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Spherical Poisson waves*

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Abstract

We introduce a model of Poisson random waves in S^2 and we study Quantitative Central Limit Theorems when both the rate of the Poisson process and the energy (i.e., frequency) of the waves (eigenfunctions) diverge to infinity. We consider finite-dimensional distributions, harmonic coefficients and convergence in law in functional spaces, and we investigate carefully the interplay between the rate of divergence of eigenvalues and Poisson governing measures.

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1 Introduction

1.1 Motivations

The analysis of Gaussian eigenfunctions on different manifolds has recently become a very attractive area of research – it started in the mathematics literature mainly about a decade ago ([34, 44]) and it has then covered a number of different questions and circumstances, including the Euclidean case (Berry’s Random Wave Model, see [3, 4, 16, 36, 43]), Random Spherical Harmonics (eigenfunctions on the sphere, see [11, 13, 12, 14, 30, 41]), Arithmetic Random Waves (eigenfunctions on the torus, see [9, 10, 22, 26, 29, 38]) and other manifolds (see [15, 17, 39]). The leading motivation for such a strong interest comes mainly from the physical sciences, and in particular from an *ansatz* by Michael Berry in a 1977 paper [3], where he claimed that Gaussian

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random waves could be taken as a universal model to approximate the behaviour even of deterministic eigenfunctions in the high-energy limit (i.e., for diverging eigenvalues) under “generic” boundary conditions.

A common argument to justify the universality of Gaussian behaviour for eigenfunctions in the physics literature is the *random phase model* (see [45] and the references therein), which we can describe as follows. Working on $\mathbb{R}^2$, assume we observe the superposition of N waves at a given frequency k , that is

$$T_{k;N}(x) = \frac{1}{\sqrt{N}} \sum_{j=1}^N \exp(ik \langle \theta_j, x \rangle + \phi_j), \quad (1.1)$$

for $x \in \mathbb{R}^2$, $k \in \mathbb{R}^+$, where $\{\theta_j\}_{j=1,\dots,N}$ are random directions on the unit circle and $\{\phi_j\}_{j=1,\dots,N}$ are random phases. By a standard Central Limit Theorem it is then immediate to show that $T_{k;N}(x)$ converges, as $N \rightarrow \infty$, in distribution to a zero mean Gaussian field $\tilde{T}_k(\cdot)$ with covariance function given by

$$\mathbb{E} [\tilde{T}_k(x_1) \tilde{T}_k(x_2)] = J_0(k \|x_1 - x_2\|_2),$$

where $J_0(\cdot)$ is the Bessel function of order 0, given by

$$J_0(u) = \sum_{m=0}^{\infty} (-1)^m \frac{u^{2m}}{2^{2m} (m!)^2}.$$

For a fixed value of the wavelength parameter, hence, the validity of a Central Limit Theorem result follows from very standard arguments.

It should be noticed, however, that the literature on random eigenfunctions has actually been developed under the implicit framework of a double asymptotic setting. Indeed, on the one hand, a diverging number of random phases is taken to ensure that the behaviour of random eigenfunctions is Gaussian; on the other hand, Gaussianity is taken for granted when investigating the asymptotic behaviour of random eigenfunctions in the high-frequency/high energy sense (i.e., for diverging eigenvalues). Some natural questions are hence the following – given that Gaussianity has been established for a *fixed* eigenvalue k , can we justify the use of this assumption in the limit as $k \rightarrow \infty$? Can we allow at the same time the eigenvalues to grow together with the number of random phases, and still have a Central Limit Theorem? Do we need some conditions that relate of the divergence for the eigenvalue k to the rate of divergence of the number of random phases N ? How many “*random phases*” do we need, in the language of Berry’s celebrated model, in order for the Gaussian approximation to hold at high frequencies?

In this paper we try to address these questions in the case of random eigenfunctions defined on the two-dimensional sphere $\mathbb{S}^2$; the choice of the sphere is motivated by the fact that it represents the most interesting case from the point of view of physical applications and it is known to exhibit the same covariance structure as the Euclidean case, in the scaling limit (due to so-called Hilb’s asymptotics, see [40, Equation 8.21.7], and [44]). The extension of these results to the planar case does not seem to pose any conceptual difficulties; it would be more interesting to explore this setting in the case of Arithmetic Random Waves, which is known to exhibit some differences with respect to Euclidean and Spherical circumstances. We leave this extension for further research.

1.2 The model

Our starting point is to reformulate Equation (1.1) as a superposition of deterministic eigenfunctions of the Laplace-Beltrami operator on the torus $\mathbb{T}^2$, centred on random

locations. More precisely, let us consider

$$T_{k,N}(x) = \frac{1}{\sqrt{N}} \sum_{j=1}^N \exp(i\langle \mathbf{k}, \mathbf{x} - \mathbf{y}_j \rangle)$$

for $x = (x_1, x_2) \in \mathbb{T}^2$, where $\mathbf{k} = (k_1, k_2) \in \mathbb{Z}$ is such that $k = \|\mathbf{k}\|_2$; here, the random locations $\mathbf{y}_j = (y_{j,1}, y_{j,2})$, $j = 1, \dots, N$ are taken independent and uniformly distributed on $\mathbb{T}^2$, so that as in Equation (1.1) for a fixed x the argument of the complex exponential $\psi_j := \langle \mathbf{k}, \mathbf{x} - \mathbf{y}_j \rangle$ can be taken uniformly distributed in $[0, 2\pi)$.

To achieve an analogous construction in the spherical case, we need to recall first that the Laplacian operator in $\mathbb{S}^2$ is defined by

$$\Delta_{\mathbb{S}^2} := \frac{1}{\sin \theta} \frac{\partial}{\partial \theta} \left(\sin \theta \frac{\partial}{\partial \theta} \right) + \frac{1}{\sin^2 \theta} \frac{\partial^2}{\partial \varphi^2};$$

in the spherical case, a deterministic eigenfunction of the spherical Laplacian centred on $y \in \mathbb{S}^2$ can be constructed by

$$e_{\ell;y}(\cdot) : \mathbb{S}^2 \rightarrow \mathbb{R}, \quad e_{\ell;y}(\cdot) := \sqrt{\frac{2\ell+1}{4\pi}} P_{\ell}(\langle \cdot, y \rangle),$$

where we have introduced the family of Legendre polynomials

$$P_{\ell}(t) := \frac{1}{2^{\ell} \ell!} \frac{d^{\ell}}{dt^{\ell}} (t^2 - 1)^{\ell}, \quad \ell = 0, 1, 2, \dots; \quad t \in [0, 1].$$

The choice of normalization ensures that $P_{\ell}(1) \equiv 1$ for all ℓ and moreover

$$\|e_{\ell;y}\|_{L^2(\mathbb{S}^2)} = \int_{\mathbb{S}^2} \frac{2\ell+1}{4\pi} P_{\ell}^2(\langle x, y \rangle) dy = P_{\ell}(\langle x, x \rangle) = 1, \quad (1.2)$$

in view of the duplication formula, see for instance, [28, Section 13.1.2]; also, we have that $\{e_{\ell;y}(\cdot)\}$ satisfies the Helmholtz equation

$$\Delta_{\mathbb{S}^2} e_{\ell;y}(x) + \lambda_{\ell} e_{\ell;y}(x) = 0, \quad \ell = 0, 1, 2, \dots,$$

where $-\lambda_{\ell} = -\ell(\ell+1)$ is the sequence of eigenvalues of the spherical Laplacian, see again [28, 44].

We convey the idea of random phases on the sphere by introducing a superposition of waves centred on Poisson distributed random points on $\mathbb{S}^2$. Here is a more formal setting.

Definition 1.1. *The Poisson spherical random wave model (with rate ν_t) is defined by*

$$T_{\ell;t}(x) := \frac{1}{\sqrt{\nu_t}} \int_{\mathbb{S}^2} \sqrt{\frac{2\ell+1}{4\pi}} P_{\ell}(\langle x, \xi \rangle) dN_t(\xi),$$

where $\{N_t(\cdot)\}$ is a Poisson process on the sphere with governing intensity measure

$$\mathbb{E}[N_t(A)] = \nu_t \times \sigma(A) \text{ for all } A \in \mathcal{B}(\mathbb{S}^2),$$

where σ is the Lebesgue measure on $\mathbb{S}^2$ defined by

$$\int_{\mathbb{S}^2} f(x) \sigma(dx) = \int_0^{2\pi} \int_0^{\pi} f(\sin \vartheta \cos \varphi, \sin \vartheta \sin \varphi, \cos \vartheta) \sin \vartheta d\vartheta d\varphi \quad (1.3)$$

in the spherical coordinates $x = (\sin \vartheta \cos \varphi, \sin \vartheta \sin \varphi, \cos \vartheta)$ for every bounded functions on $\mathbb{S}^2$ (note that $\sigma(\mathbb{S}^2) = 4\pi$).

Our model implies that for all measurable sets $A \subset \mathbb{S}^2$ and $t \geq 0$, $N_t(A)$ is a Poisson random variable with expected value equal to $\nu_t \times \sigma(A)$, and for $A_1 \cap A_2 = \emptyset$, $N_t(A_1)$ and $N_t(A_2)$ are independent.

We recall here a rigorous definition of Poisson random measures.

Definition 1.2 (Poisson random measure). *Let $(\Theta, \mathcal{A}, \rho)$ be a σ -finite measure space, such that ρ has no atoms. A Poisson random measure on Θ with intensity measure ρ is a collection of random variables $\{N(A) : A \in \mathcal{A}\}$, taking values in the space $\mathbb{Z}_+ \cup \{\infty\}$, characterized by the following two properties:*

1. *for every $A \in \mathcal{A}$, $N(A)$ has Poisson distribution with intensity $\rho(A)$;*
2. *for $A_1, \dots, A_n \in \mathcal{A}$ pairwise disjoint, $N(A_1), \dots, N(A_n)$ are independent.*

In this paper, we work with $\Theta = \mathbb{R}_+ \times \mathbb{S}^2$, while $\mathcal{A}$ is the class of Borel subsets of Θ , labeled by $\mathcal{B}(\Theta)$. We denote by N a Poisson random measure on Θ , whose homogeneous intensity is given by the product measure $\rho = \lambda \times \sigma$. The first term, which can be read as the time component, is some measure on $\mathbb{R}_+$ and σ is the Lebesgue measure on $\mathbb{S}^2$ defined in (1.3). We assume that $\lambda(\{0\}) = 0$ and that the mapping $\lambda \rightarrow \lambda([0, t])$ is strictly increasing and diverging to infinity as $t \rightarrow \infty$. We denote $\nu_t := \lambda([0, t])$, $t \geq 0$, that is $t \rightarrow \nu_t$ is the distribution function of λ .

For a fixed $t > 0$, the mapping $A \rightarrow N_t(A) := N([0, t] \times A)$ defines a Poisson random measure on $\mathbb{S}^2$, with non-atomic intensity $\rho_t(dx) = \nu_t \cdot \sigma(dx)$.

Remark 1.3. Fix some $t > 0$ and let $\{\xi_i = i \geq 1\}$ be a sequence of i.i.d. random variables (independent from $N_t(\mathbb{S}^2)$) with values in $\mathbb{S}^2$ and common probability distribution equal to $\frac{\sigma}{4\pi}$. Then, the random measure $A \rightarrow N_t(A) = N([0, t] \times A)$ has the same distribution as

$$A \rightarrow \sum_{i=1}^{N_t(\mathbb{S}^2)} \delta_{\xi_i}(A),$$

where δ_x denotes the Dirac measure at the point x (see [25, Proposition 3.8]).

To simplify the discussion, we will assume here that $\lambda(ds) = \nu \cdot \text{Leb}(ds)$, where Leb is the Lebesgue measure on $\mathbb{R}_+$ and $\nu > 0$ is a constant. We denote $\nu_t = \lambda([0, t]) = \nu t$.

