



COLUMBIA
BUSINESS
SCHOOL

Decision Models

Lecture 10

- Ski Jacket Case
 - Profit calculation
 - Spreadsheet simulation
 - Analysis of results
- Summary and Preparation for next class

Yield Management

Yield management is the process of allocating different types of capacity to different customers at different prices in order to maximize revenue.

Examples include

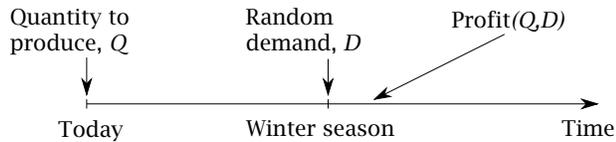
- Hotel industry
 - How many rooms to allocate for each market segment (over time)?
- Airline industry
 - How many seats to allocate for each fare class?
 - How to set fare classes and restrictions?
- Other industries
 - Cruise lines
 - Railroads
 - Car and truck rental
 - Theater and concert ticketing

In-class yield management examples:

- Retail Pricing: Quantity fixed, price over time to be decided (lecture 8)
 - Ski Jacket: Price fixed, quantity to be decided
-

Ski Jacket Production

Problem: How many ski jackets to produce given an uncertain level of demand? Because of production batch size requirements, the quantity produced must be a multiple of 2,000.



| | |
|--|-----------|
| Variable production cost per unit (C): | \$80 |
| Selling price per unit (S): | \$100 |
| Salvage value per unit (V): | \$30 |
| Fixed production cost (F): | \$100,000 |

S is the amount the manufacturer receives from the retailer. All full price jackets unsold at the end of the season can be sold for V per unit. F is the fixed cost of plant and equipment. Let Q denote the quantity of ski jackets to produce (decision variable).

Exact demand for ski jackets next winter is unknown. What can we say about the random demand D ?

Estimating Demand

Twelve managers have estimated next year's demand:

| | |
|--------|--------|
| 14,000 | 16,000 |
| 13,000 | 8,000 |
| 14,000 | 5,000 |
| 14,000 | 11,000 |
| 15,500 | 8,000 |
| 10,500 | 15,000 |

Their forecasts have mean $\mu = 12,000$ and a standard deviation $\sigma = 3,500$ (3,497 to be precise).

What distribution for demand should we use in the simulation?

Without further information about how well historical forecasts have done, we'll assume that demand is normally distributed. That is, we'll assume that

$$D \sim N(\mu = 12,000, \sigma = 3,500).$$

Profit Calculation

Suppose $Q = 12,000$. Demand is random, but suppose it turns out to be $D = 13,000$. What is the profit for the season?

$$\begin{aligned}\text{Profit} &= \text{Revenue} - \text{Variable cost} - \text{Fixed cost} \\ &= 100(12,000) - 80(12,000) - 100,000 \\ &= \$140,000\end{aligned}$$

Note: Even though $D = 13,000$, only 12,000 jackets were sold because $Q = 12,000$.

Suppose $Q = 12,000$ as before. Now suppose demand turns out to be $D = 11,000$. What is the profit for the season?

$$\begin{aligned}\text{Profit} &= \text{Rev} + \text{Salvage val} - \text{Var cost} - \text{Fixed cost} \\ &= 100(11,000) + 30(1,000) \\ &\quad - 80(12,000) - 100,000 \\ &= \$70,000\end{aligned}$$

What is a general formula?

Profit Formula

The general formula is

$$\begin{aligned}\text{Profit} &= \text{Revenue} + \text{Salvage value} \\ &\quad - \text{Var cost} - \text{Fixed cost}.\end{aligned}$$

To compute revenue, there are two cases to consider: $D < Q$ and $D \geq Q$. If $D < Q$ the revenue is $S * D$; if $D \geq Q$ the revenue is $S * Q$. This can be incorporated in a spreadsheet with the single formula

$$\text{Revenue} = S * IF(D < Q, D, Q).$$

If $D < Q$ the salvage value is $V * (Q - D)$; if $D \geq Q$ the salvage value is 0. This can be done in a spreadsheet with the formula

$$\text{Salvage value} = V * IF(D < Q, Q - D, 0).$$

The variable cost is $C * Q$ and the fixed cost is F .

