



Time-Varying Isotropic Vector Random Fields on Compact Two-Point Homogeneous Spaces

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Abstract

A general form of the covariance matrix function is derived in this paper for a vector random field that is isotropic and mean square continuous on a compact connected two-point homogeneous space and stationary on a temporal domain. A series representation is presented for such a vector random field which involves Jacobi polynomials and the distance defined on the compact two-point homogeneous space.

Keywords Covariance matrix function · Elliptically contoured random field · Gaussian random field · Isotropy · Stationarity · Jacobi polynomials

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1 Introduction

Consider the sphere \mathbb{S}^d embedded into \mathbb{R}^{d+1} as follows: $\mathbb{S}^d = \{ \mathbf{x} \in \mathbb{R}^{d+1} : \|\mathbf{x}\| = 1 \}$, and define the distance between the points \mathbf{x}_1 and \mathbf{x}_2 by $\rho(\mathbf{x}_1, \mathbf{x}_2) = \cos^{-1}(\mathbf{x}_1^\top \mathbf{x}_2)$. With this distance, any isometry between two pairs of points can be extended to an isometry of \mathbb{S}^d . A metric space with such a property is called *two-point homogeneous*. A complete classification of *connected and compact* two-point homogeneous spaces is performed in [40]. Besides spheres, the list includes projective spaces over different algebras; see Sect. 2 for details. It turns out that any such space is a *manifold*. We denote it by \mathbb{M}^d , where d is the topological dimension of the manifold. Following

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[24], denote by \mathbb{T} either the set \mathbb{R} of real numbers or the set \mathbb{Z} of integers, and call it the *temporal domain*.

Let $(\Omega, \mathfrak{F}, P)$ be a probability space.

Definition 1 An \mathbb{R}^m -valued spatio-temporal random field $\mathbf{Z}(\omega, \mathbf{x}, t) : \Omega \times \mathbb{M}^d \times \mathbb{T} \rightarrow \mathbb{R}^m$ is called (wide-sense) *isotropic* over \mathbb{M}^d and (wide-sense) *stationary* over the temporal domain \mathbb{T} , if its mean function $E[\mathbf{Z}(\mathbf{x}; t)]$ equals a constant vector, and its covariance matrix function

$$\text{cov}(\mathbf{Z}(\mathbf{x}_1; t_1), \mathbf{Z}(\mathbf{x}_2; t_2)) = E \left[(\mathbf{Z}(\mathbf{x}_1; t_1) - E[\mathbf{Z}(\mathbf{x}_1; t_1)])(\mathbf{Z}(\mathbf{x}_2; t_2) - E[\mathbf{Z}(\mathbf{x}_2; t_2)])^\top \right],$$

$$\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t_1, t_2 \in \mathbb{T},$$

depends only on the time lag $t_2 - t_1$ between t_2 and t_1 and the distance $\rho(\mathbf{x}_1, \mathbf{x}_2)$ between \mathbf{x}_1 and \mathbf{x}_2 .

As usual, we omit the argument $\omega \in \Omega$ in the notation for the random field under consideration. In such a case, the covariance matrix function is denoted by $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$,

$$C(\rho(\mathbf{x}_1, \mathbf{x}_2); t_1 - t_2) = E \left[(\mathbf{Z}(\mathbf{x}_1; t_1) - E[\mathbf{Z}(\mathbf{x}_1; t_1)])(\mathbf{Z}(\mathbf{x}_2; t_2) - E[\mathbf{Z}(\mathbf{x}_2; t_2)])^\top \right],$$

$$\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t_1, t_2 \in \mathbb{T}.$$

It is an $m \times m$ matrix function, $C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t) = (C(\rho(\mathbf{x}_1, \mathbf{x}_2); t))^\top$, and the inequality

$$\sum_{i=1}^n \sum_{j=1}^n \mathbf{a}_i^\top C(\rho(\mathbf{x}_i, \mathbf{x}_j); t_i - t_j) \mathbf{a}_j \geq 0$$

holds for every $n \in \mathbb{N}$, any $\mathbf{x}_i \in \mathbb{M}^d, t_i \in \mathbb{T}$, and $\mathbf{a}_i \in \mathbb{R}^m$ ($i = 1, 2, \dots, n$), where \mathbb{N} stands for the set of positive integers, while \mathbb{N}_0 denotes the set of nonnegative integers below. On the other hand, given an $m \times m$ matrix function with these properties, there exists an m -variate Gaussian or elliptically contoured random field $\{\mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$ with $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$ as its covariance matrix function [21].

For a scalar and purely spatial random field $\{Z(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ that is isotropic and mean square continuous, its covariance function is continuous and possesses a series representation of the form [8,14,37]

$$\text{cov}(Z(\mathbf{x}_1), Z(\mathbf{x}_2)) = \sum_{n=0}^{\infty} b_n P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \quad (1)$$

where $\{b_n : n \in \mathbb{N}_0\}$ is a sequence of nonnegative numbers with $\sum_{n=0}^{\infty} b_n P_n^{(\alpha, \beta)}(1)$ convergent, $P_n^{(\alpha, \beta)}(x)$ is a Jacobi polynomial of degree n with a pair of parameters (α, β) [1,38], shown in Table 2. A general form of the covariance matrix function and

a series representation are derived in [24] for a vector random field that is isotropic and mean square continuous on a sphere and stationary on a temporal domain. They are extended to $\mathbb{M}^d \times \mathbb{T}$ in this paper.

Isotropic random fields over \mathbb{S}^d with values in \mathbb{R}^1 and \mathbb{C}^1 were introduced in [35]. Theoretical investigations and practical applications of isotropic scalar-valued random fields on spheres may be found in [7,11,12,19,43], and vector- and tensor-valued random fields on spheres have been considered in [18,23,24,30], among others. Cosmological applications, in particular, studies of tiny fluctuations of the Cosmic Microwave Background, require development of the theory of *random sections of vector and tensor bundles* over \mathbb{S}^2 [4,15,25,27]. See also surveys of the topic in the monographs [26,31,42,44]. Isotropic random fields on connected compact two-point homogeneous spaces are studied in [2,14,28,29,33], among others.

Some important properties of \mathbb{M}^d , $\rho(\mathbf{x}_1, \mathbf{x}_2)$, and $P_n^{(\alpha,\beta)}(x)$ are reviewed in Sect. 2, and two lemmas are derived: one as a special case of the Funk–Hecke formula on \mathbb{M}^d and the other as a kind of probability interpretation. A series representation is given in Sect. 3 for an isotropic and mean square continuous vector random field on \mathbb{M}^d , and a series expression of its covariance matrix function, in terms of Jacobi polynomials. Section 4 deals with a spatio-temporal vector random field on $\mathbb{M}^d \times \mathbb{T}$, which is isotropic and mean square continuous vector random field on \mathbb{M}^d and stationary on \mathbb{T} , and obtains a series representation for the random field and a general form for its covariance matrix function. The lemmas and theorems are proved in Appendix A.

2 Compact Two-Point Homogeneous Spaces and Jacobi Polynomials

This section starts by recalling some important properties of the compact connected two-point homogeneous space \mathbb{M}^d and those of Jacobi polynomials and then establishes two useful lemmas on a special case of the Funk–Hecke formula on \mathbb{M}^d and its probability interpretation, which are conjectured in [24]. In what follows, we consider only connected compact two-point homogeneous spaces.

The compact connected two-point homogeneous spaces are shown in the first column of Table 1. Besides spheres, there are projective spaces over the fields \mathbb{R} and \mathbb{C} , over the skew field \mathbb{H} of quaternions, and over the algebra \mathbb{O} of octonions. The possible values of d are chosen in such a way that all the spaces in Table 1 are different and exhaust the list. In the lowest dimensions, we have $\mathbb{P}^1(\mathbb{R}) = \mathbb{S}^1$, $\mathbb{P}^2(\mathbb{C}) = \mathbb{S}^2$, $\mathbb{P}^4(\mathbb{H}) = \mathbb{S}^4$, and $\mathbb{P}^8(\mathbb{O}) = \mathbb{S}^8$.