Note that the Gaussian eigenfunctions of the spherical Laplacian (considered for example in [29, 31, 32], and many others) can be written as

$$T_\ell(x) := \int_{\mathbb{S}^2} \sqrt{\frac{2\ell+1}{4\pi}} P_\ell(\langle x, y \rangle) dZ(y), \quad (1.4)$$

where dZ is a Gaussian random measure with control measure given by σ , that is, for all $A \in \mathcal{B}(\mathbb{S}^2)$, $Z(A)$ is standard Gaussian and for $A_1 \cap A_2 = \emptyset$, $Z(A_1)$ and $Z(A_2)$ are independent and for any $A, B \in \mathcal{B}(\mathbb{S}^2)$, $\mathbb{E}[Z(A)Z(B)] = \sigma(A \cap B)$. Note that this is the same construction as in Definition 1.1, which is based however on a different random measure.

We can also write the Poisson spherical wave as

$$T_{\ell,t}(x) = \frac{1}{\sqrt{\nu_t}} \sum_{k=1}^{N_t(\mathbb{S}^2)} \sqrt{\frac{2\ell+1}{4\pi}} P_\ell(\langle x, \xi_k \rangle),$$

so that we can view spherical Poisson random waves as occurring from the sum of a (random) number of deterministic waves, centred at points (indexed by k) which are uniformly distributed on the sphere.

It is now convenient to introduce the standard basis for the $(2\ell+1)$ -dimensional space of eigenfunctions corresponding to the eigenvalue λ_ℓ ; the elements of the basis

are sometimes called *fully normalized spherical harmonics*, and are defined as the normalized eigenfunctions $\{Y_{\ell m}\}_{m=-\ell, \dots, \ell}$ which satisfy the further condition (in spherical coordinates)

$$Y_{\ell m} : \mathbb{S}^2 \rightarrow \mathbb{R}, \quad \frac{\partial^2}{\partial \varphi^2} Y_{\ell m}(\theta, \varphi) = -m^2 Y_{\ell m}(\theta, \varphi).$$

The elements of the real fully normalized spherical harmonics basis can be written explicitly as the normalized product of the so-called Legendre associated function $P_{\ell}^m : [-1, 1] \mapsto \mathbb{R}$ of degree ℓ and order m , which depends only on θ and is defined by

$$P_{\ell}^m(t) := (1 - t^2)^{m/2} \frac{d^m}{dt^m} P_{\ell}(t), \quad t \in [0, 1]$$

(see [28, Equation 13.7]), and a trigonometric function depending only on ϕ , that is,

$$Y_{\ell m}(\theta, \phi) = \begin{cases} \sqrt{\frac{2\ell+1}{2\pi} \frac{(\ell-m)!}{(\ell+m)!}} P_{\ell}^m(\cos \theta) \cos(m\phi) & \text{for } m \in \{1, \dots, \ell\} \\ \sqrt{\frac{2\ell+1}{4\pi}} P_{\ell}(\cos \theta) & \text{for } m = 0 \\ \sqrt{\frac{2\ell+1}{2\pi} \frac{(\ell+m)!}{(\ell-m)!}} P_{\ell}^{-m}(\cos \theta) \sin(-m\phi) & \text{for } m \in \{-\ell, \dots, -1\} \end{cases},$$

see, for example, [28, Remark 3.37].

It should be noted, however, that none of the results below depend on the specific choice of our basis; they would hold unaltered for any orthonormal system. The most important properties of the fully normalized spherical harmonics are the addition and duplication formula (see respectively [28, Eq. (3.42) and Sec. 13.1.2]), which are given respectively by

$$\sum_{m=-\ell}^{\ell} Y_{\ell m}(x) Y_{\ell m}(y) = \frac{2\ell+1}{4\pi} P_{\ell}(\langle x, y \rangle), \quad (1.5)$$

$$\int_{\mathbb{S}^2} \frac{2\ell+1}{4\pi} P_{\ell}(\langle x, z \rangle) \frac{2\ell+1}{4\pi} P_{\ell}(\langle z, y \rangle) dz = \frac{2\ell+1}{4\pi} P_{\ell}(\langle x, y \rangle), \quad (1.6)$$

for all $x, y \in \mathbb{S}^2$. Using the addition formula yields

$$T_{\ell; t}(x) = \frac{1}{\sqrt{\nu_t}} \sqrt{\frac{4\pi}{2\ell+1}} \sum_{k=1}^{N_t(\mathbb{S}^2)} \sum_{m=-\ell}^{\ell} Y_{\ell m}(x) Y_{\ell m}(\xi_k) = \sum_{m=-\ell}^{\ell} \hat{a}_{\ell, m}(t) Y_{\ell m}(x),$$

where the random spherical harmonic coefficients $\{\hat{a}_{\ell, m}(t)\}_{m=-\ell, \dots, \ell}$ are defined by

$$\hat{a}_{\ell, m}(t) := \sqrt{\frac{4\pi}{(2\ell+1)\nu_t}} \sum_{k=1}^{N_t(\mathbb{S}^2)} Y_{\ell m}(\xi_k),$$

where $\{\xi_k\}$ are the points charged by the Poisson process. Note that

$$\mathbb{E}[\hat{a}_{\ell, m}(t) \hat{a}_{\ell', m'}(t)] = \delta_m^{m'} \delta_{\ell}^{\ell'} \frac{4\pi}{(2\ell+1)}$$

and

$$\mathbb{E}[T_{\ell; t}(x) T_{\ell; t}(y)] = P_{\ell}(\langle x, y \rangle).$$

It is also easy to verify that the Parseval's identity holds, i.e.

$$\|T_{\ell; t}\|_{L^2(\mathbb{S}^2)}^2 = \int_{\mathbb{S}^2} T_{\ell; t}^2(x) dx = \sum_{m=-\ell}^{\ell} |\hat{a}_{\ell, m}(t)|^2.$$

1.3 Overview of the main results

In this work, we consider the convergence in law of the Poisson random spherical eigenfunctions to a Gaussian limit when both the rate of the Poisson process and the eigenvalue sequence λ_ℓ diverge to infinity; see for instance [21, 23, 24] and the references therein for some recent results on quantitative convergence bounds in a Poisson framework. We focus on three different cases:

a) We study the convergence of the finite-dimensional distributions for a fixed array of d -points $(x_1, x_2, \dots, x_d) \in \mathbb{S}^2$, with special emphasis on the univariate marginal distribution for $d = 1$; here we prove that a quantitative Central Limit Theorem holds insofar we have that $d^2 \sqrt{\log \ell} = o(\sqrt{\nu_t})$. In particular, for the special case $d = 1$ asymptotic Gaussianity holds for eigenvalues that increase polynomially fast with respect to the rate of occurrence of Poisson events.

b) We also study the convergence in law for the vector of spherical harmonic coefficients $\{\hat{a}_{\ell, m}\}_{m=-\ell, \dots, \ell}$; again a multivariate Central Limit Theorem would be straightforward, but here we provide a quantitative version when ℓ (and hence the dimension of the vector itself) grows with ν_t . The bound we obtain here is of order $\sqrt{\frac{\ell \log \ell}{\nu_t}}$, thus entailing that multivariate asymptotic Gaussianity holds provided $\sqrt{\ell \log \ell} = o(\sqrt{\nu_t})$. Out of this bound, it is also possible to derive an alternative rate of convergence for finite-dimensional distributions of order d , which turns out to be $d\ell \sqrt{\log \ell / \nu_t}$, see Remark 2.13. For fixed d , this is clearly worse than the bound we discussed in the previous point, but it can actually be better if one envisages d as growing with ℓ at a suitably fast rate.

c) We then consider functional convergence results, where we view the eigenfunctions $\{T_{\ell, t}\}$ as random elements $T_{\ell, t} : \Omega \rightarrow L^2(\mathbb{S}^2)$, i.e. as measurable applications with the topology induced on $L^2(\mathbb{S}^2)$ by the standard metric

$$d^2(f, g) := \|f - g\|_{L^2(\mathbb{S}^2)}^2 = \int_{\mathbb{S}^2} |f(x) - g(x)|^2 dx.$$

Exploiting some very recent and important results by [6] (see also [5]), we are able here to show that a quantitative Central Limit Theorem holds under the simple condition that $\nu_t \rightarrow \infty$. This is apparently surprising, because in this functional case it turns out that asymptotic Gaussianity will hold no matter how fast the sequence of eigenvalues diverge to infinity, on the contrary of what we have stated for the (apparently simpler) cases under b) and c). A careful inspection of the results reveals that the apparent paradox is due to the topological structure induced by the $\|\cdot\|_{L^2(\mathbb{S}^2)}$, which is much coarser than the one given for instance, by the sup norm. In particular, weak convergence with respect to $\|\cdot\|_{L^2(\mathbb{S}^2)}$ does not entail convergence of the finite-dimensional distributions, not even univariate ones.

d) Finally, we establish a quantitative Central Limit Theorem for $T_{\ell, t}$ in functional spaces which induce finer topologies; we focus in particular on Sobolev spaces (see Theorem 2.15). Here, we are able to obtain the rate

$$\frac{\sqrt{\pi} \left(1 + \sqrt{\ell(\ell+1)}\right)^{2\alpha}}{2\sqrt{\nu_t}} + \frac{2\pi \left(1 + \sqrt{\ell(\ell+1)}\right)^{3\alpha}}{\sqrt{\nu_t}},$$

which is much worse than in both the L^2 case and for marginal distributions. However, functional convergence in Sobolev spaces with sufficient regularity is, of course, a much stronger result; in particular, among others, it does imply convergence of the finite-dimensional distributions at fixed locations on the sphere, as detailed below in Section 2.3.2, Corollary 2.16.

Remark 1.4. It is well-known (see [2, Corollary 1]) that convergence in $L^2(\cdot)$ does indeed entail pointwise convergence in the case of Reproducing Kernel Hilbert Spaces (RKHS).

Because the space of spherical eigenfunctions is indeed a RKHS, again the point in c) may sound counterintuitive. There is a subtle point here, as ℓ increases, we are actually dealing with a sequence of RKHS; whereas it is indeed possible to bound the pointwise norm with the L^2 distance *up to a constant*, the “constant” does vary with ℓ , and indeed it diverges to infinity as we shall discuss below; so no contradiction arises.

1.4 Some remarks on the nature of asymptotics

At this stage, it is important to add some remarks on the nature of our asymptotic results. We note first that *both the multipole index ℓ and the Poisson rate diverge jointly to infinity in our framework*; for fixed multipoles ℓ , convergence to Gaussianity remains true but becomes rather trivial and uninteresting. As a consequence of this double asymptotics framework, the covariance functions of the processes that we study do not converge to the covariance of a well-defined, measurable function on the sphere.

Indeed it is easy to see that for any choice of fixed points on the sphere the covariance of our process converges to zero as $\ell \rightarrow \infty$. This implies that the limiting process (if it existed) would not be mean square continuous; but such a process cannot be well-defined (i.e., measurable) as proved in [27].

However, this apparent difficulty allows us to exploit the full power of quantitative central limit results. Indeed, this class of theorems does not require, by any means, that the sequences converge to a well defined limiting distribution. One can have two sequences of random variables X_n, Y_n and show that $d_W(X_n, Y_n) \rightarrow 0$ as $n \rightarrow \infty$, meaning that we can approximate the distribution of X_n arbitrarily well with the distribution of Y_n , for n large enough, independently from the fact that X_n converges or not to a limit distribution. For instance, Y_n could be a sequence of Gaussian variables with oscillating mean and variance μ_{Y_n} and $\sigma_{Y_n}^2$, and still one could use the Gaussian quantiles to approximate the distribution of X_n as $N(\mu_{Y_n}, \sigma_{Y_n}^2)$. Very much the same can be said below, where the covariance operators do not converge to meaningful limits. For ideas that are in a broad sense related, one could also think about the large p , large n framework in random matrix theory (see for instance [7, 8] where it is shown that under appropriate conditions, the laws of large Wishart random matrices become indistinguishable from the laws of the Gaussian orthogonal ensemble).

For completeness, we add that it may be possible to get some form of nondegenerate limiting behaviour for random waves: in particular, if neglecting the spherical structure and focussing only on shrinking domains around a single fixed point $x \in \mathbb{S}^2$, then it could be possible to show that the scaling limit of the waves T_ℓ when projected on the tangent plane converges locally to random eigenfunctions on $\mathbb{R}^2$ (Berry’s random waves). The price to pay for this approach would however be high: the result would no longer deal with convergence on the sphere, which is what we are studying in this work. Moreover, this approach would not allow to answer the question that we addressed here and that we consider interesting for physical applications: given a random spherical harmonic T_ℓ with ℓ suitably large, what is the order of magnitude of the governing Poisson rate that is required for the Gaussian approximation to be adequate? This is exactly the issue that we address in the sequel, under a variety of different circumstances.

