



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

Chunsheng Ma<sup>1</sup> · Anatoliy Malyarenko<sup>2</sup> 

Received: 23 April 2018 / Revised: 21 September 2018 / Published online: 15 December 2018  
© The Author(s) 2018

## 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

**Mathematics Subject Classification (2010)** 60G60 · 62M10 · 62M30

## 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

---

✉ Anatoliy Malyarenko  
anatoliy.malyarenko@mdh.se  
<https://www.mdh.se/uk/personal/maa/amo01>

Chunsheng Ma  
chunsheng.ma@wichita.edu

<sup>1</sup> Department of Mathematics, Statistics, and Physics, Wichita State University, Wichita, KS 67260-0033, USA

<sup>2</sup> Division of Applied Mathematics, Mälardalen University, 721 23 Västerås, Sweden

[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  $(\alpha, \beta)$  [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  $\mathbb{M}^d \times \mathbb{T}$  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,  $2\pi$ . 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.

**Table 1** An approach based on Lie algebras

$\mathbb{M}^d$	$G$	$K$	$p$	$q$	Zonal function
$\mathbb{S}^d, d = 1, 2, \dots$	$\mathrm{SO}(d+1)$	$\mathrm{SO}(d)$	0	$d-1$	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^d(\mathbb{R}), d = 2, 3, \dots$	$\mathrm{SO}(d+1)$	$\mathrm{O}(d)$	0	$d-1$	$R_{2n}^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})/2))$
$\mathbb{P}^d(\mathbb{C}), d = 4, 6, \dots$	$\mathrm{SU}(\frac{d}{2}+1)$	$\mathrm{S}(\mathrm{U}(\frac{d}{2}) \times \mathrm{U}(1))$	$d-2$	1	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^d(\mathbb{H}), d = 8, 12, \dots$	$\mathrm{Sp}(\frac{d}{4}+1)$	$\mathrm{Sp}(\frac{d}{4}) \times \mathrm{Sp}(1)$	$d-4$	3	$R_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{o})))$
$\mathbb{P}^{16}(\mathbb{O})$	$\mathrm{F}_4(-52)$	$\mathrm{Spin}(9)$	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  $\Delta$  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  $\Delta$  as a self-adjoint operator in the space  $H = L^2(\mathbb{M}^d, d\nu(\mathbf{x}))$ . The spectrum of  $\Delta$  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  $\Delta$  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  $f_n(\mathbf{x})$ , 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

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

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  $\pi$  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

$\mathbb{M}^d$	$p$	$q$	$\alpha$	$\beta$	$\mathbb{A}$	$i(\mathbb{M}^d)$
$\mathbb{S}^d, d = 1, 2, \dots$	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*  $\mu$ ; 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  $\mu(\mathbb{M}^d)$  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*

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

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

## 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

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

Second, for *selected values* of  $\alpha$  and  $\beta$  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  $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{M}^d$ ,*

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

where

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

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  $\mathbf{P}(\mathbf{U} \in A) = \mathbf{P}(\mathbf{U} \in gA)$ .

To construct  $\mathbf{U}$ , we start with a measure  $\sigma$  proportional to the invariant measure  $\nu$  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  $(\Omega', \mathfrak{F}', \mathbf{P}')$  as follows:  $\Omega' = \mathbb{M}^d$ ,  $\mathfrak{F}'$  is the  $\sigma$ -field of Borel subsets of  $\Omega'$ , and

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

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

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

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

*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:*

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

### 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  $\{\mathbf{Z}(\mathbf{x}) : \mathbf{x} \in \mathbb{M}^d\}$  and a series expression for its covariance matrix function, in terms of Jacobi polynomials. By mean square continuous, we mean that, for  $k = 1, \dots, m$ ,

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

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  $\mathbf{B}$  with nonnegative eigenvalues  $\lambda_1, \dots, \lambda_m$ , there is an orthogonal matrix  $\mathbf{S}$  such that  $\mathbf{S}^{-1}\mathbf{B}\mathbf{S} = \mathbf{D}$ , where  $\mathbf{D}$  is a diagonal matrix with diagonal entries  $\lambda_1, \dots, \lambda_m$ . Define the square root of  $\mathbf{B}$  by

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



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

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

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

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

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*

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

*then it takes the form*

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

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

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

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  $\Phi$  is an  $m \times m$  matrix.

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

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

*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  $\mathbb{M}^d \times \mathbb{T}$ .

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],  $\chi^2$  [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*

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

*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,

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

**Acknowledgements** We are grateful to the anonymous referee for careful reading of the manuscript and useful remarks.

**Open Access** This article is distributed under the terms of the Creative Commons Attribution 4.0 International License (<http://creativecommons.org/licenses/by/4.0/>), which permits unrestricted use, distribution, and reproduction in any medium, provided you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made.

