

KMML Homework 3

Mahdi Kallel

TOTAL POINTS

4 / 4

QUESTION 1

1 Question 1 1 / 1

✓ - **0 pts** Correct

- **1 pts** wrong or missing answer
- **0.25 pts** missing absolute value
- **0.5 pts** missing final step

QUESTION 2

2 Question 2 3 / 3

✓ - **0 pts** Correct

- **0.5 pts** Representer theorem not or wrongly justified
- **0.5 pts** K_X and K_Y are not necessarily invertible.
- **1.5 pts** Computation not finished: what is the solution to this optimisation problem?
- **2 pts** Wrong formula
- **1 pts** Computation not finished: what is the solution to this optimisation problem?
- **3 pts** Problem not solved
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Kernel Methods DM#3 :

Exercise 1:

The RKHS of the linear kernel is $\{f_w(x) = \langle w, x \rangle \mid w \in \mathbb{R}\}$ and thus it is the scalar multiplication.

By this definition $\exists f, g \in \mathbb{R}$ s.t :

$$\begin{aligned} \text{cov}_n(f(X), g(Y)) &= \frac{1}{n} \sum_i f \cdot x_i * g \cdot y_i - \frac{1}{n^2} \sum_i f \cdot x_i * \sum_i g \cdot y_i \\ &= \frac{fg}{n} (X^T Y - X^T O Y) = \frac{fg}{n} X^T (I_n - O) Y, \text{ where } O = \frac{(\mathbf{1}\mathbf{1}^T)}{n} \end{aligned}$$

From the unit ball constraint we have that $|f|, |g| \leq 1$. And thus we deduce that :

$$C_N^K(X, Y) = \max_{|f|, |g| \leq 1} \frac{fg}{n} X^T (I_n - O) Y = \frac{|X^T (I_n - O) Y|}{n}$$

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Exercise 2 :

We start by showing that we can restrict our f, g solutions to C_n^k to the form $f = \sum \alpha_i K_{x_i}$,
 $g = \sum \beta_i K_{y_i}$.

Let's $\mathcal{H}_x = \left\{ f \text{ s.t } f = \sum \alpha_i K_{x_i} \text{ } (\alpha_1 \dots \alpha_n) \in \mathbb{R}^n \right\}$ \mathcal{H}_x is a finite dimensional vector space.

Therefore $\forall f \in \mathcal{H}, f = f_x + f_\perp$.

$\forall_i f(x_i) = \langle f_x, K_{x_i} \rangle$ and by orthogonality $\|f_x\|^2 = \|f\|^2 - \|f_\perp\|^2$ therefore $\|f_x\| \leq \|f\|$.

Thus for any solution f^\star of C_n^k we can find a projection f_x^\star that can be written as $\sum \alpha_i K_{x_i}$ and is also a solution of C_n^k .

$$\text{cov}_n(f(X), g(Y)) = \frac{1}{n} \sum_i f(u_i) * g(y_i) - \frac{1}{n^2} \sum_i f(x_i) * \sum_i g(y_i)$$

$$\frac{1}{n} \sum_i [K_x F]_i [K_y G]_i - \frac{1}{n} \sum_i [K_x F]_i \frac{1}{n} \sum_i [K_y G]_i$$

$$= \frac{1}{n} (K_x F)^T (K_y G) - \frac{1}{n} (K_x F)^T O K_y G = \frac{1}{n} (K_x F)^T (I_n - O) (K_y G)$$

Since the representer theorem applies our norm constraints translate to :

$$F^T K_x F, G^T K_y G \leq 1.$$

And thus $C_n^k(X, Y) = \max_{F, G} \frac{1}{n} (K_x F)^T (I_n - O) (K_y G)$
 $s.t: F^T K_x F, G^T K_y G \leq 1$

