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

Chunsheng Ma1 · Anatoliy Malyarenko2

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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 $S^d$ embedded into $\mathbb{R}^{d+1}$ as follows: $S^d = \{ \mathbf{x} \in \mathbb{R}^{d+1} : \|\mathbf{x}\| = 1 \}$, and define the distance between the points $\mathbf{x}_1$ and $\mathbf{x}_2$ by $\rho(\mathbf{x}_1, \mathbf{x}_2) = \cos^{-1}(\mathbf{x}_1^T \mathbf{x}_2)$. With this distance, any isometry between two pairs of points can be extended to an isometry of $S^d$. 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 $M^d$, where $d$ is the topological dimension of the manifold. Following
covariance matrix function \( M \) in the temporal domain.

Let \((\Omega, \mathcal{F}, \mathbb{P})\) 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 \( \mathbb{E}[\mathbf{Z}(\mathbf{x}; t)] \) equals a constant vector, and its covariance matrix function

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

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

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 \( C(\rho(\mathbf{x}_1, \mathbf{x}_2); t) \),

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

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

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

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

holds for every \( n \in \mathbb{N} \), any \( \mathbf{x}_i \in \mathbb{M}^d, t_i \in \mathbb{T} \), and \( a_i \in \mathbb{R}^m (i = 1, 2, \ldots, n) \), 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 \( \sum_{n=0}^{\infty} b_n P_n^{(\alpha, \beta)}(\mathbf{x}) \) convergent, \( P_n^{(\alpha, \beta)}(\mathbf{x}) \) is a Jacobi polynomial of degree \( n \) with a pair of parameters \((\alpha, \beta)\) [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 $M^d \times \mathbb{T}$ in this paper.

Isotropic random fields over $S^d$ 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 $S^2$ [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 $M^d$, $\rho(x_1, x_2)$, 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 $M^d$ 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 $M^d$, 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 $M^d \times \mathbb{T}$, which is isotropic and mean square continuous vector random field on $M^d$ 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 $M^d$ and those of Jacobi polynomials and then establishes two useful lemmas on a special case of the Funk–Hecke formula on $M^d$ 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 $\mathbb{P}^1(\mathbb{R}) = S^1$, $\mathbb{P}^2(\mathbb{C}) = S^2$, $\mathbb{P}^4(\mathbb{H}) = S^4$, and $\mathbb{P}^8(\mathbb{O}) = S^8$.

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 $S^d$ 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, $2\pi$. It is natural to norm the distance in such a way that the length of any geodesic line is equal to $2\pi$, 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.
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 \( \Delta \) called the Laplace–Beltrami operator. Let \( d\nu(x) \) 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 \( \Delta \) as a self-adjoint operator in the space \( H = L^2(\mathbb{M}^d, d\nu(x)) \). The spectrum of \( \Delta \) is discrete, and the eigenvalues are

\[ \lambda_n = -\epsilon n(\epsilon n + \alpha + \beta + 1), \quad n \in \mathbb{N}_0, \]

where

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

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

Let \( H_n \) be the eigenspace of \( \Delta \) corresponding to \( \lambda_n \). 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 \( f_n(x) \), depends only on the distance \( r = \rho(x, o) \), \( f_n(x) = f_n^*(r) \).

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

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

\( x \in [-1, 1], \ n \in \mathbb{N}_0, \)

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,

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

where

\[ P_n^{(\alpha, \beta)}(1) = \frac{\Gamma(n + \alpha + 1)}{\Gamma(n + 1) \Gamma(\alpha + 1)}. \] (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 \( S^d \). An \( O(d) \)-invariant function \( f \) on \( \mathbb{P}^d(\mathbb{R}) \) can be lifted to an \( SO(d) \)-invariant function \( g \) on \( S^d \) by \( g(\mathbf{x}) = f(\pi(\mathbf{x})) \), where \( \pi \) maps a point \( \mathbf{x} \in S^d \) 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 \( \mathbb{M}^d = S^1 \), 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.
Then we have \( \cos(\rho(\text{manifold of the point } x, o) / 2) \). Then we have \( \cos(\rho(x, o)) = 2y^2 - 1 \). For the case of \( M^d = \mathbb{P}d(\mathbb{R}) \), \( \alpha = \beta = (d - 2)/2 \). 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

\[
R_n^{(\alpha, \alpha)}(\cos(\rho(x, o)/2)) = R_n^{(\alpha, -1/2)}(\cos(\rho(x, o))).
\]

For the case of \( M^d = \mathbb{P}d(\mathbb{R}) \), 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 \( R_n^{(\alpha, \beta)}(\cos(\rho(x, o))) \).