All compact two-point homogeneous spaces share the same property [6] that all of their geodesic lines are closed. Moreover, all of them are circles and have the same length. In particular, when the sphere \mathbb{S}^d is embedded into the space \mathbb{R}^{d+1} as described in Sect. 1, the length of any geodesic line is equal to that of the unit circle, that is, 2π . It is natural to norm the distance in such a way that the length of any geodesic line is equal to 2π , exactly as in the case of the unit sphere.

There are at least two different approaches to the subject of compact two-point homogeneous spaces in the literature. They are reviewed in the next two subsections.

Table 1 An approach based on Lie algebras

\mathbb{M}^d	G	K	p	q	Zonal function
$\mathbb{S}^d, d = 1, 2, \dots$	$\text{SO}(d+1)$	$\text{SO}(d)$	0	$d-1$	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^d(\mathbb{R}), d = 2, 3, \dots$	$\text{SO}(d+1)$	$\text{O}(d)$	0	$d-1$	$R_{2n}^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})/2))$
$\mathbb{P}^d(\mathbb{C}), d = 4, 6, \dots$	$\text{SU}(\frac{d}{2}+1)$	$\text{S}(\text{U}(\frac{d}{2}) \times \text{U}(1))$	$d-2$	1	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^d(\mathbb{H}), d = 8, 12, \dots$	$\text{Sp}(\frac{d}{4}+1)$	$\text{Sp}(\frac{d}{4}) \times \text{Sp}(1)$	$d-4$	3	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^{16}(\mathbb{O})$	$\text{F}_4(-52)$	$\text{Spin}(9)$	8	7	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$

2.1 An Approach Based on Lie Algebras

This approach goes back to Cartan [10]. It has been used in both the probabilistic literature [14] and the approximation theory literature [3].

Let G be the connected component of the group of isometries of \mathbb{M}^d , and let K be the stationary subgroup of a fixed point in \mathbb{M}^d , call it \mathbf{o} . Cartan [10] defined and calculated the numbers p and q , which are dimensions of some root spaces connected with the Lie algebras of the groups G and K . The groups G and K are listed in the second and the third columns of Table 1, while the numbers p and q are listed in the fourth and fifth columns of the table.

By [17, Theorem 11], if \mathbb{M}^d is a two-point homogeneous space, then the only differential operators on \mathbb{M}^d that are invariant under all isometries of \mathbb{M}^d are the polynomials in a special differential operator Δ called the *Laplace–Beltrami operator*. Let $d\nu(\mathbf{x})$ be the measure which is induced on the homogeneous space $\mathbb{M}^d = G/K$ by the *probabilistic* invariant measure on G . It is possible to define Δ as a self-adjoint operator in the space $H = L^2(\mathbb{M}^d, d\nu(\mathbf{x}))$. The spectrum of Δ is discrete, and the eigenvalues are

$$\lambda_n = -\varepsilon n(\varepsilon n + \alpha + \beta + 1), \quad n \in \mathbb{N}_0,$$

where

$$\alpha = (p + q - 1)/2, \quad \beta = (q - 1)/2, \quad (2)$$

and where $\varepsilon = 2$ if $\mathbb{M}^d = \mathbb{P}^d(\mathbb{R})$ and $\varepsilon = 1$ otherwise.

Let H_n be the eigenspace of Δ corresponding to λ_n . The space H is the Hilbert direct sum of its subspaces H_n , $n \in \mathbb{N}_0$. The space H_n is finite-dimensional with

$$\dim H_n = \frac{(2n + \alpha + \beta + 1)\Gamma(\beta + 1)\Gamma(n + \alpha + \beta + 1)\Gamma(n + \alpha + 1)}{\Gamma(\alpha + 1)\Gamma(\alpha + \beta + 2)\Gamma(n + 1)\Gamma(n + \beta + 1)}.$$

Each of the spaces H_n contains a unique one-dimensional subspace whose elements are *K-spherical functions*; that is, functions invariant under the action of K on \mathbb{M}^d . Such a function, say $f_n(\mathbf{x})$, depends only on the distance $r = \rho(\mathbf{x}, \mathbf{o})$, $f_n(\mathbf{x}) = f_n^*(r)$. A spherical function is called *zonal* if $f_n^*(0) = 1$.

The zonal spherical functions of all compact connected two-point homogeneous spaces are listed in the last column of Table 1. To explain notation, we recall that the *Jacobi polynomials*

$$P_n^{(\alpha,\beta)}(x) = \frac{\Gamma(\alpha + n + 1)}{n! \Gamma(\alpha + \beta + n + 1)} \sum_{k=0}^n \binom{n}{k} \frac{\Gamma(\alpha + \beta + n + k + 1)}{\Gamma(\alpha + k + 1)} \left(\frac{x - 1}{2}\right)^k,$$

$$x \in [-1, 1], \quad n \in \mathbb{N}_0,$$

are the eigenfunctions of the *Jacobi operator* [38, Theorem 4.2.1]

$$\Delta_x = \frac{1}{(1 - x)^\alpha (1 + x)^\beta} \frac{d}{dx} \left((1 - x)^{\alpha+1} (1 + x)^{\beta+1} \frac{d}{dx} \right).$$

In the last column of Table 1, the *normalised Jacobi polynomials* are introduced,

$$R_n^{(\alpha,\beta)}(x) = \frac{P_n^{(\alpha,\beta)}(x)}{P_n^{(\alpha,\beta)}(1)}, \quad n \in \mathbb{N}_0,$$

where

$$P_n^{(\alpha,\beta)}(1) = \frac{\Gamma(n + \alpha + 1)}{\Gamma(n + 1) \Gamma(\alpha + 1)}. \tag{3}$$

The reason for the exceptional behaviour of the real projective spaces is as follows; see [14, 16]. The space $\mathbb{P}^d(\mathbb{R})$ may be constructed by identification of antipodal points on the sphere \mathbb{S}^d . An $O(d)$ -invariant function f on $\mathbb{P}^d(\mathbb{R})$ can be lifted to an $SO(d)$ -invariant function g on \mathbb{S}^d by $g(\mathbf{x}) = f(\pi(\mathbf{x}))$, where π maps a point $\mathbf{x} \in \mathbb{S}^d$ to the pair of antipodal points $\pi(\mathbf{x}) \in \mathbb{P}^d(\mathbb{R})$. This simply means that a function on $[0, 1]$ can be extended to an even function on $[-1, 1]$. Only the even polynomials can be functions on the so constructed manifold. By [38, Equation (4.1.3)], we have

$$P_n^{(\alpha,\beta)}(x) = (-1)^n P_n^{(\beta,\alpha)}(-x).$$

For the real projective spaces $\alpha = \beta$, and the corresponding normalised Jacobi polynomials are even if and only if n is even.

Remark 1 If two Lie groups have the same connected component of identity, then they have the same Lie algebra. For example, the groups $SO(d)$ and $O(d)$ have the same Lie algebra $\mathfrak{so}(d)$. That is, the approach based on Lie algebras gives the same values of p and q for spheres and real projective spaces of equal dimensions. Only zonal spherical functions can distinguish between the two cases.

In the only case of $\mathbb{M}^d = \mathbb{S}^1$, we have $p = q = 0$. The reason is that only in this case the Lie algebra $\mathfrak{so}(2)$ is commutative rather than semisimple, and does not have nonzero root spaces at all.

