

Positive Definite Multi-Kernels for Scattered Data Interpolations

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Abstract

In this article, we use the knowledge of positive definite tensors to develop a concept of positive definite multi-kernels to construct the kernel-based interpolants of scattered data. By the theorems of reproducing kernel Banach spaces, the optimal recoveries and error analysis of the kernel-based interpolants are shown for a special class of strictly positive definite multi-kernels.

Keywords: Kernel-based approximation method, scattered data interpolation, positive definite multi-kernel, reproducing kernel Banach space, positive definite tensor, multi-linear system

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1. Introduction

In many areas of practical applications, we often face a problem of reconstructing an unknown function f from the scattered data. The scattered data consist of high-dimension data points $\{\mathbf{x}_i\}_{i=1}^n$ and data values $\{y_i\}_{i=1}^n$ such that $y_i = f(\mathbf{x}_i)$ for all $i = 1, 2, \dots, n$. The reconstruction is to find an estimate function s to approximate f . Generally, s is sought to interpolate the scattered data, that is, $s(\mathbf{x}_i) = y_i$ for all $i = 1, 2, \dots, n$. It is well-known that the kernel-based approximation method is a fundamental approach of scattered data interpolations. The classical kernel-based approximation method is mainly dependent of the positive definite kernels. Here, we develop a concept of positive definite multi-kernels in Definition 3.1 which is a generalization of positive definite kernels. The positive definite multi-kernels can be also used to reconstruct f from the scattered data. For examples, we compare

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the interpolations by a positive definite kernel $K_2 : \otimes_{k=1}^2 \mathbb{R}^d \rightarrow \mathbb{R}$ and a positive definite multi-kernel $K_4 : \otimes_{k=1}^4 \mathbb{R}^d \rightarrow \mathbb{R}$. The classical interpolant s_2 is composed of a kernel basis

$$K_2(\cdot, \mathbf{x}_i), \quad \text{for } i = 1, 2, \dots, n,$$

that is,

$$s_2(\mathbf{x}) := \sum_{i=1}^n c_i K_2(\mathbf{x}, \mathbf{x}_i), \quad \text{for } \mathbf{x} \in \mathbb{R}^d,$$

where the coefficients c_1, c_2, \dots, c_n are solved by a linear system

$$\sum_{i_2=1}^n K_2(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}) c_{i_2} = y_{i_1}, \quad \text{for } i_1 = 1, 2, \dots, n.$$

The new interpolant s_4 is composed of another kernel basis

$$K_4(\cdot, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \mathbf{x}_{i_3}), \quad \text{for } i_1, i_2, i_3 = 1, 2, \dots, n,$$

that is,

$$s_4(\mathbf{x}) := \sum_{i_1, i_2, i_3=1}^{n, n, n} c_{i_1} c_{i_2} c_{i_3} K_4(\mathbf{x}, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \mathbf{x}_{i_3}), \quad \text{for } \mathbf{x} \in \mathbb{R}^d,$$

where the coefficients c_1, c_2, \dots, c_n are solved by a multi-linear system

$$\sum_{i_2, i_3, i_4=1}^{n, n, n} K_4(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \mathbf{x}_{i_3}, \mathbf{x}_{i_4}) c_{i_2} c_{i_3} c_{i_4} = y_{i_1}, \quad \text{for } i_1 = 1, 2, \dots, n.$$

In this article, we mainly discuss how to use a special class of strictly positive definite multi-kernel Φ_m in Equation (3.3) to construct the kernel-based interpolant s_m in Theorem 4.1 which shows that the related multi-linear system exists the unique solution. By the theorems of reproducing kernel Banach spaces in [5], we can obtain the advanced properties of s_m including optimal recoveries in Theorem 4.3 and error analysis in Theorems 4.5 and 4.6.

2. Positive Definite Tensors and Multi-Linear Systems

In this section, we review the theory of positive definite tensors which will be used to define the positive definite multi-kernels in Section 3. For convenience of the readers, the notations and operations of tensors are defined as in the book [2]. We say $T_{m,n}$ a collection of all m th order n th dimensional real tenors, for example,

$T_{3,n} = \mathbb{R}^{n \times n \times n}$, $T_{2,n} = \mathbb{R}^{n \times n}$, and $T_{1,n} = \mathbb{R}^n$. For $\mathbf{A}_m \in T_{m,n}$, we say \mathbf{A}_m a symmetric tensor if all entries $a_{i_1 \dots i_m}$ of \mathbf{A}_m are invariant under any permutation of the indices. For $\mathbf{v} \in \mathbb{R}^n$, the tensor outer product is defined as

$$\mathbf{v}^{\otimes m} = \mathbf{v} \otimes \dots \otimes \mathbf{v} := (v_{i_1} \dots v_{i_m}) \in T_{m,n}.$$

