

HW1. Steepest edge

資工四 408410098 蔡嘉祥

Meaning of Steepest edge

$$X = \begin{bmatrix} X_B \\ \hline X_N \end{bmatrix} = \begin{bmatrix} B^{-1}b \\ \hline 0 \end{bmatrix} = \begin{bmatrix} x_{B_1} \\ x_{B_2} \\ \dots \\ x_{B_m} \\ \hline x_{N_1}(0) \\ \dots \\ x_{N_{n-m}}(0) \end{bmatrix}$$

$$z = \mathbf{c}_B B^{-1} N - \sum_{j \in \text{Nonbasic}} \underbrace{(\mathbf{c}_B B^{-1} A_j - c_j)}_{z_j} x_j$$

Gradient of z

$$\nabla z_j = \frac{\partial z}{\partial x_j} = -(z_j - c_j)$$

Length of ΔX if changing x_j

Minimum ratio test :

- $\beta = B^{-1}b = X_B$
- $\alpha_k = B^{-1}A_k$
- entering variable :
 - $\theta_k = \min(\frac{\beta_r}{\alpha_{r,k}} | \forall \alpha_{r,k} > 0, k \in \text{Nonbasic})$

if entering x_r to θ_k by minimum ratio test:

$$\bullet X^* = \begin{bmatrix} x_{B_1} - \theta_k * \alpha_{1,k} \\ x_{B_2} - \theta_k * \alpha_{2,k} \\ \dots \\ x_{B_m} - \theta_k * \alpha_{m,k} \\ \hline x_{N_1}(0) \\ \dots \\ \theta_k \\ \dots \\ x_{N_{n-m}}(0) \end{bmatrix}$$

different (distance, in L_2 norm) between 2 solution vectors:

$$\begin{aligned} \bullet \Delta X = |X - X^*|_2 &= \left\| \begin{bmatrix} \theta_k * \alpha_{1,k} \\ \theta_k * \alpha_{2,k} \\ \dots \\ \theta_k * \alpha_{m,k} \\ \hline 0 \\ \dots \\ -\theta_k \\ \dots \\ 0 \end{bmatrix} \right\|_2 \\ &= \sqrt{\theta_k^2 + \sum_{j \in \text{Nonbasic}} (\theta_k * \alpha_{j,k})^2} = \theta_k \sqrt{1 + \sum_{j \in \text{Nonbasic}} (\alpha_{j,k})^2} = \\ &\theta_k \sqrt{1 + |\alpha_k|_2^2} \end{aligned}$$

Steepest edge

if change x_{j_k} to θ_k ,

$$\Delta z = \nabla z_k \theta_k \text{ and hence change per unit length} = \frac{\nabla z_k \theta_k}{\theta_k \sqrt{1 + |\alpha_k|_2^2}}$$

Entering the most steepest one:

(i.e. argument maximum of **change per unit length**)

$$\operatorname{argmax}_k \left(\frac{\nabla z_k \theta_k}{\theta_k \sqrt{1 + |\alpha_k|_2^2}} \right) = \operatorname{argmax}_k \left(\frac{\nabla z_k}{|\alpha_k|_2} \right) = \operatorname{argmin}_k \left(\frac{z_k - c_k}{|B^{-1} A_k|_2} \right)$$

Comparison

- disadvantage:
 - Computationally expensive. But still, less than greatest improvement method.
- advantage:

- In practice, it works well.
 - Leading to fewer pivots overall than greatest improvement method.

Reference:

Steepest-edge rule and its number of simplex iterations for a nondegenerate LP
(<https://www.sciencedirect.com/science/article/pii/S0167637718304346>)

<https://people.orie.cornell.edu/dpw/orie6300/Lectures/lec13.pdf>