

## Problem Set 729

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**Estimate Single Kernel Effect** Let's start by estimating single kernel effect. The objective function is,

$$\|\mathbf{y} - \mathbf{X}\boldsymbol{\beta} - K_1\boldsymbol{\alpha}_1\|^2 + \lambda\boldsymbol{\alpha}_1^\top K_1\boldsymbol{\alpha}_1. \quad (1)$$

Differentiating (1) with respect to  $\boldsymbol{\beta}$  and  $\boldsymbol{\alpha}_1$ , we obtain,

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top (\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1), \quad (2)$$

$$\hat{\boldsymbol{\alpha}}_1 = (K_1 + \lambda \mathbf{I})^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}}). \quad (3)$$

Substituting (3) into (2), we get

$$\begin{aligned} \hat{\boldsymbol{\beta}} &= \{\mathbf{X}^\top [K_1(K_1 + \lambda \mathbf{I})^{-1} - \mathbf{I}] \mathbf{X}\}^{-1} \mathbf{X}^\top [K_1(K_1 + \lambda \mathbf{I})^{-1} - \mathbf{I}] \mathbf{y} \\ &= \{\mathbf{X}^\top (\mathbf{I} + \lambda^{-1} K_1)^{-1} \mathbf{X}\}^{-1} \mathbf{X}^\top (\mathbf{I} + \lambda^{-1} K_1)^{-1} \mathbf{y} \quad (\text{Woodbury matrix identity}) \\ &= \{\mathbf{X}^\top V_1^{-1} \mathbf{X}\}^{-1} \mathbf{X}^\top V_1^{-1} \mathbf{y}, \end{aligned} \quad (4)$$

where we denote  $V_1 = \mathbf{I} + \lambda^{-1} K_1$  and therefore  $(\lambda V_1)^{-1} = (K_1 + \lambda \mathbf{I})^{-1}$ .

**Estimate Multiple Kernel Effects** Now we move on to estimating multiple kernel effects. Here we consider two kernel effects, and the objective function becomes,

$$\|\mathbf{y} - \mathbf{X}\boldsymbol{\beta} - K_1\boldsymbol{\alpha}_1 - K_2\boldsymbol{\alpha}_2\|^2 + \lambda\boldsymbol{\alpha}_1^\top K_1\boldsymbol{\alpha}_1 + \lambda\boldsymbol{\alpha}_2^\top K_2\boldsymbol{\alpha}_2. \quad (5)$$

Similarly, differentiating (1) with respect to  $\boldsymbol{\beta}$ ,  $\boldsymbol{\alpha}_1$  and  $\boldsymbol{\alpha}_2$ , we obtain,

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top (\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1 - K_2 \hat{\boldsymbol{\alpha}}_2), \quad (6)$$

$$\hat{\boldsymbol{\alpha}}_1 = (K_1 + \lambda \mathbf{I})^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}} - K_2 \hat{\boldsymbol{\alpha}}_2) = (\lambda V_1)^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}} - K_2 \hat{\boldsymbol{\alpha}}_2), \quad (7)$$

$$\hat{\boldsymbol{\alpha}}_2 = (K_2 + \lambda \mathbf{I})^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}} - K_1 \hat{\boldsymbol{\alpha}}_1) = (\lambda V_2)^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}} - K_1 \hat{\boldsymbol{\alpha}}_1). \quad (8)$$

Plugging (8) into (6), we obtain,

$$\hat{\boldsymbol{\beta}} = \{\mathbf{X}^\top V_2^{-1} \mathbf{X}\}^{-1} \mathbf{X}^\top V_2^{-1} (\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1). \quad (9)$$

Plugging (8) into (7), we obtain,

$$\begin{aligned} \hat{\boldsymbol{\alpha}}_1 &= \lambda [\lambda^2 \mathbf{I} - V_1^{-1} K_2 V_2^{-1} K_1]^{-1} V_1^{-1} V_2^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}}) \\ &= \lambda W_1^{-1} (\mathbf{y} - \mathbf{X} \hat{\boldsymbol{\beta}}) \\ &= \lambda W_1^{-1} \{\mathbf{y} - \mathbf{X} \{\mathbf{X}^\top V_2^{-1} \mathbf{X}\}^{-1} \mathbf{X}^\top V_2^{-1} (\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1)\}, \end{aligned} \quad (\text{plug in (9)})$$