For $\mathbf{c} \in \mathbb{R}^n$, the $(m-1)$ -mode and m -mode products are defined as

$$\mathbf{A}_m \mathbf{c}^{m-1} := \left(\sum_{i_2, \dots, i_m=1}^{n, \dots, n} a_{i_1 i_2 \dots i_m} c_{i_2} \dots c_{i_m} \right)_{i_1=1}^n \in \mathbb{R}^n,$$

and

$$\mathbf{A}_m \mathbf{c}^m := \sum_{i_1, \dots, i_m=1}^{n, \dots, n} a_{i_1 \dots i_m} c_{i_1} \dots c_{i_m} \in \mathbb{R},$$

respectively. For $\mathbf{A}_m \in T_{m,n}$, we say \mathbf{A}_m a *semi-positive definite tensor* if

$$\mathbf{A}_m \mathbf{c}^m \geq 0, \quad \text{for all } \mathbf{c} \in \mathbb{R}^n.$$

If further $\mathbf{A}_m \mathbf{c}^m = 0$ if and only if $\mathbf{c} = 0$, then \mathbf{A}_m is said a *positive definite tensor* same as in [3, Definition 4.1]. Specially, the positive definite tensor \mathbf{A}_2 is a positive definite matrix.

For any $\mathbf{b} \in \mathbb{R}^n$, we look at the multi-linear system

$$\mathbf{A}_m \mathbf{c}^{m-1} = \mathbf{b}. \tag{2.1}$$

If \mathbf{A}_2 is a symmetric positive definite matrix, then Equation (2.1) has a unique solution $\mathbf{c} \in \mathbb{R}^n$. For the general cases, we will study the solutions of Equation (2.1) by the theorems of the tensor variational inequality (TVI) in [3]. Let

$$\mathcal{S}(\mathbf{A}_m, \mathbf{b}) := \left\{ \mathbf{c} \in \mathbb{R}^n : (\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{b}) \geq 0, \text{ for all } \mathbf{d} \in \mathbb{R}^n \right\}.$$

Then $\mathcal{S}(\mathbf{A}_m, \mathbf{b})$ is said the solutions set of TVI with \mathbb{R}^n , \mathbf{A}_m , and $-\mathbf{b}$. If \mathbf{A}_m is a positive definite tensor, then [3, Theorem 4.2] guarantees that $\mathcal{S}(\mathbf{A}_m, \mathbf{b})$ is nonempty and compact.

Proposition 2.1. *If \mathbf{A}_m is a positive definite tensor, then the solutions set of Equation (2.1) is nonempty and compact.*

Proof. If we verify that $\mathcal{S}(\mathbf{A}_m, \mathbf{b})$ is equal to the solutions set of Equation (2.1), then the proof is completed by [3, Theorem 4.2].

We first suppose that \mathbf{c} is a solution of Equation (2.1). Thus, we conclude from $\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{b} = 0$ that $(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{b}) = 0$ for all $\mathbf{d} \in \mathbb{R}^n$, hence that $\mathbf{c} \in \mathcal{S}(\mathbf{A}_m, \mathbf{b})$.

In the other side, we suppose that $\mathbf{c} \in \mathcal{S}(\mathbf{A}_m, \mathbf{b})$. Let $\mathbf{v} := \mathbf{c} - \mathbf{d}$ and $\mathbf{u} := \mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{b}$. Thus, $\mathbf{v}^T \mathbf{u} \geq 0$ for any $\mathbf{v} \in \mathbb{R}^n$. Since the range of \mathbf{v} is equal to \mathbb{R}^n , we take $\mathbf{v} := -\mathbf{u}$ such that $0 \geq -\|\mathbf{u}\|_2^2 = (-\mathbf{u})^T \mathbf{u} \geq 0$. This shows that $\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{b} = \mathbf{u} = 0$. Therefore, \mathbf{c} is a solution of Equation (2.1). \square

Same as in [3, Definition 4.1], we say \mathbf{A}_m a *strictly positive definite tensor* if $(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{A}_m \mathbf{d}^{m-1}) > 0$ for all $\mathbf{c}, \mathbf{d} \in \mathbb{R}^n$ with $\mathbf{c} \neq \mathbf{d}$. Obviously, if \mathbf{A}_m is a strictly positive definite tensor, then \mathbf{A}_m is a positive definite tensor. Specially, the positive definite matrix \mathbf{A}_2 is a strictly positive definite tensor. If \mathbf{A}_m is a strictly positive definite tensor, then [3, Theorem 4.2] guarantees that $\mathcal{S}(\mathbf{A}_m, \mathbf{b})$ is singleton.

Proposition 2.2. *If \mathbf{A}_m is a strictly positive definite tensor, then Equation (2.1) exists a unique solution.*

Proof. As in the proof of Proposition 2.1, $\mathcal{S}(\mathbf{A}_m, \mathbf{b})$ is equal to the solutions set of Equation (2.1). Thus, the proof is completed by [3, Theorem 4.2]. \square

In [1, 6], there are many kinds of multi-linear systems such as M -tensors, Z^+ -tensors, and P -tensors to introduce the same results. In this article, we only discuss the positive definite tensors for the scattered data interpolations.

