

# Necessary condition for matrix-valued kernels

Miaoyan Wang, April 26, 2020

**Theorem 0.1** (Necessary condition). *Suppose  $\mathbf{K}: \mathbb{R}^{d' \times d} \times \mathbb{R}^{d' \times d} \mapsto \mathbb{R}^{d \times d}$  is a function that takes as input a pair of matrices and produces a matrix. Let  $\{\mathbf{X}_i \in \mathbb{R}^{d' \times d}: i \in [n]\}$  denote a set of input matrices, and let  $\mathcal{K}$  denote an order-4  $(n, n, d, d)$ -dimensional tensor,*

$$\mathcal{K} = \llbracket \mathcal{K}(i, i', p, p') \rrbracket, \quad \text{where } \mathcal{K}(i, i', p, p') \text{ is the } (p, p')\text{-th entry of the matrix } \mathbf{K}(\mathbf{X}_i, \mathbf{X}_{i'}).$$

*Then, the factorization  $\mathbf{K}(\mathbf{X}_i, \mathbf{X}_{i'}) = \mathbf{h}(\mathbf{X}_i)^T \mathbf{h}(\mathbf{X}_{i'})$  exists for some mapping  $\mathbf{h}$ , only if both of the following conditions hold:*

- (1) *For every index  $i \in [n]$ , the matrix  $\mathcal{K}(i, i, :, :) \in \mathbb{R}^{d \times d}$  is positive semidefinite.*
- (2) *For every index  $p \in [d]$ , the matrix  $\mathcal{K}(:, :, p, p) \in \mathbb{R}^{n \times n}$  is positive semidefinite.*

*Proof.* (1) Let  $i \in [n]$  be a fixed index. For any vector  $\mathbf{a} \in \mathbb{R}^d$ ,

$$\mathbf{a}^T \mathcal{K}(i, i, :, :) \mathbf{a} = \mathbf{a}^T \mathbf{h}(\mathbf{X}_i)^T \mathbf{h}(\mathbf{X}_i) \mathbf{a} = \langle \mathbf{h}(\mathbf{X}_i) \mathbf{a}, \mathbf{h}(\mathbf{X}_i) \mathbf{a} \rangle = \|\mathbf{h}(\mathbf{X}_i) \mathbf{a}\|_2^2 \geq 0$$

- (2) Let  $p \in [d]$  be a fixed index. We use  $[\cdot]_{(k,p)}$  to denote the  $(k, p)$ -th entry of the matrix. For any vector  $\mathbf{b} = (b_1, \dots, b_n)^T \in \mathbb{R}^n$ ,

$$\begin{aligned} \mathbf{b}^T \mathcal{K}(:, :, p, p) \mathbf{b} &= \sum_{ij} b_i b_j [\mathbf{h}(\mathbf{X}_i)^T \mathbf{h}(\mathbf{X}_j)]_{(p,p)} \\ &= \sum_{ij} b_i b_j \sum_k [\mathbf{h}(\mathbf{X}_i)]_{(k,p)} [\mathbf{h}(\mathbf{X}_j)]_{(k,p)} \\ &= \sum_k \left( \sum_i [\mathbf{h}(\mathbf{X}_i)]_{(k,p)} b_i \right) \left( \sum_j [\mathbf{h}(\mathbf{X}_j)]_{(k,p)} b_j \right) \\ &= \sum_k \left( \sum_i [\mathbf{h}(\mathbf{X}_i)]_{(k,p)} b_i \right)^2 \geq 0. \end{aligned}$$

□

Updated on April 29, 2020. Generalization of Mercer's theorem to matrix-valued predictors.

**Definition 1** (Validity and Admissibility). We call the matrix-valued kernel  $\mathbf{K}$  a valid kernel if there exists a feature mapping  $\mathbf{h}$  such that  $\mathbf{K}(\mathbf{X}, \mathbf{X}') = \mathbf{h}(\mathbf{X}) \mathbf{h}^T(\mathbf{X}')$  for all  $\mathbf{X}, \mathbf{X}' \in \mathbb{R}^{d \times d'}$ . We call  $\mathbf{K}$  an admissible kernel if the equality holds under the trace operation; i.e.,  $\text{tr} [\mathbf{K}(\mathbf{X}, \mathbf{X}')] = \text{tr} [\mathbf{h}(\mathbf{X}) \mathbf{h}^T(\mathbf{X}')] = \|\mathbf{h}(\mathbf{X})\|_F^2$  for all  $\mathbf{X}, \mathbf{X}' \in \mathbb{R}^{d \times d'}$ .

**Theorem 0.2** (Characterization of admissible kernels). *Let  $\mathbf{K}: \mathbb{R}^{d' \times d} \times \mathbb{R}^{d' \times d} \mapsto \mathbb{R}^{d \times d}$  be a function that takes as input a pair of matrices and produces a matrix. Define a function  $\mathcal{F}: \mathbb{R}^{d' \times d} \times \mathbb{R}^{d' \times d} \mapsto \mathbb{R}$  as follows:*

$$\mathcal{F}(\mathbf{X}, \mathbf{X}') = \text{tr}[\mathbf{K}(\mathbf{X}, \mathbf{X}')], \text{ for all } \mathbf{X}, \mathbf{X}' \in \mathbb{R}^{d' \times d}.$$

*Then, the following two statements are equivalent:*

1. *The function  $\mathbf{K}$  is an admissible kernel.*
2. *The function  $\mathcal{F}$  is positive semidefinite.*

**Remark 1.** Recall that earlier we have defined two types of kernel  $\mathbf{K}$ :

- Hadamard-product type:  $\mathbf{K}(\mathbf{X}, \mathbf{X}') = \underbrace{(\mathbf{X}^T \mathbf{X}' + \mathbf{1}\mathbf{1}^T) \circ \cdots \circ (\mathbf{X}^T \mathbf{X}' + \mathbf{1}\mathbf{1}^T)}_{d \text{ times}}.$
- Matrix-polynomial type:  $\mathbf{K}(\mathbf{X}, \mathbf{X}') = (\mathbf{X}^T \mathbf{X}' + \mathbf{1}\mathbf{1}^T)^d.$

Theorem 0.2 provides a practical way to verify the non-existence of feature mapping for a given  $\mathbf{K}$ . Note that being admissible is a necessary condition for validity. Straightforward calculation shows that  $\mathcal{F}$  defined by the matrix-polynomial  $\mathbf{K}$  is not positive semidefinite, so the kernel  $\mathbf{K}$  is non-valid.