## 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

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

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.  $\square$

**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

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

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

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

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

□

**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  $S(\mathbf{x}) = P_n^{(\alpha, \beta)}(\cos(\rho(\mathbf{x}, \mathbf{y})))$ ,

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

The covariance function is

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

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

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

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

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

Two distinct terms of (8) are obviously uncorrelated each other.  $\square$

**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),

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

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

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

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  $\{\tilde{\mathbf{Z}}(\mathbf{x}; t) : \mathbf{x} \in \mathbb{M}^d, t \in \mathbb{T}\}$  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} \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) d\mathbf{u} \\
&= \frac{1}{\omega_d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \int_{\mathbb{M}^d} \text{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{C(\rho(\mathbf{x}, \mathbf{y}); t_1 - t_2) + 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} 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} 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} 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} 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} \\
&= 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  $B_n(t)$  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  $P_n^{(\alpha, \beta)}(\cos \rho(\mathbf{x}_1, \mathbf{x}_2))$  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  $\sum_{n=0}^{\infty} B_n(0) P_n^{(\alpha, \beta)}(1)$  results from the existence of  $C(0; 0) = \text{Var}[Z(\mathbf{x}; t)]$ . 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

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

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  $\mathbb{T}$ .  $\square$

**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

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

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

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

## References

- Andrews, G.E., Askey, R., Roy, R.: Special functions, Encyclopedia of Mathematics and its Applications, vol. 71. Cambridge University Press, Cambridge (1999)
- Askey, R., Bingham, N.H.: Gaussian processes on compact symmetric spaces. *Z. Wahrscheinlichkeitstheorie und Verw. Gebiete* **37**(2), 127–143 (1976/77)
- Azevedo, D., Barbosa, V.S.: Covering numbers of isotropic reproducing kernels on compact two-point homogeneous spaces. *Math. Nachr.* **290**(16), 2444–2458 (2017)
- Baldi, P., Rossi, M.: Representation of Gaussian isotropic spin random fields. *Stoch. Process. Appl.* **124**(5), 1910–1941 (2014)
- Berger, M., Gauduchon, P., Mazet, E.: Le spectre d'une variété riemannienne. *Lecture Notes in Mathematics*, vol. 194. Springer, Berlin (1971)
- Besse, A.L.: Manifolds all of whose geodesics are closed. With appendices. In: Epstein, D.B.A., Bourguignon, J.-P., Bérard-Bergery, L., Berger, M., Kazdan, J.L. (eds.) *Ergebnisse der Mathematik und ihrer Grenzgebiete [Results in Mathematics and Related Areas]*, vol. 93. Springer, Berlin (1978)
- Bingham, N.H.: Positive definite functions on spheres. *Proc. Cambridge Philos. Soc.* **73**, 145–156 (1973)
- Bochner, S.: Hilbert distances and positive definite functions. *Ann. Math.* **2**(42), 647–656 (1941)
- Brown, G., Dai, F.: Approximation of smooth functions on compact two-point homogeneous spaces. *J. Funct. Anal.* **220**(2), 401–423 (2005)
- Cartan, E.: Sur certaines formes Riemanniennes remarquables des géométries à groupe fondamental simple. *Ann. Sci. Éc. Norm. Supér.* **3**(44), 345–467 (1927)
- Cheng, D., Xiao, Y.: Excursion probability of Gaussian random fields on sphere. *Bernoulli* **22**(2), 1113–1130 (2016)
- Cohen, S., Lifshits, M.A.: Stationary Gaussian random fields on hyperbolic spaces and on Euclidean spheres. *ESAIM Probab. Stat.* **16**, 165–221 (2012)
- Colzani, L., Tenconi, M.: Localization for Riesz means on the compact rank one symmetric spaces. In: *Proceedings of the AMSI/AustMS 2014 Workshop in Harmonic Analysis and its Applications*, Proc. Centre Math. Appl. Austral. Nat. Univ., vol. 47, pp. 26–49. Austral. Nat. Univ., Canberra (2017)
- Gangolli, R.: Positive definite kernels on homogeneous spaces and certain stochastic processes related to Lévy's Brownian motion of several parameters. *Ann. Inst. H. Poincaré Sect. B (N.S.)* **3**, 121–226 (1967)
- Geller, D., Marinucci, D.: Spin wavelets on the sphere. *J. Fourier Anal. Appl.* **16**(6), 840–884 (2010)
- González Vieli, F.J.: Pointwise Fourier inversion on rank one compact symmetric spaces using Cesàro means. *Acta Sci. Math. (Szeged)* **68**(3–4), 783–795 (2002)
- Helgason, S.: Differential operators on homogeneous spaces. *Acta Math.* **102**, 239–299 (1959)
- Leonenko, N., Sakhno, L.: On spectral representations of tensor random fields on the sphere. *Stoch. Anal. Appl.* **30**(1), 44–66 (2012)
- Leonenko, N.N., Shieh, N.R.: Rényi function for multifractal random fields. *Fractals* **21**(2), 1350,009 (2013). 13
- Ma, C.: Covariance matrix functions of vector  $\chi^2$  random fields in space and time. *IEEE Trans. Commun.* **59**(9), 2554–2561 (2011). <https://doi.org/10.1109/TCOMM.2011.063011.100528>
- Ma, C.: Vector random fields with second-order moments or second-order increments. *Stoch. Anal. Appl.* **29**(2), 197–215 (2011)