3. Positive Definite Multi-Kernels and Reproducing Kernel Banach Spaces

In this section, we develop a concept of positive definite multi-kernels. Let a domain $\Omega \subseteq \mathbb{R}^d$. By [4, Definition 6.24], a symmetric kernel $K_2 : \otimes_{k=1}^2 \Omega \rightarrow \mathbb{R}$ is said a positive definite kernel if, for all $n \in \mathbb{N}$ and all pairwise distinct points $\{\mathbf{x}_i\}_{i=1}^n \subseteq \Omega$, the quadratic form

$$\sum_{i_1, i_2=1}^{n, n} K_2(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}) c_{i_1} c_{i_2},$$

for all $\mathbf{c} := (c_1, c_2, \dots, c_n)^T \in \mathbb{R}^n \setminus \{0\}$. Let $\mathbf{A}_2 := (K_2(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}))_{i_1, i_2=1}^{n, n} \in T_{2, n}$. Thus, \mathbf{A}_2 is a symmetric positive definite matrix.

A multi-kernel of order $m \in \mathbb{N}$ is defined as $K_m : \otimes_{k=1}^m \Omega \rightarrow \mathbb{R}$ for $m \in \mathbb{N}$. We say K_m a *symmetric multi-kernel* if $K_m(\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_m) = K_m(\mathbf{z}_{i_1}, \mathbf{z}_{i_2}, \dots, \mathbf{z}_{i_m})$ for all permutation of the pairwise distinct indices $i_1, i_2, \dots, i_m \in \{1, 2, \dots, m\}$.

Definition 3.1. A symmetric multi-kernel K_m is said a *semi-positive definite multi-kernel* if, for all $n \in \mathbb{N}$ and all pairwise distinct points $\{\mathbf{x}_i\}_{i=1}^n \subseteq \Omega$, the multiple product form

$$\sum_{i_1, i_2, \dots, i_m=1}^{n, n, \dots, n} K_m(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_m}) c_{i_1} c_{i_2} \dots c_{i_m} \geq 0,$$

for all $\mathbf{c} := (c_1, c_2, \dots, c_n)^T \in \mathbb{R}^n$. If further the multiple product is equal to 0 if and only if $\mathbf{c} = 0$, then K_m is said a *positive definite multi-kernel*.

A symmetric multi-kernel K_m is said a *strictly positive definite multi-kernel* if, for all $n \in \mathbb{N}$ and all pairwise distinct points $\{\mathbf{x}_i\}_{i=1}^n \subseteq \Omega$, the multiple product form

$$\sum_{i_1, i_2, \dots, i_m=1}^{n, n, \dots, n} K_m(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_m}) (c_{i_1} - d_{i_1}) (c_{i_2} \dots c_{i_m} - d_{i_2} \dots d_{i_m}) > 0,$$

for all $\mathbf{c} := (c_1, c_2, \dots, c_n)^T, \mathbf{d} := (d_1, d_2, \dots, d_n)^T \in \mathbb{R}^n$ with $\mathbf{c} \neq \mathbf{d}$.

Obviously, the symmetric strictly positive definite multi-kernel K_m is a symmetric positive definite multi-kernel. Specially, the symmetric positive definite multi-kernel K_2 is a symmetric positive definite kernel and a symmetric strictly positive definite multi-kernel. Given any pairwise distinct points $\{\mathbf{x}_i\}_{i=1}^n \subseteq \Omega$, we define

$$\mathbf{A}_m := (K_m(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_m}))_{i_1, i_2, \dots, i_m=1}^{n, n, \dots, n} \in T_{m, n}. \quad (3.1)$$

Proposition 3.2. *If K_m is a symmetric (strictly) positive definite multi-kernel, then \mathbf{A}_m in Equation (3.1) is a symmetric (strictly) positive definite tensor.*

Proof. The proof is straightforward by the definitions of (strictly) positive definite multi-kernels and (strictly) positive definite tensors. \square

Now we look at a special class of multi-kernels related to the p -norm reproducing kernel Banach spaces in [4]. Same as in [4, Section 10.4], we suppose that the domain Ω is *compact* and the symmetric positive definite kernel Φ_2 is *continuous*. Thus, the Mercer theorem assures that Φ_2 possesses the absolutely uniformly convergent representation

$$\Phi_2(\mathbf{z}_1, \mathbf{z}_2) = \sum_{k=1}^{\infty} \lambda_k e_k(\mathbf{z}_1) e_k(\mathbf{z}_2), \quad \text{for } \mathbf{z}_1, \mathbf{z}_2 \in \Omega,$$

where $\{\lambda_k : k \in \mathbb{N}\}$ and $\{e_k : k \in \mathbb{N}\}$ are the related positive eigenvalues and continuous eigenfunctions of Φ_2 , respectively. Let $\phi_k := \lambda_k^{1/2} e_k$ for all $k \in \mathbb{N}$. Same as [5, Section 4.2], we further suppose that

$$\sum_{k=1}^{\infty} |\phi_k(\mathbf{x})| < \infty, \quad \text{for all } \mathbf{x} \in \Omega. \quad (3.2)$$

Thus, by [5, Theorem 5.10], Equation (3.2) assures that the special multi-kernel

$$\Phi_m(\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_m) := \sum_{k=1}^{\infty} \prod_{i=1}^m \phi_k(\mathbf{z}_i), \quad \text{for } \mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_m \in \Omega, \quad (3.3)$$

is well-defined for all $m \geq 2$.

Remark 3.3. In this article, the positive definite multi-kernels are consistent with the positive definite tensors in [2, 3]. The semi-positive definite kernels discussed here have the same meaning as the positive definite kernels in [5]. For convenience, we will *always* suppose that K_2 is a continuous symmetric positive definite kernel on the compact domain Ω . Moreover, the eigenvalues and eigenfunctions of Φ_2 are fixed and satisfy the condition in Equation (3.2) such that Φ_m is fixed and well-defined in the following discussions.