Table 2 A geometric approach

\mathbb{M}^d	p	q	α	β	\mathbb{A}	$i(\mathbb{M}^d)$
$\mathbb{S}^d, d = 1, 2, \dots$	0	$d - 1$	$\frac{d-2}{2}$	$\frac{d-2}{2}$	\mathbb{S}^0	1
$\mathbb{P}^d(\mathbb{R}), d = 2, 3, \dots$	$d - 1$	0	$\frac{d-2}{2}$	$-\frac{1}{2}$	$\mathbb{P}^{d-1}(\mathbb{R})$	2^{d-1}
$\mathbb{P}^d(\mathbb{C}), d = 4, 6, \dots$	$d - 2$	1	$\frac{d-2}{2}$	0	$\mathbb{P}^{d-2}(\mathbb{C})$	$\binom{d-1}{d/2-1}$
$\mathbb{P}^d(\mathbb{H}), d = 8, 12, \dots$	$d - 4$	3	$\frac{d-2}{2}$	1	$\mathbb{P}^{d-4}(\mathbb{H})$	$\frac{1}{d/2+1} \binom{d-1}{d/2-1}$
$\mathbb{P}^{16}(\mathbb{O})$	8	7	7	3	$\mathbb{P}^8(\mathbb{O})$	39

2.2 A Geometric Approach

There is a trick that allows us to write down *all* zonal spherical functions of *all* compact two-point homogeneous spaces in the same form, which is used in probabilistic literature [2,26,28,29,33] and in approximation theory [9,13]. Denote $y = \cos(\rho(\mathbf{x}, \mathbf{o})/2)$. Then we have $\cos(\rho(\mathbf{x}, \mathbf{o})) = 2y^2 - 1$. For the case of $\mathbb{M}^d = \mathbb{P}^d(\mathbb{R})$, $\alpha = \beta = (d - 2)/2$. By [38, Theorem 4.1],

$$P_{2n}^{(\alpha,\alpha)}(y) = \frac{\Gamma(2n + \alpha + 1)\Gamma(n + 1)}{\Gamma(n + \alpha + 1)\Gamma(2n + 1)} P_n^{(\alpha,-1/2)}(2y^2 - 1).$$

In terms of the normalised Jacobi polynomials, we obtain

$$R_{2n}^{(\alpha,\alpha)}(\cos(\rho(\mathbf{x}, \mathbf{o})/2)) = R_n^{(\alpha,-1/2)}(\cos(\rho(\mathbf{x}, \mathbf{o}))).$$

For the case of $\mathbb{M}^d = \mathbb{P}^d(\mathbb{R})$, if we redefine $\alpha = (d - 2)/2$, $\beta = -1/2$, then *all* zonal spherical functions of *all* compact two-point homogeneous spaces are given by the same expression $R_n^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$.

It easily follows from (2) that the new values for p and q in the case of $\mathbb{M}^d = \mathbb{P}^d(\mathbb{R})$ are $p = d - 1$ and $q = 0$. It is interesting to note that the new values of p and q for the real projective spaces together with their old values for the rest of spaces still have a meaning; see [13] and Table 2. This time, the values of p and q are connected with the *geometry* of the space \mathbb{M}^d rather than with Lie algebras.

Specifically, let $\mathbb{A} = \{\mathbf{x} \in \mathbb{M}^d : \rho(\mathbf{x}, \mathbf{o}) = \pi\}$. This set is called the *antipodal manifold* of the point \mathbf{o} . The antipodal manifolds are listed in the sixth column of Table 2. Geometrically, if $\mathbb{M}^d = \mathbb{S}^d$ and \mathbf{o} is the North pole, then $\mathbb{A} = \mathbb{S}^0$ is the South pole. Otherwise, \mathbb{A} is the *space at infinity* of the point \mathbf{o} in the terms of projective geometry. The new number p turns out to be the *dimension of the antipodal manifold*, while the number $p + q + 1$ is, as before, the dimension of the space \mathbb{M}^d itself.

In what follows, we use the geometric approach. It turns out that all the spaces \mathbb{M}^d are *Riemannian manifolds*, as is defined in [5]. Each Riemannian manifold carries the *canonical measure* μ ; see [5, pp. 10–11]. The measure μ is proportional to the measure ν constructed in Sect. 2.1. The coefficient of proportionality or the total measure $\mu(\mathbb{M}^d)$ of the compact manifold \mathbb{M}^d is called the *volume* of \mathbb{M}^d .

Lemma 1 *The volume of the space \mathbb{M}^d is*

$$\omega_d = \mu(\mathbb{M}^d) = \frac{(4\pi)^{\alpha+1} \Gamma(\beta + 1)}{\Gamma(\alpha + \beta + 2)}. \tag{4}$$

In what follows, we write just dx instead of $d\mu(\mathbf{x})$.

2.3 Orthogonal Properties of Jacobi Polynomials

The set of Jacobi polynomials $\{P_n^{(\alpha,\beta)}(x) : n \in \mathbb{N}_0, x \in \mathbb{R}\}$ possesses two types of orthogonal properties. First, for each pair of $\alpha > -1$ and $\beta > -1$, this set is a complete orthogonal system on the interval $[-1, 1]$ with respect to the weight function $(1 - x)^\alpha (1 + x)^\beta$, in the sense that

$$\int_{-1}^1 P_i^{(\alpha,\beta)}(x) P_j^{(\alpha,\beta)}(x) (1-x)^\alpha (1+x)^\beta dx = \begin{cases} \frac{2^{\alpha+\beta+1}}{2j+\alpha+\beta+1} \frac{\Gamma(j+\alpha+1)\Gamma(j+\beta+1)}{j!\Gamma(j+\alpha+\beta+1)}, & i = j, \\ 0, & i \neq j. \end{cases} \tag{5}$$

Second, for *selected values* of α and β given by (2) with p and q given in Table 2, they are orthogonal over \mathbb{M}^d , as the following lemma describes, which is derived from the Funk–Hecke formula recently established in [3]. In the particular case $\mathbb{M}^d = \mathbb{S}^d$, the Funk–Hecke formula may be found in classical references such as [1,34].

Lemma 2 *For $i, j \in \mathbb{N}_0$, and $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d$,*

$$\int_{\mathbb{M}^d} P_i^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}))) P_j^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_2, \mathbf{x}))) d\mathbf{x} = \frac{\delta_{ij} \omega_d}{a_i^2} P_i^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))),$$

where

$$a_n = \left(\frac{\Gamma(\beta + 1)(2n + \alpha + \beta + 1)\Gamma(n + \alpha + \beta + 1)}{\Gamma(\alpha + \beta + 2)\Gamma(n + \beta + 1)} \right)^{\frac{1}{2}}, \quad n \in \mathbb{N}_0. \tag{6}$$

The probabilistic interpretation of zonal spherical functions on \mathbb{M}^d is provided in Lemma 3. The spherical case is given in [23].

Definition 2 A random vector \mathbf{U} is said to be *uniformly distributed* on \mathbb{M}^d if, for every Borel set $A \subseteq \mathbb{M}^d$ and every isometry g we have $P(\mathbf{U} \in A) = P(\mathbf{U} \in gA)$.