The ski jacket production problem can be formulated as an optimization model:

$$\max_Q E[\text{Profit}(Q, D)],$$

i.e., find the Q among 6000, 8000, ..., 14000, etc., which maximizes expected profit.

Ski Jacket Simulation Spreadsheet

Assumption cell

| | A | B | C | D | E | F | G |
|----|-------------------|-----------|-----------------------|---------------|--------------------|---|--------|
| 1 | SKI.XLS | | Ski Jacket Simulation | | | | |
| 2 | | | | | | | |
| 3 | Variable Cost (C) | | 80 | | Mean demand | | 12,000 |
| 4 | Selling Price (S) | | 100 | | Standard deviation | | 3,500 |
| 5 | Salvage value (V) | | 30 | | Demand (D) | | 12,000 |
| 6 | Fixed Cost (F) | | 100,000 | | | | |
| 7 | | | | | | | |
| 8 | | | Salvage Value | Variable Cost | Profit | | |
| 9 | Quantity (Q) | Revenue | | | | | |
| 10 | 6,000 | 600,000 | 0 | 480,000 | 20,000 | | |
| 11 | 8,000 | 800,000 | 0 | 640,000 | 60,000 | | |
| 12 | 10,000 | 1,000,000 | 0 | 800,000 | 100,000 | | |
| 13 | 12,000 | 1,200,000 | 0 | 960,000 | 140,000 | | |
| 14 | 14,000 | 1,200,000 | 60,000 | 1,120,000 | 40,000 | | |
| 15 | | | | | | | |

Forecast cells

Cell B10: = $\$C\$4*IF(\$G\$5 < A10, \$G\$5, A10)$

Cell C10: = $\$C\$5*IF(\$G\$5 < A10, A10 - \$G\$5, 0)$

Cell D10: = $\$C\$3*A10$

Cell E10: = $B10 + C10 - D10 - \$C\6

To compare the five production quantities with a Crystal Ball simulation:

- (1) Define assumption cell (cell G5 is normal with mean =G3 and standard deviation =G4)
- (2) Define forecast cells (E10 through E14)*
- (3) Set simulation run preferences (500 trials and seed 123)
- (4) Run the simulation

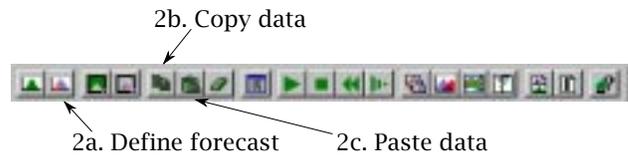
*See next slide for details of step 2.

Defining Multiple Forecasts in Crystal Ball

In step 2 (define forecasts) we want to define cells E10:E14 as forecast cells. A quick way to do this is:

- (2a) Define forecast cell E10 (move cursor to cell E10 and then click on the Crystal Ball “Define Forecast” icon)
- (2b) With the cursor on cell E10, click on the Crystal Ball “Copy Data” icon
- (2c) Highlight cells E11:E14. Then click on the Crystal Ball “Paste Data” icon

This procedure defines cells E10:E14 as forecast cells.



After setting the run preferences, click on the “Start Simulation” icon.

Ski Jacket Simulation Results — 500 Trials, Seed 123



Extract data

To compare the results of the five forecasts simultaneously, use the “Extract Data” icon from the Crystal Ball toolbar (choose “Statistics” as the Type of Data) to get summary results:

| Quantity (Q) | Average Profit | Std. Error |
|--------------|----------------|------------|
| 6,000 | 15,945 | 1,202 |
| 8,000 | 46,228 | 2,351 |
| 10,000 | 58,936 | 4,105 |
| 12,000 | 45,729 | 6,260 |
| 14,000 | 3,645 | 8,321 |

According to the simulation, the optimal quantity to produce is

$$Q^* = 10,000.$$

Confidence Interval Computation

Give a 95% confidence interval for the average profit corresponding to $Q = 10,000$.

