- Taguchi Quality Control
  - what is quality?
  - the quality loss function
  - expected loss of quality
- Factor Analysis
  - design of experiments
  - orthogonal design of experiments

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## quality as defined by Taguchi

#### Some obvious points about what it is not:

- product quality ≠ product quantity
  This is a cultural issue. The U.S. is a typical throw-away society
  with a diverse market, whereas Japan is a homogeneous society.
  The outcome of world war II promoted mass production and
  consumption in the U.S., while Japan suffered.
- quality ≠ value
   Value is subjective and related to the supply-demand-marketing chain. An object (like a worn out teddy bear) can have a tremendous personal value, but no quality.

#### Definition (Genichi Taguchi, 1986)

An article of good quality performs its intended functions without variability, and causes little loss through harmful side effects, including the cost of using it.

# the key is without variability

#### Some observations on this definition:

- The key point is <u>without variability</u>.
  - Prior to the ideas of Taguchi, people tought the production was okay as long as the products were within the tolerances.
  - The reduction of variability is the goal of Taguchi quality control.
- To enhance the quality we want to reduce loss.
  But we only care about loss caused by variability in the product, not by loss caused by harmful side effects.
  - Consider for example liquor. The quality of a bottle of liquor is its percentage of alcohol, because its intended effect is intoxication. Harmful side effects are the accidents or fights, etc...
- Modeling the loss by a quality loss function is central.

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#### an example

Consider for example the production of shirts, measured by neck size.

Let *y* be the size of the produced shirt,

i.e.: its neck size; and *m* the buyer's exact neck size.

We denote by L(y) the loss due to the difference between y and m.

- When y < m (the shirt is too small), return (or discard) the item.
- When y > m (the shirt is too wide), we have to tailor.

Consider the mathematical Taylor expansion of L(y) about m:

$$L(y) = L(m) + \frac{L'(m)}{1!}(y-m) + \frac{L''(m)}{2!}(y-m)^2 + \text{higher order terms}.$$

#### a quality loss function

$$L(y) = L(m) + \frac{L'(m)}{1!}(y-m) + \frac{L''(m)}{2!}(y-m)^2 + \text{higher order terms.}$$

Why does L(y) model the loss well? Consider:

- L(m) = 0, if y equals m, the neck size, then there is no loss.
- L'(m) = 0, the loss is minimal for y = m.

So, we may represent the loss by

$$L(y) \approx k(y-m)^2$$
, for some constant k.

The constant k is determined by interpolation.

#### a numerical example

Suppose the critical deviation from *m* 

- for the shirt being too small occurs at  $\Delta m^- = 0.5$ cm,
- with corresponding cost of rejection being  $L^- = $40$ .

Furthermore, the shirt is found too wide

- when it deviates more than one cm from m, i.e.,  $\Delta m^+ = 1$  cm,
- with an associated cost for tailoring set at  $L^+ = $20$ .

Since we have different losses when too small or too wide, our quality loss function is piecewise quadratic:

$$L(y) = \begin{cases} k^+(y-m)^2 & \text{for } y \ge m, \\ k^-(y-m)^2 & \text{for } y < m. \end{cases}$$

#### a numerical quality loss function

Our quality loss function is piecewise quadratic:

$$L(y) = \begin{cases} k^+(y-m)^2 & \text{for } y \ge m, \\ k^-(y-m)^2 & \text{for } y < m. \end{cases}$$

With critical deviations and corresponding costs:

- When a shirt is too small:  $\Delta m^- = 0.5$ cm,  $L^- = $40$ .
- When a shirt is too wide:  $\Delta m^+ = 1$ cm,  $L^+ = $20$ .

We determine  $k^+$  as follows ( $\Delta m^+ = y - m, y \ge m$ ):

$$L^{+} = k^{+}(\Delta m^{+})^{2} \Rightarrow 20 = k^{+}(1.0)^{2} \Rightarrow k^{+} = 20.$$

Similarly,  $k^-$  is computed as follows ( $\Delta m^- = m - y, y < m$ ):

$$L^{-} = k^{-} (\Delta m^{-})^{2} \Rightarrow 40 = k^{-} (0.5)^{2} \Rightarrow k^{-} = 160.$$

#### the complete piecewise quality loss function

Let m = 40cm be the neck size of the buyer.

$$L(y) = \begin{cases} 20 & \text{for } y \ge 41\\ 20(y - 40)^2 & \text{for } 40 \le y < 41\\ 160(y - 40)^2 & \text{for } 39.5 < y < 41\\ 40 & \text{for } y \le 39.5 \end{cases}$$

```
function loss(y)

if y >= 41  # $20 cost for too wide

return 20

elseif y >= 40

return 20*(y-40)^2

elseif y > 39.5

return 160*(y-40)^2

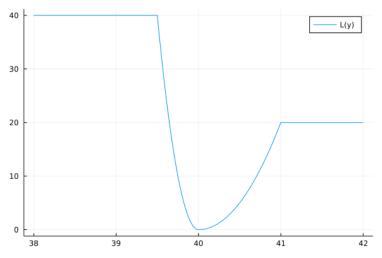
else  # $40 cost for too small

return 40
```

end

end

# the plot of the quality loss function L(y)



Missing the target causes loss.