To construct \mathbf{U} , we start with a measure σ proportional to the invariant measure ν of Sect. 2.1. Let $T_{\mathbf{o}}$ be the tangent space to \mathbb{M}^d at the point \mathbf{o} . Choose a Cartesian coordinate system in $T_{\mathbf{o}}$ and identify this space with the space \mathbb{R}^d . Construct a chart $\varphi: \mathbb{M}^d \setminus \mathbb{A} \rightarrow \mathbb{R}^d$ as follows. Put $\varphi(\mathbf{o}) = \mathbf{0} \in \mathbb{R}^d$. For any other point $\mathbf{x} \in \mathbb{M}^d \setminus \mathbb{A}$, draw the unique geodesic line connecting \mathbf{o} and \mathbf{x} . Let $\mathbf{r} \in \mathbb{R}^d$ be the unit tangent vector to the above geodesic line. Define

$$\varphi(\mathbf{x}) = \mathbf{r} \tan(\rho(\mathbf{x}, \mathbf{o})/2),$$

and, for each Borel set $B \subseteq \mathbb{M}^d$,

$$\sigma(B) = \int_{\varphi^{-1}(B \setminus \mathbb{A})} \frac{d\mathbf{x}}{(1 + \|\mathbf{x}\|^2)^{\alpha+\beta+2}}.$$

This measure is indeed invariant [39, p. 113]. Finally, define a probability space $(\Omega', \mathfrak{F}', P')$ as follows: $\Omega' = \mathbb{M}^d$, \mathfrak{F}' is the σ -field of Borel subsets of Ω' , and

$$P'(B) = \frac{\sigma(B)}{\sigma(\mathbb{M}^d)}, \quad B \in \mathfrak{B}'.$$

The random variable $U(\omega) = \omega$ is then uniformly distributed on \mathbb{M}^d .

Lemma 3 *Let U be a random vector uniformly distributed on \mathbb{M}^d . For $n \in \mathbb{N}$,*

$$Z_n(\mathbf{x}) = a_n P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, U))), \quad \mathbf{x} \in \mathbb{M}^d,$$

is a centred isotropic random field with covariance function

$$\text{cov}(Z_n(\mathbf{x}_1), Z_n(\mathbf{x}_2)) = P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d,$$

where a_n is given by (6). Moreover, for $k \neq n$, the random fields $\{Z_k(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ and $\{Z_n(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ are uncorrelated:

$$\text{cov}(Z_k(\mathbf{x}_1), Z_n(\mathbf{x}_2)) = 0, \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d. \tag{7}$$

3 Isotropic Vector Random Fields on \mathbb{M}^d

In the purely spatial case, this section presents a series representation for an m -variate isotropic and mean square continuous random field $\{Z(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ and a series expression for its covariance matrix function, in terms of Jacobi polynomials. By mean square continuous, we mean that, for $k = 1, \dots, m$,

$$E \left[|Z_k(\mathbf{x}_1) - Z_k(\mathbf{x}_2)|^2 \right] \rightarrow 0, \quad \text{as } \rho(\mathbf{x}_1, \mathbf{x}_2) \rightarrow 0, \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d.$$

It implies the continuity of each entry of the associated covariance matrix function in terms of $\rho(\mathbf{x}_1, \mathbf{x}_2)$.

In what follows, d is assumed to be greater than 1, while \mathbb{M}^d reduces to the unit circle \mathbb{S}^1 when $d = 1$, over which the treatment of isotropic vector random fields may be found in [23,24]. For an $m \times m$ symmetric and nonnegative definite matrix B with nonnegative eigenvalues $\lambda_1, \dots, \lambda_m$, there is an orthogonal matrix S such that $S^{-1}BS = D$, where D is a diagonal matrix with diagonal entries $\lambda_1, \dots, \lambda_m$. Define the square root of B by

$$B^{\frac{1}{2}} = SD^{\frac{1}{2}}S^{-1},$$

where $D^{\frac{1}{2}}$ is a diagonal matrix with diagonal entries $\sqrt{\lambda_1}, \dots, \sqrt{\lambda_m}$. Clearly, $B^{\frac{1}{2}}$ is symmetric, nonnegative definite, and $(B^{\frac{1}{2}})^2 = B$. Denote by I_m an $m \times m$ identity matrix. For a sequence of $m \times m$ matrices $\{B_n : n \in \mathbb{N}_0\}$, the series $\sum_{n=0}^{\infty} B_n$ is said to be convergent, if each of its entries is convergent.

Theorem 1 *Suppose that $\{\mathbf{V}_n : n \in \mathbb{N}_0\}$ is a sequence of independent m -variate random vectors with $E(\mathbf{V}_n) = \mathbf{0}$ and $\text{cov}(\mathbf{V}_n, \mathbf{V}_n) = a_n^2 I_m$, \mathbf{U} is a random vector uniformly distributed on \mathbb{M}^d and is independent of $\{\mathbf{V}_n : n \in \mathbb{N}_0\}$, and that $\{B_n : n \in \mathbb{N}_0\}$ is a sequence of $m \times m$ symmetric nonnegative definite matrices. If the series $\sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(1)$ converges, then*

$$\mathbf{Z}(\mathbf{x}) = \sum_{n=0}^{\infty} B_n^{\frac{1}{2}} \mathbf{V}_n P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})), \quad \mathbf{x} \in \mathbb{M}^d, \tag{8}$$

is a centred m -variate isotropic random field on \mathbb{M}^d , with covariance matrix function

$$\text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2)) = \sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d. \tag{9}$$

The terms of (8) are uncorrelated; more precisely,

$$\text{cov}\left(B_i^{\frac{1}{2}} \mathbf{V}_i P_i^{(\alpha, \beta)}(\rho(\mathbf{x}_1, \mathbf{U})), B_j^{\frac{1}{2}} \mathbf{V}_j P_j^{(\alpha, \beta)}(\rho(\mathbf{x}_2, \mathbf{U}))\right) = \mathbf{0}, \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \quad i \neq j.$$

Since $|P_n^{(\alpha, \beta)}(\cos \vartheta)| \leq P_n^{(\alpha, \beta)}(1), n \in \mathbb{N}_0$, the convergent assumption of the series $\sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(1)$ ensures not only the mean square convergence of the series at the right-hand side of (8), but also the uniform and absolute convergence of the series at the right-hand side of (9).

When $\mathbb{M}^d = \mathbb{S}^2$ and $m = 1$, we have $\dim H_n = 2n + 1$, and (9) takes the form

$$\text{cov}(Z(\mathbf{x}_1), Z(\mathbf{x}_2)) = \sum_{n=0}^{\infty} b_n P_n(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)),$$

where $P_n(x)$ are Legendre polynomials. In the theory of Cosmic Microwave Background, this equation is traditionally written in the form

$$\text{cov}(Z(\mathbf{x}_1), Z(\mathbf{x}_2)) = \sum_{\ell=0}^{\infty} (2\ell + 1) C_{\ell} P_{\ell}(\mathbf{x}_1 \cdot \mathbf{x}_2),$$

and the sequence $\{C_{\ell} : \ell \geq 0\}$ is called the *angular power spectrum*. In the general case, define the angular power spectrum by

$$C_n = \frac{1}{\dim H_n} B_n.$$

A lot of examples of the angular power spectrum for general compact two-point homogeneous spaces may be found in [2].

As the next theorem indicates, (9) is a general form that the covariance matrix function of an m -variate isotropic and mean square continuous random field on \mathbb{M}^d must take.

Theorem 2 *For an m -variate isotropic and mean square continuous random field $\{Z(\mathbf{x}): \mathbf{x} \in \mathbb{M}^d\}$, its covariance matrix function $\text{cov}(Z(\mathbf{x}_1), Z(\mathbf{x}_2))$ is of the form*

$$C(\mathbf{x}_1, \mathbf{x}_2) = \sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \quad (10)$$

where $\{B_n: n \in \mathbb{N}_0\}$ is a sequence of $m \times m$ nonnegative definite matrices and the series $\sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(1)$ converges.

Conversely, if an $m \times m$ matrix function $C(\mathbf{x}_1, \mathbf{x}_2)$ is of the form (10), then it is the covariance matrix function of an m -variate isotropic Gaussian or elliptically contoured random field on \mathbb{M}^d .

Examples of covariance matrix functions on \mathbb{S}^d may be found in, for instance, [23, 24]. We would call for parametric and semi-parametric covariance matrix structures on \mathbb{M}^d .