2 Main results

Before we proceed with the statement of our results, we need to recall briefly the probability metrics that we are going to exploit, which are defined by

- (a) Kolmogorov metric: for any two random variables $X, Y : \Omega \rightarrow \mathbb{R}$

$$d_{Kol}(X, Y) = \sup_{x \in \mathbb{R}} |\mathbb{E} [\mathbb{1}_{(-\infty, x]}(X)] - \mathbb{E} [\mathbb{1}_{(-\infty, x]}(Y)]| = \sup_{x \in \mathbb{R}} |F_X(x) - F_Y(x)|,$$

$F(\cdot)$ denoting as usual the distribution function.

(b) Wasserstein metric: for any two random vectors $X, Y : \Omega \rightarrow \mathbb{R}^d$

$$d_W(X, Y) = \sup_{h \in Lip(1)} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|,$$

$$h \in Lip(1) \Leftrightarrow h : \mathbb{R}^d \rightarrow \mathbb{R} : M_1(h) \leq 1.$$

where $M_1(h)$ is defined by

$$M_1(h) := \sup_{\substack{x, y \in \mathbb{R}^d, \\ x \neq y}} \frac{|h(x) - h(y)|}{\|x - y\|_{\mathbb{R}^d}}. \quad (2.1)$$

(c) d_3 metric: for any two random vectors $X, Y : \Omega \rightarrow \mathbb{R}^d$ such that $\mathbb{E}\|X\|_{\mathbb{R}^d}^2, \mathbb{E}\|Y\|_{\mathbb{R}^d}^2 < \infty$,

$$d_3(X, Y) = \sup_{h \in \mathcal{I}} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|$$

where $\mathcal{I}$ indicates the collection of all functions $h \in C^3(\mathbb{R}^d)$ such that $\|h''\|_{\infty} \leq 1$ and $\|h'''\|_{\infty} \leq 1$, $C^3(\mathbb{R}^d)$ denoting the space of three times continuously differentiable functions.

(d) Functional d_3 metric: for a general function space K we have that $C_b^3(K)$ is the class of real-valued functions on K that have bounded Fréchet derivatives up to order three. This space is equipped with the norm

$$\|h\|_{C_b^3(K)} = \sup_{j=1,2,3} \sup_{x \in K} \|D^j h(x)\|_{K^{\otimes j}}.$$

Then, given a Hilbert space K and any two random elements $X, Y : \Omega \rightarrow K$

$$d_3(X, Y) = \sup_{h \in C_b^3(K)} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|.$$

Remark 2.1. Clearly, (c) can be viewed as a special case of (d) for $K = \mathbb{R}^d$. We refer to [35, Appendix C] for more discussion and examples on probability metrics and their mutual relationships.

We divide our results below in three subsections, referring respectively to finite-dimensional distributions, harmonic coefficients and functional convergence.

2.1 Convergence of the finite dimensional distributions

We start from a simple univariate case; this is of course implied by the d -dimensional result that we give below, but we prefer to treat it on its own for clarity of exposition and to optimize the value of the relevant constants.

Theorem 2.2 (One-dimensional case). *Let the notation above prevail and $Z \sim \mathcal{N}(0, 1)$. For all $x \in \mathbb{S}^2$ we have that, as $\ell \rightarrow \infty$,*

$$d_W(T_{\ell,t}(x), Z) \leq \left(\frac{\sqrt{3}}{2\pi^2} + \frac{\sqrt{2}}{\sqrt{3\pi^3}} \right) \sqrt{\frac{\log \ell}{\nu_t}} + o_{\ell \rightarrow \infty} \left(\sqrt{\frac{\log \ell}{\nu_t}} \right).$$

Proof. Note first that, because $T_{\ell,t}(x) = \frac{1}{\sqrt{\nu_t}} \int_{\mathbb{S}^2} \sqrt{\frac{2\ell+1}{4\pi}} P_{\ell}(\langle x, \xi \rangle) dN_t(\xi)$ we are in the domain of validity of Fourth Moment Theorems for integral functionals of Poisson processes, see for instance [37] and many subsequent papers. In particular, we shall

exploit [19, Theorem 1.7], which we recall in Section A.1 below, see Theorem A.2. To apply this result, we need to compute the fourth moment of $T_{\ell;t}$, which is given by

$$\mathbb{E}[T_{\ell;t}^4(x)] = \sum_{m_1, \dots, m_4 = -\ell}^{\ell} \mathbb{E}[\hat{a}_{\ell, m_1}(t) \hat{a}_{\ell, m_2}(t) \hat{a}_{\ell, m_3}(t) \hat{a}_{\ell, m_4}(t)] Y_{\ell m_1}(x) Y_{\ell m_2}(x) Y_{\ell m_3}(x) Y_{\ell m_4}(x);$$

substituting the value of $\hat{a}_{\ell, m}$ we have that

$$\mathbb{E}[T_{\ell;t}^4(x)] = \left(\frac{4\pi}{\nu_t(2\ell+1)} \right)^2 \sum_{m_1, \dots, m_4 = -\ell}^{\ell} \mathbb{E} \left[\sum_{k_1, \dots, k_4 = 1}^{N_t(\mathbb{S}^2)} Y_{\ell m_1}(\xi_{k_1}) Y_{\ell m_2}(\xi_{k_2}) Y_{\ell m_3}(\xi_{k_3}) Y_{\ell m_4}(\xi_{k_4}) \right] \\ \times Y_{\ell m_1}(x) Y_{\ell m_2}(x) Y_{\ell m_3}(x) Y_{\ell m_4}(x).$$

Exploiting the addition formula (1.5), we get

$$\mathbb{E}[T_{\ell;t}^4(x)] = \left(\frac{2\ell+1}{4\pi\nu_t} \right)^2 \mathbb{E} \left[\sum_{k_1, \dots, k_4 = 1}^{N_t(\mathbb{S}^2)} P_{\ell}(\langle \xi_{k_1}, x \rangle) P_{\ell}(\langle \xi_{k_2}, x \rangle) P_{\ell}(\langle \xi_{k_3}, x \rangle) P_{\ell}(\langle \xi_{k_4}, x \rangle) \right] \\ = \left(\frac{2\ell+1}{4\pi\nu_t} \right)^2 \left(4\pi\nu_t \mathbb{E} [P_{\ell}(\langle \xi_{k_1}, x \rangle)^4] + 3(4\pi\nu_t)^2 \mathbb{E} [P_{\ell}(\langle \xi_{k_1}, x \rangle)^2]^2 \right).$$

In [32, Lemma 2.3], it has been shown that

$$\int_0^1 P_{\ell}^4(t) dt \sim \frac{3}{2\pi^2} \frac{\log \ell}{\ell^2},$$

where for any two positive sequences $\{a_{\ell}, b_{\ell}\}_{\ell=1,2,\dots}$ we write

$$a_{\ell} \sim b_{\ell} \Leftrightarrow \lim_{\ell \rightarrow \infty} \frac{a_{\ell}}{b_{\ell}} = 1.$$

Thus we get

$$\mathbb{E} [P_{\ell}(\langle \xi_{k_1}, x \rangle)^4] = \frac{1}{4\pi} \int_{\mathbb{S}^2} P_{\ell}(\langle z, x \rangle)^4 dz = \int_0^1 P_{\ell}^4(t) dt \sim \frac{3}{2\pi^2} \frac{\log \ell}{\ell^2}, \text{ as } \ell \rightarrow \infty.$$

Moreover, since

$$\int_0^1 P_{\ell}(t)^2 dt = \frac{1}{2\ell+1},$$

we also have that

$$\mathbb{E}[P_{\ell}(\langle \xi_{k_1}, x \rangle)^2] = \frac{1}{4\pi} \int_{\mathbb{S}^2} P_{\ell}(\langle z, x \rangle)^2 dz = \int_0^1 P_{\ell}(t)^2 dt = \frac{1}{2\ell+1}.$$

It follows that

$$\mathbb{E}[T_{\ell;t}^4(x)] = 3 + \frac{3}{2\pi^3} \frac{\log \ell}{\nu_t} + o\left(\frac{\log \ell}{\nu_t}\right). \quad (2.2)$$

Applying Theorem A.2 concludes the proof. $\square$

Remark 2.3. Theorem A.3 can also be applied to obtain a Quantitative Central Limit Theorem in the Kolmogorov distance instead of the Wasserstein one. Indeed, by combining the second part of Theorem A.2 with the estimate of the fourth moment of $T_{\ell;t}$

obtained in (2.2), we get that, for any fixed $x \in \mathbb{S}^2$,

$$\begin{aligned} d_{Kol}(T_{\ell;t}(x), Z) &\leq \left[11 + \sqrt{3 + \frac{3}{2\pi^3} \frac{\log \ell}{\nu_t} + o\left(\frac{\log \ell}{\nu_t}\right)} \right. \\ &\quad \left. + \sqrt[4]{3 + \frac{3}{2\pi^3} \frac{\log \ell}{\nu_t} + o\left(\frac{\log \ell}{\nu_t}\right)} \right] \sqrt{\frac{3}{2\pi^3} \frac{\log \ell}{\nu_t} + o\left(\frac{\log \ell}{\nu_t}\right)} \\ &= O\left(\sqrt{\frac{\log \ell}{\nu_t}}\right), \end{aligned}$$

where $Z \sim \mathcal{N}(0, 1)$.

Remark 2.4 (A Comparison with Needlet/Wavelet Coefficients). We note here that the constraint on the rate of convergence of the eigenvalues with respect to the rate in the Poisson governing intensity measure is very weak; Theorem 2.2 shows that asymptotic Gaussianity will continue to hold even if we allow λ_ℓ to grow as any polynomial function of the rate ν_t . This is in sharp contrast with what is observed in related circumstances for the behaviour of spherical wavelet/needlet coefficients (see, for example, [20]). To compare those results with the ones presented here, we recall that needlet coefficients corresponding to $\xi \in \mathbb{S}^2$ in the notation of this can be considered as equivalent to (after normalization)

$$\begin{aligned} \beta_j(\xi) &:= \sum_{\ell=2^{j-1}}^{2^{j+1}} b\left(\frac{\ell}{2^j}\right) T_{\ell;t}(\xi) \\ &= \frac{1}{\sqrt{\nu_t}} \int_{\mathbb{S}^2} \psi_j(\langle x, \xi \rangle) dN_t(\xi), \\ \psi_j(\langle x, \xi \rangle) &:= \sum_{\ell=2^{j-1}}^{2^{j+1}} b\left(\frac{\ell}{2^j}\right) \frac{2\ell+1}{4\pi} P_\ell(\langle x, \xi \rangle), \end{aligned}$$

where $\{b(\frac{\ell}{2^j})\}_{\ell=2^{j-1}, \dots, 2^{j+1}}$ is a sequence of suitably constructed weights (see [1, 33]), normalized here so that the coefficients have unit variance. It can be shown that (see [20])

$$d_3(\beta_j(\xi), Z) = O\left(\sqrt{\frac{2^{2j}}{\nu_t}}\right) = O\left(\sqrt{\frac{\ell_j^2}{\nu_t}}\right), \quad \ell_j := 2^j,$$

so that asymptotic Gaussianity follows only for multipoles which grow sub-linearly with respect to $\sqrt{\nu_t}$. Heuristically, the kernel $\{\psi_j(\langle \xi, \cdot \rangle)\}$ is characterized by a very fast decay, as opposed to Legendre polynomials (see [33, 1]); its support can be considered to shrink as ℓ_j^{-2} , and hence the “effective” Poisson rate behaves as $\ell_j^{-2} \times \nu_t$. This is very different from what we observe in this paper for Poisson random waves, because as we mentioned above the support of Legendre polynomials does not shrink in any similar way as ℓ grows, which makes asymptotic Gaussianity much simpler to achieve.