It easily follows from (2) that the new values for \( p \) and \( q \) in the case of \( M^d = \mathbb{P}d(\mathbb{R}) \) 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 \( M^d \) rather than with Lie algebras.

Specifically, let \( A = \{ x \in M^d : \rho(x, o) = \pi \} \). This set is called the antipodal manifold of the point \( o \). The antipodal manifolds are listed in the sixth column of Table 2. Geometrically, if \( M^d = \mathbb{S}^d \) and \( o \) is the North pole, then \( A = \mathbb{S}^0 \) is the South pole. Otherwise, \( A \) is the space at infinity of the point \( o \) 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 \( M^d \) itself.

In what follows, we use the geometric approach. It turns out that all the spaces \( M^d \) are Riemannian manifolds, as is defined in [5]. Each Riemannian manifold carries the canonical measure \( \mu \); see [5, pp. 10–11]. The measure \( \mu \) is proportional to the measure \( \nu \) constructed in Sect. 2.1. The coefficient of proportionality or the total measure \( \mu(M^d) \) of the compact manifold \( M^d \) is called the volume of \( M^d \).
Lemma 1  The volume of the space \( M^d \) is

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

(4)

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

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

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

(5)

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

Lemma 2  For \( i, j \in \mathbb{N}_0 \), and \( x_1, x_2 \in M^d \),

\[
\int_{M^d} P_i^{(\alpha,\beta)}(\cos(\rho(x_1, x))) P_j^{(\alpha,\beta)}(\cos(\rho(x_2, x))) \, dx = \frac{\delta_{ij} \omega_d}{a_n^2} P_i^{(\alpha,\beta)}(\cos(\rho(x_1, x_2))),
\]

where

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

(6)

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

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

To construct \( U \), we start with a measure \( \sigma \) proportional to the invariant measure \( \nu \) of Sect. 2.1. Let \( T_0 \) be the tangent space to \( M^d \) at the point \( o \). Choose a Cartesian coordinate system in \( T_0 \) and identify this space with the space \( \mathbb{R}^d \). Construct a chart \( \varphi : M^d \setminus A \to \mathbb{R}^d \) as follows. Put \( \varphi(o) = 0 \in \mathbb{R}^d \). For any other point \( x \in M^d \setminus A \), draw the unique geodesic line connecting \( o \) and \( x \). Let \( r \in \mathbb{R}^d \) be the unit tangent vector to the above geodesic line. Define

\[
\varphi(x) = r \tan(\rho(x, o)/2),
\]
and, for each Borel set $B \subseteq \mathbb{M}^d$, 

$$\sigma(B) = \int_{\phi^{-1}(B \setminus A)} \frac{dx}{(1 + \|x\|^2)^{\alpha+\beta+2}}.$$ 

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

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

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 $n \in \mathbb{N}$, 

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

is a centred isotropic random field with covariance function 

$$\text{cov}(Z_n(x_1), Z_n(x_2)) = \mathcal{P}_n^{(\alpha,\beta)}(\cos(\rho(x_1, x_2))), \quad x_1, x_2 \in \mathbb{M}^d,$$

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

$$\text{cov}(Z_k(x_1), Z_n(x_2)) = 0, \quad x_1, x_2 \in \mathbb{M}^d.$$ (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 $\{Z(x) : x \in \mathbb{M}^d\}$ and a series expression for its covariance matrix function, in terms of Jacobi polynomials. By mean square continuous, we mean that, for $k = 1, \ldots, m$, 

$$E\left[|Z_k(x_1) - Z_k(x_2)|^2\right] \to 0, \quad \text{as } \rho(x_1, x_2) \to 0, \quad x_1, x_2 \in \mathbb{M}^d.$$ 

It implies the continuity of each entry of the associated covariance matrix function in terms of $\rho(x_1, x_2)$.