## definition of the quality loss function

#### Definition (quality loss function)

Denote by *X* the random variable modeling the outcome of production.

Let  $\theta$  denote the target value for the production.

Then the *quality loss function* is

$$L(X, \theta) = k(X - \theta)^2$$
, for some constant  $k$ .

We call the constant *k* the *loss coefficient*.

The cumulative probability distribution function is denoted by F(x),

- with mean  $\mu = E[X]$  and
- standard deviation  $\sigma^2 = E[(X \mu)^2]$ .

Given  $\mu$  and  $\sigma$ , what is the expected loss?



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## expected loss of quality

Given a quality loss function L, the expected loss  $E[L(X, \theta)]$ 

$$= \int_{-\infty}^{+\infty} k(x-\theta)^{2} dF(x) = \int_{-\infty}^{+\infty} k(x-\mu+\mu-\theta)^{2} dF(x)$$

$$= \int_{-\infty}^{+\infty} k[(x-\mu)+(\mu-\theta)]^{2} dF(x)$$

$$= \int_{-\infty}^{+\infty} k(x-\mu)^{2} dF(x) + \int_{-\infty}^{+\infty} 2k(x-\mu)(\mu-\theta) dF(x)$$

$$+ \int_{-\infty}^{+\infty} k(\mu-\theta)^{2} dF(x)$$

$$= k \int_{-\infty}^{+\infty} (x-\mu)^{2} dF(x) + 2k(\mu-\theta) \left( \int_{-\infty}^{+\infty} x dF(x) - \mu \int_{-\infty}^{+\infty} dF(x) \right)$$

$$+ k(\mu-\theta)^{2} \int_{-\infty}^{+\infty} dF(x)$$

### the terms in the expected loss

$$k \int_{-\infty}^{+\infty} (x - \mu)^2 dF(x) + 2k(\mu - \theta) \left( \int_{-\infty}^{+\infty} x dF(x) - \mu \int_{-\infty}^{+\infty} dF(x) \right) + k(\mu - \theta)^2 \int_{-\infty}^{+\infty} dF(x) \quad \text{is simplified with}$$

$$\sigma^2 = \int_{-\infty}^{+\infty} (x - \mu)^2 dF(x), \, \mu = \int_{-\infty}^{+\infty} x dF(x), \, \text{and } \int_{-\infty}^{+\infty} dF(x) = 1.$$

$$E[L(X,\theta)] = k\sigma^2 + 2k(\mu - \theta)(\mu - \mu) + k(\mu - \theta)^2$$
  
=  $k\sigma^2 + k(\mu - \theta)^2$ 

The expected loss consists of two terms:

- the loss due to variability  $k\sigma^2$  and
- 2  $k(\mu \theta)^2$  is the loss due to missing the target.

## expected loss of quality

#### Definition (Expected Loss)

Let  $L(X) = k(X - \theta)^2$  be the quality loss function, where

- k is the loss coefficient,
- $\theta$  is the target value of the production,
- X is a random variable which models the production, with
  - $\mu = E[X]$ , the mean value of the production,
  - $\sigma^2 = E[(X \mu)^2]$ , the standard deviation.

The expected loss is  $E[L(X, \theta)] = k\sigma^2 + k(\mu - \theta)^2$ .

#### an example on the cost of variance

We consider the color density of televisions.

The loss coefficient k is \$1.25, so  $L(y) = 1.25(y - m)^2$ , where y is the produced color density, m is the target color density. The expected loss is  $1.25(\sigma^2 + (\mu - m)^2)$ .

For televisions produced in San Diego and Tokyo,  $\mu-m=0$ , but

- $\sigma^2 = 8.33$  for the San Diego plant, and
- $\sigma^2 = 2.78$  for the Tokyo plant.

The expected loss per television is then computed as

- 1.25(8.33 + 0) = \$10.41 per unit for the San Diego plant, and
- 1.25(2.78 + 0) = \$3.48 per unit for the Tokyo plant.

#### an exercise

#### Exercise 1:

Suppose an item costs \$117 to manufacture.

