Necessary condition for matrix-valued kernels

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

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

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

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

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

(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 $\mathbf{b} = (b_1, \dots, b_n)^T \in \mathbb{R}^n$,

$$\begin{aligned} \boldsymbol{b}^{T} \mathcal{K}(:,:,p,p) \boldsymbol{b} &= \sum_{ij} b_{i} b_{j} \left[\boldsymbol{h}(\boldsymbol{X}_{i})^{T} \boldsymbol{h}(\boldsymbol{X}_{j}) \right]_{(p,p)} \\ &= \sum_{ij} b_{i} b_{j} \sum_{k} \left[\boldsymbol{h}(\boldsymbol{X}_{i}) \right]_{(k,p)} \left[\boldsymbol{h}(\boldsymbol{X}_{j}) \right]_{(k,p)} \\ &= \sum_{k} \left(\sum_{i} \left[\boldsymbol{h}(\boldsymbol{X}_{i}) \right]_{(k,p)} b_{i} \right) \left(\sum_{j} \left[\boldsymbol{h}(\boldsymbol{X}_{j}) \right]_{(k,p)} b_{j} \right) \\ &= \sum_{k} \left(\sum_{i} \left[\boldsymbol{h}(\boldsymbol{X}_{i}) \right]_{(k,p)} b_{i} \right)^{2} \geq 0. \end{aligned}$$

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

Theorem 0.2 (Necessary and Sufficient condition (sample version)). Suppose $K: \mathbb{R}^{d' \times d} \times \mathbb{R}^{d' \times d} \mapsto \mathbb{R}^{d \times d}$ is a function that takes as input a pair of matrices and produces a matrix. Let $\mathcal{X} = \{X_i \in \mathbb{R}^{d' \times d}: i \in [n]\}$ denote a set of input matrices, and let \mathcal{K} denote an order-4 (n, n, d, d)-dimensional tensor,

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

Then, the following two statements are equivalent:

- There exists a feature mapping $h(\cdot)$ over the input set \mathcal{X} such that $K(X_i, X_{i'}) = h(X_i)^T h(X_{i'})$.
- The matrix $\mathcal{K}_{(13)(24)} \in \mathbb{R}^{nd \times nd}$ is positive semidefinite. Here $\mathcal{K}_{(13)(24)}$ denotes the square unfolding of the tensor \mathcal{K} .

Remark 1. Recall that earlier we have defined two types of kernel K:

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

Theorem 0.2 implies that the existence of feature mapping can be verified using the positive-definiteness of $\mathcal{K}_{(13)(24)}$. Straightforward calculation shows that the Hadamard-product type kernel K is valid whereas the matrix-polynomial type kernel is not.

Proof of Theorem 0.2. " \Leftarrow " Suppose $\mathcal{K}_{(13)(24)} \in \mathbb{R}^{nd \times nd}$ is a positive semidefinite matrix. Then there exists a matrix $\mathbf{A} \in \mathbb{R}^{m \times (nd)}$ such that

$$\mathcal{K}_{(13)(24)} = \mathbf{A}^T \mathbf{A}.\tag{1}$$

We reshape A into an (m, n, d)-dimensional tensor, and with a little abuse of notation, we still use A to denote the resulting object. We define the feature mapping $h: \mathbb{R}^{d' \times d} \mapsto \mathbb{R}^{m \times d}$ as follows:

$$m{h} \colon \mathbb{R}^{d' \times d} \mapsto \mathbb{R}^{m \times d}$$

 $m{X}_i \mapsto m{h}(m{X}_i) \stackrel{ ext{def}}{=} m{A}(:,i,:), \quad ext{for all } i \in [n].$

The defined mapping satisfies that, for all $i, i' \in [n]$,

$$h(\boldsymbol{X}_i)^T h(\boldsymbol{X}_{i'}) = \boldsymbol{A}(:,i,:)^T \boldsymbol{A}(:,i',:)$$
$$= \mathcal{K}(i,:,i',:)$$
$$= \boldsymbol{K}(\boldsymbol{X}_i,\boldsymbol{X}_{i'}),$$

where the second line follows form (1) and the third line follows from the definition of \mathcal{K} .

" \Rightarrow " Let $\boldsymbol{a} = (a_{11}, \dots, a_{1d}, a_{21}, \dots, a_{2d}, \dots, a_{nd})^T$ denote an arbitrary vector in \mathbb{R}^{nd} . Note that

$$\mathbf{a}^{T} \mathcal{K}_{(13)(24)} \mathbf{a} = \sum_{i,i',p,p'} a_{ip} \mathcal{K}(i,i',p,p') a_{i'p'}$$

$$= \sum_{i,i',p,p'} a_{ip} \left[\mathbf{K}(\mathbf{X}_{i},\mathbf{X}_{i'}) \right]_{pp'} a_{i'p'}$$

$$= \sum_{i,i',p,p'} a_{ip} \left(\sum_{k} \left[\mathbf{h}(\mathbf{X}_{i})^{T} \right]_{pk} \left[\mathbf{h}(\mathbf{X}_{i'}) \right]_{kp'} \right) a_{i'p'}$$

$$= \sum_{k} \left(\sum_{i,p} a_{ip} \left[\boldsymbol{h}(\boldsymbol{X}_{i}) \right]_{kp} \right) \left(\sum_{i',p'} a_{i'p'} \left[\boldsymbol{h}(\boldsymbol{X}_{i'}) \right]_{kp'} \right)$$
$$= \sum_{k} \left(\sum_{i,p} a_{ip} \left[\boldsymbol{h}(\boldsymbol{X}_{i}) \right]_{kp} \right)^{2} \ge 0.$$

Therefore, $\mathcal{K}_{(13)(24)}$ is positive semi-definite.