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 $\lambda_1, \ldots, \lambda_m$, there is an orthogonal matrix $S$ such that $S^{-1}BS = D$, where $D$ is a diagonal matrix with diagonal entries $\lambda_1, \ldots, \lambda_m$. 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 $\sqrt{\lambda_1}, \ldots, \sqrt{\lambda_m}$. 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 \{ $V_n : n \in \mathbb{N}_0$ \} is a sequence of independent $m$-variate random vectors with $E(V_n) = 0$ and $\text{cov}(V_n, V_n) = a^2nI_m$. $U$ is a random vector uniformly distributed on $M^d$ and is independent of \{ $V_n : n \in \mathbb{N}_0$ \}, 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

$$Z(x) = \sum_{n=0}^{\infty} B_n^\frac{1}{2} V_n P_{n}^{(\alpha, \beta)}(\cos \rho(x, U)), \quad x \in M^d,$$

(8)

is a centred $m$-variate isotropic random field on $M^d$, with covariance matrix function

$$\text{cov}(Z(x_1), Z(x_2)) = \sum_{n=0}^{\infty} B_n P_{n}^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in M^d. \quad (9)$$

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

$$\text{cov} \left( B_i^\frac{1}{2} V_i P_{i}^{(\alpha, \beta)}(\rho(x_1, U)), \ B_j^\frac{1}{2} V_j P_{j}^{(\alpha, \beta)}(\rho(x_2, U)) \right) = 0, \quad x_1, x_2 \in M^d, \ i \neq j.$$

Since $\left| P_{n}^{(\alpha, \beta)}(\cos \theta) \right| \leq P_{n}^{(\alpha, \beta)}(1), n \in \mathbb{N}_0$, 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 $M^d = S^2$ and $m = 1$, we have $\dim H_n = 2n + 1$, and (9) takes the form

$$\text{cov}(Z(x_1), Z(x_2)) = \sum_{n=0}^{\infty} b_n P_{n}(\cos \rho(x_1, x_2)),$$

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

$$\text{cov}(Z(x_1), Z(x_2)) = \sum_{\ell=0}^{\infty} (2\ell + 1) C_{\ell} P_{\ell}(x_1 \cdot x_2),$$

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 \( \{Z(x) : x \in \mathbb{M}^d\} \), its covariance matrix function \( \text{cov}(Z(x_1), Z(x_2)) \) is of the form

\[
C(x_1, x_2) = \sum_{n=0}^{\infty} B_n P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in \mathbb{M}^d, \tag{10}
\]

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 \( C(x_1, x_2) \) 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 \( \{Z(x; t) : x \in \mathbb{M}^d, t \in \mathbb{T}\} \) 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 $C(\rho(x_1, x_2); t)$, which is a continuous function of $\rho(x_1, x_2)$ 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 \( \{Z(x; t), x \in \mathbb{M}^d, t \in \mathbb{T}\} \) is isotropic and mean square continuous over $\mathbb{M}^d$ and stationary on $\mathbb{T}$, then

\[
C(\rho(x_1, x_2); -t) = (C(\rho(x_1, x_2); t))^\top,
\]

and \( \frac{C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)}{2} \) is of the form

\[
\frac{C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)}{2} = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in \mathbb{M}^d, t \in \mathbb{T}, \tag{11}
\]

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 \( \frac{C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)}{2} \), instead of \( C(\rho(x_1, x_2); t) \) itself, is given in Theorem 3, that of \( C(\rho(x_1, x_2); t) \) can be obtained in certain special cases, such as spatio-temporal symmetric, and purely spatial.