In order to focus on the more general finite dimensional distributions case, we need first to introduce some additional notation. Let us fix d points $x_1, x_2, \dots, x_d$ on $\mathbb{S}^2$ and introduce the random vector

$$F_d = (T_{\ell;t}(x_1), T_{\ell;t}(x_2), \dots, T_{\ell;t}(x_d)); \quad (2.3)$$

the elements of the covariance matrix of F_d , which we denote by $\Gamma_d := \Gamma_d^{(\ell)}(F_d)$, are easily seen to be given by

$$\Gamma_{d;ij} := \mathbb{E}[T_{\ell;t}(x_i)T_{\ell;t}(x_j)] = P_\ell(\langle x_i, x_j \rangle), \quad i, j = 1, \dots, d.$$

Note that the elements on the diagonal $\Gamma_{d;ii}$, $i = 1, \dots, d$, are exactly equal to 1 (cf. Eq. (1.2)).

Before we state our next result, some words on notation. Similarly to what was done for the definition of Wasserstein distance, where the supremum was taken with respect to Lipschitz functions with constant no larger than one, we might have defined the $d_3(\cdot, \cdot)$ distance with respect to a more definite class of functions, such that the two factors $M_2(g), M_3(g)$ are smaller than one (say). These constants are explicitly given by

$$M_k(g) := \sup_{x \neq y} \frac{\|D^{k-1}g(x) - D^{k-1}g(y)\|_{op}}{\|x - y\|_{\mathbb{R}^d}}, \quad k \in \mathbb{N}, \quad g \in C^{k-1}(\mathbb{R}^d),$$

where $D^{k-1}g(x)$ is the $(k-1)$ -th derivative of g at any point $x \in \mathbb{R}^d$ (see also Section A.1). We also recall that for a vector $x = (x_1, \dots, x_d)^T \in \mathbb{R}^d$, we denote by $\|x\|_2$ its Euclidean norm and for a matrix $A \in \mathbb{R}^{d \times d}$, we denote by $\|A\|_{op}$ the operator norm induced by the Euclidean norm, i.e.,

$$\|A\|_{op} := \sup\{\|Ax\|_2 : \|x\|_2 = 1\}.$$

We prefer however the current formulation which is more general and flexible, although slightly more cumbersome. We write Z_d for a Gaussian vector of dimension d with zero mean and covariance matrix equal to Γ_d .

Theorem 2.5. *We have that*

$$d_3(F_d, Z_d) \leq \sup_{g \in C^3} B_3(g; d) \sqrt{\frac{3}{2\pi^3}} d \sqrt{\frac{\log \ell}{\nu_t}} + o\left(d^2 \sqrt{\frac{\log \ell}{\nu_t}}\right)$$

where

$$B_3(g; d) := \frac{\sqrt{2d}}{4} M_2(g) + \frac{2d}{9} M_3(g).$$

Proof. First of all we note that all the components of F_d belong to the same first-order Poisson Wiener chaos and then we can apply Theorem A.3 (see below). Moreover, from Theorem 2.2 we have that

$$\mathbb{E}[T_{\ell;t}^4] = 3 + \frac{3}{2\pi^3} \frac{\log \ell}{\nu_t} + o\left(\frac{\log \ell}{\nu_t}\right).$$

Hence we obtain

$$\begin{aligned} |\mathbb{E}[g(F_d)] - \mathbb{E}[g(Z_d)]| &\leq B_3(g; d) \sum_{i=1}^d (\text{cum}_4(F_{d;i}))^{1/2} \\ &\sim B_3(g; d) \sqrt{\frac{3}{2\pi^3}} \sum_{i=1}^d \left(\frac{\log \ell}{\nu_t}\right)^{1/2} \end{aligned}$$

where $F_{d;i}$ is the i -th component of the vector F_d and

$$B_3(g; d) = \frac{\sqrt{2d}}{4} M_2(g) + \frac{2\sqrt{d \text{Tr}(\Gamma_d)}}{9} M_3(g).$$

We recall that for a zero mean random variable F , the fourth-cumulant $\text{cum}_4(F)$ is given by

$$\text{cum}_4(F) = \mathbb{E}[F^4] - 3(\mathbb{E}[F^2])^2,$$

see for instance [35] for more discussions and details. Noting that $\text{Tr}(\Gamma_d) = d$, the theorem is proved. $\square$

Remark 2.6. Note that we have established the bound

$$d_3(F_d, Z_d) = O_{\ell \rightarrow \infty} \left(d^2 \sqrt{\frac{\log \ell}{\nu_t}} \right);$$

the fact that the dependence on the dimension d can be made explicit allows in principle also to consider asymptotic regimes cases where d may itself vary with ℓ and ν_t .

Remark 2.7. We point out that because the proof is based on multidimensional results in Theorem A.3 (which was proved in [19]), in Theorem 2.5 we obtain upper bounds on the d_3 distance rather than the (stronger) Wasserstein distance that was considered for the one-dimensional case in Theorem 2.2. Of course, this weaker norm is still sufficient to imply convergence in distribution.

2.2 Convergence of spherical harmonic coefficients

Let us consider the vector

$$V_{\ell;t} := (\hat{a}_{\ell,-\ell}(t), \dots, \hat{a}_{\ell,\ell}(t)) = \{\hat{a}_{\ell,m}(t)\}_{m=-\ell, \dots, \ell},$$

where

$$\hat{a}_{\ell,m}(t) = \sqrt{\frac{4\pi}{(2\ell+1)\nu_t}} \sum_{k=1}^{N_t(\mathbb{S}^2)} Y_{\ell,m}(\xi_k).$$

Observe that each entry of $V_{\ell;t}$ is built by evaluating a different element of the fully normalized spherical harmonic basis $\{Y_{\ell m}\}$ over the same set of random points $\{\xi_k\}$. As a consequence, the random coefficients are neither independent nor identically distributed, although they are still uncorrelated. Indeed we have that, for all $m, m' = -\ell, \dots, \ell$

$$\mathbb{E}[\hat{a}_{\ell,m}(t)] = \int_{\mathbb{S}^2} Y_{\ell m}(z) dz = 0$$

and

$$\begin{aligned} \mathbb{E}[\hat{a}_{\ell,m}(t) \hat{a}_{\ell,m'}(t)] &= \frac{4\pi}{(2\ell+1)} \int_{\mathbb{S}^2} Y_{\ell m}(z) Y_{\ell m'}(z) dz \\ &= \delta_m^{m'} \frac{4\pi}{(2\ell+1)}. \end{aligned} \quad (2.4)$$

Theorem 2.8. Let $Z_{2\ell+1}$ be a Gaussian vector of dimension $2\ell+1$ with zero mean and diagonal variance/covariance matrix equal to $\frac{4\pi}{2\ell+1} I_{2\ell+1}$. Then we have that

$$d_3(V_{\ell;t}, Z_{2\ell+1}) \leq \sup_{g \in C^3} B_3(g; \ell) \sqrt{64 \frac{1.539 \log \ell}{\nu_t} + O\left(\frac{1}{\nu_t}\right)},$$

where

$$B_3(g; \ell) := \frac{\sqrt{2(2\ell+1)}}{4} M_2(g) + \frac{2}{9} \sqrt{(2\ell+1)4\pi} M_3(g).$$

It should be noted that the resulting bound is of order $\sqrt{\frac{\ell \log \ell}{\nu_t}}$. Before proving the theorem we need some lemmas.

Lemma 2.9. We have that

$$\mathbb{E}[\hat{a}_{\ell,m}^4(t)] = \left(\frac{4\pi}{2\ell+1} \right)^2 \frac{4\pi}{\nu_t} \mathbb{E}[|Y_{\ell m}(\xi_1)|^4] + 3 \left(\frac{4\pi}{2\ell+1} \right)^2.$$

Remark 2.10. From (A.9) (see also [31], p. 23) we have:

$$\mathbb{E} [Y_{\ell m}(\xi_1)^4] = \frac{(2\ell+1)^2}{4\pi} \frac{4}{4\pi} \left[\sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2 \left(C_{\ell,-m;\ell,m}^{L,0}\right)^2}{2L+1} \right],$$

where $\{C_{\ell_1,m_1;\ell_2,m_2}^{\ell_3,m_3}\}$ are the Clebsch-Gordan coefficients, defined in Appendix A.2 (see [42, Chapter 8] and [28, p. 77 cap. 3.5]). The factor 4 arises because (A.9) applies to complex spherical harmonics while here we deal with the real ones.

Note that as $\ell \rightarrow \infty$ we have (see [31, p. 16])

$$\sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^4}{2L+1} = \int_0^1 P_\ell(t)^4 dt \sim \frac{3}{2\pi^2} \frac{\log \ell}{\ell^2}, \quad (2.5)$$

Corollary 2.11. We have that

$$\text{cum}_4(\hat{a}_{\ell,m}(t)) = \frac{16\pi}{\nu_t} \left[\sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2 \left(C_{\ell,-m;\ell,m}^{L,0}\right)^2}{2L+1} \right]$$

and as $\ell \rightarrow \infty$

$$\begin{aligned} \text{cum}_4(\hat{a}_{\ell,0}(t)) &= \frac{16\pi}{\nu_t} \left[\sum_L \frac{\{C_{\ell,0;\ell,0}^{L,0}\}^4}{2L+1} \right] \\ &\sim 24 \frac{1}{\pi} \frac{\log \ell}{\ell^2 \nu_t}. \end{aligned}$$

The first result of Corollary 2.11 follows by exploiting Remark 2.10 together with Lemma 2.9 and recalling the definition of the fourth cumulant of a zero mean random variable. The second result is due to (2.5).

We also need the following lemma.

Lemma 2.12. We have that

$$\frac{16\pi}{(2\ell+1)\nu_t} \leq \sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t)) \leq 32 \frac{1.539 \log \ell}{\ell \nu_t} + O\left(\frac{1}{\ell \nu_t}\right). \quad (2.6)$$

Proof of Lemma 2.12. From Corollary 2.11 we get

$$\begin{aligned} \sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t)) &= \sum_{m=-\ell}^{\ell} \frac{16\pi}{\nu_t} \left[\sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2 \left(C_{\ell,-m;\ell,m}^{L,0}\right)^2}{2L+1} \right] \\ &= \frac{16\pi}{\nu_t} \sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} \sum_{m=-\ell}^{\ell} \left(C_{\ell,-m;\ell,m}^{L,0}\right)^2. \end{aligned}$$

In view of the unitary property (A.5) recalled in Appendix A.2 (see for example eq. (3.62) [28]), we have

$$\sum_{m=-\ell}^{\ell} \left(C_{\ell,-m;\ell,m}^{L,0}\right)^2 = 1$$

and then we obtain

$$\sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t)) = \frac{16\pi}{\nu_t} \sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1}. \quad (2.7)$$

The relation (A.3) between the Clebsch-Gordan coefficients and the 3j Wigner coefficients (defined in Appendix A.2), leads to

$$\frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} = \begin{pmatrix} \ell & \ell & L \\ 0 & 0 & 0 \end{pmatrix}^2.$$

From Lemma A.1 in [32] we have that

$$\begin{pmatrix} \ell & \ell & L \\ 0 & 0 & 0 \end{pmatrix}^2 = \frac{2}{\pi} \gamma_{\ell L} \frac{1}{L(2\ell-L)^{1/2}(2\ell+L)^{1/2}}; \quad (2.8)$$

where $\{\gamma_{\ell L}\}_{\ell=1,2,\dots}$ is a deterministic sequence such that $0.596 \leq \gamma_{\ell L} \leq 1.539$ uniformly in ℓ and L . Then

$$\sum_{L=0}^{2\ell} \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} = \left(C_{\ell,0;\ell,0}^{0,0}\right)^2 + \sum_{L=1}^{2\ell-1} \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} + \frac{\left(C_{\ell,0;\ell,0}^{2\ell,0}\right)^2}{4\ell+1}.$$

In view of (A.7), the first term of this sum is

$$\left(C_{\ell,0;\ell,0}^{0,0}\right)^2 = \frac{1}{2\ell+1},$$

and from (A.8) we have that $\left(C_{\ell,0;\ell,0}^{2\ell,0}\right)^2 \leq \frac{1}{(4\ell+1)}$ so that the last term of the sum above satisfies

$$\frac{\left(C_{\ell,0;\ell,0}^{2\ell,0}\right)^2}{4\ell+1} \leq \frac{1}{(4\ell+1)^2}.$$

In view of (2.8), we also get

$$\sum_{L=1}^{2\ell-1} \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} = \frac{2}{\pi} \sum_{L=1}^{2\ell-1} \frac{\gamma_{\ell L}}{L(2\ell-L)^{1/2}(2\ell+L)^{1/2}}.$$