Proposition 3.4. *If m is a positive even integer and $K_m := \Phi_m$, then \mathbf{A}_m in Equation (3.1) is a symmetric strictly positive definite tensor.*

Proof. Equation (3.3) shows that Φ_m is a symmetric multi-kernel. Thus, \mathbf{A}_m is a symmetric tensor. We take any $\mathbf{c}, \mathbf{d} \in \mathbb{R}^n$ with $\mathbf{c} \neq \mathbf{d}$. By the definition of strictly positive definite tensors, if we prove that $(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{A}_m \mathbf{d}^{m-1}) > 0$, then the proof is complete.

Let $\mathbf{v}_k := (\phi_k(\mathbf{x}_1), \phi_k(\mathbf{x}_2), \dots, \phi_k(\mathbf{x}_n))^T$ for all $k \in \mathbb{N}$. We conclude from Equation (3.3) that $\mathbf{A}_m = \sum_{k=1}^{\infty} \mathbf{v}_k^{\otimes m}$, hence that $\mathbf{A}_m \mathbf{c}^{m-1} = \sum_{k=1}^{\infty} (\mathbf{v}_k^T \mathbf{c})^{m-1} \mathbf{v}_k$ and $\mathbf{A}_m \mathbf{d}^{m-1} = \sum_{k=1}^{\infty} (\mathbf{v}_k^T \mathbf{d})^{m-1} \mathbf{v}_k$. Let $\alpha_k := \mathbf{v}_k^T \mathbf{c}$ and $\beta_k := \mathbf{v}_k^T \mathbf{d}$ for all $k \in \mathbb{N}$. Thus,

$$(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{A}_m \mathbf{d}^{m-1}) = \sum_{k=1}^{\infty} (\alpha_k - \beta_k) (\alpha_k^{m-1} - \beta_k^{m-1}).$$

Now we verify that $(\alpha_k - \beta_k) (\alpha_k^{m-1} - \beta_k^{m-1})$ is nonnegative for all $k \in \mathbb{N}$. Since m is even, $m - 1$ is odd. If $\alpha_k \geq 0$ and $\beta_k \geq 0$, then $\alpha_k - \beta_k$ and $\alpha_k^{m-1} - \beta_k^{m-1}$ have the same sign. If $\alpha_k \geq 0$ and $\beta_k \leq 0$, then $\alpha_k - \beta_k \geq 0$ and $\alpha_k^{m-1} - \beta_k^{m-1} \geq 0$.

0. Thus, $(\alpha_k - \beta_k)(\alpha_k^{m-1} - \beta_k^{m-1}) \geq 0$. Moreover, since $(\alpha_k - \beta_k)(\alpha_k^{m-1} - \beta_k^{m-1}) = (\beta_k - \alpha_k)(\beta_k^{m-1} - \alpha_k^{m-1})$, the assertion also holds. Therefore, we conclude that $(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{A}_m \mathbf{d}^{m-1}) \geq 0$.

Since Φ_2 is a symmetric positive definite kernel, \mathbf{A}_2 is a symmetric positive definite matrix. Thus, we conclude from $\sum_{k=1}^{\infty} (\alpha_k - \beta_k)^2 = (\mathbf{c} - \mathbf{d})^T (\mathbf{A}_2 \mathbf{c} - \mathbf{A}_2 \mathbf{d}) = \mathbf{A}_2 (\mathbf{c} - \mathbf{d})^2 > 0$ that $\alpha_k - \beta_k \neq 0$ such that $(\alpha_k - \beta_k)(\alpha_k^{m-1} - \beta_k^{m-1}) \neq 0$ for at least one k , hence that $(\mathbf{c} - \mathbf{d})^T (\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{A}_m \mathbf{d}^{m-1})$ is positive. \square

Proposition 3.5. *If m is a positive even integer, then Φ_m is a symmetric strictly positive definite multi-kernel.*

Proof. Proposition 3.4 guarantees that \mathbf{A}_m is a symmetric strictly positive definite tensor. Thus, the proof is straightforward by Definition 3.1. \square

Example 3.6. Let

$$\Phi_2(\mathbf{z}_1, \mathbf{z}_2) := \prod_{j=1}^d R(z_{1j}, z_{2j}),$$

for $\mathbf{z}_1 := (z_{11}, z_{12}, \dots, z_{1d})^T, \mathbf{z}_2 := (z_{21}, z_{22}, \dots, z_{2d})^T \in (0, 1)^d$, where

$$R(z_1, z_2) := \begin{cases} -z_1^3 + z_1^3 z_2 + z_1 z_2^3 - 3z_1 z_2^2 + 2z_1 z_2, & 0 < z_1 \leq z_2 < 1, \\ -z_2^3 + z_1 z_2^3 + z_1^3 z_2 - 3z_1^2 z_2 + 2z_1 z_2, & 0 < z_2 \leq z_1 < 1. \end{cases}$$