**Corollary 1** If \( C(\rho(x_1, x_2); t) \) is spatio-temporal symmetric in the sense that

\[
C(\rho(x_1, x_2); -t) = C(\rho(x_1, x_2); t), \quad x_1, x_2 \in \mathbb{M}^d, \quad t \in \mathbb{T},
\]

then it takes the form

\[
C(\rho(x_1, x_2); t) = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in \mathbb{M}^d, \quad t \in \mathbb{T}.
\]

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

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

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

**Theorem 4** An \( m \times m \) matrix function

\[
C(\rho(x_1, x_2); t) = \sum_{n=0}^{\infty} B_n(t) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in \mathbb{M}^d, \quad t \in \mathbb{T},
\]

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

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

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
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, $\chi^2$, K-distributed, 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

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

is isotropic and mean square continuous on $\mathbb{M}^d$, stationary on $\mathbb{T}$, and possesses mean 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

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

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,

$$\text{cov} \left( \mathbf{V}_i(t) P_i^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})), \mathbf{V}_j(t) P_j^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) \right) = 0,$$

$\mathbf{x} \in \mathbb{M}^d, \ t \in \mathbb{T}, \ i \neq j,$

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,

$$\int_{\mathbb{M}^d} \mathbf{Z}(\mathbf{x}; t) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}$$

$$= \sum_{k=0}^{\infty} \mathbf{V}_n(t) \int_{\mathbb{M}^d} P_k^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}, \mathbf{U})) d\mathbf{x}$$

$$= \frac{1}{a_n^2} P_n^{(\alpha, \beta)}(1) \mathbf{V}_n(t).$$
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A Proofs

Proof of Lemma 1 To calculate \( \mu(M^d) \), 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

\[
i(M) = \frac{\mu(M^d)}{L^n \mu(S^d)}
\]

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

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

The Weinstein’s integers \( i(M^d) \) are shown in the last column of Table 2. Following [36], consider all the geodesics from \( o \) to a point in \( A \). 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 \( S^d \), 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

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

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

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

and (4) easily follows. \( \square \)

Proof of Lemma 2 In Theorem 2.1 of [3], put \( K(x) = P_i^{(\alpha,\beta)}(x) \) and \( S(x) = P_j^{(\alpha,\beta)}(\cos(\rho(x_2, x))) \). We obtain

\[
\int_{M^d} P_i^{(\alpha,\beta)}(\cos(\rho(x_1, x))) P_j^{(\alpha,\beta)}(\cos(\rho(x_2, x))) \, dx
\]

\[
= \omega_d P_j^{(\alpha,\beta)}(\cos(\rho(x_1, x_2))) \int_{-1}^{1} \frac{P_i^{(\alpha,\beta)}(x)}{P_i^{(\alpha,\beta)}(1)} P_j^{(\alpha,\beta)}(x) d\nu_{\alpha,\beta}(x)
\]
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

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

\[\square\]

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

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

The covariance function is

\[
\text{cov}(Z_n(x_1), Z_n(x_2)) = \omega_d^{-1} a_n^2 \int_{M^d} P_n^{(\alpha,\beta)}(\cos(\rho(x_1,z)))P_n^{(\alpha,\beta)}(\cos(\rho(x_2,z))) \, dz
\]

\[= P_n^{(\alpha,\beta)}(\cos(\rho(x_1,x_2))),\]

by Lemma 2. Equation (7) easily follows from the same lemma. \[\square\]

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

\[
E \left[ \left( \sum_{i=1}^{n_1+n_2} B_i^\top V_i P_i^{(\alpha,\beta)}(\cos(\rho(x,U))) \right) \left( \sum_{j=1}^{n_1+n_2} B_j^\top V_j P_j^{(\alpha,\beta)}(\cos(\rho(x,U))) \right) \right]^\top
\]

\[= \sum_{i=1}^{n_1+n_2} \sum_{j=1}^{n_1+n_2} B_i^\top B_j^\top E[(V_i V_j^\top)]E \left[ \left( P_i^{(\alpha,\beta)}(\cos(\rho(x,U)))P_j^{(\alpha,\beta)}(\cos(\rho(x,U))) \right) \right]
\]

\[= \sum_{i=1}^{n_1+n_2} B_i \sigma_i^2 E \left[ \left( P_i^{(\alpha,\beta)}(\cos(\rho(x,U)))P_i^{(\alpha,\beta)}(\rho(x,U))) \right) \right]
\]

