## KMR: Estimate Multiple Kernel Effects

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Problem Set 0

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# 0.1 Single Kernel

Let's start by estimating single kernel effect. The objective function is,

$$\|\mathbf{y} - \mathbf{X}\boldsymbol{\beta} - K_1\boldsymbol{\alpha}_1\|^2 + \lambda \boldsymbol{\alpha}_1^{\mathsf{T}} K_1\boldsymbol{\alpha}_1. \tag{1}$$

Differentiating (1) with respect to  $\beta$  and  $\alpha_1$ , we obtain,

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

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

Substituting (3) into (2), we get

$$\hat{\boldsymbol{\beta}} = \{ \mathbf{X}^{\top} [K_1(K_1 + \lambda \mathbf{I})^{-1} - \mathbf{I}] \mathbf{X} \}^{-1} \mathbf{X}^{\top} [K_1(K_1 + \lambda \mathbf{I})^{-1} - \mathbf{I}] \mathbf{y}$$

$$= \{ \mathbf{X}^{\top} (\mathbf{I} + \lambda^{-1} K_1)^{-1} \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\mathbf{I} + \lambda^{-1} K_1)^{-1} \mathbf{y}$$
 (Woodbury matrix identity)
$$= \{ \mathbf{X}^{\top} V_1^{-1} \mathbf{X} \}^{-1} \mathbf{X}^{\top} V_1^{-1} \mathbf{y},$$

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

$$(\mathbf{I} + \lambda^{-1}K)^{-1} = \mathbf{I} - K(K + \lambda \mathbf{I})^{-1}$$

And finally,

$$\hat{\boldsymbol{\alpha}}_{1} = \{\lambda V_{1} - \mathbf{X}(\mathbf{X}^{\top}\mathbf{X})^{-1}\mathbf{X}^{\top}K_{1}\}^{-1}\{\mathbf{I} - \mathbf{X}(\mathbf{X}^{\top}\mathbf{X})^{-1}\mathbf{X}^{\top}\}\mathbf{y}.$$
(4)

#### 0.2 Two Kernels

Now we move on to estimating multiple kernel effects. Here we consider two kernel effects, and the objective function becomes,

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

Similarly, differentiating (5) with respect to  $\beta$ ,  $\alpha_1$  and  $\alpha_2$ , we obtain,

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

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

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

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

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

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

$$\hat{\boldsymbol{\alpha}}_{1} = \{\lambda^{2}V_{2}V_{1} + K_{1} - \lambda V_{2}K_{1}\}^{-1}(\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}})$$

$$= W_{21}^{-1}(\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}})$$

$$= W_{21}^{-1}\{\mathbf{y} - \mathbf{X}\{\mathbf{X}^{\top}V_{2}^{-1}\mathbf{X}\}^{-1}\mathbf{X}^{\top}V_{2}^{-1}(\mathbf{y} - K_{1}\hat{\boldsymbol{\alpha}}_{1})\}, \qquad (\text{plug in } (9))$$

$$\hat{\boldsymbol{\alpha}}_{1} = \{W_{21} - \mathbf{X}\{\mathbf{X}^{\top}V_{2}^{-1}\mathbf{X}\}^{-1}\mathbf{X}^{\top}V_{2}^{-1}K_{1}\}^{-1}\{\mathbf{I} - \mathbf{X}\{\mathbf{X}^{\top}V_{2}^{-1}\mathbf{X}\}^{-1}\mathbf{X}^{\top}V_{2}^{-1}\}\mathbf{y}.$$
 (10)

where we denote

$$W_{21} = \lambda^2 V_2 V_1 + K_1 - \lambda V_2 K_1.$$

#### 0.3 Three Kernels

We consider three kernel effects, and the objective function becomes,

$$\|\mathbf{y} - \mathbf{X}\boldsymbol{\beta} - K_1\boldsymbol{\alpha}_1 - K_2\boldsymbol{\alpha}_2 - K_3\boldsymbol{\alpha}_3\|^2 + \lambda\boldsymbol{\alpha}_1^{\top}K_1\boldsymbol{\alpha}_1 + \lambda\boldsymbol{\alpha}_2^{\top}K_2\boldsymbol{\alpha}_2 + \lambda\boldsymbol{\alpha}_3^{\top}K_3\boldsymbol{\alpha}_3.$$
(11)