Now we note that

$$\begin{aligned} \sum_{L=1}^{2\ell-1} \frac{1}{L(2\ell-L)^{1/2}(2\ell+L)^{1/2}} &\leq \frac{1}{(2\ell+1)^{1/2}} \sum_{L=1}^{2\ell-1} \frac{1}{L(2\ell-L)^{1/2}} \\ &= \frac{1}{(2\ell+1)^{1/2}} \sum_{L=1}^{\ell} \frac{1}{L(2\ell-L)^{1/2}} + \frac{1}{(2\ell+1)^{1/2}} \sum_{L=\ell+1}^{2\ell-1} \frac{1}{L(2\ell-L)^{1/2}}. \end{aligned} \quad (2.9)$$

The first sum can be bounded as follows:

$$\frac{1}{(2\ell+1)^{1/2}} \sum_{L=1}^{\ell} \frac{1}{L(2\ell-L)^{1/2}} \leq \frac{1}{(2\ell+1)^{1/2}(\ell)^{1/2}} \sum_{L=1}^{\ell} \frac{1}{L} \leq \frac{\log \ell}{\ell}.$$

For the second sum in (2.9) we have that

$$\frac{1}{(2\ell+1)^{1/2}} \sum_{L=\ell+1}^{2\ell-1} \frac{1}{L(2\ell-L)^{1/2}} \leq \frac{1}{(2\ell+1)^{1/2}(\ell+1)} \sum_{L=\ell+1}^{2\ell-1} \frac{1}{(2\ell-L)^{1/2}}$$

and changing variable $L' = 2\ell - L$ we get

$$\frac{1}{(2\ell+1)^{1/2}(\ell+1)} \sum_{L'=1}^{\ell-1} \frac{1}{(L')^{1/2}} \leq \frac{1}{(2\ell+1)^{1/2}(\ell+1)} (\ell)^{1/2} \leq \frac{1}{\ell}.$$

Finally we have

$$\sum_{L=0}^{2\ell} \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} \leq \frac{2}{\pi} \left(\frac{1.539 \log \ell}{\ell} + \frac{1.539}{\ell} + \frac{\pi}{2(4\ell+1)^2} + \frac{\pi}{2(2\ell+1)} \right)$$

and exploiting this bound in (2.7) we conclude that

$$\sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t)) \leq 32 \frac{1.539 \log \ell}{\ell \nu_t} + O\left(\frac{1}{\ell^2 \nu_t}\right).$$

On the other hand it can be easily seen that

$$\begin{aligned} \sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t)) &= \frac{16\pi}{\nu_t} \sum_L \frac{\left(C_{\ell,0;\ell,0}^{L,0}\right)^2}{2L+1} \geq \frac{16\pi}{\nu_t} \left(C_{\ell,0;\ell,0}^{0,0}\right)^2 \\ &= \frac{16\pi}{\nu_t} \frac{1}{2\ell+1} = \frac{16\pi}{(2\ell+1)\nu_t}, \end{aligned}$$

where we used (A.7) in the second-last equality, and that leads to the thesis of the lemma. $\square$

Proof of Theorem 2.8. We exploit the multidimensional Fourth Moment Theorem in [19], in particular Theorem A.3 in Section A, to get

$$|\mathbb{E}[g(V_{\ell;t})] - \mathbb{E}[g(Z_{2\ell+1})]| \leq B_3(g; \ell) \sum_{m=-\ell}^{\ell} \sqrt{\text{cum}_4(\hat{a}_{\ell,m}(t))}. \quad (2.10)$$

Applying the following Cauchy-Schwarz inequality

$$\sum_{i=1}^d \sqrt{a_i} \leq d^{\frac{1}{2}} \left(\sum_{i=1}^d a_i \right)^{\frac{1}{2}}, \quad (2.11)$$

it follows that

$$|\mathbb{E}[g(V_{\ell;t})] - \mathbb{E}[g(Z_{2\ell+1})]| \leq B_3(g; \ell) \sqrt{2\ell+1} \sqrt{\sum_{m=-\ell}^{\ell} \text{cum}_4(\hat{a}_{\ell,m}(t))}. \quad (2.12)$$

In view of the definition of $B_3(g; d)$ in (A.1) and using (2.4) we find

$$B_3(g; \ell) = A_2(g; \ell) + \frac{2\sqrt{(2\ell+1)4\pi}}{9} M_3(g),$$

with

$$A_2(g; \ell) = \frac{\sqrt{2(2\ell+1)}}{4} M_2(g).$$

Exploiting the upper bound of (2.6) in Lemma 2.12, the thesis of the theorem follows. $\square$

Remark 2.13. Since the covariance matrix of the vector $V_{\ell;t}$ is positive definite, we can also apply the second part of Theorem A.3 in Section A.1 to get a quantitative Central Limit Theorem in terms of the metric d_2 , defined as follows. For any two random vectors $X, Y : \Omega \rightarrow \mathbb{R}^d$ such that $\mathbb{E}[\|X\|_{\mathbb{R}^d}^2], \mathbb{E}[\|Y\|_{\mathbb{R}^d}^2] < \infty$, we have that

$$d_2(X, Y) = \sup_{h \in \mathcal{I}} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|$$

where $\mathcal{I}$ indicates the collection of all functions $h \in \mathbb{C}^2(\mathbb{R}^d)$ such that $\|h\|_{Lip} \leq 1$ and $M_2(h) \leq 1$.

Similarly to the proof of Theorem 2.8, we find

$$d_2(V_{\ell;t}, Z_{2\ell+1}) \leq \sup_{g \in C^2} B_2(g; \ell) \sqrt{64 \frac{1.539 \log \ell}{\nu_t} + O\left(\frac{1}{\ell \nu_t}\right)}$$

where

$$B_2(g; \ell) = A_1(g) + \frac{\sqrt{2\pi}}{6} \sqrt{2\ell+1} M_2(g) \text{ and } A_1(g) = \sqrt{\frac{2\ell+1}{4\pi}} \frac{1}{\sqrt{\pi}} M_1(g).$$

A natural question concerns the relationship between the results of this section and the quantitative bounds for the convergence of the finite-dimensional distributions provided in the previous pages. To this aim, we recall the definition of $F_d = (T_{\ell;t}(x_1), \dots, T_{\ell;t}(x_d))$ and we simply note that

$$\begin{aligned} F_d &= \begin{pmatrix} Y_{\ell,-\ell}(x_1) & Y_{\ell,-\ell+1}(x_1) & \dots & Y_{\ell,\ell}(x_1) \\ \vdots & \vdots & \vdots & \vdots \\ Y_{\ell,-\ell}(x_d) & Y_{\ell,-\ell+1}(x_d) & \dots & Y_{\ell,\ell}(x_d) \end{pmatrix} \begin{pmatrix} \widehat{a}_{\ell,-\ell}(t) \\ \vdots \\ \widehat{a}_{\ell,\ell}(t) \end{pmatrix} \\ &=: \Psi_{\ell;d}(a_{\ell,\cdot}), \end{aligned}$$

where $\Psi_{\ell;d} : \mathbb{R}^{2\ell+1} \rightarrow \mathbb{R}^d$ is a linear function and hence obviously continuous (and bounded). Indeed, because $|Y_{\ell,m}| \leq \sqrt{\frac{2\ell+1}{2\pi}}$ uniformly over the sphere, for all functions $h : \mathbb{R}^d \rightarrow \mathbb{R}$ we can write

$$h(T_{\ell;t}(x_1), \dots, T_{\ell;t}(x_d)) = h \circ \Psi_{\ell;d}(a_{\ell,\cdot}) =: \tilde{h}(a_{\ell,\cdot}).$$

Hence we have that

$$\begin{aligned} &\sup_{h \in C^3} |\mathbb{E}[h(T_{\ell;t}(x_1), \dots, T_{\ell;t}(x_d))] - \mathbb{E}[h(Z_1, \dots, Z_d)]| \\ &= \sup_{h \in C^3} |\mathbb{E}[(h \circ \Psi_{\ell;d})(a_{\ell,\cdot})] - \mathbb{E}[(h \circ \Psi_{\ell;d})(Z_1, \dots, Z_{2\ell+1})]| \\ &\leq \sup_{h \in C^3} B_3(h \circ \Psi_{\ell;d}) \sqrt{64 \frac{1.539 \log \ell}{\nu_t} + O\left(\frac{1}{\ell \nu_t}\right)} \end{aligned}$$

where

$$\begin{aligned} B_3(h \circ \Psi_{\ell;d}) &= \frac{\sqrt{2(2\ell+1)}}{4} M_2(h \circ \Psi_{\ell;d}) + \frac{2}{9} \sqrt{(2\ell+1)4\pi} M_3(h \circ \Psi_{\ell;d}) \\ &\leq \frac{\sqrt{2(2\ell+1)}}{4} \sqrt{\frac{2\ell+1}{2\pi}} d M_2(h) + \frac{2}{9} \sqrt{(2\ell+1)4\pi} \sqrt{\frac{2\ell+1}{2\pi}} d M_3(h) \\ &= d \frac{2\ell+1}{4\sqrt{\pi}} M_2(h) + \frac{2\sqrt{2}d}{9} (2\ell+1) M_3(h). \end{aligned}$$

Here we have used the simple fact that

$$\sup_{h: \mathbb{R}^d \rightarrow \mathbb{R}} M_k(h \circ \Psi_{\ell;d}) \leq d \sqrt{\frac{2\ell+1}{2\pi}} \sup_h M_k(h), k \in \mathbb{N}.$$

Summing up, we have here a bound of order $\frac{d\ell\sqrt{\log \ell}}{\sqrt{\nu_t}}$, to be compared with the bound of order $d^2\sqrt{\frac{\log \ell}{\nu_t}}$ which was obtained in Theorem 2.5. Overall, we can claim that

$$\begin{aligned} & \sup_{h \in \mathcal{C}^3, \|h\|_{\mathcal{C}^3} < 1} |\mathbb{E}[h(T_{\ell,t}(x_1), \dots, T_{\ell,t}(x_d))] - \mathbb{E}[h(Z_1, \dots, Z_d)]| \\ &= O\left(d \times (\ell \wedge d) \times \sqrt{\frac{\log \ell}{\nu_t}}\right). \end{aligned}$$

2.3 Functional convergence

In the previous subsections we presented a number of quantitative convergence results for sequences of random vectors, such as vectors of spherical harmonic coefficients or points evaluations over a subset of d points. It is natural to ask whether we can also obtain results for the sequence of eigenfunctions $\{T_{\ell,t}(\cdot)\}$ considered as random elements in functional spaces; the answer is affirmative, thanks to some very recent results in this direction in [6]. We shall consider in particular $L^2(\mathbb{S}^2)$ and the Sobolev space $W_{\alpha,2}(\mathbb{S}^2)$, $\alpha > 0$, to distinguish the probability metric in the two cases, we shall write $d_{3,L^2(\mathbb{S}^2)}$ and $d_{3,W_{\alpha,2}(\mathbb{S}^2)}$, respectively. Let us recall also that for a general function space K we have that $C_b^3(K)$ is the class of real-valued functions on K that have bounded Fréchet derivatives up to order three. This space is equipped with the norm

$$\|h\|_{C_b^3(K)} = \sup_{j=1,2,3} \sup_{x \in K} \|D^j h(x)\|_{K^{\otimes j}}.$$

2.3.1 Quantitative central limit theorems in $L^2(\mathbb{S}^2)$

We start by considering the space of $L^2(\mathbb{S}^2)$. Our main result is the following.