The mean profit is 58,936 with a standard error of 4,105.

General formula for a $(1 - \alpha)$ confidence interval:

$$\bar{X} \pm z_{\alpha/2} \text{std error},$$

where the standard error is s/\sqrt{n} .

For a 95% confidence interval: $\alpha = 0.05$, and $z_{\alpha/2} = 1.96$. In general, $z_{\alpha/2}$ can be computed in a spreadsheet using $\text{NORMSINV}(1 - \alpha/2)$, e.g., $\text{NORMINV}(0.975) = 1.959961$.

For our example, the 95% confidence interval based on $n = 500$ trials is:

$$58,936 \pm 1.96(4,105),$$

i.e., the 95% confidence interval is

$$[50890, 66982].$$

To obtain a narrower confidence interval, run more simulation trials.

Results From 10,000 Simulation Trials

| Quantity (Q) | 500 trials | 10,000 trials | |
|------------------|-------------|---------------|---------|
| | Avg. Profit | Avg. Profit | Std Dev |
| 6,000 | 15,945 | 15,744 | 27,606 |
| 8,000 | 46,228 | 44,789 | 54,799 |
| 10,000 | 58,936 | 57,056 | 94,940 |
| 12,000 | 45,729 | 42,743 | 142,503 |
| 14,000 | 3,645 | -2,684 | 186,741 |

With many more simulation trials, the estimates of the mean profit change slightly, but the optimal quantity to produce still appears to be

$$Q^* = 10,000.$$

Now the 95% confidence interval for the average profit at $Q = 10,000$ is:

$$57,056 \pm 1.96(949),$$

i.e., the 95% confidence interval based on $n = 10,000$ trials is

$$[55196, 58916].$$

Is $Q = 10,000$ Better Than $Q = 12,000$?

Another way to check that the profit from $Q = 10,000$ is greater than $Q = 12,000$: define a new forecast cell in the spreadsheet which is the difference in their profits. Run the simulation and see that the average difference in profit is positive and compute a confidence interval to show that the difference is statistically significant.

We observe that $Q^* < \mu$, i.e., the optimal order quantity is less than the mean demand of 12,000. Furthermore, this result does *not* appear to be due to simulation error.

So why is

$$Q^* < \mu?$$

Intuition Behind the Result

Observation: $Q^* < \mu$.

Intuition: Compare the marginal profit of stocking 1 more ski jacket with the marginal cost. If the extra ski jacket is sold, the added profit is $S - C = \$20$. If the extra jacket remains unsold, the added profit is $V - C = -\$50$, i.e., a \$50 cost.

The *cost* of not selling one more jacket is *much greater* than the *benefit* of selling one more jacket. Hence the optimal Q will be less than μ .

Analytical Solution

- The ski jacket optimization model is:

$$\begin{aligned} \max \quad & E[\text{Profit}(Q, D)] \\ \text{s.t.} \quad & Q \geq 0 \end{aligned}$$

- A formula which finds Q^* for all costs and demand distributions:

