

Representer theorem for RKHS

$K: \mathcal{X} \times \mathcal{X} \rightarrow \mathbb{R}$ be a PDS kernel

H is its RKHS \hookrightarrow Moore-Aronson theorem

$g: \mathbb{R} \rightarrow \mathbb{R}$ increasing function

$l: \mathbb{R}^m \rightarrow \mathbb{R}$ loss function

$$\underset{h \in H}{\operatorname{argmin}} \underbrace{g(\|h\|_H)} + \underbrace{l(h(x_1), \dots, h(x_m))}_{\in \operatorname{span}\{\{K(x_i, \cdot) : i \in [m]\}\}}$$

Proof: $H_S = \left\{ \sum_{i=1}^m a_i K(x_i, \cdot) : a_i \in \mathbb{R}, x_i \in S \right\}$

Any $h \in H$ can be written as

$$h = h_0 + h_0^\perp \quad \text{where } h_0 \in H_S \text{ and } h_0^\perp \in H_S^\perp$$

By reproducing property,

$$h(x_i) = \langle h, k(x_i, \cdot) \rangle$$

$$= \langle h_0 + h_0^\perp, k(x_i, \cdot) \rangle$$

$$= \langle h_0, k(x_i, \cdot) \rangle + \underbrace{\langle h_0^\perp, k(x_i, \cdot) \rangle}_0$$

(h_0^\perp is ortho. to any function in H_S)

$$= \langle h_0, k(x_i, \cdot) \rangle$$

$$= h_0(x_i)$$

$$l(h(x_1), \dots, h(x_m)) = l(h_0(x_1), \dots, h_0(x_m))$$

$$\|h\| = \|h_0 + h_0^\perp\|$$

$$(\|h_0 + h_0^\perp\|^2 = \|h_0\|^2 + \|h_0^\perp\|^2 + 2\langle h_0, h_0^\perp \rangle)$$

$$\text{if } \|h\| > \|h_0\|, \text{ then } = \|h_0\|^2 + \|h_0^\perp\|^2$$

$$g(\|h\|) > g(\|h_0\|) \quad (g \text{ is increasing})$$

$$\Rightarrow \underset{h \in H}{\operatorname{argmin}} l(h(x_1), \dots, h(x_m)) + g(\|h\|)$$

$$= \underset{h \in H_S}{\operatorname{argmin}} l(h(x_1), \dots, h(x_m)) + g(\|h\|)$$

For ML Mohri 6.22

Anomaly detection applications

\rightarrow given a data point, is it an outlier?

e.g. climate, credit card fraud

without knowing Φ .

\rightarrow comparing data sets

$$a) \mathcal{L}(x, c, \alpha) = x^2 + \sum_{i=1}^m \alpha_i (\| \Phi(x_i) - c \|^2 - x^2)$$

KKT

$$1) \partial_x \mathcal{L}(x, c, \alpha) = 0 \quad \checkmark$$

$$2) \partial_c \mathcal{L}(x, c, \alpha) = 0 \quad \leftarrow$$

3) Complementarity

$$\sum_{i=1}^m \alpha_i (\| \Phi(x_i) - c \|^2 - x^2) = 0$$

$$\alpha_i \geq 0$$

For each i ,

$$\alpha_i (\| \Phi(x_i) - c \|^2 - x^2) = 0$$

From

$$1), \sum_{i=1}^m \alpha_i = 1$$

$$H_S = \left\{ \sum_{i=1}^m a_i \Phi(x_i) : a_i \in \mathbb{R}, x_i \in S \right\}$$

For any $c \in H$, $c = c_0 + c_0^\perp$

$$\| \Phi(x_i) - c_0 - c_0^\perp \|^2$$

$$= \| \Phi(x_i) - c_0 \|^2 + \| c_0^\perp \|^2$$

$$+ 2 \langle \underbrace{\Phi(x_i)}_{H_S} - \underbrace{c_0}_{H_S}, \underbrace{c_0^\perp}_{H_S^\perp} \rangle$$

$$= \| \Phi(x_i) - c_0 \|^2 + \| c_0^\perp \|^2$$

$$\underset{x, a \in \mathbb{R}^m}{\operatorname{argmin}} x^2 \text{ st. } \left(\| \Phi(x_i) - \sum_{j=1}^m a_j \Phi(x_j) \|^2 - x^2 \right) \leq 0$$

$$a = [a_1, \dots, a_m]^T$$