Theorem 2.14. *Let Z be a centred Gaussian process with the same covariance operator as $T_{\ell,t}$. We have that*

$$d_{3,L^2(\mathbb{S}^2)}(T_{\ell,t}, Z) \leq \left(\frac{1}{4} + \sqrt{\pi}\right) \sqrt{\frac{4\pi}{\nu_t}}$$

Proof. In view of Theorem A.4 (see [6], Theorem 3 and Corollary 1), we need to compute the quantity

$$\mathbb{E}[\|T_{\ell,t}\|_{L^2(\mathbb{S}^2)}^4] - (\mathbb{E}[\|T_{\ell,t}\|_{L^2(\mathbb{S}^2)}^2])^2 - 2\|S_{\ell,t}\|_{HS(L^2(\mathbb{S}^2))}^2,$$

where $S_{\ell,t}$ is the covariance operator of $T_{\ell,t}$ and $\|\cdot\|_{HS}$ denotes the Hilbert-Schmidt norm (see the end of Appendix A.1). First, we have that

$$\begin{aligned} & \mathbb{E}[\|T_{\ell,t}\|_{L^2(\mathbb{S}^2)}^2] \\ &= \mathbb{E}\left[\int_{\mathbb{S}^2} |T_{\ell,t}(x)|^2 dx\right] = \int_{\mathbb{S}^2} \sum_{m_1=-\ell}^{\ell} \sum_{m_2=-\ell}^{\ell} \mathbb{E}[\hat{a}_{\ell,m_1}(t) \hat{a}_{\ell,m_2}(t)] Y_{\ell m_1}(x) Y_{\ell m_2}(x) dx \\ &= \int_{\mathbb{S}^2} \frac{4\pi}{2\ell+1} \sum_{m=-\ell}^{\ell} Y_{\ell m}(x) Y_{\ell m}(x) dx \\ &= \frac{4\pi}{2\ell+1} \sum_{m=-\ell}^{\ell} \int_{\mathbb{S}^2} Y_{\ell m}(x) Y_{\ell m}(x) dx = 4\pi. \end{aligned}$$

It follows that $(\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^2])^2 = (4\pi)^2$. Now we compute $\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4]$, which gives

$$\begin{aligned}\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4] &= \mathbb{E}\left[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^2 \|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^2\right] = \mathbb{E}\left[\sum_{m_1=-\ell}^{\ell} |\hat{a}_{\ell,m_1}(t)|^2 \sum_{m_2=-\ell}^{\ell} |\hat{a}_{\ell,m_2}(t)|^2\right] \\ &= \left(\frac{4\pi}{(2\ell+1)\nu_t}\right)^2 \mathbb{E}\left[\sum_{m_1=-\ell}^{\ell} \sum_{k_1 k_2} Y_{\ell m_1}(\xi_{k_1}) Y_{\ell m_1}(\xi_{k_2}) \sum_{m_2=-\ell}^{\ell} \sum_{k_3 k_4} Y_{\ell m_2}(\xi_{k_3}) Y_{\ell m_2}(\xi_{k_4})\right].\end{aligned}$$

Applying the addition formula we get

$$\begin{aligned}\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4] &= \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=1}^{N_t(\mathbb{S}^2)} \sum_{k_2=1}^{N_t(\mathbb{S}^2)} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle) \sum_{k_3=1}^{N_t(\mathbb{S}^2)} \sum_{k_4=1}^{N_t(\mathbb{S}^2)} P_{\ell}(\langle \xi_{k_3}, \xi_{k_4} \rangle)\right] \\ &= \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=1}^{N_t(\mathbb{S}^2)} P_{\ell}(\langle \xi_{k_1}, \xi_{k_1} \rangle)^2\right] \\ &\quad + \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=k_2 \neq k_3=k_4} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle) P_{\ell}(\langle \xi_{k_3}, \xi_{k_4} \rangle)\right] \\ &\quad + \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=k_3 \neq k_2=k_4} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle) P_{\ell}(\langle \xi_{k_3}, \xi_{k_4} \rangle)\right] \\ &\quad + \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=k_4 \neq k_3=k_2} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle) P_{\ell}(\langle \xi_{k_3}, \xi_{k_4} \rangle)\right]\end{aligned}$$

and since $P_{\ell}(0) = 1$ for all ℓ we obtain

$$\begin{aligned}\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4] &= \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=1}^{N_t(\mathbb{S}^2)} 1\right] + \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=k_2 \neq k_3=k_4} 1\right] \\ &\quad + 2 \left(\frac{1}{\nu_t}\right)^2 \mathbb{E}\left[\sum_{k_1=k_3 \neq k_2=k_4} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle)^2\right] \\ &= \frac{4\pi}{\nu_t} + (4\pi)^2 \left(\frac{1}{\nu_t}\right)^2 \nu_t^2 + \left(\frac{1}{\nu_t}\right)^2 2\nu_t^2 \int_{(\mathbb{S}^2)^2} P_{\ell}(\langle \xi_{k_1}, \xi_{k_2} \rangle)^2 d\xi_{k_1} d\xi_{k_2} \\ &= \frac{4\pi}{\nu_t} + (4\pi)^2 + 2 \frac{(4\pi)^2}{2\ell+1}.\end{aligned}$$

The covariance operator $S_{\ell;t}$ is such that

$$\|S_{\ell;t}\|_{HS(L^2(\mathbb{S}^2))}^2 = \sum_{m=-\ell}^{\ell} \sum_{m'=-\ell}^{\ell} \mathbb{E}[a_{\ell,m}(t) a_{\ell,m'}(t)]^2 = \sum_{m=-\ell}^{\ell} \sum_{m'=-\ell}^{\ell} \left(\delta_m^{m'} \frac{4\pi}{2\ell+1}\right)^2 = \frac{(4\pi)^2}{2\ell+1},$$

and then we finally obtain

$$\begin{aligned}\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4] &- (\mathbb{E}[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^2])^2 - 2 \|S_{\ell;t}\|_{HS(L^2(\mathbb{S}^2))}^2 \\ &= \frac{4\pi}{\nu_t} + (4\pi)^2 + 2 \frac{(4\pi)^2}{2\ell+1} - (4\pi)^2 - 2 \frac{(4\pi)^2}{2\ell+1} = \frac{4\pi}{\nu_t}.\end{aligned}$$

Exploiting Theorem A.4 (see also [6]) we get the thesis of the theorem. $\square$

As mentioned above, it may come at first sight as a surprise that the rate of convergence in this functional setting (i.e., $1/\sqrt{\nu_t}$) does not depend on the index ℓ and it is indeed faster than in the finite-dimensional case. The apparent paradox is solved noting that the topology here is too coarse to imply convergence of the finite-dimensional distributions. In the next subsection, we investigate convergence in functional spaces with a finer topological structure.

2.3.2 Quantitative central limit theorems in $W_{\alpha,2}(\mathbb{S}^2)$

Now we consider the random eigenfunctions taking values in Sobolev spaces $W_{\alpha,2}(\mathbb{S}^2)$, $\alpha > 0$, on the sphere, i.e., the spaces of functions $f \in L(\mathbb{S}^2)$, $f = \sum_{\ell,m} a_{\ell,m} Y_{\ell,m}$, with finite norm

$$\|f\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 = \sum_{\ell \geq 0} \sum_{m=-\ell}^{\ell} \left(1 + \sqrt{\ell(\ell+1)}\right)^{2\alpha} |a_{\ell,m}|^2.$$

Our main result here is the following.

Theorem 2.15. *Let Z be a centred Gaussian process with the same covariance operator as $T_{\ell;t}$. We have that*

$$d_{3,W_{\alpha,2}}(T_{\ell;t}, Z) \leq \frac{\sqrt{\pi} \left(1 + \sqrt{\ell(\ell+1)}\right)^{2\alpha}}{2\sqrt{\nu_t}} + \frac{2\pi \left(1 + \sqrt{\ell(\ell+1)}\right)^{3\alpha}}{\sqrt{\nu_t}}.$$

Proof. First note that

$$\mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^4 \right] = \left(1 + \sqrt{\ell(\ell+1)}\right)^{4\alpha} \mathbb{E} \left[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4 \right]$$

and

$$\mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \right] = \left(1 + \sqrt{\ell(\ell+1)}\right)^{2\alpha} \mathbb{E} \left[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^2 \right]. \quad (2.13)$$

Indeed, we have that

$$\begin{aligned} \mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^4 \right] &= \mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \right] \\ &= \mathbb{E} \left[\sum_{m=-\ell}^{\ell} \sum_{m'=-\ell}^{\ell} \left(1 + \sqrt{\ell(\ell+1)}\right)^{4\alpha} |\widehat{a}_{\ell,m}|^2 |\widehat{a}_{\ell,m'}|^2 \right] \\ &= \left(1 + \sqrt{\ell(\ell+1)}\right)^{4\alpha} \mathbb{E} \left[\sum_{m=-\ell}^{\ell} \sum_{m'=-\ell}^{\ell} |\widehat{a}_{\ell,m}|^2 |\widehat{a}_{\ell,m'}|^2 \right] \\ &= \left(1 + \sqrt{\ell(\ell+1)}\right)^{4\alpha} \mathbb{E} \left[\|T_{\ell;t}\|_{L^2(\mathbb{S}^2)}^4 \right], \end{aligned}$$

where in the last equation we used Parseval's identity. Similarly (2.13) holds. In view of the computations of the previous section and this remark, we conclude that

$$\mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^4 \right] = \left(1 + \sqrt{\ell(\ell+1)}\right)^{4\alpha} \left(\frac{4\pi}{\nu_t} + (4\pi)^2 + 2(4\pi) \frac{4\pi}{2\ell+1} \right)$$

and

$$\mathbb{E} \left[\|T_{\ell;t}\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \right] = 4\pi \left(1 + \sqrt{\ell(\ell+1)}\right)^{2\alpha}.$$

Furthermore, letting $\{e_i: i \geq 1\}$ be an orthonormal basis of $W_{\alpha,2}(\mathbb{S}^2)$, we can compute $\|S_{\ell,t}\|_{HS(W_{\alpha,2})}^2$ as

$$\begin{aligned} \|S_{\ell,t}\|_{HS(W_{\alpha,2})}^2 &= \sum_{i \geq 1} \left\| \mathbb{E} \left(\langle T_{\ell,t}, e_i \rangle_{W_{\alpha,2}(\mathbb{S}^2)} T_{\ell,t} \right) \right\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \\ &= \sum_{i \geq 1} \left\| \sum_{m=-\ell}^{\ell} \mathbb{E} \left(\widehat{a}_{\ell,m}(t)^2 \right) \langle Y_{\ell m}, e_i \rangle_{W_{\alpha,2}(\mathbb{S}^2)} Y_{\ell m} \right\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \\ &= \sum_{m=-\ell}^{\ell} \left(1 + \sqrt{\ell(\ell+1)} \right)^{2\alpha} \left(\frac{4\pi}{2\ell+1} \right)^2 \sum_{i \geq 1} \langle Y_{\ell m}, e_i \rangle_{W_{\alpha,2}(\mathbb{S}^2)}^2 \\ &= \sum_{m=-\ell}^{\ell} \left(1 + \sqrt{\ell(\ell+1)} \right)^{4\alpha} \left(\frac{4\pi}{2\ell+1} \right)^2 \\ &= \left(1 + \sqrt{\ell(\ell+1)} \right)^{4\alpha} \frac{(4\pi)^2}{2\ell+1}. \end{aligned}$$

We now have all the necessary elements to apply Theorem A.4 as in the previous subsection, from which the result follows after elementary algebraic manipulations. $\square$

As a final result we want to show that, for $\alpha > \frac{3}{2}$, a quantitative Central Limit Theorem in Sobolev space does indeed imply the quantitative Central Limit Theorem for the marginal distribution at every given location on the sphere. We start by noting that

$$\begin{aligned} \|f\|_{L^\infty(\mathbb{S}^2)} &= \sup_x \left| \sum_{\ell} \sum_m a_{\ell m}(f) Y_{\ell m}(x) \right| \\ &\leq \sum_{\ell} \sum_m |a_{\ell m}(f)| \sup_x |Y_{\ell m}(x)| \\ &\leq \sum_{\ell} \sum_m |a_{\ell m}(f)| \sqrt{\frac{2\ell+1}{2\pi}}, \end{aligned}$$

whence

$$\|f\|_{L^\infty(\mathbb{S}^2)}^2 \leq \frac{1}{2\pi} \left\{ \sum_{\ell} \sum_m |a_{\ell m}(f)| \sqrt{2\ell+1} \right\}^2.$$

