

$$D = \begin{bmatrix} \text{--- } D_1 \text{ ---} \\ \text{--- } D_2 \text{ ---} \\ \vdots \\ \text{--- } D_m \text{ ---} \end{bmatrix}$$

$$D\beta \text{ sparse} \iff D_j \beta = 0 \text{ many } j$$

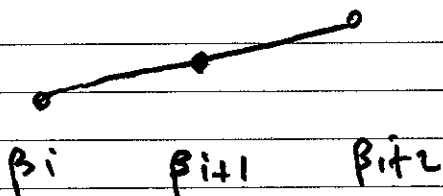
$$D = \begin{bmatrix} \text{--- } e_1 \text{ ---} \\ \text{--- } e_2 \text{ ---} \\ \vdots \\ \text{--- } e_k \text{ ---} \end{bmatrix}$$

$$\sum (y_i - \beta_i)^2 + \lambda \sum |\beta_i - \beta_{i+1}|$$

$$\beta_i = \frac{e^{-\beta_i}}{1 + e^{-\beta_i}}$$

$$\beta_i - 2\beta_{i+1} + \beta_{i+2} = 0$$

$$\iff \beta_{i+1} = \frac{\beta_i + \beta_{i+2}}{2}$$





$$\beta = D\theta \iff A\beta = 0$$

for some  $\theta \in \mathbb{R}^k$

$$\iff f(D\theta) + \lambda \|D\theta\|,$$

$$\iff g(\theta) + \lambda \|D\theta\|,$$

$$\hat{\beta} = D\hat{\theta}.$$


---

$$\nabla f(\beta) + \lambda D^T \gamma$$

$$\gamma \in \|\cdot\|, \quad | \quad x = D\beta$$

Chain rule:

$$g(x) = f(Ax+b)$$

$$\partial g(x) = A^T \partial f(Ax+b)$$

$$\beta \leftarrow \beta + \nabla f(\beta) + \lambda D^T \gamma$$


---

$$\min_{\beta} f(\beta) + \lambda \|D\beta\|$$

$$\min_{z, \beta} f(\beta) + \lambda \|z\|, \quad \text{s.t.} \quad z = D\beta$$

$$(DD^T)^{-1} b$$

$$\nabla f(\hat{\beta}) + D^T \hat{u} = 0.$$

$$\hat{u}_i \in \begin{cases} \{\lambda\} & \text{when } (D\hat{\beta})_i > 0 \\ \{-\lambda\} & \text{when } (D\hat{\beta})_i < 0 \\ [-\lambda, \lambda] & \text{else} \end{cases}$$

$$-\lambda < \hat{u}_i < \lambda \implies (D\hat{\beta})_i = 0.$$

$$\|u\|_\infty \leq \lambda \iff -\lambda \leq u_i \leq \lambda \quad i=1, \dots, m$$

$$\min_u \left[ f^*(-D^T u) + \frac{1}{\epsilon} \phi(u) \right]$$

$$\phi(u) = - \sum \left[ \log(\lambda + u_i) + \log(\lambda - u_i) \right]$$

$$F(u) = f^*(-D^T u) + \frac{1}{\epsilon} \phi(u)$$

$$\nabla^2 F(u) = D \nabla^2 f^* D^T + \frac{1}{\epsilon} \nabla^2 \phi$$

$$\nabla F(u) = D \nabla f^* + \frac{1}{\epsilon} \nabla \phi$$