

x is coord. wise minimizer

$$\begin{aligned} f(y) - f(x) &= g(y) + h(y) - (g(x) + h(x)) \\ &\geq \nabla g(x)^T (y - x) + \sum_{i=1}^n h_i(x_i) - h_i(y_i) \\ &= \sum_{i=1}^n \left(\underbrace{\nabla_i g(x) (y_i - x_i) + h_i(y_i) - h_i(x_i)}_{\geq 0} \right) \\ &\geq 0. \end{aligned}$$

$$\min_{\beta} \|y - \sum_{j=1}^p X_j \beta_j\|_2^2$$

min over β_i (all others fixed)

$$\min_{\beta_i} \|y - \sum_{j \neq i} X_j \beta_j - X_i \beta_i\|_2^2$$

$$\Leftrightarrow \min_{\beta_i} \|X_i\|_2^2 \beta_i^2 - 2(y - \sum_{j \neq i} X_j \beta_j)^T X_i \beta_i + \text{const}$$

$$\Leftrightarrow \beta_i = \frac{X_i^T (y - \sum_{j \neq i} X_j \beta_j)}{\|X_i\|_2^2} = \frac{X_i^T (y - X_{-i} \beta_{-i})}{\|X_i\|_2^2}$$

cycle $i = 1, \dots, p$

scheme for full cycle
costs $O(np)$ flops

Gradient update:

$$\beta^+ = \beta + t X^T (y - X\beta) \quad \left. \vphantom{\beta^+ = \beta + t X^T (y - X\beta)} \right\} \text{costs } O(np)$$

incremental
gradient

$$\min \sum_{i=1}^n f_i(\beta) \quad \text{linear regression}$$

$$\sum_{i=1}^n (y_i - x_i^T \beta)^2$$

row of matrix X .

coordinate
descent

$$\|y - \sum X_j \beta_j\|_2^2$$

col of matrix X

Rewrite i^{th} coord update as

$$\beta_i \leftarrow \frac{X_i^T r}{\|X_i\|_2^2}$$

single update given r ,
takes $O(n)$ flops

$$r = y - X_{-i} \beta_{-i} \quad i^{\text{th}} \text{ step}$$

$$r \leftarrow r + X_i \beta_i - X_{i+1} \beta_{i+1} \quad (i+1)^{\text{st}} \text{ step}$$

update to r also costs $O(n)$ flops

therefore done cleverly, full c.d. cycle takes $O(n^2)$!

$$\min_{\beta} \|y - X\beta\|_2^2 + \lambda \sum_{i=1}^p \|\beta_i\|_1$$

$$\min_{\beta_i} \|y - \sum_{j \neq i} X_j \beta_j - X_i \beta_i\|_2^2 + \lambda \|\beta_i\|_1$$

$$\Leftrightarrow \beta_i = \frac{S_{\lambda}}{\|X_i\|_2^2} \left(\frac{X_i^T (y - X_{-i} \beta_{-i})}{\|X_i\|_2^2} \right)$$

$$\min_{\beta} \|y - X\beta\|_2^2 \quad \text{st} \quad \|\beta\|_{\infty} \leq s$$

$$\Leftrightarrow \min_{\beta} \underbrace{\|y - X\beta\|_2^2}_{g(\beta)} + \underbrace{\sum_{i=1}^p \mathbb{1}\{|\beta_i| \leq s\}}_{h_i(\beta_i)}$$