Similarly, differentiating (11) with respect to  $\beta$ ,  $\alpha_1$ ,  $\alpha_2$  and  $\alpha_3$ , we obtain,

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^{\top} \mathbf{X})^{-1} \mathbf{X}^{\top} (\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1 - K_2 \hat{\boldsymbol{\alpha}}_2 - K_3 \hat{\boldsymbol{\alpha}}_3), \tag{12}$$

$$\hat{\boldsymbol{\alpha}}_1 = (\lambda V_1)^{-1} (\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}} - K_2\hat{\boldsymbol{\alpha}}_2 - K_3\hat{\boldsymbol{\alpha}}_3), \tag{13}$$

$$\hat{\boldsymbol{\alpha}}_2 = (\lambda V_2)^{-1} (\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}} - K_1\hat{\boldsymbol{\alpha}}_1 - K_3\hat{\boldsymbol{\alpha}}_3), \tag{14}$$

$$\hat{\boldsymbol{\alpha}}_3 = (\lambda V_3)^{-1} (\mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}} - K_1\hat{\boldsymbol{\alpha}}_1 - K_2\hat{\boldsymbol{\alpha}}_2). \tag{15}$$

Plugging (14) into (15), we obtain,

$$\hat{\boldsymbol{\alpha}}_3 = W_{23}^{-1} \{ \mathbf{y} - \mathbf{X}\hat{\boldsymbol{\beta}} - K_1 \hat{\boldsymbol{\alpha}}_1 \}, \tag{16}$$

where we denote

$$W_{23} = \lambda^2 V_2 V_3 + K_3 - \lambda V_2 K_3.$$

Plugging (14) into (12), we obtain.

$$\mathbf{X}^{\top}(\lambda V_2)^{-1}\mathbf{X}\hat{\boldsymbol{\beta}} = \mathbf{X}^{\top}(\lambda V_2)^{-1}[\mathbf{y} - K_1\hat{\boldsymbol{\alpha}}_1 - K_3\hat{\boldsymbol{\alpha}}_3]. \tag{17}$$

Plugging (16) into (17), we obtain,

$$\hat{\boldsymbol{\beta}} = \{ \mathbf{X}^{\top} (\lambda V_2)^{-1} [\mathbf{I} - K_3 W_{23}^{-1}] \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\lambda V_2)^{-1} [\mathbf{I} - K_3 W_{23}^{-1}] [\mathbf{y} - K_1 \hat{\boldsymbol{\alpha}}_1].$$
 (18)

Plugging (14), (16) and (18) subsequently into (13), we obtain,

$$\begin{cases}
[W_{21} - K_3 W_{21}^{-1} K_1] - \gamma \mathbf{X} \{ \mathbf{X}^{\top} (\lambda V_2)^{-1} \gamma \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\lambda V_2)^{-1} \gamma K_1 \} \hat{\boldsymbol{\alpha}}_1 \\
= \gamma \{ \mathbf{I} - \mathbf{X} \{ \mathbf{X}^{\top} (\lambda V_2)^{-1} \gamma \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\lambda V_2)^{-1} \gamma \} \mathbf{y},
\end{cases}$$

where we denote

$$\gamma = [\mathbf{I} - K_3 W_{23}^{-1}].$$

$$\hat{\boldsymbol{\alpha}}_{1} = \left\{ [W_{21} - K_{3}W_{21}^{-1}K_{1}] - \gamma \mathbf{X} \{ \mathbf{X}^{\top} (\lambda V_{2})^{-1} \gamma \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\lambda V_{2})^{-1} \gamma K_{1} \right\}^{-1} \cdot \gamma \{ \mathbf{I} - \mathbf{X} \{ \mathbf{X}^{\top} (\lambda V_{2})^{-1} \gamma \mathbf{X} \}^{-1} \mathbf{X}^{\top} (\lambda V_{2})^{-1} \gamma \} \mathbf{y}.$$