Thus, Φ_2 is a positive definite kernel on $(0, 1)^d$. Let

$$\lambda_{\mathbf{k}} := \prod_{j=1}^d \frac{6}{k_j^4 \pi^4}, \quad e_{\mathbf{k}}(\mathbf{x}) := 2^{d/2} \prod_{j=1}^d \sin(k_j \pi x_j),$$

for all $\mathbf{k} := (k_1, k_2, \dots, k_d)^T \in \mathbb{N}^d$. Thus, $\lambda_{\mathbf{k}}$ and $e_{\mathbf{k}}$ are the eigenvalues and eigenfunctions of K_2 and $\phi_{\mathbf{k}} := \lambda_{\mathbf{k}}^{1/2} e_{\mathbf{k}}$ satisfy the condition in Equation (3.2). By Proposition 3.5, we obtain the strictly positive definite multi-kernel

$$\Phi_m(\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_m) := \sum_{\mathbf{k} \in \mathbb{N}^d} \prod_{i=1}^m \phi_{\mathbf{k}}(\mathbf{z}_i), \quad \text{for } \mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_m \in (0, 1)^d.$$

Another examples of strictly positive definite multi-kernels can be constructed by Gaussian kernels and power series kernels such as in [5, Sections 4.4 and 4.5].

Finally, we introduce the reproducing kernel Banach spaces defined by the positive definite kernels such as in [5, Chapter 4]. Since Φ_2 is a symmetric positive

definite kernel, the eigenfunctions $\{e_k : k \in \mathbb{N}\}$ of Φ_2 can be an orthonormal basis of $L_2(\Omega)$. Thus, $\{\phi_k : k \in \mathbb{N}\}$ is a basis of $L_2(\Omega)$. Let $\boldsymbol{\phi} := (\phi_1, \phi_2, \dots, \phi_k, \dots)^T$. We define a normed space

$$\mathcal{B}^p := \left\{ f := \boldsymbol{\alpha}^T \boldsymbol{\phi} : \boldsymbol{\alpha} \in \mathbb{R}^{\mathbb{N}} \text{ and } \|\boldsymbol{\alpha}\|_p < \infty \right\},$$

equipped with the norm $\|f\|_{\mathcal{B}^p} := \|\boldsymbol{\alpha}\|_p$, where $1 < p < \infty$. By Equation (3.2), [5, Theorem 4.3] guarantees that \mathcal{B}^p is a two-sided reproducing kernel Banach space and Φ_2 is the two-sided reproducing kernel of \mathcal{B}^p , more precisely, (i) $\Phi_2(\mathbf{z}_1, \cdot) \in \mathcal{B}^q$, (ii) $\langle f, \Phi_2(\mathbf{z}_1, \cdot) \rangle = f(\mathbf{z}_1)$, (iii) $\Phi_2(\cdot, \mathbf{z}_2) \in \mathcal{B}^p$, and (iv) $\langle \Phi_2(\cdot, \mathbf{z}_2), g \rangle = g(\mathbf{z}_2)$, for all $\mathbf{z}_1, \mathbf{z}_2 \in \Omega$, all $f \in \mathcal{B}^p$, and all $g \in \mathcal{B}^q$, where $q := p/(p-1)$, the dual space of \mathcal{B}^p is isometrically equivalent to \mathcal{B}^q , and $\langle \cdot, \cdot \rangle$ represents the dual bilinear product of $\mathcal{B}^p \times \mathcal{B}^q$, that is, $\langle f, g \rangle = \boldsymbol{\alpha}^T \boldsymbol{\beta}$ for $f = \boldsymbol{\alpha}^T \boldsymbol{\phi}$ and $g = \boldsymbol{\beta}^T \boldsymbol{\phi}$. Moreover, [5, Proposition 4.4] guarantees that $\mathcal{B}^p \subseteq C(\Omega)$. The Gâteaux derivative of $\|\cdot\|_{\mathcal{B}^p}$ at $f \neq 0$ has the form

$$d_G \|\cdot\|_{\mathcal{B}^p}(f) = \sum_{k=1}^{\infty} \left(\frac{\alpha_k |\alpha_k|^{p-2}}{\|\boldsymbol{\alpha}\|_p^{p-1}} \right) \phi_k, \quad \text{for } f = \boldsymbol{\alpha}^T \boldsymbol{\phi} \in \mathcal{B}^p. \quad (3.4)$$

In Section 4, we will use the positive definite multi-kernel Φ_m to construct the interpolant and analyze its properties in the reproducing kernel Banach space $\mathcal{B}^{m/(m-1)}$ such as optimal recoveries and error analysis when m is a positive even integer.