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$$\hat{\alpha}_1 = \{\lambda^{-1}W_1 - \mathbf{X}\{\mathbf{X}^\top V_2^{-1}\mathbf{X}\}^{-1}\mathbf{X}^\top V_2^{-1}K_1\}^{-1}\{\mathbf{I} - \mathbf{X}\{\mathbf{X}^\top V_2^{-1}\mathbf{X}\}^{-1}\mathbf{X}^\top V_2^{-1}\}\mathbf{y}. \quad (10)$$

where we denote

$$W_1 = V_2 V_1 (\lambda^2 \mathbf{I} - V_1^{-1} K_2 V_2^{-1} K_1) = \lambda^2 V_2 V_1 - V_2 K_2 V_2^{-1} K_1.$$

Similarly,

$$\hat{\alpha}_2 = \{\lambda^{-1}W_2 - \mathbf{X}\{\mathbf{X}^\top V_1^{-1}\mathbf{X}\}^{-1}\mathbf{X}^\top V_1^{-1}K_2\}^{-1}\{\mathbf{I} - \mathbf{X}\{\mathbf{X}^\top V_1^{-1}\mathbf{X}\}^{-1}\mathbf{X}^\top V_1^{-1}\}\mathbf{y}. \quad (11)$$

**Alternative** Following the same calculation as constructing block matrices. Letting

$$K_3 = \begin{bmatrix} X & K_1 & K_2 \\ n \times p & n \times n & n \times n \end{bmatrix}, \quad K_4 = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & K_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & K_2 \end{bmatrix}, \quad \theta = \begin{bmatrix} \beta \\ \alpha_1 \\ \alpha_2 \end{bmatrix}.$$

Then (5) becomes

$$\|\mathbf{y} - K_3 \theta\|^2 + \lambda \theta^\top K_4 \theta. \quad (12)$$

Differentiating (12) with respect to  $\theta$ , we obtain,

$$\begin{aligned} (\lambda K_4 + K_3^\top K_3) \theta &= K_3^\top \mathbf{y}, \\ \theta &= (\lambda K_4 + K_3^\top K_3)^{-1} K_3^\top \mathbf{y}. \end{aligned}$$

Specifically,

$$\begin{aligned} & \left\{ \lambda \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & K_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & K_2 \end{bmatrix} + \begin{bmatrix} \mathbf{X}^\top \mathbf{X} & \mathbf{X}^\top K_1 & \mathbf{X}^\top K_2 \\ K_1^\top X & K_1^\top K_1 & K_1^\top K_2 \\ K_2^\top X & K_2^\top K_1 & K_2^\top K_2 \end{bmatrix} \right\} \begin{bmatrix} \beta \\ \alpha_1 \\ \alpha_2 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{X}^\top \mathbf{X} & \mathbf{X}^\top K_1 & \mathbf{X}^\top K_2 \\ K_1 X & K_1(K_1 + \lambda \mathbf{I}) & K_1 K_2 \\ K_2 X & K_2 K_1 & K_2(K_2 + \lambda \mathbf{I}) \end{bmatrix} \begin{bmatrix} \beta \\ \alpha_1 \\ \alpha_2 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{X}^\top \mathbf{y} \\ K_1 \mathbf{y} \\ K_2 \mathbf{y} \end{bmatrix}, \\ & \begin{bmatrix} \hat{\beta} \\ \hat{\alpha}_1 \\ \hat{\alpha}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{X}^\top \mathbf{X} & \mathbf{X}^\top K_1 & \mathbf{X}^\top K_2 \\ K_1 X & K_1(K_1 + \lambda \mathbf{I}) & K_1 K_2 \\ K_2 X & K_2 K_1 & K_2(K_2 + \lambda \mathbf{I}) \end{bmatrix}^{-1} \begin{bmatrix} \mathbf{X}^\top \mathbf{y} \\ K_1 \mathbf{y} \\ K_2 \mathbf{y} \end{bmatrix}. \end{aligned}$$