22. Ma, C.: K-distributed vector random fields in space and time. *Stat. Probab. Lett.* **83**(4), 1143–1150 (2013). <https://doi.org/10.1016/j.spl.2013.01.004>
23. Ma, C.: Stochastic representations of isotropic vector random fields on spheres. *Stoch. Anal. Appl.* **34**(3), 389–403 (2016)
24. Ma, C.: Time varying isotropic vector random fields on spheres. *J. Theor. Probab.* **30**(4), 1763–1785 (2017)
25. Malyarenko, A.: Invariant random fields in vector bundles and application to cosmology. *Ann. Inst. Henri Poincaré Probab. Stat.* **47**(4), 1068–1095 (2011)
26. Malyarenko, A.: Invariant Random Fields on Spaces with a Group Action. Probability and its Applications (New York). Springer, Heidelberg (2013). **(With a foreword by Nikolai Leonenko)**
27. Malyarenko, A.: Spectral expansions of random sections of homogeneous vector bundles. *Teor. Ėmovir Mat. Stat.* **97**, 142–156 (2017)
28. Malyarenko, A.A.: Local properties of Gaussian random fields on compact symmetric spaces, and Jackson-type and Bernstein-type theorems. *Ukraĭn. Mat. Zh.* **51**(1), 60–68 (1999)
29. Malyarenko, A.A.: Abelian and Tauberian theorems for random fields on two-point homogeneous spaces. *Teor. Ėmovir Mat. Stat.* **69**, 106–118 (2003)
30. Malyarenko, A.A., Olenko, A.Y.: Multidimensional covariant random fields on commutative locally compact groups. *Ukraĭn. Mat. Zh.* **44**(11), 1505–1510 (1992)
31. Marinucci, D., Peccati, G.: Random fields on the sphere. Representation, limit theorems and cosmological applications, London Mathematical Society Lecture Note Series, vol. 389. Cambridge University Press, Cambridge (2011)
32. Matheron, G.: The internal consistency of models in geostatistics. In: Armstrong, M. (ed.) *Geostatistics*, pp. 21–38. Springer, Dordrecht (1989)
33. Molčan, G.M.: Homogeneous random fields on symmetric spaces of rank one. *Teor. Veroyatnost. i Mat. Statist.* **21**, 123–148, 167 (1979)
34. Müller, C.: Analysis of Spherical Symmetries in Euclidean Spaces, Applied Mathematical Sciences, vol. 129. Springer, New York (1998)
35. Obukhov, A.M.: Statistically homogeneous fields on a sphere. *Usp. Mat. Nauk* **2**(2), 196–198 (1947)
36. Sakamoto, K.: Helical minimal immersions of compact Riemannian manifolds into a unit sphere. *Trans. Am. Math. Soc.* **288**(2), 765–790 (1985)
37. Schoenberg, I.J.: Positive definite functions on spheres. *Duke Math. J.* **9**, 96–108 (1942)
38. Szegő, G.: Orthogonal polynomials, vol. XXIII, 4th edn. American Mathematical Society, Colloquium Publications, Providence (1975)
39. Volchkov, V.V., Volchkov, V.V.: Offbeat Integral Geometry on Symmetric Spaces. Birkhäuser, Basel (2013). <https://doi.org/10.1007/978-3-0348-0572-8>
40. Wang, H.C.: Two-point homogeneous spaces. *Ann. Math.* **2**(55), 177–191 (1952)
41. Weinstein, A.: On the volume of manifolds all of whose geodesics are closed. *J. Differ. Geom.* **9**, 513–517 (1974)
42. Yadrenko, M.Ĭ.: Spectral theory of random fields. Translation Series in Mathematics and Engineering. Optimization Software, Inc., Publications Division, New York (1983). **(Translated from the Russian)**
43. Yaglom, A.M.: Second-order homogeneous random fields. In: *Proceedings of 4th Berkeley Symposium on Mathematical Statistics and Probability*, vol. II, pp. 593–622. University of California Press, Berkeley (1961)
44. Yaglom, A.M.: Correlation Theory of Stationary and Related Random Functions, vol. I: Basic Results. Springer Series in Statistics. Springer, New York (1987)