4. Interpolations, Optimal Recoveries, and Error Analysis

In this section, we discuss how to construct the interpolant s_m from the scattered data by the strictly positive definite multi-kernel Φ_m in Equation (3.3). Suppose that the data $(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_n, y_n)$ compose of the pairwise distinct points $\{\mathbf{x}_i\}_{i=1}^n \subseteq \Omega \subseteq \mathbb{R}^d$ and the values $\{y_i\}_{i=1}^n \subseteq \mathbb{R}$ evaluated by some function $f \in C(\Omega)$, that is,

$$y_1 := f(\mathbf{x}_1), y_2 := f(\mathbf{x}_2), \dots, y_n := f(\mathbf{x}_n). \quad (4.1)$$

Let $\mathbf{y} := (y_1, y_2, \dots, y_n)^T \in \mathbb{R}^n$. Let \mathbf{A}_m be a tensor defined in Equation (3.1) by the multi-kernel $K_m := \Phi_m$ and the data points $\{\mathbf{x}_i\}_{i=1}^n$. Different from the classical kernel-based interpolations, the basis of s_m composes of $\Phi_m(\cdot, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_{m-1}})$ for $i_1, i_2, \dots, i_{m-1} = 1, 2, \dots, n$. Let a tensor function

$$\mathbf{B}_m(\mathbf{x}) := (\Phi_m(\mathbf{x}, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_{m-1}}))_{i_1, i_2, \dots, i_{m-1}=1}^{n, n, \dots, n} \in T_{m-1, n}, \quad \text{for } \mathbf{x} \in \Omega.$$

Thus, $\mathbf{A}_m = (\mathbf{B}_m(\mathbf{x}_i))_{i=1}^n$.

Theorem 4.1. *If m is a positive even integer, then s_m has the form*

$$s_m(\mathbf{x}) := \mathbf{B}_m(\mathbf{x})\mathbf{c}^{m-1}, \quad \text{for } \mathbf{x} \in \Omega,$$

such that

$$s_m(\mathbf{x}_1) = y_1, s_m(\mathbf{x}_2) = y_2, \dots, s_m(\mathbf{x}_n) = y_n,$$

where the coefficients $\mathbf{c} \in \mathbb{R}^n$ are uniquely solved by the multi-linear system

$$\mathbf{A}_m \mathbf{c}^{m-1} = \mathbf{y}. \quad (4.2)$$

Proof. Proposition 3.4 guarantees that \mathbf{A}_m is a symmetric strictly positive definite tensor. Thus, by Proposition 2.2, multi-linear system (4.2) has the unique solution \mathbf{c} . This assures that s_m satisfies the interpolation conditions. \square

Remark 4.2. If s_m is constructed by the general positive definite multi-kernel K_m , then \mathbf{A}_m is a symmetric positive definite tensor such that multi-linear system (4.2) still has the solution \mathbf{c} . But, the solution \mathbf{c} may not be unique by Proposition 2.1 such that the interpolant s_m may not be unique.

Equation (3.2) shows that $\Phi_m(\cdot, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_{m-1}}) \in \mathcal{B}^{m/(m-1)}$; hence $s_m \in \mathcal{B}^{m/(m-1)}$. Now we show the optimal recovery of s_m in $\mathcal{B}^{m/(m-1)}$ by the theorems in [5].

Theorem 4.3. *If m is a positive even integer, then s_m is the minimizer of the norm-minimal interpolation*

$$\min_{f \in \mathcal{B}^{m/(m-1)}} \|f\|_{\mathcal{B}^{m/(m-1)}} \text{ subjected to } f(\mathbf{x}_i) = y_i \text{ for all } i = 1, 2, \dots, n. \quad (4.3)$$

Proof. If $\mathbf{y} = 0$, then $s_m = 0$ and s_m is the unique minimizer of minimization (4.3). If $\mathbf{y} \neq 0$, then the minimizer of minimization (4.3) is nonzero. For convenience, we suppose that $\mathbf{y} \neq 0$. [5, Lemma 2.22] guarantees that minimization (4.3) has the unique minimizer s such that its Gateaux derivative $d_G \|\cdot\|_{\mathcal{B}^{m/(m-1)}}(s) = \sum_{i=1}^n \beta_i \Phi_2(\mathbf{x}_i, \cdot)$, where the constants $\beta_1, \dots, \beta_n \in \mathbb{R}$. By the expansions of Φ_2 , the Gateaux derivative can be rewritten as

$$d_G \|\cdot\|_{\mathcal{B}^{m/(m-1)}}(s) = \sum_{k=1}^{\infty} \left(\sum_{i=1}^n \beta_i \phi_k(\mathbf{x}_i) \right) \phi_k. \quad (4.4)$$

Combining Equations (3.4) and (4.4), the coefficients α of s has the form

$$\frac{\alpha_k |\alpha_k|^{m/(m-1)-2}}{\|\alpha\|_{m/(m-1)}^{m/(m-1)-1}} = \sum_{i=1}^n \beta_i \phi_k(\mathbf{x}_i), \quad \text{for } k \in \mathbb{N}. \quad (4.5)$$

Let $c_i := \|\alpha\|_{m/(m-1)}^{m/(m-1)-1} \beta_i$ for all $i = 1, 2, \dots, n$. Thus, Equation (4.5) shows that

$$s(\mathbf{x}) = \sum_{k=1}^{\infty} \left(\sum_{i=1}^n c_i \phi_k(\mathbf{x}_i) \left| \sum_{j=1}^n c_j \phi_k(\mathbf{x}_j) \right|^{m-2} \right) \phi_k(\mathbf{x}), \quad \text{for } \mathbf{x} \in \Omega. \quad (4.6)$$