4 Time-Varying Isotropic Vector Random Fields on \mathbb{M}^d

For an m -variate random field $\{\mathbf{Z}(\mathbf{x}; t): \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$ that is isotropic and mean square continuous over \mathbb{M}^d and stationary on \mathbb{T} , this section presents the general form of its covariance matrix function $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$, which is a continuous function of $\rho(\mathbf{x}_1, \mathbf{x}_2)$ and is also a continuous function of $t \in \mathbb{R}$ if $\mathbb{T} = \mathbb{R}$. A series representation is given in the following theorem for such a random field, as an extension of that on $\mathbb{S}^d \times \mathbb{T}$.

Theorem 3 *If an m -variate random field $\{\mathbf{Z}(\mathbf{x}; t), \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$ is isotropic and mean square continuous over \mathbb{M}^d and stationary on \mathbb{T} , then*

$$C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t) = (C(\rho(\mathbf{x}_1, \mathbf{x}_2); t))^{\top},$$

and $\frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)}{2}$ is of the form

$$\begin{aligned} & \frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)}{2} \\ &= \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t \in \mathbb{T}, \end{aligned} \quad (11)$$

where, for each fixed $n \in \mathbb{N}_0$, $B_n(t)$ is a stationary covariance matrix function on \mathbb{T} , and, for each fixed $t \in \mathbb{T}$, $B_n(t)$ ($n \in \mathbb{N}_0$) are $m \times m$ symmetric matrices and $\sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(1)$ converges.

While a general form of $\frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)}{2}$, instead of $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$ itself, is given in Theorem 3, that of $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$ can be obtained in certain special cases, such as spatio-temporal symmetric, and purely spatial.

Corollary 1 *If $C(\rho(\mathbf{x}_1, \mathbf{x}_2); t)$ is spatio-temporal symmetric in the sense that*

$$C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t) = C(\rho(\mathbf{x}_1, \mathbf{x}_2); t), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t \in \mathbb{T},$$

then it takes the form

$$C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t \in \mathbb{T}.$$

In contrast to those in (11), the $m \times m$ matrices $B_n(t)$ ($n \in \mathbb{N}_0$) in the next theorem are not necessarily symmetric. One simple such example is

$$B(t) = \begin{cases} \Sigma + \Phi \Sigma \Phi^\top, & t = 0, \\ \Phi \Sigma, & t = -1, \\ \Sigma \Phi^\top, & t = 1, \\ 0, & t = \pm 2, \pm 3, \dots, \end{cases}$$

which is the covariance matrix function of an m -variate first order moving average time series $\mathbf{Z}(t) = \boldsymbol{\varepsilon}(t) + \Phi \boldsymbol{\varepsilon}(t - 1)$, $t \in \mathbb{Z}$, where $\{\boldsymbol{\varepsilon}(t) : t \in \mathbb{Z}\}$ is m -variate white noise with $E[\boldsymbol{\varepsilon}(t)] = \mathbf{0}$ and $\text{Var}[\boldsymbol{\varepsilon}(t)] = \Sigma$, and Φ is an $m \times m$ matrix.

Theorem 4 *An $m \times m$ matrix function*

$$C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t \in \mathbb{T}, \quad (12)$$

is the covariance matrix function of an m -variate Gaussian or elliptically contoured random field on $\mathbb{M}^d \times \mathbb{T}$ if and only if $\{B_n(t) : n \in \mathbb{N}_0\}$ is a sequence of stationary covariance matrix functions on \mathbb{T} and $\sum_{n=0}^{\infty} B_n(0) P_n^{(\alpha, \beta)}(1)$ converges.

As an example of (12), let

$$B_n(t) = \begin{cases} \Sigma_n + \Phi \Sigma_n \Phi^\top, & t = 0, \\ \Phi \Sigma_n, & t = -1, \\ \Sigma_n \Phi^\top, & t = 1, \\ 0, & t = \pm 2, \pm 3, \dots, \quad n \in \mathbb{N}_0, \end{cases}$$

where $\{\Sigma_n : n \in \mathbb{N}_0\}$ is a sequence of $m \times m$ nonnegative definite matrices and $\sum_{n=0}^{\infty} \Sigma_n P_n^{(\alpha, \beta)}(1)$ converges. In this case, (12) is the covariance matrix function of an m -variate Gaussian or elliptically contoured random field on $\mathbb{M}^d \times \mathbb{Z}$.

Gaussian and second-order elliptically contoured random fields form one of the largest sets, if not the largest set, which allows any possible correlation structure

[21]. The covariance matrix functions developed in Theorem 4 can be adopted for a Gaussian or elliptically contoured vector random field. However, they may not be available for other non-Gaussian random fields, such as a log-Gaussian [32], χ^2 [20], K-distributed [22], or skew-Gaussian one, for which admissible correlation structure must be investigated on a case-by-case basis. A series representation is given in the following theorem for an m -variate spatio-temporal random field on $\mathbb{M}^d \times \mathbb{T}$.

Theorem 5 *An m -variate random field*

$$\mathbf{Z}(\mathbf{x}; t) = \sum_{n=0}^{\infty} \mathbf{V}_n(t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})), \quad \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}, \tag{13}$$

is isotropic and mean square continuous on \mathbb{M}^d , stationary on \mathbb{T} , and possesses mean $\mathbf{0}$ and covariance matrix function (12), where $\{\mathbf{V}_n(t) : n \in \mathbb{N}_0\}$ is a sequence of independent m -variate stationary stochastic processes on \mathbb{T} with

$$E(\mathbf{V}_n) = \mathbf{0}, \quad \text{cov}(\mathbf{V}_n(t_1), \mathbf{V}_n(t_2)) = a_n^2 B_n(t_1 - t_2), \quad n \in \mathbb{N}_0,$$

the random vector \mathbf{U} is uniformly distributed on \mathbb{M}^d and is independent with $\{\mathbf{V}_n(t) : n \in \mathbb{N}_0\}$, and $\sum_{n=0}^{\infty} B_n(0) P_n^{(\alpha, \beta)}(1)$ converges.

The distinct terms of (13) are uncorrelated each other,

$$\begin{aligned} \text{cov} \left(\mathbf{V}_i(t) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})), \mathbf{V}_j(t) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) &= \mathbf{0}, \\ \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}, i \neq j, \end{aligned}$$

due to Lemma 3 and the independent assumption among $\mathbf{U}, \mathbf{V}_i(t), \mathbf{V}_j(t)$. The vector stochastic process $\mathbf{V}_n(t)$ can be expressed as, in terms of $\mathbf{Z}(\mathbf{x}; t)$ and \mathbf{U} ,

$$\mathbf{V}_n(t) = \frac{a_n^2}{\omega_d P_n^{(\alpha, \beta)}(1)} \int_{\mathbb{M}^d} \mathbf{Z}(\mathbf{x}; t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}, \quad t \in \mathbb{T}, n \in \mathbb{N}_0,$$

where the integral is understood as a Bochner integral of a function taking values in the Hilbert space of random vectors $\mathbf{Z} \in \mathbb{R}^m$ with $E[\|\mathbf{Z}\|_{\mathbb{R}^m}^2] < \infty$.