\[= \sum_{i=1}^{n_1+n_2} B_i P_i^{(\alpha,\beta)}(1)
\]

\[\rightarrow 0, \quad \text{as } n_1, n_2 \rightarrow \infty,\]

where the second equality follows from the independent assumption between \( \{ V_n : n \in \mathbb{N}_0 \} \) and $U$, and the third from Lemma 3. Thus, (8) is an $m$-variate second-order random field. Its mean function is clearly identical to $0$, and it covariance function is
The covariance function has to be of the form (1), so that its covariance function is (10). To this end, suppose that we already construct an $m$-variate isotropic and mean square continuous random field. Then, for an arbitrary $\{ a_i \}_{i=0}^{\infty}$, where $\{ a_i \}_{i=0}^{\infty}$ is a scalar isotropic and mean square continuous random field, $\{ a_i \}_{i=0}^{\infty}$ is an $m$-variate isotropic and mean square continuous random field on $M^d$. Then, for an arbitrary $a \in \mathbb{R}^m$, $\{ a^T Z(x) : x \in M^d \}$ is a scalar isotropic and mean square continuous random field, so that its covariance function has to be of the form (1),

$$
\text{cov} \left( \sum_{i=0}^{\infty} B_i^\frac{1}{2} V_i P_i^{(\alpha, \beta)}(\cos \rho(x_1, U)), \sum_{j=0}^{\infty} B_j^\frac{1}{2} V_j P_j^{(\alpha, \beta)}(\cos \rho(x_2, U)) \right) = \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} B_i^\frac{1}{2} B_j^\frac{1}{2} \mathbb{E} \left[ (V_i V_j^T) \mathbb{E} \left( P_i^{(\alpha, \beta)}(\cos \rho(x, U)) P_j^{(\alpha, \beta)}(\cos \rho(x, U)) \right) \right] = \sum_{i=0}^{\infty} B_i \sigma_i^2 \mathbb{E} \left[ P_i^{(\alpha, \beta)}(\cos \rho(x_1, U)) P_i^{(\alpha, \beta)}(\cos \rho(x_2, U)) \right] = \sum_{i=0}^{\infty} B_i P_i^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in M^d.
$$

Two distinct terms of (8) are obviously uncorrelated each other. \hfill \Box

**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 $M^d$ whose covariance matrix function is (10). To this end, suppose that $\{ Z(x) : x \in M^d \}$ is an $m$-variate isotropic and mean square continuous random field. Then, for an arbitrary $a \in \mathbb{R}^m$, $\{ a^T Z(x) : x \in M^d \}$ is a scalar isotropic and mean square continuous random field, so that its covariance function has to be of the form (1),

$$
\text{cov} \left( a^T Z(x_1), a^T Z(x_2) \right) = \sum_{n=0}^{\infty} b_n(a) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in M^d, \quad (16)
$$

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

$$
\frac{1}{4} \text{cov} \left( (a + b)^T Z(x_1), (a + b)^T Z(x_2) \right) = \sum_{n=0}^{\infty} b_n(a + b) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)),
$$

$$
\frac{1}{4} \text{cov} \left( (a - b)^T Z(x_1), (a - b)^T Z(x_2) \right) = \sum_{n=0}^{\infty} b_n(a - b) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in M^d.
$$

Taking the difference between the last two equations yields

$$
\frac{1}{2} \left( \text{cov}(Z(x_1), Z(x_2)) b + b^T \text{cov}(Z(x_1), Z(x_2)) a \right) = \frac{1}{2} \left( \text{cov}(a^T Z(x_1), b^T Z(x_2)) + \text{cov}(b^T Z(x_1), a^T Z(x_2)) \right) = \sum_{n=0}^{\infty} (b_n(a + b) - b_n(a - b)) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in M^d.
$$
or

\[ a^\top \text{cov}(Z(x_1), Z(x_2))b = \sum_{n=0}^{\infty} (b_n(a+b) - b_n(a-b)) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \ x_1, x_2 \in \mathbb{M}^d, \]  

