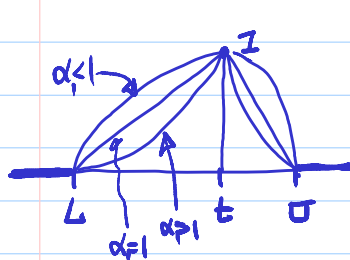


Analysis Strategies for Multiple Responses

Desirability function

- For nominal-the-best problem $\hat{f}(x)$



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

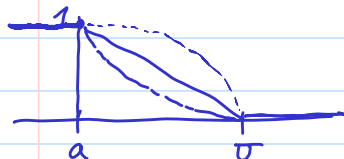
lower bound of acceptable \hat{y} value

larger α : importance of being close to t

smaller α : do not have to be very close to the target t .

upper bound

- For smaller-the-better problem

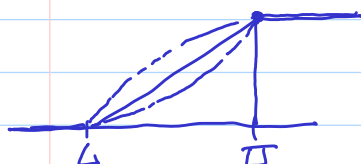


$$d = \begin{cases} \frac{U - \hat{y}}{U - a}^{\alpha}, & a \leq \hat{y} \leq U \\ 0, & \hat{y} > U \end{cases}$$

upper bound

Smallest possible value of \hat{y}

- For larger-the-better problem



$$d = \begin{cases} \frac{\hat{y} - L}{U - L}^{\alpha}, & L \leq \hat{y} \leq U \\ 1, & \hat{y} > U \\ 0, & \hat{y} < L \end{cases}$$

upper bound

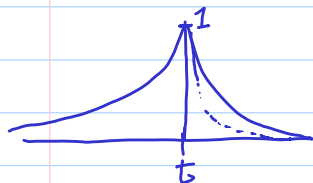
lower bound

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.

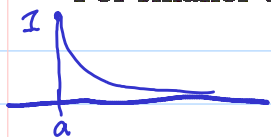
- For nominal-the-best problem:



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

Smaller α : make the desirability function drop more slowly from its peak.

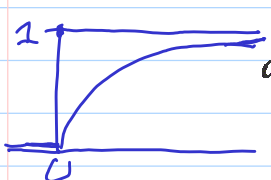
- For smaller-the-better problem



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

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

- For larger-the-better problem



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

lower bound

Analysis Strategies for Multiple Responses

• Overall desirability function D

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

Weighted geometric mean

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

weight

suggestion: better to understand the relationship between $\hat{y}_1, \dots, \hat{y}_m$ before perform desirability analysis.

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

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

weighted average of $\ln(d_i)$'s

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

\hat{y}_i undesirable/unacceptable

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

e.g. do PCA on $\hat{y}_1, \dots, \hat{y}_m$

– Weighted arithmetic mean

$$D^{\bar{x}} = w_1 d_1 + w_2 d_2 + \cdots + w_m d_m$$

• 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