Multiplying and dividing by $(1 + \sqrt{\ell(\ell+1)})^\alpha \sqrt{2\ell+1}$ and then applying the Cauchy-Schwarz inequality twice, we get

$$\begin{aligned} \|f\|_{L^\infty(\mathbb{S}^2)}^2 &\leq \frac{1}{2\pi} \sum_{\ell} (2\ell+1) \sum_m |a_{\ell m}(f)|^2 \frac{(1 + \sqrt{\ell(\ell+1)})^{2\alpha}}{(2\ell+1)} \sum_{\ell} \frac{(2\ell+1)^2}{(1 + \sqrt{\ell(\ell+1)})^{2\alpha}} \\ &= \frac{1}{2\pi} \|f\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \sum_{\ell} \frac{(2\ell+1)^2}{(1 + \sqrt{\ell(\ell+1)})^{2\alpha}} \\ &\leq \frac{2}{\pi} \|f\|_{W_{\alpha,2}(\mathbb{S}^2)}^2 \zeta(2\alpha - 2), \end{aligned}$$

where as usual

$$\zeta(2\alpha - 2) = \sum_{\ell=1}^{\infty} \frac{1}{\ell^{2\alpha-2}} < \infty$$

as $\alpha > \frac{3}{2}$. Hence, we have that

$$\|f\|_{L^\infty(\mathbb{S}^2)}^2 < \frac{2}{\pi} \zeta(2\alpha - 2) \times \|f\|_{W_{\alpha,2}(\mathbb{S}^2)}^2.$$

Because of this inequality, the topology induced by the norm $\|\cdot\|_{W_{\alpha,2}(\mathbb{S}^2)}$ is finer than the topology generated by $\|\cdot\|_{L^\infty(\mathbb{S}^2)}$; hence a function continuous with respect to the latter is certainly continuous with respect to the former as well. Therefore

$$\sup_{h \text{ continuous w.r.t. } \|\cdot\|_{L^\infty(\mathbb{S}^2)}} |\mathbb{E}h(X) - \mathbb{E}h(Y)| \leq \sup_{h \text{ continuous w.r.t. } \|\cdot\|_{W_{\alpha,2}(\mathbb{S}^2)}} |\mathbb{E}h(X) - \mathbb{E}h(Y)|.$$

Now, given X_ℓ a random element taking values in the Sobolev space $W_{\alpha,2}(\mathbb{S}^2)$ and Z_ℓ a Gaussian process taking values in the same Sobolev space and with the same covariance operator as X_ℓ , we show that $d_{3,W_{\alpha,2}}(X_\ell, Z_\ell) \rightarrow 0$ implies

$$\mathbb{E}g(X_\ell(x)) \rightarrow \mathbb{E}g(Z_\ell(x)) \text{ for all } g \in C_b^3(\mathbb{R}),$$

for fixed $x \in \mathbb{S}^2$, which in turn implies $X_\ell(x) \rightarrow_d N(0, 1)$ because d_3 metrizes convergence in distribution, in particular on $\mathbb{R}$. Actually we are able to prove the following, slightly stronger result.

Proposition 2.16. *For $\alpha > \frac{3}{2}$, we have that*

$$d_3(X_\ell(x), Z_\ell(x)) = \sup_{g \in C_b^3(\mathbb{R})} |\mathbb{E}g(X_\ell(x)) - \mathbb{E}g(Z_\ell(x))| \leq C(\alpha)d_{3,W_{\alpha,2}}(X_\ell, Z_\ell),$$

where the term $C(\alpha)$ does not depend on ℓ .

Proof. We can write

$$\mathbb{E}g(X_\ell(x)) = \mathbb{E}(g \circ \pi_x(X_\ell(\cdot))) \text{ where } g \circ \pi_x \in C_b,$$

where the evaluation map $\pi_x : \pi_x(X_\ell) = X_\ell(x)$ is continuous with respect to the Sobolev norm (because it is continuous with respect to the sup norm).

Note that the Gateaux derivatives of the evaluation functionals are given by

$$\frac{|\pi_x(X_\ell + tH) - \pi_x(X_\ell)|}{t} = H(x) = \pi_x(H), \quad \forall H \in W_{\alpha,2},$$

so that the Fréchet derivative $((D\pi_x)(X_\ell))H = H(x)$, that is $((D\pi_x)(X_\ell)) = \pi_x$. Note also that the (dual) norm of π_x is bounded, indeed by its definition we have that

$$\begin{aligned} \|\pi_x(\cdot)\|_{W_{\alpha,2}^*} &:= \sup_{h: \|h\|_{W_{\alpha,2}}=1} |\pi_x(h)| = \sup_h \frac{|h(x)|}{\|h\|_{W_{\alpha,2}(\mathbb{S}^2)}} \\ &\leq \sup_h \frac{\|h\|_{L^\infty(\mathbb{S}^2)}}{\|h\|_{W_{\alpha,2}(\mathbb{S}^2)}} \leq \frac{2}{\pi} \zeta(2\alpha - 2). \end{aligned}$$

Similar results are obtained if we take the second or third order Fréchet derivatives, with the same bound. Therefore we have that

$$d_3(X_\ell(x), Z_\ell(x)) = \sup_{g \in C_b^3(\mathbb{R})} |\mathbb{E}g(X_\ell(x)) - \mathbb{E}g(Z_\ell(x))| \leq C(\alpha)d_{3,W_{\alpha,2}}(X_\ell, Z_\ell),$$

which proves the claim with $C(\alpha) := \frac{2}{\pi} \zeta(2\alpha - 2)$. $\square$

Remark 2.17. As usual with Sobolev spaces, the coefficient α in $W_{\alpha,2}$ characterizes regularity; in particular, $\alpha > \frac{3}{2}$ implies that the elements in this space are not only continuous, but also differentiable, in the L^2 sense.

A Appendix

In this Appendix, we collect for convenience a number of background results on Fourth Moment Theorems in a Poisson environment and on integrals of spherical harmonics. We start introducing some notation and definitions.

A.1 Wiener chaos in a Poisson environment

We now present, in a form properly adapted to our goals, some upper bounds related to random variables living in the first Wiener chaos of a Poisson random measure. The first two bounds have been proved in [19] and provide a Fourth Moment Theorem on the Poisson space for the univariate and the multivariate case respectively. The third bound appears in [6] and concerns a quantitative and functional Central Limit Theorem for convergence to a Gaussian process.

We start by recalling some basic concepts on Poisson random measures and Wiener chaos.

Let us consider $(\Theta, \mathcal{A}, \rho)$ a Probability space as given in Definition 1.2 and N a Poisson random measure on Θ with intensity measure ρ ; let $\hat{N}$ be defined by $\hat{N} := N - \rho$.

For the sake of brevity, we will make use of the shorthand notation $L^p(\rho)$ to denote the space $L^p(\Theta, \mathcal{A}, \rho)$, $p \geq 1$.

Definition A.1 (First Wiener chaos). *For every deterministic function $h \in L^2(\rho)$, the Wiener-Itô integral of h with respect to $\hat{N}$ is given by*

$$I_1(h) = \int_{\Theta} h(z) \hat{N}(dz).$$

The Hilbert space composed of the random variables of the form $I_1(h)$, where $h \in L^2(\rho)$, and labeled by W_1 , is called the first Wiener chaos associated with the Poisson measure N .

Theorem A.2 (Quantitative Fourth Moment Theorem (unidimensional case), [19, Theorem 2.1 and Corollary 1.3] and [18, Theorem 1.3]). *For $\ell \in \mathbb{N}$, let $F \in W_1$, while $Z \sim \mathcal{N}(0, 1)$ denote a standard normal random variable. Moreover, assume that $\text{Var}(F) = 1$ and $\mathbb{E}[F^4] < \infty$. Then it holds that:*

$$d_W(F, Z) \leq c_1 \sqrt{\mathbb{E}[F^4] - 3},$$

where

$$c_1 := \frac{1}{\sqrt{2\pi}} + \frac{2}{3}.$$

Moreover, it holds that

$$d_{Kol}(F, Z) \leq \left(11 + (\mathbb{E}[F^4])^{1/2} + (\mathbb{E}[F^4])^{1/4}\right) \sqrt{\mathbb{E}[F^4] - 3}.$$

Before stating the next result, we need some additional notation. For any $\ell \in \mathbb{N}$, fixed an integer $d \geq 2$, we consider the centred random vector $F = (F_1, \dots, F_d)^T$ where $F_j \in W_1$, for $1 \leq j \leq d$. For $j = 1, \dots, d$. We denote by Γ_d the covariance matrix of F , i.e. $\Gamma_{d;ij} = \mathbb{E}[F_i F_j]$ for $i, j = 1, \dots, d$.

For a k -multilinear form $\psi : (\mathbb{R}^d)^k \rightarrow \mathbb{R}$, $k \in \mathbb{N}$, we define the operator norm

$$\|\psi\|_{op} := \sup\{|\psi(u_1, \dots, u_k)| : u_j \in \mathbb{R}^d, \|u_j\|_2 = 1, j = 1, \dots, k\}.$$

Furthermore, for $g \in C^{k-1}(\mathbb{R}^d)$, note that $D^{k-1}g(x)$, the $(k-1)$ -derivative of g at the point x , can be read as a multilinear form. In this setting, we can define the generalization of

the minimum Lipschitz constant for any derivative of order $k - 1$ of $g(\cdot)$ as follows: for fixed $k \geq 1$ take

$$M_k(g) := \sup_{x \neq y} \frac{\|D^{k-1}g(x) - D^{k-1}g(y)\|_{op}}{\|x - y\|_{\mathbb{R}^d}},$$

see again [19].

Theorem A.3 (Quantitative Fourth Moment Theorem (multidimensional case), [19, Theorem 1.7, Corollary 1.8 and Remark 4.3]). *Under the above notation, let Z_d be a centred Gaussian random vector of dimension d with covariance matrix Γ_d . Then, for every $g \in C^3(\mathbb{R}^d)$, we have that*

$$|\mathbb{E}[g(F)] - \mathbb{E}[g(Z_d)]| \leq B_3(g; d) \sum_{i=1}^d \sqrt{\mathbb{E}[F_i^4] - 3\mathbb{E}[F_i^2]^2}$$

where

$$B_3(g; d) = A_2(g; d) + \frac{2\sqrt{d} \operatorname{Tr}(\Gamma_d)}{9} M_3(g), \quad A_2(g; d) = \frac{\sqrt{2d}}{4} M_2(g). \quad (\text{A.1})$$

If in addition Γ_d is positive definite, then for every $g \in C^2(\mathbb{R}^d)$, it holds that

$$|\mathbb{E}[g(F)] - \mathbb{E}[g(Z)]| \leq B_2(g; d) \sum_{i=1}^d \sqrt{\mathbb{E}[F_i^4] - 3\mathbb{E}[F_i^2]^2}$$

with

$$B_2(g; d) = A_1(g; d) + \frac{\sqrt{2\pi} \|\Gamma_d^{-\frac{1}{2}}\|_{op} \operatorname{Tr}(\Gamma_d)}{6} M_2(g), \quad A_1(g; d) = \frac{\|\Gamma_d^{-\frac{1}{2}}\|_{op}}{\sqrt{\pi}} M_1(g).$$

Let K be a separable Hilbert space and X a K -valued random variable in $L^2(\rho)$. We recall that if $X \in L^2(\rho)$, with $\mathbb{E}[\|X\|_K^2] < \infty$, the covariance operator $S : K \rightarrow K$ of X is defined by

$$Su = \mathbb{E}[\langle X, u \rangle_K X].$$

S is a positive, self-adjoint trace-class operator that verifies the identity

$$\operatorname{Tr} S = \mathbb{E}[\|X\|_K^2].$$

We consider the Banach space of all trace-class operators on K , equipped with norm $\operatorname{Tr}|A|$, where $|A| = \sqrt{A^*A}$ and A^* denotes the adjoint of A . The subspace of Hilbert-Schmidt operators on K is denoted by $\operatorname{HS}(K)$, associated to the norm $\|A\|_{\operatorname{HS}(K)} = \sqrt{\operatorname{Tr}(AA^*)}$, $A \in \operatorname{HS}(K)$.

Now, assume that X is a K -valued random variable which belongs to the first Wiener chaos with finite fourth moment, i.e. $\mathbb{E}[\|X\|_K^4] < \infty$, and with covariance operator S . We denote by Z a Gaussian process taking values in the same separable Hilbert space as X and having the same covariance operator S . The following result holds.