K_x, K_y are positive semi definite and thus have a root $\sqrt{K_x}, \sqrt{K_y}$

$$C_n^k(X, Y) = \max_{F, G} \frac{1}{n} (F^T K_x) (I_n - O) (K_y G) \text{ st...}$$

$$= \max_{F, G} \frac{1}{n} F^T \sqrt{K_x} \sqrt{K_x} (I_n - O) \sqrt{K_y} \sqrt{K_y} G$$

$$s.t \ ||F^T \sqrt{K_x}||^2, ||G^T \sqrt{K_y}||^2 \leq 1 \Leftrightarrow ||F^T \sqrt{K_x}||, ||G^T \sqrt{K_y}|| \leq 1$$

We now want to show that this is equivalent to the following problem :

$$C_n^k(X, Y) = \max_{\tilde{F}, \tilde{G}} \frac{1}{n} \tilde{F}^T \sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}$$

$$s.t \ ||\tilde{F}||, ||\tilde{G}|| \leq 1$$

→ If F, G are solutions of the original problem, then we can define $\tilde{F} = \sqrt{K_x} F, \tilde{G} = \sqrt{K_y} G$
and we get $||\tilde{F}||, ||\tilde{G}|| \leq 1$ and $\frac{1}{n} \tilde{F}^T \sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G} = \frac{1}{n} (F^T K_x) (I_n - O) (K_y G)$.

← If \tilde{F}, \tilde{G} are solutions of the second problem, since K_x, K_y are p.d it's diagonalizable in an orthogonal

$$\Rightarrow E = \text{Im}(K_x) \oplus \text{Ker}(K_x) = \text{Im}(K_y) \oplus \text{Ker}(K_y)$$

$\exists F_k, F_k$ st $\tilde{F} = \sqrt{K_x} F + F_k$ and $\sqrt{K_x} F_k = 0$. (and the same for \tilde{G})

$$\frac{1}{n} \tilde{F}^T \sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G} = \frac{1}{n} F^T \sqrt{K_x} \sqrt{K_x} (I_n - O) \sqrt{K_y} \sqrt{K_y} G$$

$$= \frac{1}{n} F^T K_x (I_n - O) K_y G.$$

By orthogonality : $\|F\|^2 = \|\tilde{F}\|^2 - \|F_k\|^2 \leq 1 - \|F_k\|^2 \leq 1$

Thus we've shown that our optimization problem can be rewritten as :

$$C_n^k(X, Y) = \max_{\tilde{F}, \tilde{G}} \frac{1}{n} \tilde{F}^T \sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}$$

$$\text{s.t. } \|\tilde{F}\|, \|\tilde{G}\| \leq 1$$

For a fixed F , we call $A_G = \sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}$

$$\text{The problem is then : } \max_{\|\tilde{F}\| \leq 1} \tilde{F}^T A_G = \max_{\tilde{F}} \frac{\tilde{F}^T A_G}{\|\tilde{F}\|}$$

which is solved for $\tilde{F} = \frac{A_G}{\|A_G\|}$ (using cauchy schwarz).

And thus the problem can be rewritten as :

$$\max_{\|\tilde{G}\| \leq 1} \frac{A_G^T A_G}{\|A_G\|} = \max_{\|\tilde{G}\| \leq 1} \|A_G\| = \max_{\|\tilde{G}\| \leq 1} \|\sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}\|$$

One can show that : $\max_{\|\tilde{G}\| \leq 1} \|\sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}\| = \max_{\|\tilde{G}\| = 1} \|\sqrt{K_x} (I_n - O) \sqrt{K_y} \tilde{G}\|$

$\left(\text{It suffices to multiply any solution } G^* \text{ by } \frac{1}{\|G^*\|} \text{ to get a better solution} \right)$

Thus we recognize the spectral norm problem and thus :

$$C_n^k(X, Y) = \frac{\lambda_{\max}}{n} \text{ where } \lambda_{\max} \text{ is the maximum eigenvalue of } \sqrt{K_x} (I_n - O) \sqrt{K_y}.$$

2 Question 2 3 / 3

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