

## Statistical Computing: Homework 4

Due on May 8 (Monday) 23:30

1. Consider the following normal model:

$$Y_i \sim N(\mu_i, \sigma_i^2), \quad i = 1, 2, \dots, n;$$

$$\mu_i = \beta_0 + \beta_1 x_i, \quad \sigma_i^2 = e^{\alpha_0 + \alpha_1 x_i}, \quad \boldsymbol{\alpha} = (\alpha_0, \alpha_1)' \in R^2, \quad \boldsymbol{\beta} = (\beta_0, \beta_1)' \in R^2.$$

Using DataA, solve the MLE of  $(\boldsymbol{\beta}, \boldsymbol{\alpha})$  based on the (block) coordinate descent method. (You may treat  $\boldsymbol{\beta}$  as a block and  $\boldsymbol{\alpha}$  as a block.)

DataA (extract from Boston Housing data):

$Y_i = \log(\text{medv}_i)$  for the  $i$ th suburb region of Boston; and  $x_i = \text{lstat}_i$ . Detail variable descriptions can be found in “Boston {ISLR2}”.

2. Consider an alternative fused lasso problem:

$$\min_{\boldsymbol{\beta}} \left\{ \frac{1}{2} \sum_{t=1}^n (y_t - \beta_t)^2 + \tau \sum_{t=3}^n |\beta_t - 2\beta_{t-1} + \beta_{t-2}| \right\}, \quad \boldsymbol{\beta} = (\beta_1, \dots, \beta_n)',$$

which can be reformulated as

$$\min_{\boldsymbol{\beta}} \left\{ \frac{1}{2} \|\mathbf{y} - \boldsymbol{\beta}\|^2 + \tau \sum_{t=3}^n |\delta_t| \right\},$$

where

$$\delta_t \equiv \beta_t - 2\beta_{t-1} + \beta_{t-2} = (0, \dots, 0, 1, -2, \underbrace{1}_{t\text{-th}}, 0, \dots, 0)\boldsymbol{\beta}, \quad t = 3, \dots, n;$$

$$\underbrace{\boldsymbol{\delta}}_{(n-2) \times 1} \equiv (\delta_3, \delta_4, \dots, \delta_n)' = \underbrace{D}_{(n-2) \times n} \boldsymbol{\beta}, \quad D = \begin{bmatrix} 1 & -2 & 1 & \dots & 0 \\ 0 & 1 & -2 & 1 & \dots \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 1 & -2 & 1 \end{bmatrix}.$$

Using ADMM to solve  $\{\beta_t\}$  for DataB.

DataB:  $\{(t, y_t) : t = 1, 2, \dots, 100\}$  (same data shown in R Lab7-2 Example 5)

3. Consider the following function:

$$f(\mathbf{x}) = - \sum_{j=1}^{10} \sin(\pi x_j) [\sin(j\pi x_j^2)]^{20}, \quad \mathbf{x} = (x_1, x_2, \dots, x_{10})', \quad 0 \leq x_j \leq 1, \quad j = 1, 2, \dots, 10.$$

Use any optimization method to find the minimum of  $f(\mathbf{x})$  for  $\mathbf{x} \in [0, 1]^{10}$ .