Expanding of Equation (4.6), we have

$$\begin{aligned} s(\mathbf{x}) &= \sum_{k=1}^{\infty} \sum_{i_1, \dots, i_{m-1}=1}^{n, \dots, n} \phi_k(\mathbf{x}) \prod_{j=1}^{m-1} c_{i_j} \phi_n(\mathbf{x}_{i_j}) = \sum_{i_1, \dots, i_{m-1}=1}^{n, \dots, n} \prod_{j=1}^{m-1} c_{i_j} \sum_{k=1}^{\infty} \phi_k(\mathbf{x}) \prod_{l=1}^{m-1} \phi_n(\mathbf{x}_{i_l}) \\ &= \sum_{i_1, \dots, i_{m-1}=1}^{n, \dots, n} c_{i_1} c_{i_2} \cdots c_{i_{m-1}} \Phi_m(\mathbf{x}, \mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_{m-1}}) = \mathbf{B}_m(\mathbf{x}) \mathbf{c}^{m-1}. \end{aligned}$$

Since $s(\mathbf{x}_i) = y_i$ for all $i = 1, 2, \dots, n$, we have $\mathbf{A}_m \mathbf{c}^{m-1} = \mathbf{y}$. Moreover, Proposition 4.1 guarantees that the solution of multi-linear system (4.2) is unique. This assures that $s(\mathbf{x}) = s_m(\mathbf{x})$. \square

Theorem 4.4. *If m is a positive even integer, then*

$$\|s_m\|_{\mathcal{G}^{m/(m-1)}} = (\mathbf{A}_m \mathbf{c}^m)^{1-1/m}.$$

Proof. If $s_m = 0$, then the proof is straightforward. For convenience, we suppose that $s_m \neq 0$. As in the proof of Theorem 4.3, we have

$$\|s_m\|_{\mathcal{G}^{m/(m-1)}} = \langle s_m, d_G \|\cdot\|_{\mathcal{G}^{m/(m-1)}}(s_m) \rangle = \sum_{i=1}^n \beta_i \langle s_m, \Phi_2(\mathbf{x}_i, \cdot) \rangle,$$

and

$$\beta_i = \|\alpha\|_{m/(m-1)}^{1-m/(m-1)} c_i = \|s_m\|_{\mathcal{G}^{m/(m-1)}}^{-1/(m-1)} c_i, \quad \text{for } i = 1, \dots, n.$$

By the reproducing properties of $\mathcal{B}^{m/(m-1)}$, we compute the norm

$$\|s_m\|_{\mathcal{G}^{m/(m-1)}} = \sum_{i=1}^n \beta_i s_m(\mathbf{x}_i) = \|s_m\|_{\mathcal{G}^{m/(m-1)}}^{-1/(m-1)} \sum_{i=1}^n c_i \mathbf{B}_m(\mathbf{x}_i) \mathbf{c}^{m-1}.$$

Therefore,

$$\|s_m\|_{\mathcal{G}^{m/(m-1)}}^{m/(m-1)} = \mathbf{A}_m \mathbf{c}^m.$$

\square

For an approximate problem of $f(\mathbf{x}_1) \approx y_1, \dots, f(\mathbf{x}_n) \approx y_n$, the estimate function s is solved by the regularization

$$\min_{f \in \mathcal{B}^{m/(m-1)}} \sum_{i=1}^n (f(\mathbf{x}_i) - y_i)^2 + \sigma \|f\|_{\mathcal{B}^{m/(m-1)}}^{m/(m-1)}, \quad (4.7)$$

where σ is a positive parameter. In the same manner of Theorems 4.3 and 4.4, the minimizer of optimization (4.7) has the form as $s = B_m \mathbf{c}^{m-1}$, where the coefficients \mathbf{c} are uniquely solved by the minimization

$$\min_{\mathbf{c} \in \mathbb{R}^n} \|\mathbf{A}_m \mathbf{c}^{m-1} - \mathbf{y}\|_2^2 + \sigma \mathbf{A}_m \mathbf{c}^m.$$

Finally, we study the error analysis of $|s_m(\mathbf{x}) - f(\mathbf{x})|$ when $f \in \mathcal{B}^{m/(m-1)}$. We define the generalized power function

$$P_m(\mathbf{x}) := \min_{\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}} \left\| \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \right\|_{\mathcal{B}^m}, \quad \text{for } \mathbf{x} \in \Omega. \quad (4.8)$$

Let $\mathbf{x}_0 := \mathbf{x}$ and $\tilde{\mathbf{A}}_m := (\Phi_m(\mathbf{x}_{i_1}, \mathbf{x}_{i_2}, \dots, \mathbf{x}_{i_m}))_{i_1, i_2, \dots, i_m=0}^{n, n, \dots, n} \in T_{m, n+1}$. As in the proof of Theorems 4.3 and 4.4, we can compute

$$P_m(\mathbf{x}) = \min_{\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}} \left(\tilde{\mathbf{A}}_m \tilde{\boldsymbol{\lambda}}^m \right)^{1/m},$$

where $\tilde{\boldsymbol{\lambda}} := (1, -\lambda_1, -\lambda_2, \dots, -\lambda_n)^T \in \mathbb{R}^{n+1}$. Since \mathcal{B}^2 is a reproducing kernel Hilbert space with the reproducing kernel Φ_2 , P_2 is equal to the classical power function in [4, Section 11.1] such that $P_2(\mathbf{x}) = \left(\Phi_2(\mathbf{x}, \mathbf{x}) - \mathbf{A}_2^{-1} \mathbf{B}_2(\mathbf{x})^2 \right)^{1/2}$.

