

4. Constructing Kernels

⇒ Choosing a kernel $K(x, z) = \phi(x)^T \phi(z)$, we can implicitly map data to a high dimensional space.

↳ One way to generate kernel is to explicitly define the mapping ϕ to a higher dimensional space, and then work with the corresponding K .

⇒ Here are interested in directly constructing kernels.

That we think gives an appropriate similarity measure for our learning problem.

*Let:

→ K_1, K_2 are kernels over $\mathbb{R}^n \times \mathbb{R}^n$

→ $a \in \mathbb{R}^+$

→ $f: \mathbb{R}^n \mapsto \mathbb{R}$

→ $\phi: \mathbb{R}^n \rightarrow \mathbb{R}^d$

→ K_3 be a kernel over $\mathbb{R}^d \times \mathbb{R}^d$

→ $P(x) \rightarrow$ Polynomial over x with positive coefficients.

Ⓚ For each K , state whether it is necessarily a kernel.

②

$$K(x, z) = K_1(x, z) + K_2(x, z)$$

$$\downarrow \qquad \qquad \downarrow$$

$$\phi_1^T(x) \phi_1(z) + \phi_2^T(x) \phi_2(z)$$

$$K(x, z) = K(z, x)$$

$$K(x, x) = K_1(x, x) + K_2(x, x) \geq 0$$

\Rightarrow Let K^i be same matrix for kernel K_i .

$$K = K^1 + K^2 \quad \{ \text{it is symmetric} \}$$

$$z^T K z = z^T K^1 z + z^T K^2 z \geq 0 \quad \left\{ \begin{array}{l} \text{Positive Semi} \\ \text{definite} \end{array} \right\}$$

~~and K^1, K^2 are also positive semi-definite~~

\Rightarrow Hence K is a **Valid Kernel**.

⑥ $K(x, z) = K_1(x, z) - K_2(x, z)$

$$K(z, x) = K^1(z, x) - K^2(z, x)$$

$$= K^1(x, z) - K^2(x, z)$$

$$= K(x, z) \quad \{ \text{Hence symmetric} \}$$

$$z^T K z = z^T K^1 z - z^T K^2 z$$

$\{ \text{Not Positive Semi-definite} \}$

⑦ $K(x, z) = a K_1(x, z)$

$$K(z, x) = a K^1(z, x) = a K^1(x, z) = K(x, z) \quad \{ \text{Symmetric} \}$$

$$z^T K z = z^T a K^1 z = a z^T K^1 z \quad \left\{ \begin{array}{l} \text{as } a \in \mathbb{R}^+, \text{ it is} \\ \text{Positive Semi-definite} \end{array} \right\}$$

Valid Kernel

③

$$d) K(x, z) = -a K_1(x, z)$$

Similar to c), but as $a \in \mathbb{R}^+$, $-az^T K_1 z$ is not positive semidefinite.

\Rightarrow Hence K is ^{Not a} Valid Kernel.

$$e) K(x, z) = K_1(x, z) K_2(x, z)$$

$$K(z, x) = K_1^T(z, x) K_2^T(z, x)$$

$$= K_1^T(x, z) K_2^T(x, z)$$

$$= K(x, z) \quad \left\{ \text{Symmetric} \right\}$$

$$z^T K z = z^T K_1 K_2 z$$

$$= \sum_{i,j} z_i \phi_1'(x^{(i)})^T \phi_1'(x^{(j)}) \phi_2^2(x^{(i)})^T \phi_2^2(x^{(j)}) z_j$$

$$= \sum_{i,j} z_i \sum_{k=1}^m \phi_1'(x^{(i)})_k \phi_1'(x^{(j)})_k \sum_{p=1}^m \phi_2^2(x^{(i)})_p \phi_2^2(x^{(j)})_p z_j$$

$$= \sum_{k=1}^m \sum_{p=1}^m \left(\sum_{i=1}^m z_i \phi_1'(x^{(i)})_k \phi_2^2(x^{(i)})_p \right) \left(\sum_{j=1}^m z_j \phi_1'(x^{(j)})_k \phi_2^2(x^{(j)})_p \right)$$

$$= \sum_{k=1}^m \sum_{p=1}^m \left(\sum_{i=1}^m z_i \phi_1'(x^{(i)})_k \phi_2^2(x^{(i)})_p \right)^2 \geq 0$$

$\left\{ \begin{array}{l} \text{It is positive} \\ \text{semidefinite} \end{array} \right\}$

Valid Kernel

⑧ $K(x, z) = f(x) f(z)$

$$\begin{aligned} K(z, x) &= f(z) f(x) \\ &= f(x) f(z) \\ &= K(x, z) \quad \{ \text{Symmetric} \} \end{aligned}$$

$$z^T K z = \sum_{j,i} z_i f(x^{(i)}) f(x^{(i)}) z_j = (z^T x)^2$$

valid kernel

$\{ \text{If } x_i = f(x^{(i)}) \}$

⑨ $K(x, z) = K_3(\phi(x), \phi(z))$

~~valid kernel~~ valid kernel

⑩ $K(x, z) = P(K_1(x, z))$

~~valid kernel~~

$$P(K_1) = \underbrace{b_m}_{b_m} K_1^m + \underbrace{b_{m-1}}_{b_{m-1}} K_1^{m-1} + \dots + \underbrace{b_1}_{b_1} K_1$$

* Power of kernel is valid kernel

* aK is valid kernel

* $K_1 + K_2$ is valid kernel

So $P(K_1)$ is valid kernel