If the item misses the target of 12 by a margin of 3, then the item must be discarded.

Determine the quality loss function.

The production has a mean of 11.7 and standard deviation of 1.0.

Suppose an effort of \$20 per item reduces the standard deviation to 0.95. Is this a worthwhile effort?

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#### factor analysis

Which factors have the most effect on quality?

Examining a production line, we could do better

- if only we had better machines,
- if only we had better trained people,
- if only we had more production time,
- ...

Of the factors listed above, which is most important?

How do we relate the factors to quality?

## loss of quality

Suppose we have three factors  $x_1$ ,  $x_2$ ,  $x_3$ .

 $L(x_1, x_2, x_3)$  is the loss of quality as a function of the three factors.

Apply Taylor expansion again and use a linear approximation:

$$L(x_1, x_2, x_3) = \alpha_0 + \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3.$$

The factor with the largest coefficient has the largest impact.

How do we determine the coefficients  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ ?

#### experimental factorial design

To determine the coefficients in

$$L(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) = \alpha_0 + \alpha_1 \mathbf{x}_1 + \alpha_2 \mathbf{x}_2 + \alpha_3 \mathbf{x}_3$$

we can run the production many times, for

$$-1 \le x_i \le +1, \quad i=1,2,3,$$

setting each factor once low (-1) and once high (+1).

For example:

$$-L(-1, x_2 = c_2, x_3 = c_3) = \alpha_0 - \alpha_1 + \alpha_2 c_2 + \alpha_3 c_3 +L(+1, x_2 = c_2, x_3 = c_3) = \alpha_0 + \alpha_1 + \alpha_2 c_2 + \alpha_3 c_3$$

$$L(+1, x_2 = c_2, x_3 = c_3) = 2\alpha_1$$

Problem:  $2^3 = 8$  runs needed.

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#### an orthogonal array of experiments

We run four experiments and observe  $L_1$ ,  $L_2$ ,  $L_3$ ,  $L_4$ :

	<i>X</i> <sub>1</sub>	<i>X</i> <sub>2</sub>	<i>X</i> 3	
1:	+1	-1	-1	<i>L</i> <sub>1</sub>
2:	-1	+1	-1	$L_2$
3:	+1	+1	+1	$L_3$
4:	-1	-1	+1	$L_4$

#### Observe:

- Every column sums up to zero.
- The inner product of every column with every other column is zero.

The columns are orthogonal to each other.

#### apply linear algebra

Apply the array to  $L(x_1, x_2, x_3) = \alpha_0 + \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3$ .

$$\begin{array}{lll} L(+1,-1,-1) & = & \alpha_0 + \alpha_1 - \alpha_2 - \alpha_3 = L_1 \\ L(-1,+1,-1) & = & \alpha_0 - \alpha_1 + \alpha_2 - \alpha_3 = L_2 \\ L(+1,-1,+1) & = & \alpha_0 + \alpha_1 - \alpha_2 + \alpha_3 = L_3 \\ L(-1,+1,+1) & = & \alpha_0 - \alpha_1 + \alpha_2 + \alpha_3 = L_4 \end{array}$$

$$4\alpha_0 = L_1 + L_2 + L_3 + L_4$$

In matrix-vector notation:

$$\begin{bmatrix} +1 & -1 & -1 \\ -1 & +1 & -1 \\ +1 & -1 & +1 \\ -1 & +1 & +1 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \begin{bmatrix} L_1 - \alpha_0 \\ L_2 - \alpha_0 \\ L_3 - \alpha_0 \\ L_4 - \alpha_0 \end{bmatrix}.$$

#### an orthogonal matrix

In matrix-vector notation,  $A\alpha = \mathbf{b}$ :

$$\begin{bmatrix} +1 & -1 & -1 \\ -1 & +1 & -1 \\ +1 & -1 & +1 \\ -1 & +1 & +1 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \begin{bmatrix} L_1 - \alpha_0 \\ L_2 - \alpha_0 \\ L_3 - \alpha_0 \\ L_4 - \alpha_0 \end{bmatrix}.$$

By the orthogonality  $A^T A = 4I$ , where I is the identity matrix.

So we simply read off the values for  $\alpha$ .

#### a second exercise

#### Exercise 2:

Construct a 9-by-4 orthogonal array using an equal amount of -1, 0, and +1 per column.

- the sum of the element in each column is zero, and
- the columns are orthogonal to each other.

Explain your work.

## bibliography

- G. Taguchi. Introduction to Quality Engineering. Designing Quality into Products and Processes. Asian Productivity Organization, 1986.
- Handbook of Total Quality Management, edited by Christian N. Madu. Springer-Verlag, 1998.