Theorem 4.5. *If m is a positive even integer and $f \in \mathcal{B}^{m/(m-1)}$ satisfies interpolation condition (4.1), then*

$$|s_m(\mathbf{x}) - f(\mathbf{x})| \leq 2 \|f\|_{\mathcal{B}^{m/(m-1)}} P_m(\mathbf{x}), \quad \text{for } \mathbf{x} \in \Omega.$$

Proof. By the reproducing properties of $\mathcal{B}^{m/(m-1)}$, the interpolation conditions of f and s_m show that

$$\langle f - s_m, \Phi_2(\mathbf{x}_i, \cdot) \rangle = f(\mathbf{x}_i) - s_m(\mathbf{x}_i) = y_i - y_i = 0,$$

for $i = 1, 2, \dots, n$. Thus, we have

$$\begin{aligned}
f(\mathbf{x}) - s_m(\mathbf{x}) &= \langle f - s_m, \Phi_2(\mathbf{x}, \cdot) \rangle \\
&= \langle f - s_m, \Phi_2(\mathbf{x}, \cdot) \rangle - \sum_{i=1}^n \lambda_i \langle f - s_m, \Phi_2(\mathbf{x}_i, \cdot) \rangle \\
&= \langle f - s_m, \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \rangle,
\end{aligned} \tag{4.9}$$

for any $\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}$. Since $\frac{m-1}{m} + \frac{1}{m} = 1$, Equation (4.9) shows that

$$|f(\mathbf{x}) - s_m(\mathbf{x})| \leq \|f - s_m\|_{\mathcal{B}^{m/(m-1)}} \left\| \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \right\|_{\mathcal{B}^m}. \tag{4.10}$$

Moreover, since Theorem 4.3 guarantees that $\|f\|_{\mathcal{B}^{m/(m-1)}} \geq \|s_m\|_{\mathcal{B}^{m/(m-1)}}$, we have

$$\|f - s_m\|_{\mathcal{B}^{m/(m-1)}} \leq \|f\|_{\mathcal{B}^{m/(m-1)}} + \|s_m\|_{\mathcal{B}^{m/(m-1)}} \leq 2 \|f\|_{\mathcal{B}^{m/(m-1)}}. \tag{4.11}$$

We conclude from Equations (4.10) and (4.11) that

$$|f(\mathbf{x}) - s_m(\mathbf{x})| \leq 2 \|f\|_{\mathcal{B}^{m/(m-1)}} \min_{\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}} \left\| \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \right\|_{\mathcal{B}^m},$$

hence that the proof is completed by Equation (4.8). \square

Let the fill distance $h := \sup_{\mathbf{x} \in \Omega} \min_{i=1,2,\dots,n} \|\mathbf{x} - \mathbf{x}_i\|_2$. If $\Phi_2 \in C^{2\gamma}(\Omega \times \Omega)$ for $\gamma \in \mathbb{N}$, then [4, Theorem 11.13] guarantees that

$$P_2(\mathbf{x}) \leq C_\Omega h^\gamma, \quad \text{for } \mathbf{x} \in \Omega, \tag{4.12}$$

where C_Ω is a positive constant independent of h and \mathbf{x} .

Theorem 4.6. *If m is a positive even integer, $f \in \mathcal{B}^{m/(m-1)}$ satisfies interpolation condition (4.1), and $\Phi_2 \in C^{2\gamma}(\Omega \times \Omega)$ for $\gamma \in \mathbb{N}$, then*

$$|f(\mathbf{x}) - s_m(\mathbf{x})| \leq Ch^\gamma, \quad \text{for } \mathbf{x} \in \Omega,$$

where C is a positive constant independent of h and \mathbf{x} .

Proof. The main idea of the proof is to compute the upper bound of $P_m(\mathbf{x})$ by the fill distance h . The constructions of \mathcal{B}^m and \mathcal{B}^2 shows that $\|f\|_{\mathcal{B}^m} \leq \|f\|_{\mathcal{B}^2}$ for $f \in \mathcal{B}^m \subseteq \mathcal{B}^2$; hence

$$\left\| \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \right\|_{\mathcal{B}^m} \leq \left\| \Phi_2(\mathbf{x}, \cdot) - \sum_{i=1}^n \lambda_i \Phi_2(\mathbf{x}_i, \cdot) \right\|_{\mathcal{B}^2}, \quad (4.13)$$

for any $\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{R}$. Substituting Equations (4.13) into (4.8), we have

$$P_m(\mathbf{x}) \leq P_2(\mathbf{x}). \quad (4.14)$$

Combining Equations (4.12) and (4.14), we also have

$$P_m(\mathbf{x}) \leq C_\Omega h^\gamma.$$

Let $C := 2 \|f\|_{\mathcal{B}^{m/(m-1)}} C_\Omega$. Therefore, the proof is completed by Theorem 4.5. \square

Theorem 4.6 guarantees that $s_m(\mathbf{x}) \rightarrow f(\mathbf{x})$ when $h \rightarrow 0$ for $\mathbf{x} \in \Omega$, and moreover, $s_m \rightarrow f$ uniformly when $h \rightarrow 0$.

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