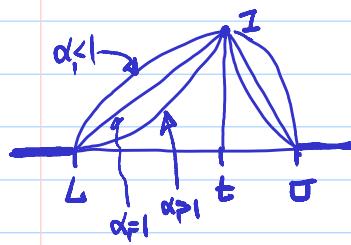


## Analysis Strategies for Multiple Responses

- Desirability function

- For nominal-the-best problem



$$d = \begin{cases} \frac{1}{1 + (\frac{y - t}{\alpha_1})^{\alpha_1}}, & L \leq y \leq t \\ \frac{1}{1 + (\frac{t - y}{\alpha_2})^{\alpha_2}}, & t \leq y \leq U \\ 0, & \text{otherwise} \end{cases}$$

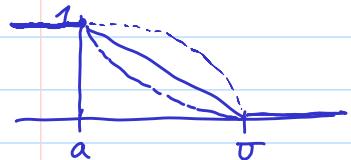
lower bound of acceptable  $\hat{y}$  value

target value

larger  $\alpha$ : importance of being close to  $t$

smaller  $\alpha$ : do not have to be very close to the target  $t$ .

- For smaller-the-better problem

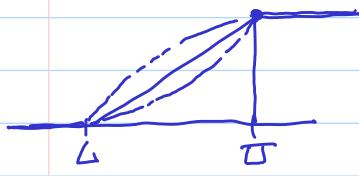


$$d = \begin{cases} 1, & a \leq y \leq U \\ 0, & y > U \end{cases}$$

upper bound

smallest possible value of  $\hat{y}$

- For larger-the-better problem



$$d = \begin{cases} 0, & L \leq y \leq U \\ 1, & y > U \\ 0, & y < L \end{cases}$$

upper bound

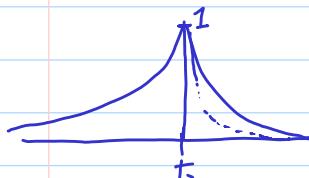
lower bound

## Analysis Strategies for Multiple Responses

- Desirability function II

- In some practical situations, the  $L$  and  $U$  values in the desirability function I cannot be properly chosen.

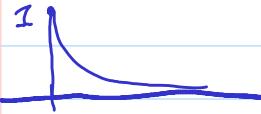
- For nominal-the-best problem:



$$d = \begin{cases} \exp\{-c(\hat{y} - t)^\alpha\}, & -\infty < \hat{y} \leq t \\ \exp\{-c(\hat{y} - t)^\alpha\}, & t \leq \hat{y} < \infty \end{cases}$$

smaller  $\alpha$ : make the desirability function drop more slowly from its peak.

- For smaller-the-better problem

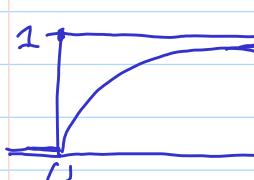


$$d = \exp\{-c(\hat{y} - a)^\alpha\}, \quad a \leq \hat{y} < \infty$$

smallest possible value.

smaller  $C$ : increase the spread of the desirability function by lowering the whole exponential curve between 0 & 1.

- For larger-the-better problem



$$d = \begin{cases} 1 - \exp\{-c(\hat{y} - L)^\alpha\} / \exp\{-c(L)^\alpha\}, & L \leq \hat{y} < \infty \\ 0, & \hat{y} \leq L \end{cases}$$

lower bound

## Analysis Strategies for Multiple Responses

- Overall desirability function  $D$

transform multiple weighted geometric mean response problem into single response problem. (why reasonable?)

$$(x) D = d_1^{w_1} d_2^{w_2} \cdots d_m^{w_m},$$

Suggestion:

better to understand the relationship between  $y_1, \dots, y_m$  before perform desirability analysis.

where the weights  $w_i$ 's satisfy  $0 < w_i < 1$  and  $w_1 + \dots + w_m = 1$ .

$\ln D = \sum_{i=1}^m w_i \ln d_i$

If any  $d_i = 0$ , then  $D = 0$

When  $w_1 = \dots = w_m = 1/m$ ,  $D$  is the geometric mean

- Weighted arithmetic mean

$$(x)$$

$$D = w_1 d_1 + w_2 d_2 + \cdots + w_m d_m$$

e.g. do PCA on  $y_1, \dots, y_m$

- Any setting for the input factors that maximizes the  $D$  value is chosen to be one which achieves an optimal balance over the  $m$  responses.
  - After finding such a setting, it should be verified whether all constraints on the responses are satisfied can do confirm expt Some constraint may not be explicitly spelled out in the beginning.

❖ Reading: textbook, 10.6