(17)

noticing that \( \text{cov}(Z(x_1), Z(x_2)) \) is a symmetric matrix. The form (10) of \( \text{cov}(Z(x_1), Z(x_2)) \) is now confirmed by letting the 1th entry of \( a \) and the 2th entry of \( b \) 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 \( a^\top \) from the left and \( a \) from the right, and obtain

\[ a^\top C(x_1, x_2)a = \sum_{n=0}^{\infty} a^\top B_n a P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \ x_1, x_2 \in \mathbb{M}^d, \]

comparing which with (16) results in that \( a^\top B_n a \geq 0 \) or the nonnegative definiteness of \( B_n, n \in \mathbb{N}_0 \), and the convergence of \( \sum_{n=0}^{\infty} a^\top B_n a P_n^{(\alpha, \beta)}(1) \) 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 \( \{Z(x; 0) + Z(x; t) : x \in \mathbb{M}^d\} \). It is isotropic and mean square continuous on \( \mathbb{M}^d \), with covariance matrix function

\[
\begin{align*}
\text{cov} (Z(x_1; 0) + Z(x_1; t), Z(x_2; 0) + Z(x_2; t)) &= 2C(\rho(x_1, x_2); 0) + C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t) \\
&= \sum_{n=0}^{\infty} B_{n+} (t) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)), \ x_1, x_2 \in \mathbb{M}^d,
\end{align*}
\]

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

\[
\begin{align*}
\text{cov} (Z(x_1; 0) - Z(x_1; t), Z(x_2; 0) - Z(x_2; t)) &= 2C(\rho(x_1, x_2); 0) - C(\rho(x_1, x_2); t) - C(\rho(x_1, x_2); -t) \\
&= \sum_{n=0}^{\infty} B_{n-} (t) P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)),
\end{align*}
\]

and thus,

\[
\frac{C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)}{2} = \frac{1}{4} [2C(\rho(x_1, x_2); 0) + C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)]
\]
which confirms the format (11) for \( \frac{C(\rho(x_1, x_2); t) + C(\rho(x_1, x_2); -t)}{2} \), with \( B_n(t) = \frac{B_{n+1}(t) - B_{n-1}(t)}{4}, n \in \mathbb{N}_0 \). 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 function of an \( m \)-variate isotropic random field \( \{ Z(x; t): x \in \mathbb{M}^d, t \in \mathbb{T} \} \), where \( \{ \tilde{Z}(x; t): x \in \mathbb{M}^d, t \in \mathbb{T} \} \) is an independent copy of \( \{ Z(x; t): x \in \mathbb{M}^d, t \in \mathbb{T} \} \). In fact,

\[
\text{cov} \left( \frac{Z(x_1; t_1) + \tilde{Z}(x_1; -t_1)}{\sqrt{2}}, \frac{Z(x_2; t_2) + \tilde{Z}(x_2; -t_2)}{\sqrt{2}} \right) = \frac{C(\rho(x_1, x_2); t_1 - t_2) + C(\rho(x_1, x_2); t_2 - t_1)}{2} = \sum_{k=0}^{\infty} B_k(t_1 - t_2) P_k^{(\alpha, \beta)}(\cos \rho(x_1, x_2))
\]

with \( x_1, x_2 \in \mathbb{M}^d, t_1, t_2 \in \mathbb{T} \).

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

\[
W_n(t) = \int_{\mathbb{M}^d} \frac{Z(x; t) + \tilde{Z}(x; -t)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(x, U)) dx, \quad t \in \mathbb{T},
\]

