2\pi U}} e^{-\frac{(v-M)^{2}}{2\sigma^{2}}}$ 

D: Normal ( $\mu$ , $\sigma$ )

$$F_D(Q^*) = \frac{c_u}{c_u + c_o}$$

Define  $Z=(D-\mu)/\sigma$ , we are looking for  $z^*$  such that:

$$F_Z(z^*) = \frac{c_u}{c_u + c_o}$$



Interpretation :  $z^*$  – safety factor (depends only on the financial parameters)

optimal order quantity = mean demand ( $\mu$ )+ safety stock ( $z^*\sigma$ )

$$\int_{2}^{2} (2) = \frac{1}{\sqrt{2\pi}} e^{-2^{2}/2} - \infty < 2 < \infty$$

$$\overline{\Phi}(z^*) = \int_{-\infty}^{2^n} \int_{2^n} (z) dz \qquad \overline{\Phi}(z^*) = 1 - \overline{\Phi}(z^*)$$

some properties

1. 
$$f_2(z) = \int_2(-z)$$

2. 
$$\overline{0}(-2) = 1 - \overline{0}(2) = \overline{0}(4)$$

Let's now calculate 
$$\mathcal{L}$$

$$\begin{bmatrix} (Z-z^*)^{\dagger} \end{bmatrix} = \int (z-z^*) \int_{\mathbb{R}^n} h(z) dz$$

$$= \int_{\mathbb{R}^n} 2 \int_{\mathbb{R}^n} (z) dz - z^* \int_{\mathbb{R}^n} f_z(z) dz$$

$$= \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} f(z) dz - \int_{\mathbb{R}^n} f(z^*)$$

$$= \int_{\mathbb{R}^n} f(z^*) - \int_{\mathbb{R}^n} f(z^*) dz$$

$$E[(2^{*}-2)] = \int_{-\infty}^{2^{*}} (2^{*}-2) f_{2}(dx)$$

$$= 2^{*}F(2^{*}) + f_{2}(2^{*})$$

Typically we take 
$$D \sim N(M, \sigma)$$
  
and  $M > 4\sigma$   
how, let's por for  $F[(D-Q^*)^{\dagger}]$   
but first  
 $P(D>d) = \int \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{(\pi-M)^2}{2\sigma^2}} dx$   
 $d = \frac{\pi-M}{\sigma} \Rightarrow dx = \sigma dx$   
 $d = \frac{\pi-M}{\sigma} \Rightarrow dx = \sigma dx$ 

Now 
$$E((0-0^{\circ})]^{\frac{1}{2}} = \int_{0^{\circ}}^{\infty} (\gamma_{2}-0^{\circ}) \frac{1}{2^{\circ}} e^{-(\kappa_{1}m)^{2}} dx$$

$$Z = \frac{\gamma_{2}-M}{6} \Rightarrow dx = TdZ$$

$$= \int_{0^{\circ}}^{\infty} (M_{1} \times \sigma - M_{1} \times \sigma) \frac{1}{2^{\circ}} e^{-2^{\circ}/2} dx$$

$$= \sigma E[(2-2^{\circ})^{\circ}]$$

$$= \sigma E[(2-2^{\circ})^{\circ}]$$

$$Similarly : E((0^{\circ}-0)^{\circ}] = \sigma, E[(2^{\circ}-2)^{\circ}]$$

And finally
$$E[C(0^*)] = Cu E[(D-Q^*)^*] + GE[(Q^*-D)^*]$$
and because  $f(Q^*) = \frac{Cu}{a_0 + C_0}$ 

$$\Rightarrow E[C(Q^*)] = (cu + G_0) Tf_2(z^*)$$

$$E[C(Q^*)] increasing in T[$$

re call that
$$Q^* = M + 2^n T$$

$$E[((l^*)]]$$
 foes not depend on  $M$   
but recall that  
 $E[T(Q^*)] = CuM - E[C(Q^*)]$ 

Let  $f_z$  be the pdf of Z (normal (0,1)):

$$Q^* = \mu + z^* \sigma,$$

The optimal cost:

$$E[C(Q^*)] = E[c_u(D - Q^*)^+ + c_o(Q^* - D)^+] = (c_u + c_o)f_Z(z^*)\sigma$$

This is a simplification that only works for a normal distribution.

The optimal cost does not depend on the average demand!

How can we improve performance (reduce costs)?

Reduce variance of demand prediction!
The MSE of the predictor plays a crucial role.

#### **Expected Profit**

The expected optimal profit is:

$$E[\Pi(Q^*)] = (c_u)\mu - E[C(Q^*)]$$

$$= (c_u)\mu - E[c_u(D - Q^*)^+ + c_o(Q^* - D)^+]$$

$$= (c_u)\mu - (c_u + c_o)f(z^*)\sigma$$

Note that the expected profit depends on the average demand. Moreover, it can be negative if:

$$(c_u)\mu < (c_u + c_o)f(z^*)\sigma$$

# Prediction based on a regression

• Assume that the prediction of demand is based on a regression model:

$$d_{t} = \beta_{0} + \beta_{1}x_{1t} + \beta_{2}x_{2t} + \dots + \beta_{p}x_{pt} + \epsilon_{t}$$

• Assume that the most recent values of the predictors are:  $x'_1, x'_2, ..., x'_p$ . If the regression errors are normally distributed with standard deviation  $\hat{\sigma}$  We then have:

$$Q^* = (\beta_0 + \beta_1 x_1' + \beta_2 x_2' + \dots + \beta_p x_p') + z^* \hat{\sigma}$$

• If we let  $\beta_{00} = \beta_0 + z^* \hat{\sigma}$ , we can note that optimal order quantity is linear in the predictors:

$$Q^* = \beta_{00} + \beta_1 x_1' + \beta_2 x_2' + \dots + \beta_p x_p'$$

# Prediction based on a regression

We can also get a closed-form expression for the optimal profit:

$$E[\Pi(Q^*)] = c_u(\beta_0 + \beta_1 x_1' + \beta_2 x_2' + \dots + \beta_p x_p') - (c_u + c_o) f_Z(z^*) \hat{\sigma}$$

- We can the see that there are some general statements to make:
  - If  $\beta_i > 0$ , then when predictor  $x'_i$  increases, expected optimal profit increases
  - ② If  $\beta_i < 0$ , then when predictor  $x_i'$  increases, expected optimal profit decreases
  - **3** Expected optimal profit always decreases in  $\hat{\sigma}$

## Discrete Demand

- When demand arrives in small discrete quantities, the continuous approximation is not reasonable
- There are many practically relevant discrete distributions
  - Poisson (discrete analogue to Normal distribution)
  - Negative binomial (known to be a good model of retail demand)
  - And of course, discrete empirical distributions (each value from a sample of n past demand observations has equal probability 1/n).

### The Newsvendor Problem: discrete demand

Now assume that D is a discrete random variable taking values restricted to integers  $Z^+ = \{0,1,2,...\}$ .

$$Q^* = \min \left\{ F_D(Q) \ge \frac{c_u}{c_u + c_0}, \ Q \in \mathbf{Z}^+ \right\}$$