It is obtained after we multiply both sides of (13) by $P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U}))$, integrate over \mathbb{M}^d , and apply Lemma 3,

$$\begin{aligned} &\int_{\mathbb{M}^d} \mathbf{Z}(\mathbf{x}; t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x} \\ &= \sum_{k=0}^{\infty} \mathbf{V}_k(t) \int_{\mathbb{M}^d} P_k^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x} \\ &= \frac{1}{a_n^2} P_n^{(\alpha, \beta)}(1) \mathbf{V}_n(t). \end{aligned}$$

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A Proofs

Proof of Lemma 1 To calculate $\mu(\mathbb{M}^d)$, we use the result of [41]. If all the geodesics on a d -dimensional Riemannian manifold M are closed and have length $2\pi L$, then the ratio

$$i(M) = \frac{\mu(\mathbb{M}^d)}{L^n \mu(\mathbb{S}^d)}$$

is an integer. With our convention $L = 1$, we obtain $\mu(\mathbb{M}^d) = i(\mathbb{M}^d)\mu(\mathbb{S}^d)$. It is well known that

$$\mu(\mathbb{S}^d) = \frac{2\pi^{(d+1)/2}}{\Gamma((d+1)/2)} = \frac{2\pi^{\alpha+3/2}}{\Gamma(\alpha+3/2)}. \tag{14}$$

The *Weinstein’s integers* $i(\mathbb{M}^d)$ are shown in the last column of Table 2. Following [36], consider all the geodesics from \mathbf{o} to a point in \mathbb{A} . Draw a tangent line to each of them and denote by e the dimension of the linear space generated by these lines. We have $e = d$ for \mathbb{S}^d , 1 for $P^d(\mathbb{R})$, 2 for $P^d(\mathbb{C})$, 4 for $P^d(\mathbb{H})$, and 8 for $P^2(\mathbb{O})$. It is proved in [36] that

$$i(\mathbb{M}^d) = \frac{2^{d-1} \Gamma((d+1)/2) \Gamma(e/2)}{\sqrt{\pi} \Gamma((d+e)/2)}$$

We know that $d = 2\alpha + 2$. It is easy to check that $e = 2\beta + 2$, then we obtain

$$i(\mathbb{M}^d) = \frac{2^{2\alpha+1} \Gamma(\alpha+3/2) \Gamma(\beta+1)}{\sqrt{\pi} \Gamma(\alpha+\beta+2)},$$

and (4) easily follows. □

Proof of Lemma 2 In Theorem 2.1 of [3], put $K(x) = P_i^{(\alpha,\beta)}(x)$ and $S(\mathbf{x}) = P_j^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_2, \mathbf{x})))$. We obtain

$$\begin{aligned} & \int_{\mathbb{M}^d} P_i^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}))) P_j^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_2, \mathbf{x}))) \, d\mathbf{x} \\ &= \omega_d P_j^{(\alpha,\beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))) \int_{-1}^1 \frac{P_i^{(\alpha,\beta)}(x)}{P_i^{(\alpha,\beta)}(1)} P_j^{(\alpha,\beta)}(x) \, d\nu_{\alpha,\beta}(x) \end{aligned}$$

$$= \omega_d \frac{\delta_{ij}}{a_i^2} P_i^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))),$$

where the last equality follows from (3), (5), and the following well-known result: the probabilistic measure $\nu_{\alpha, \beta}$ on $[-1, 1]$, proportional to $(1 - x)^\alpha(1 + x)^\beta dx$, is

$$d\nu_{\alpha, \beta}(x) = \frac{\Gamma(\alpha + \beta + 2)}{2^{\alpha + \beta + 1} \Gamma(\alpha + 1) \Gamma(\beta + 1)} (1 - x)^\alpha (1 + x)^\beta dx. \tag{15}$$

□

Proof of Lemma 3 The mean function of $\{Z_n(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ is obtained by applying of [3, Theorem 2.1] to $K(x) = 1$ and $S(\mathbf{x}) = P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{y})))$,

$$E[Z_n(\mathbf{x})] = a_n \omega_d \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{y}))) d\mathbf{y} = a_n \cdot 0 = 0.$$

The covariance function is

$$\begin{aligned} \text{cov}(Z_n(\mathbf{x}_1), Z_n(\mathbf{x}_2)) &= \omega_d^{-1} a_n^2 \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{z}))) P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_2, \mathbf{z}))) dz \\ &= P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}_1, \mathbf{x}_2))), \end{aligned}$$

by Lemma 2. Equation (7) easily follows from the same lemma. □

Proof of Theorem 1 The series at the right-hand side of (8) converges in mean square for every $\mathbf{x} \in \mathbb{M}^d$ since

$$\begin{aligned} &E \left[\left(\sum_{i=n_1}^{n_1+n_2} B_i^{\frac{1}{2}} \mathbf{V}_i P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) \left(\sum_{j=n_1}^{n_1+n_2} B_j^{\frac{1}{2}} \mathbf{V}_j P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right)^\top \right] \\ &= \sum_{i=n_1}^{n_1+n_2} \sum_{j=n_1}^{n_1+n_2} B_i^{\frac{1}{2}} B_j^{\frac{1}{2}} E[(\mathbf{V}_i \mathbf{V}_j^\top)] E \left[\left(P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) \right] \\ &= \sum_{i=n_1}^{n_1+n_2} B_i \sigma_i^2 E \left[\left(P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) \right] \\ &= \sum_{i=n_1}^{n_1+n_2} B_i P_i^{(\alpha, \beta)}(1) \\ &\rightarrow \mathbf{0}, \quad \text{as } n_1, n_2 \rightarrow \infty, \end{aligned}$$

where the second equality follows from the independent assumption between $\{\mathbf{V}_n : n \in \mathbb{N}_0\}$ and \mathbf{U} , and the third from Lemma 3. Thus, (8) is an m -variate second-order random field. Its mean function is clearly identical to $\mathbf{0}$, and its covariance function is

$$\begin{aligned}
 & \text{cov} \left(\sum_{i=0}^{\infty} B_i^{\frac{1}{2}} \mathbf{V}_i P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{U})), \sum_{j=0}^{\infty} B_j^{\frac{1}{2}} \mathbf{V}_j P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_2, \mathbf{U})) \right) \\
 &= \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} B_i^{\frac{1}{2}} B_j^{\frac{1}{2}} \mathbb{E} \left[(\mathbf{V}_i \mathbf{V}_j^{\top}) \mathbb{E} \left[\left(P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) \right] \right] \\
 &= \sum_{i=0}^{\infty} B_i \sigma_i^2 \mathbb{E} \left[\left(P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{U})) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_2, \mathbf{U})) \right) \right] \\
 &= \sum_{i=0}^{\infty} B_i P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d.
 \end{aligned}$$

Two distinct terms of (8) are obviously uncorrelated each other. □

Proof of Theorem 2 It suffices to verify (10) to be a general form, since in Theorem 1 we already construct an m -variate isotropic random field on \mathbb{M}^d whose covariance matrix function is (10). To this end, suppose that $\{\mathbf{Z}(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ is an m -variate isotropic and mean square continuous random field. Then, for an arbitrary $\mathbf{a} \in \mathbb{R}^m$, $\{\mathbf{a}^{\top} \mathbf{Z}(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$ is a scalar isotropic and mean square continuous random field, so that its covariance function has to be of the form (1),

$$\text{cov} \left(\mathbf{a}^{\top} \mathbf{Z}(\mathbf{x}_1), \mathbf{a}^{\top} \mathbf{Z}(\mathbf{x}_2) \right) = \sum_{n=0}^{\infty} b_n(\mathbf{a}) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \quad (16)$$

where $\{b_n(\mathbf{a}) : n \in \mathbb{N}_0\}$ is a sequence of nonnegative constants and $\sum_{n=0}^{\infty} b_n(\mathbf{a}) P_n^{(\alpha, \beta)}$ (1) converges. Similarly, for $\mathbf{b} \in \mathbb{R}^m$, we obtain

$$\begin{aligned}
 & \frac{1}{4} \text{cov}((\mathbf{a} + \mathbf{b})^{\top} \mathbf{Z}(\mathbf{x}_1), (\mathbf{a} + \mathbf{b})^{\top} \mathbf{Z}(\mathbf{x}_2)) \\
 &= \sum_{n=0}^{\infty} b_n(\mathbf{a} + \mathbf{b}) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \\
 & \frac{1}{4} \text{cov}((\mathbf{a} - \mathbf{b})^{\top} \mathbf{Z}(\mathbf{x}_1), (\mathbf{a} - \mathbf{b})^{\top} \mathbf{Z}(\mathbf{x}_2)) \\
 &= \sum_{n=0}^{\infty} b_n(\mathbf{a} - \mathbf{b}) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d.
 \end{aligned}$$