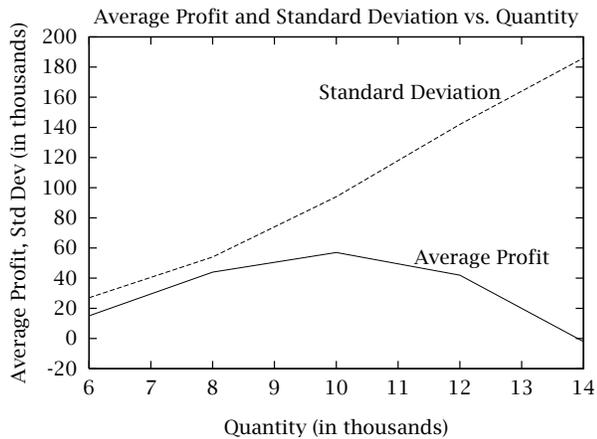
$$Q^* = F^{-1}\left(\frac{C_u}{C_u + C_o}\right) \quad (Q_{opt})$$

where $C_u = S - C$, $C_o = C - V$, and where $F^{-1}(x)$ is the inverse of $F(x) = P(D \leq x)$ (see, e.g., the W&A text, p.503).

Analytical Solution

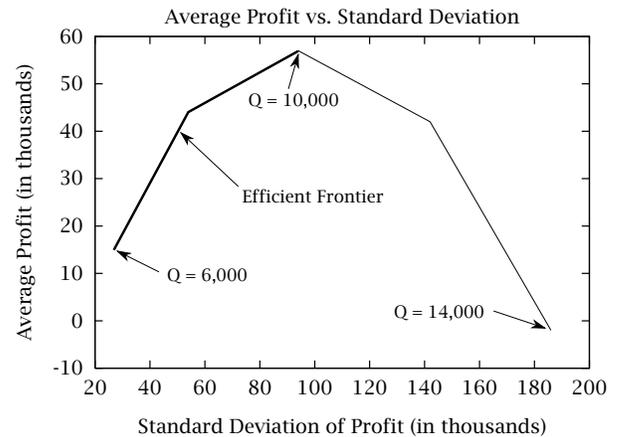
- We have $C_u = 100 - 80 = 20$, $C_o = 80 - 30 = 50$, and $C_u / (C_u + C_o) = 20 / (20 + 50) = 2/7$. Since demand is distributed $N(\mu = 12,000, \sigma = 3,500)$, $F^{-1}(2/7) = \text{NORMINV}(2/7, 12000, 3500) = 10019.18$, i.e., $P(D \leq 10019.18) = 2/7$.
 - The optimal (continuous) quantity is $Q^* = 10019.18$. Why bother simulating?
 - Optimal average profit: The formula (Q_{opt}) gives the optimal Q , but not the optimal average profit.
 - The formula (Q_{opt}) gives the optimal value if Q is allowed to be continuous, but does not handle batch size requirements as easily. (In this case, the optimal Q turned out to be very close to an allowable batch size of 10,000.)
 - Maximizing average profit does not consider the *distribution* of profits that will actually occur. For example, the table of average profit vs. quantity shows that $Q = 12,000$ and $Q = 8,000$ have the same average profit of about \$44,000. Would management be indifferent between these two production quantities?
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Average Profit, Std Dev vs. Quantity Q



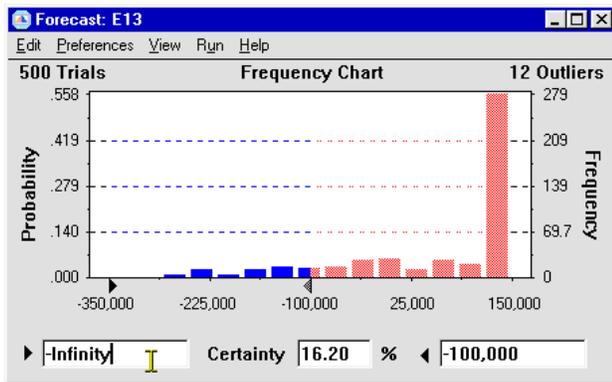
The standard deviation of profit for $Q = 8,000$ is approximately \$55,000. The standard deviation of profit for $Q = 12,000$ is approximately \$143,000. Even though both quantities have the same expected profit, the risk is nearly *3 times greater* for $Q = 12,000$ compared to $Q = 8,000$.