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

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

and its covariance matrix function is by Lemmas 2 and 3

\[
\text{cov}(W_n(t_1), W_n(t_2)) = \frac{1}{\omega_d} \text{cov} \left( \int_{\mathbb{M}^d} \frac{Z(x; t_1) + \tilde{Z}(x; -t_1)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(x, U)) dx, \int_{\mathbb{M}^d} \frac{Z(y; t_2) + \tilde{Z}(y; -t_2)}{\sqrt{2}} P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dy \right)
\]
\[
= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \text{cov} \left( \int_{\mathbb{M}^d} Z(x; t_1) + \tilde{Z}(x; -t_1) P_n^{(\alpha, \beta)}(\cos \rho(x, U)) dx, \
\int_{\mathbb{M}^d} Z(y; t_2) + \tilde{Z}(y; -t_2) P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dy \right) du \\
= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \text{cov} \left( \frac{Z(x; t_1) + \tilde{Z}(x; -t_1)}{\sqrt{2}}, \frac{Z(y; t_2) + \tilde{Z}(y; -t_2)}{\sqrt{2}} \right) \times P_n^{(\alpha, \beta)}(\cos \rho(x, U)) P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dx dy du \\
= \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos \rho(x, U)) P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dx dy du \\
= \frac{1}{\omega_d} \sum_{k=0}^{\infty} B_k(t_1 - t_2) \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos \rho(x, y)) \times P_n^{(\alpha, \beta)}(\cos \rho(x, U)) dx P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dy du \\
= \frac{1}{\omega_d} B_n(t_1 - t_2) \int_{\mathbb{M}^d} \frac{1}{a_n^2} \int_{\mathbb{M}^d} P_n^{(\alpha, \beta)}(\cos \rho(y, U)) P_n^{(\alpha, \beta)}(\cos \rho(y, U)) dy du \\
= \frac{1}{\omega_d} B_n(t_1 - t_2) \left( \frac{\omega_d}{a_n^2} \right)^2 P_n^{(\alpha, \beta)}(1) du \\
= B_n(t_1 - t_2) \left( \frac{\omega_d}{a_n^2} \right)^2 P_n^{(\alpha, \beta)}(1), \quad t_1, t_2 \in \mathbb{T},
\]

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

**Proof of Theorem 4** The convergent assumption of \( \sum_{n=0}^{\infty} B_n(0) P_n^{(\alpha, \beta)}(1) \) ensures the uniform and absolute convergence of the series at the right-hand side of (12). If \( \{ B_n(t) : n \in \mathbb{N}_0 \} \) 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 \( B_n(t) \) on \( \mathbb{T} \) and an isotropic covariance function \( P_n^{(\alpha, \beta)}(\cos \rho(x_1, x_2)) \) 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 \( \{ Z(x; t) : x \in \mathbb{M}^d, t \in \mathbb{T} \} \). The convergence of \( \sum_{n=0}^{\infty} B_n(0) P_n^{(\alpha, \beta)}(1) \) results from the existence of \( C(0; 0) = \text{Var}(Z(x; t)) \). In order to show that \( B_n(t) \) is a stationary covariance matrix function on \( \mathbb{T} \) for each fixed \( n \in \mathbb{N}_0 \), consider an \( m \)-variate stochastic process
\[ W_n(t) = \int_{\mathbb{M}^d} Z(x; t) P_n^{(\alpha, \beta)}(\cos \rho(x, U)) dx, \quad t \in T, \]

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

\[
\text{cov}(W_n(t_1), W_n(t_2)) = B_n(t_1 - t_2) \left( \frac{\omega_d}{\alpha_n^2} \right)^2 P_n^{(\alpha, \beta)}(1), \quad t_1, t_2 \in T,
\]

which implies that \( B_n(t) \) is a stationary covariance matrix function on \( T \).

**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

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

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

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

and

\[
\text{cov}(Z(x_1; t_1), Z(x_2; t_2)) \\
= \text{cov} \left( \sum_{i=0}^{\infty} V_i(t_1) P_i^{(\alpha, \beta)}(\cos \rho(x_1, U)), \sum_{j=0}^{\infty} V_j(t_2) P_j^{(\alpha, \beta)}(\cos \rho(x_2, U)) \right)
\]
\[
= \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} \mathbb{E}[V_i(t_1)V_j^T(t_2)]\mathbb{E}\left[P_i^{(\alpha,\beta)}(\cos \rho(x_1, U)) P_j^{(\alpha,\beta)}(\cos \rho(x_2, U))\right]
\]

\[
= \sum_{n=0}^{\infty} B_n(t_1 - t_2) \frac{1}{a_n^2} P_n^{(\alpha,\beta)}(\cos \rho(x_1, x_2)), \quad x_1, x_2 \in \mathbb{M}^d, \ t_1, t_2 \in \mathbb{T}.
\]

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

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