Taking the difference between the last two equations yields

$$\begin{aligned}
 & \frac{1}{2} \left(\mathbf{a}^{\top} \text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2)) \mathbf{b} + \mathbf{b}^{\top} \text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2)) \mathbf{a} \right) \\
 &= \frac{1}{2} \left(\text{cov}(\mathbf{a}^{\top} \mathbf{Z}(\mathbf{x}_1), \mathbf{b}^{\top} \mathbf{Z}(\mathbf{x}_2)) + \text{cov}(\mathbf{b}^{\top} \mathbf{Z}(\mathbf{x}_1), \mathbf{a}^{\top} \mathbf{Z}(\mathbf{x}_2)) \right) \\
 &= \sum_{n=0}^{\infty} (b_n(\mathbf{a} + \mathbf{b}) - b_n(\mathbf{a} - \mathbf{b})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d,
 \end{aligned}$$

or

$$\mathbf{a}^\top \text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2))\mathbf{b} = \sum_{n=0}^\infty (b_n(\mathbf{a}+\mathbf{b}) - b_n(\mathbf{a}-\mathbf{b})) P_n^{(\alpha,\beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \tag{17}$$

noticing that $\text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2))$ is a symmetric matrix. The form (10) of $\text{cov}(\mathbf{Z}(\mathbf{x}_1), \mathbf{Z}(\mathbf{x}_2))$ is now confirmed by letting the i th entry of \mathbf{a} and the j th entry of \mathbf{b} be 1 and the rest vanish in (17). It remains to verify the nonnegative definiteness of each B_n in (10). To do so, we multiply its both sides by \mathbf{a}^\top from the left and \mathbf{a} from the right, and obtain

$$\mathbf{a}^\top C(\mathbf{x}_1, \mathbf{x}_2)\mathbf{a} = \sum_{n=0}^\infty \mathbf{a}^\top B_n \mathbf{a} P_n^{(\alpha,\beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d,$$

comparing which with (16) results in that $\mathbf{a}^\top B_n \mathbf{a} \geq 0$ or the nonnegative definiteness of B_n , $n \in \mathbb{N}_0$, and the convergence of $\sum_{n=0}^\infty \mathbf{a}^\top B_n \mathbf{a} P_n^{(\alpha,\beta)}(1)$ or that of each entry of the matrix $\sum_{n=0}^\infty B_n P_n^{(\alpha,\beta)}(1)$. \square

Proof of Theorem 3 For a fixed $t \in \mathbb{T}$, consider a random field $\{\mathbf{Z}(\mathbf{x}; 0) + \mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d\}$. It is isotropic and mean square continuous on \mathbb{M}^d , with covariance matrix function

$$\begin{aligned} &\text{cov}(\mathbf{Z}(\mathbf{x}_1; 0) + \mathbf{Z}(\mathbf{x}_1; t), \mathbf{Z}(\mathbf{x}_2; 0) + \mathbf{Z}(\mathbf{x}_2; t)) \\ &= 2C(\rho(\mathbf{x}_1, \mathbf{x}_2); 0) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t) \\ &= \sum_{n=0}^\infty B_{n+}(t) P_n^{(\alpha,\beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, \end{aligned}$$

where the last equality follows from Theorem 2, $\{B_{n+}(t) : n \in \mathbb{N}_0\}$ is a sequence of nonnegative definite matrices, and $\sum_{n=0}^\infty B_{n+}(t) P_n^{(\alpha,\beta)}(1)$ converges. Similarly, we have

$$\begin{aligned} &\text{cov}(\mathbf{Z}(\mathbf{x}_1; 0) - \mathbf{Z}(\mathbf{x}_1; t), \mathbf{Z}(\mathbf{x}_2; 0) - \mathbf{Z}(\mathbf{x}_2; t)) \\ &= 2C(\rho(\mathbf{x}_1, \mathbf{x}_2); 0) - C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) - C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t) \\ &= \sum_{n=0}^\infty B_{n-}(t) P_n^{(\alpha,\beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \end{aligned}$$

and thus,

$$\begin{aligned} &\frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)}{2} \\ &= \frac{1}{4}[2C(\rho(\mathbf{x}_1, \mathbf{x}_2); 0) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)] \end{aligned}$$

$$\begin{aligned}
 & -\frac{1}{4}[2C(\rho(\mathbf{x}_1, \mathbf{x}_2); 0) - C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) - C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)] \\
 & = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d,
 \end{aligned}$$

which confirms the format (11) for $\frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); -t)}{2}$, with $B_n(t) = \frac{B_{n+(t)} - B_{n-(t)}}{4}$, $n \in \mathbb{N}_0$. Obviously, $B_n(t)$ is symmetric, and $\sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(1)$ converges. Moreover, (11) is the covariance matrix function of an m -variate isotropic random field $\left\{ \frac{\mathbf{Z}(\mathbf{x}; t) + \tilde{\mathbf{Z}}(\mathbf{x}; -t)}{\sqrt{2}} : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \right\}$, where $\{ \tilde{\mathbf{Z}}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \}$ is an independent copy of $\{ \mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \}$. In fact,

$$\begin{aligned}
 & \text{cov} \left(\frac{\mathbf{Z}(\mathbf{x}_1; t_1) + \tilde{\mathbf{Z}}(\mathbf{x}_1; -t_1)}{\sqrt{2}}, \frac{\mathbf{Z}(\mathbf{x}_2; t_2) + \tilde{\mathbf{Z}}(\mathbf{x}_2; -t_2)}{\sqrt{2}} \right) \\
 & = \frac{C(\rho(\mathbf{x}_1, \mathbf{x}_2); t_1 - t_2) + C(\rho(\mathbf{x}_1, \mathbf{x}_2); t_2 - t_1)}{2} \\
 & = \sum_{k=0}^{\infty} B_k(t_1 - t_2) P_k^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2))
 \end{aligned}$$

with $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t_1, t_2 \in \mathbb{T}$.

For each fixed $n \in \mathbb{N}_0$, in order to verify that $B_n(t)$ is a stationary covariance matrix function on \mathbb{T} , we consider an m -variate stochastic process

$$\mathbf{W}_n(t) = \int_{\mathbb{M}^d} \frac{\mathbf{Z}(\mathbf{x}; t) + \tilde{\mathbf{Z}}(\mathbf{x}; -t)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}, \quad t \in \mathbb{T},$$

where $\{ \tilde{\mathbf{Z}}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \}$ is an independent copy of $\{ \mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \}$, \mathbf{U} is a random vector uniformly distributed on \mathbb{M}^d , and $\mathbf{U}, \{ \mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T} \}$ and $\{ \tilde{\mathbf{Z}}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{S}^d, t \in \mathbb{T} \}$ are independent. By Lemma 2, the mean function of $\{ \mathbf{W}_n(t) : t \in \mathbb{T} \}$ is

$$\mathbb{E}[\mathbf{W}_n(t)] = \begin{cases} \sqrt{2} P_0^{(\alpha, \beta)}(1) \omega_d \mathbb{E}[\mathbf{Z}(\mathbf{x}; t)], & n = 0, \\ 0, & n \in \mathbb{N}, \end{cases}$$