Efficient Frontier



Production quantities $Q \leq 10,000$ are efficient; production quantities $Q > 10,000$ are inefficient. Management might prefer $Q < 10,000$ because of the reduced risk.

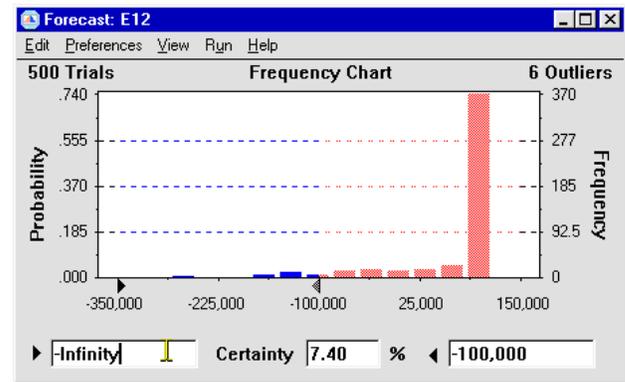
Average profit and standard deviation do not tell the whole story though. A frequency histogram shows the entire distribution of profit for a fixed Q . Since demand is normally distributed, is profit also normally distributed?

Distribution of Profit ($Q = \mu = 12,000$)

The histogram shows that the distribution of profit is highly skewed to the left. Losses greater than \$100,000 happen about 16% of the time. (In the cell E13 frequency chart, drag the right arrow to $-100,000$ and read the probability in the window titled “Certainty.”)

Even though demand is normally distributed, profit is *not* normally distributed.

With $Q = \mu = 12,000$, the most likely outcome is a profit of \$140,000. This occurs whenever $D \geq \mu$, which happens 50% of the time.

Distribution of Profit ($Q^* = 10,000$)

The histogram shows that the distribution of profit is also highly skewed to the left. Losses greater than \$100,000 happen about 7% of the time. (In the E12 frequency chart, drag the right arrow to $-100,000$ and read the probability in the window titled “Certainty.”)

Even though demand is normally distributed, profit is *not* normally distributed.

With $Q^* = 10,000$, the most likely outcome is a profit of \$100,000. This occurs whenever $D \geq 10,000$, which happens about 75% of the time.

Modifying Histograms in Crystal Ball

The histograms on the previous two slides were automatically generated when the simulation was run. However, some adjustments were made to make them more presentable. Specifically in the E12 and E13 forecast windows:

To change the number of bins: Choose “Preferences” and then choose “Chart.” Change the number of groups (bins) to 15.

To change the x-axis range: Choose “Preferences” and then choose “Display Range.” In the section “Using Fixed End-Points” change the “Min” to -350,000 and the “Max” to 150,000.

To print the histogram: Choose “Edit” and then “Copy.” Return to the Excel spreadsheet (or word processor) and select “Paste.”

Crystal Ball note: It is a good idea to have only one simulation spreadsheet open at a time when a Crystal Ball simulation is run (because Crystal Ball will simulate all open spreadsheets, and this will be slower and also means that the random number seed will not work as expected.)

Other Versions of the “Newsvendor” Model

- The main characteristics of the ski jacket case are:
 - A decision needs to be made with imperfect information
 - The item is “perishable” in that it cannot be restocked after better information about demand becomes known
 - There are costs for overstocking (or overestimating demand) and costs for understocking (or underestimating demand)

Other applications:

- Fashion retailing
- PC manufacturing
- Calendar / date book sales
- Number of H&H bagels to stock in the Uris deli each morning

For next class

- Read “Identifying, Measuring, and Hedging Currency Risk at Merck,” “Merck’s 1995 Annual Report,” and “Managing Risk” in the readings book.
 - At this point we have covered enough material on simulation for you to begin the “Yield Management at American Airlines” case (this is instead of the “Piedmont Airlines” case that was mentioned in the syllabus). Even though the case isn’t due until March 6, it is a good idea to start this case now!
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