## 课程作业

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## 题目 1. 5.9

解答.

$$\hat{f}(x_i) = N(N^T N + \lambda \Omega_N)^{-1} N^T y = S_{\lambda} y$$
$$S_{\lambda} = N(N^T N + \lambda \Omega_N)^{-1} N^T$$

当 N 可逆:

$$S_{\lambda} = ((N^{T})^{-1}(N^{T}N + \lambda\Omega_{N})N^{-1})^{-1}$$

$$= ((N^{T})^{-1}N^{T}NN^{-1} + \lambda(N^{T})^{-1}\Omega_{N}N^{-1})^{-1}$$

$$= (I + \lambda(N^{T})^{-1}\Omega_{N}N^{-1})^{-1}$$

所以:

$$K = (N^T)^{-1} \Omega_N N^{-1}$$

K 不依赖于 λ

题目 2. 5.15

解答. (a)

$$\langle K(\bullet, x_i), f \rangle_{H_k} = \sum_{i=1}^{\infty} \frac{c_i}{r_i} \langle K(\bullet, x_i), \phi_i(\bullet) \rangle = \sum_{i=1}^{\infty} \frac{c_i}{r_i} [r_i \phi_i(x_i)] = \sum_{i=1}^{\infty} c_i \phi_i(x_i) = f(x_i)$$

(b) 
$$\langle K(\bullet, x_i), K(\bullet, x_j) \rangle_{H_k} = \sum_{k=1}^{\infty} \frac{1}{r_k} \langle K(\bullet, x_i), K(\bullet, x_j) \rangle$$

$$= \sum_{k=1}^{\infty} \frac{1}{r_k} [r_k \phi_k(x_i) r_k \phi_k(x_j)] = \sum_{k=1}^{\infty} r_k \phi_k(x_i) r_k \phi_k(x_j) = K(x_i, x_j)$$
(c)

$$J(g) = \langle g(x), g(x) \rangle_{H_k} = \sum_{i=1}^{N} \sum_{i=1}^{N} \langle K(x, x_i), K(x, x_i) \rangle_{H_k} \, \partial i \partial j = \sum_{i=1}^{N} \sum_{i=1}^{N} K(x_i, x_j) \partial i \partial j$$

(d) 
$$\rho$$
 在  $H_k$  中正交,所以  $< K(\bullet, x_i), \rho(x) >= 0$ 

$$\tilde{g}(x_i) = \langle K(\bullet, \tilde{g}(x_i)) \rangle_{H_k} = \langle K(\bullet, x_i), g(x_i) \rangle_{H_k} + \langle K(\bullet, x_i), \rho(x_i) \rangle_{H_k}$$

$$q(x_i)$$

$$\begin{split} J(\tilde{g}) = <\tilde{g}(x), \tilde{g}(x)>_{H_k} = _{H_k} + 2 < g(x), \rho(x)>_{H_k} + <\rho(x), \rho(x)>_{H_k} \\ = J(g) + 2\sum_{i=1}^N \partial_i < K(x,x_i), \rho(x)>_{H_k} + J(\rho) \\ = J(g) + J(\rho) \end{split}$$

所以最后可以推出:

$$\sum_{i=1}^{N} L(y_i, \tilde{g}(x_i)) + \lambda J(\tilde{g})$$

$$= \sum_{i=1}^{N} L(y_i, g(x_i)) + \lambda (J(g) + J(\rho)) \ge \sum_{i=1}^{N} L(y_i, g(x_i)) + \lambda J(\rho)$$

题目 3. 5.16

解答. (a)

$$K(x,y) = \sum_{m=1}^{M} h_m(x)h_m(y) = \sum_{i=1}^{\infty} \gamma_i \phi_i(x)\phi_i(y)$$

$$\Rightarrow \sum_{m=1}^{M} (\int h_m(x)\phi_k(x)dx)h_m(y) = \sum_{m=1}^{M} \gamma_i (\int \phi_i(x)\phi_k(x)dx)\phi_i(y)$$

因为  $\phi_i(x)$  有如下性质:

$$\int \phi_i(x)\phi_k(x)dx = \langle \phi_i(x), \phi_k(x) \rangle = \begin{cases} 1 & i = k \\ 0 & i \neq k \end{cases}$$

所以:

$$\sum_{m=1}^{M} \left( \int h_m(x)\phi_k(x)dx \right) h_m(y) = \gamma_k \phi_k(y)$$

 $\cdots$  (b)

ESL 书中公式 (5.63) 对  $\beta$  求偏导, 并使其等于 0, 可得:

$$-H^{T}(y - H(\hat{\beta})) + \lambda \hat{\beta} = 0$$

$$\hat{\beta} = (H^{T}H + \lambda I)^{-1}H^{T}y$$

$$\Rightarrow \hat{f} = H\hat{\beta} = H(H^{T}H + \lambda I)^{-1}H^{T}y = (I + \lambda K^{-1})^{-1}y$$

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

(c)

$$\hat{f} = H\hat{\beta} = K\hat{\alpha} = K(\alpha_1, \alpha_2, \cdots, \alpha_N)^T$$
$$\hat{f}(x) = h(x)^T\hat{\beta} = \sum_{i=1}^N K(x, x_i)\hat{\alpha_i}$$