and its covariance matrix function is by Lemmas 2 and 3

$$\begin{aligned}
 & \text{cov}(\mathbf{W}_n(t_1), \mathbf{W}_n(t_2)) \\
 & = \frac{1}{\omega_d} \text{cov} \left(\int_{\mathbb{M}^d} \frac{\mathbf{Z}(\mathbf{x}; t_1) + \tilde{\mathbf{Z}}(\mathbf{x}; -t_1)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}, \right. \\
 & \quad \left. \int_{\mathbb{M}^d} \frac{\mathbf{Z}(\mathbf{y}; t_2) + \tilde{\mathbf{Z}}(\mathbf{y}; -t_2)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{U})) d\mathbf{y} \right)
 \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \operatorname{cov} \left(\int_{\mathbb{M}^d} \frac{\mathbf{Z}(\mathbf{x}; t_1) + \tilde{\mathbf{Z}}(\mathbf{x}; -t_1)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}, \right. \\
&\quad \left. \int_{\mathbb{M}^d} \frac{\mathbf{Z}(\mathbf{y}; t_2) + \tilde{\mathbf{Z}}(\mathbf{y}; -t_2)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{y} \right) d\mathbf{u} \\
&= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \operatorname{cov} \left(\frac{\mathbf{Z}(\mathbf{x}; t_1) + \tilde{\mathbf{Z}}(\mathbf{x}; -t_1)}{\sqrt{2}}, \frac{\mathbf{Z}(\mathbf{y}; t_2) + \tilde{\mathbf{Z}}(\mathbf{y}; -t_2)}{\sqrt{2}} \right) \\
&\quad \times P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{u})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{x} d\mathbf{y} d\mathbf{u} \\
&= \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \frac{\mathbf{C}(\rho(\mathbf{x}, \mathbf{y}); t_1 - t_2) + \mathbf{C}(\rho(\mathbf{x}, \mathbf{y}); t_2 - t_1)}{2\omega_d} \\
&\quad \times P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{u})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{x} d\mathbf{y} d\mathbf{u} \\
&= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \sum_{k=0}^{\infty} \mathbf{B}_k(t_1 - t_2) P_k^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{y})) \\
&\quad \times P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{u})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{x} d\mathbf{y} d\mathbf{u} \\
&= \frac{1}{\omega_d} \sum_{k=0}^{\infty} \mathbf{B}_k(t_1 - t_2) \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} P_k^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{y})) \\
&\quad \times P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{u})) d\mathbf{x} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{y} d\mathbf{u} \\
&= \frac{1}{\omega_d} \mathbf{B}_n(t_1 - t_2) \int_{\mathbb{M}^d} \frac{1}{a_n^2} \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{y}, \mathbf{u})) d\mathbf{y} d\mathbf{u} \\
&= \frac{1}{\omega_d} \mathbf{B}_n(t_1 - t_2) \int_{\mathbb{M}^d} \left(\frac{\omega_d}{a_n^2} \right)^2 P_n^{(\alpha, \beta)}(1) d\mathbf{u} \\
&= \mathbf{B}_n(t_1 - t_2) \left(\frac{\omega_d}{a_n^2} \right)^2 P_n^{(\alpha, \beta)}(1), \quad t_1, t_2 \in \mathbb{T},
\end{aligned}$$

which implies that $\mathbf{B}_n(t)$ is a stationary covariance matrix function on \mathbb{T} . \square

Proof of Theorem 4 The convergent assumption of $\sum_{n=0}^{\infty} \mathbf{B}_n(0) P_n^{(\alpha, \beta)}(1)$ ensures the uniform and absolute convergence of the series at the right-hand side of (12). If $\{\mathbf{B}_n(t) : n \in \mathbb{N}_0\}$ is a sequence of stationary covariance matrix function on \mathbb{T} , then each term of the series at the right-hand side of (12) is the product of a stationary covariance matrix function $\mathbf{B}_n(t)$ on \mathbb{T} and an isotropic covariance function $P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2))$ on \mathbb{M}^d , and thus, (12) can be treated [21] as the covariance matrix function of an m -variate random field on $\mathbb{M}^d \times \mathbb{T}$.

On the other hand, assume that (12) is the covariance matrix function of an m -variate random field $\{\mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$. The convergence of $\sum_{n=0}^{\infty} \mathbf{B}_n(0) P_n^{(\alpha, \beta)}(1)$ results from the existence of $\mathbf{C}(0; 0) = \operatorname{Var}[Z(\mathbf{x}; t)]$. In order to show that $\mathbf{B}_n(t)$ is a stationary covariance matrix function on \mathbb{T} for each fixed $n \in \mathbb{N}_0$, consider an m -variate stochastic process

$$\mathbf{W}_n(t) = \int_{\mathbb{M}^d} \mathbf{Z}(\mathbf{x}; t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}, \quad t \in \mathbb{T},$$

where \mathbf{U} is a random vector uniformly distributed on \mathbb{M}^d and is independent with $\{\mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$. Similar to that in the proof of Theorem 3, applying Lemmas 2 and 3 we obtain that the covariance matrix function of $\{\mathbf{W}_n(t) : t \in \mathbb{T}\}$ is positively propositional to $B_n(t)$; more precisely,

$$\text{cov}(\mathbf{W}_n(t_1), \mathbf{W}_n(t_2)) = B_n(t_1 - t_2) \left(\frac{\omega_d}{a_n^2}\right)^2 P_n^{(\alpha, \beta)}(1), \quad t_1, t_2 \in \mathbb{T},$$

which implies that $B_n(t)$ is a stationary covariance matrix function on \mathbb{T} . □

Proof of Theorem 5 The convergent assumption of $\sum_{n=0}^\infty B_n(0) P_n^{(\alpha, \beta)}(1)$ ensures the mean square convergence of the series at the right-hand side of (13), since

$$\begin{aligned} & \mathbb{E} \left[\left(\sum_{i=n_1}^{n_1+n_2} \mathbf{V}_i(t) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) \left(\sum_{j=n_1}^{n_1+n_2} \mathbf{V}_j(t) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right)^\top \right] \\ &= \mathbb{E} \left[\sum_{i=n_1}^{n_1+n_2} \sum_{j=n_1}^{n_1+n_2} \mathbf{V}_i(t) \mathbf{V}_j^\top(t) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right] \\ &= \sum_{i=n_1}^{n_1+n_2} \sum_{j=n_1}^{n_1+n_2} \mathbb{E}[\mathbf{V}_i(t) \mathbf{V}_j^\top(t)] \mathbb{E} \left[P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right] \\ &= \omega_d \sum_{i=n_1}^{n_1+n_2} B_i(0) P_i^{(\alpha, \beta)}(1) \\ &\rightarrow 0, \quad \text{as } n_1, n_2 \rightarrow \infty, \end{aligned}$$

where the second equality follows from the independent assumption between \mathbf{U} and $\{\mathbf{V}_n(t) : n \in \mathbb{N}_0\}$, and the third one from Lemma 3. Applying Lemma 3 we obtain the mean and covariance matrix functions of $\{\mathbf{Z}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$, under the independent assumption among \mathbf{U} and $\{\mathbf{V}_n(t) : n \in \mathbb{N}_0\}$,

$$\mathbb{E}[\mathbf{Z}(\mathbf{x}; t)] = \sum_{n=0}^\infty \mathbb{E}[\mathbf{V}_n(t)] \mathbb{E} \left[P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right] = \mathbf{0}, \quad \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T},$$

and

$$\begin{aligned} & \text{cov}(\mathbf{Z}(\mathbf{x}_1; t_1), \mathbf{Z}(\mathbf{x}_2; t_2)) \\ &= \text{cov} \left(\sum_{i=0}^\infty \mathbf{V}_i(t_1) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{U})), \sum_{j=0}^\infty \mathbf{V}_j(t_2) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_2, \mathbf{U})) \right) \end{aligned}$$

$$\begin{aligned}
&= \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} E[\mathbf{V}_i(t_1) \mathbf{V}_j^{\top}(t_2)] E \left[P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{U})) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_2, \mathbf{U})) \right] \\
&= \sum_{n=0}^{\infty} B_n(t_1 - t_2) \frac{1}{a_n^2} P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2)), \quad \mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d, t_1, t_2 \in \mathbb{T}.
\end{aligned}$$

The latter is obviously isotropic and continuous on \mathbb{M}^d and stationary on \mathbb{T} . \square

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