

Q2. $\min \frac{1}{2} \|w\|^2$ sub to. $y^t (w x^t) \geq 1, t=1, \dots, N$

(a) bias $b=0$,

$$x^{(1)} = (1, 1)^T, \quad x^{(2)} = (1, 0)^T;$$

$$y^{(1)} = 1, \quad y^{(2)} = -1;$$

$$w = ?, \quad r = ?$$

transfer to dual problem:

$$\min_a \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N a_i a_j y_i y_j (x_i \cdot x_j) - \sum_{i=1}^N a_i$$

$$= \frac{1}{2} (2a_1^2 + a_2^2 - 2a_1 a_2) - a_1 a_2$$

$$\text{s.t. } a_i \geq 0, \quad i=1, 2.$$

$$\min_a \quad a_1^2 + \frac{1}{2} a_2^2 - a_1 a_2 - a_1 - a_2 = A$$

$$\text{let } \frac{dA}{da_1} = 0 \quad \frac{dA}{da_2} = 0$$

$$a_1 = 2, \quad a_2 = 3,$$

$$a_1 > 0, \quad a_2 > 0, \quad \text{condition fulfilled}$$

$$w = \sum_i a_i y_i x_i = (-1, 2).$$

$$r = \frac{1}{\|w\|} = \frac{1}{\sqrt{5}}$$

Q2

(2) bias b can be non-zero:

transfer to dual problem:

$$\min_a \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N a_i a_j y_i y_j (x_i \cdot x_j) - \sum_{i=1}^N a_i$$

$$= \frac{1}{2} (2a_1^2 + a_2^2 - 2a_1 a_2) - a_1 a_2$$

$$\text{s.t. } a_i \geq 0, \quad i=1,2. \quad \underline{a_1 - a_2 = 0}$$

$$\min_a \quad \frac{1}{2} a_1^2 - 2a = B$$

$$\frac{dB}{da_1} = 0 \quad \frac{dB}{da_2} = 0$$

$$a_1 = 2 = a_2 > 0, \text{ condition satisfied}$$

$$w = \sum a_i y_i x_i = (0, 2)$$

for $a_j > 0$:

$$b = y_1 - \sum_{i=1}^N a_i y_i (x_i \cdot x_j) = -1$$

$$r = \frac{1}{\|w\|} = 1$$

Q3. 1. $k(x, z) = k_1(x, z) k_2(x, z)$

$$k_1(x, y) = a(x)^T a(y), \text{ where } a(z) = [a_1(z) \dots a_m(z)]$$

$$k_2(x, y) = b(x)^T b(y), \text{ where } b(z) = [b_1(z) \dots b_n(z)]$$

where $a(z), b(z)$ projects an n -d, m -d vector respectively.

$$k(x, z) = k_1(x, z) k_2(x, z)$$

$$= a(x)^T a(y) \cdot b(x)^T b(y)$$

$$= \sum_{m=1}^M \sum_{n=1}^N [a_m(x) b_n(x)] [a_m(y) b_n(y)]$$

$$\text{let } c_{mn}(z) = a_m(z) b_n(z)$$

where $C(z)$ projects an $m \times n$ d vector.

Hence, $k(x, z)$ is a kernel function.

2. $k(x, z) = a k_1(x, z) + b k_2(x, z), a, b > 0, \in \mathbb{R}$

By Mercer's Theorem, if K is semi-definite, positive, symmetric continuous function,

there exists $\{\phi_i\}_{i \in \mathbb{N}}, \{\lambda_i\}_{i \in \mathbb{N}}$ that $k(x, z) = \sum_{i=1}^{\infty} \lambda_i \phi_i(x) \phi_i(z)$, a kernel func.

Since $k_1(x, z), k_2(x, z)$ are kernel func, k_1, k_2 are symmetric, positive, semi-definitive.

Now we need to prove k is also semi-definitive, positive, symmetric.

Since k_1, k_2 is symmetric, a, b is constant

$a k_1, b k_2$ is symmetric.

By theorem of symmetric matrix, the sum of 2 symmetric matrix is still symmetric:

proof: $(A+B)^T = A^T + B^T = (A+B)$

Hence: $k(x, z)$ is symmetric.

since a, b are ~~not~~ positive real, $a k_1(x, z)$, $b k_2(x, z)$ ^{also} semi-positive

hence $h(a k_1(x, z)) h^T + h(b k_2(x, z)) h^T$

$= h(k_1(x, z) + k_2(x, z)) h^T.$

Hence. $a k_1(x, z) + b k_2(x, z) = k(x, z)$ ^{positive semi-definite}

hence, k is kernel func.

3. When $a=1$, $b=2$, $k_1=k_2$

$k(x, z) = -k_2(x, z)$

k now has semi-negative kernel matrix
by counter example

k is not.

4. by kernel definition,

$k(x, z) = \phi(x) \phi(z)$, when $\phi(z): \mathbb{R}^n \rightarrow \mathbb{R}_1$,

if $f: \mathbb{R}^n \rightarrow \mathbb{R}$,

$k(x, z) = f(x) f(z)$, is a kernel