$$(19)$$

# 0.4 Summary

Finally, if we put all those  $\hat{\alpha}_{1}$ s together,

$$\hat{\boldsymbol{\alpha}}_{1} = \{\lambda V_{1} - \mathbf{X}(\mathbf{X}^{\top}\mathbf{X})^{-1}\mathbf{X}^{\top}K_{1}\}^{-1}\{\mathbf{I} - \mathbf{X}(\mathbf{X}^{\top}\mathbf{X})^{-1}\mathbf{X}^{\top}\}\mathbf{y}, \qquad \text{(single kernel)}$$

$$\hat{\boldsymbol{\alpha}}_{1} = \{W_{21} - \mathbf{X}\{\mathbf{X}^{\top}V_{2}^{-1}\mathbf{X}\}^{-1}\mathbf{X}^{\top}V_{2}^{-1}K_{1}\}^{-1}\{\mathbf{I} - \mathbf{X}\{\mathbf{X}^{\top}V_{2}^{-1}\mathbf{X}\}^{-1}\mathbf{X}^{\top}V_{2}^{-1}\}\mathbf{y}, \qquad \text{(two kernels)}$$

$$\hat{\boldsymbol{\alpha}}_{1} = \{[W_{21} - K_{3}W_{21}^{-1}K_{1}] - \gamma\mathbf{X}\{\mathbf{X}^{\top}(\lambda V_{2})^{-1}\gamma\mathbf{X}\}^{-1}\mathbf{X}^{\top}(\lambda V_{2})^{-1}\gamma K_{1}\}^{-1}$$

$$\cdot \gamma\{\mathbf{I} - \mathbf{X}\{\mathbf{X}^{\top}(\lambda V_{2})^{-1}\gamma\mathbf{X}\}^{-1}\mathbf{X}^{\top}(\lambda V_{2})^{-1}\gamma\}\mathbf{y}. \qquad \text{(three kernels)}$$

where,

$$V_1 = \mathbf{I} + \lambda^{-1} K_1,$$

$$V_2 = \mathbf{I} + \lambda^{-1} K_2,$$

$$W_{21} = \lambda^2 V_2 V_1 + K_1 - \lambda V_2 K_1,$$

$$W_{23} = \lambda^2 V_2 V_3 + K_3 - \lambda V_2 K_3,$$

$$\gamma = [\mathbf{I} - K_3 W_{23}^{-1}].$$

## 0.5 Alternative for Two Kernels

Following the same calculation as constructing block matrices. Letting

$$K_{3}_{n\times(p+n+n)} = \begin{bmatrix} X, K_{1}, K_{2} \end{bmatrix}, \quad K_{4}_{(p+n+n)\times(p+n+n)} = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & K_{1} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & K_{2} \end{bmatrix}, \quad \boldsymbol{\theta} = \begin{bmatrix} \boldsymbol{\beta} \\ p\times 1 \\ \boldsymbol{\alpha}_{1} \\ n\times 1 \\ \boldsymbol{\alpha}_{2} \\ n\times 1 \end{bmatrix}.$$

Then (5) becomes

$$\|\mathbf{y} - K_3 \boldsymbol{\theta}\|^2 + \lambda \boldsymbol{\theta}^\top K_4 \boldsymbol{\theta}. \tag{20}$$

Differentiating (20) with respect to  $\theta$ , we obtain,

$$(\lambda K_4 + K_3^{\mathsf{T}} K_3) \boldsymbol{\theta} = K_3^{\mathsf{T}} \mathbf{y},$$
  
$$\boldsymbol{\theta} = (\lambda K_4 + K_3^{\mathsf{T}} K_3)^{-1} K_3^{\mathsf{T}} \mathbf{y}.$$

Specifically,

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