Theorem A.4 (Functional Quantitative Fourth Moment Theorem, [6, Theorem 3 and Corollary 1]). *Under the above notation and assumptions, it holds that*

$$d_3(X, Z) \leq \left(\frac{1}{4} + \frac{1}{2} \sqrt{\mathbb{E}[\|X\|_K^2]} \right) \sqrt{\mathbb{E}[\|X\|_K^4] - \mathbb{E}[\|X\|_K^2]^2 - 2\|S\|_{\operatorname{HS}(K)}^2}.$$

A.2 Wigner's and Clebsch-Gordan coefficients

In this section we review briefly some background facts and notation about Wigner's 3j and Clebsch-Gordan coefficients; see [42] and [28] for a much more detailed discussion, in particular concerning the relationships with the quantum theory of angular momentum and group representation properties of $SO(3)$.

We start recalling the following analytic expression for the Wigner's 3j coefficients (valid for $m_1 + m_2 + m_3 = 0$, see [42], eq. (8.2.1.5))

$$\begin{aligned} \begin{pmatrix} \ell_1 & \ell_2 & \ell_3 \\ m_1 & m_2 & m_3 \end{pmatrix} &:= (-1)^{\ell_1+m_1} \sqrt{2\ell_3+1} \left[\frac{(\ell_1+\ell_2-\ell_3)!(\ell_1-\ell_2+\ell_3)!(\ell_1-\ell_2+\ell_3)!}{(\ell_1+\ell_2+\ell_3+1)!} \right]^{1/2} \\ &\times \left[\frac{(\ell_3+m_3)!(\ell_3-m_3)!}{(\ell_1+m_1)!(\ell_1-m_1)!(\ell_2+m_2)!(\ell_2-m_2)!} \right]^{1/2} \\ &\times \sum_z \frac{(-1)^z (\ell_2+\ell_3+m_1-z)!(\ell_1-m_1+z)!}{z!(\ell_2+\ell_3-\ell_1-z)!(\ell_3+m_3-z)!(\ell_1-\ell_2-m_3+z)!}, \end{aligned}$$

where the summation runs over all z 's such that the factorials are non-negative, and $\ell_1, \ell_2, \ell_3, m_1, m_2, m_3 \in \mathbb{Z}$. This expression becomes simpler when $m_1 = m_2 = m_3 = 0$, where we have

$$\begin{pmatrix} \ell_1 & \ell_2 & \ell_3 \\ 0 & 0 & 0 \end{pmatrix} = \begin{cases} 0, & \text{for } \ell_1 + \ell_2 + \ell_3 \text{ odd} \\ \frac{(-1)^{\frac{\ell_1+\ell_2-\ell_3}{2}} [(\ell_1+\ell_2+\ell_3)/2]!}{[(\ell_1+\ell_2-\ell_3)/2]![(\ell_1-\ell_2+\ell_3)/2]![(-\ell_1+\ell_2+\ell_3)/2]!} \left\{ \frac{(\ell_1+\ell_2-\ell_3)!(\ell_1-\ell_2+\ell_3)!(-\ell_1+\ell_2+\ell_3)!}{(\ell_1+\ell_2+\ell_3+1)!} \right\}^{1/2}, & \text{for } \ell_1 + \ell_2 + \ell_3 \text{ even.} \end{cases} \quad (\text{A.2})$$

On the other hand the so-called Clebsch-Gordan coefficients, denoted by $\{C_{\ell_1, m_1; \ell_2, m_2}^{\ell_3, m_3}\}$, are defined by the identities (see [42], Chapter 8)

$$\begin{pmatrix} \ell_1 & \ell_2 & \ell_3 \\ m_1 & m_2 & m_3 \end{pmatrix} = (-1)^{\ell_3+m_3} \frac{1}{\sqrt{2\ell_3+1}} C_{\ell_1, -m_1; \ell_2, -m_2}^{\ell_3, m_3} \quad (\text{A.3})$$

$$C_{\ell_1, m_1; \ell_2, m_2}^{\ell_3, m_3} = (-1)^{\ell_1-\ell_2+m_3} \sqrt{2\ell_3+1} \begin{pmatrix} \ell_1 & \ell_2 & \ell_3 \\ m_1 & m_2 & -m_3 \end{pmatrix}. \quad (\text{A.4})$$

Note that the Clebsch-Gordan coefficients vanish unless the so-called triangular conditions

$$|\ell_1 - \ell_2| \leq \ell_3 \leq \ell_1 + \ell_2, \quad \text{and the equation} \quad m_1 + m_2 = m_3$$

are satisfied (see [42], Chapter 8).

The following orthonormality properties hold and are exploited in this paper:

$$\sum_{m_1=-\ell_1}^{\ell_1} \sum_{m_2=-\ell_2}^{\ell_2} C_{\ell_1, m_1; \ell_2, m_2}^{\ell_3, m_3} C_{\ell_1, m_1; \ell_2, m_2}^{\ell'_3, m'_3} = \delta_{\ell'_3}^{\ell_3} \delta_{m'_3}^{m_3}, \quad (\text{A.5})$$

$$\sum_{\ell_3=|\ell_1-\ell_2|}^{\ell_1+\ell_2} \sum_{m_3=-\ell_3}^{\ell_3} C_{\ell_1, m_1; \ell_2, m_2}^{\ell_3, m_3} C_{\ell_1, m'_1; \ell_2, m'_2}^{\ell_3, m_3} = \delta_{m'_1}^{m_1} \delta_{m'_2}^{m_2}. \quad (\text{A.6})$$

For some special values of the arguments, namely if $\ell_3 = 0$ or $\ell_2 = 0$, one has more explicit forms of these coefficients:

$$C_{\ell_1, m_1; \ell_2, m_2}^{0,0} = (-1)^{\ell_1-m_1} \frac{\delta_{\ell_1}^{\ell_2} \delta_{m_1}^{-m_2}}{\sqrt{2\ell_1+1}}. \quad (\text{A.7})$$

From Remark 8.9 in [28] we also have the inequality

$$\left(C_{\ell,0;\ell,0}^{L,0}\right)^2 \leq \frac{1}{(2L+1)}. \quad (\text{A.8})$$

Now we recall the general formula ([42], eqs. 5.6.2.12-13, or [28] eqs 3.64 and 6.46) for the evaluation of multiple integrals of spherical harmonics, the so-called *Gaunt integrals*, given by

$$\begin{aligned} & \int_{\mathbb{S}^2} Y_{\ell_1 m_1}(x) \cdots Y_{\ell_n m_n}(x) dx \\ &= \sqrt{\frac{4\pi}{2\ell_n+1}} \sum_{L_1=|\ell_1-\ell_2|}^{\ell_1+\ell_2} \sum_{L_2=|L_1-\ell_3|}^{L_1+\ell_3} \cdots \sum_{L_{n-3}=|L_{n-4}-\ell_{n-2}|}^{L_{n-4}+\ell_{n-2}} \sum_{M_1=-L_1}^{L_1} \sum_{M_2=-L_2}^{L_2} \cdots \sum_{M_{n-3}=-L_{n-3}}^{L_{n-3}} \\ & \times \left[C_{\ell_1, m_1; \ell_2, m_2}^{L_1, M_1} C_{L_1, M_1; \ell_3, m_3}^{L_2, M_2} \cdots C_{L_{n-3}, M_{n-3}; \ell_{n-1}, m_{n-1}}^{\ell_n, -m_n} \right. \\ & \left. \times \sqrt{\frac{\prod_{i=1}^{n-1} (2\ell_i+1)}{(4\pi)^{n-1}}} \{C_{\ell_1, 0; \ell_2, 0}^{L_1, 0} C_{L_1, 0; \ell_3, 0}^{L_2, 0} \cdots C_{L_{n-3}, 0; \ell_{n-1}, 0}^{\ell_n, 0}\} \right]. \end{aligned}$$

The most important case dealt with in this paper is given by

$$\begin{aligned} & \int_{\mathbb{S}^2} Y_{\ell m_1}(x) Y_{\ell m_2}(x) Y_{\ell m_3}(x) Y_{\ell m_4}(x) dx \\ &= \frac{2\ell+1}{4\pi} \sum_{L=0}^{2\ell} \sum_{M=-L}^L C_{\ell, 0; \ell, 0}^{L, 0} C_{L, 0; \ell, 0}^{\ell, 0} C_{\ell, m_1; \ell, m_2}^{L, M} C_{L, M; \ell, m_3}^{\ell, -m_4} \\ &= \frac{(2\ell+1)^2}{4\pi} \sum_{L=0}^{2\ell} \sum_{M=-L}^L (-1)^{L-M} \{C_{\ell, 0; \ell, 0}^{L, 0}\}^2 \frac{C_{\ell, m_1; \ell, m_2}^{L, M} C_{\ell, m_3; \ell, m_4}^{L, -M}}{2L+1}. \quad (\text{A.9}) \end{aligned}$$

Finally, similarly, as shown in [32, Eq. (30)], the following identity hold

$$\begin{aligned} \int_0^1 P_l(t)^4 dt &= \frac{1}{2\ell+1} \sum_{L=0}^{2\ell} \{C_{\ell, 0; \ell, 0}^{L, 0} C_{L, 0; \ell, 0}^{\ell, 0}\}^2 \\ &= \sum_{L=0}^{2\ell} (2L+1) \begin{pmatrix} l & l & L \\ 0 & 0 & 0 \end{pmatrix}^4, \end{aligned}$$

see [42, Eq. (8.9.4.20)].

References

- [1] Baldi, P., Kerkycharian, G., Marinucci, D. and Picard, D. (2009). Asymptotics for spherical needlets. *Ann. Statist.*, 37, no. 3, 1150–1171. MR2509070
- [2] Berlinet, A. and Thomas-Agnan, C. (2004). *Reproducing Kernel Hilbert Spaces in Probability and Statistics*. Kluwer. MR2239907
- [3] Berry, M. V. (1977). Regular and irregular semiclassical wavefunctions. *J. Phys. A*, 10, no. 12, 2083–2091. MR0489542
- [4] Berry, M. V. (2002). Statistics of nodal lines and points in chaotic quantum billiards: perimeter corrections, fluctuations, curvature. *J. Phys. A*, 35, 3025–3038. MR1913853
- [5] Bourguin, S. and Campese, S. (2020). Approximation of Hilbert-valued Gaussians on Dirichlet structures. *Electron. J. Probab.*, 25, 1–30. MR4193891
- [6] Bourguin, S., Campese, S. and Dang, T. (2021). Functional Gaussian approximations on Hilbert-Poisson spaces. arXiv:2110.04877v2.

- [7] Bourguin, S. and Dang, T. (2022). High dimensional regimes of non-stationary Gaussian correlated Wishart matrices. *Random Matrices Theory Appl.*, 1, Paper No. 2250006, 43 pp. MR4391829
- [8] Bourguin, S., Diez, C. P. and Tudor, C. A. (2021). Limiting behavior of large correlated Wishart matrices with chaotic entries. *Bernoulli*, 27, no. 2, 1077–1102. MR4255227
- [9] Cammarota, V. (2019). Nodal area distribution for arithmetic random waves. *Trans. Amer. Math. Soc.*, 372, no. 5, 353. MR3988618
- [10] Cammarota, V., Klurman, O. and Wigman, I. (2020). Boundary effect on the nodal length for arithmetic random waves, and spectral semi-correlations. *Comm. Math. Phys.*, 376, no. 2, 1261–1310. MR4103968
- [11] Cammarota, V., Marinucci, D. and Wigman, I. (2016). On the distribution of the critical values of random spherical harmonics. *J. Geom. Anal.*, 4, 3252–3324. MR3544960
- [12] Cammarota, V. and Marinucci, D., (2018). A quantitative central limit theorem for the Euler-Poincaré Characteristic of random spherical eigenfunctions. *Ann. Probab.*, 46, no. 6, 3188–3228. MR3857854
- [13] Cammarota, V. and Marinucci, D. (2020). A reduction principle for the critical values of random spherical harmonics. *Stoch. Process. Their Appl.*, 130, no. 4, 2433–2470. MR4074706
- [14] Cammarota, V. and Wigman, I. (2017). Fluctuations of the total number of critical points of random spherical harmonics. *Stoch. Process. Their Appl.*, 127, no. 12, 3825–3869. MR3718098
- [15] Canzani, Y. and Hanin, B. (2020). Local universality for zeros and critical points of monochromatic random waves, *Comm. Math. Phys.*, 378, no. 3, 1677–1712. MR4150887
- [16] Dalmao, F., Estrade, A. and León, J. (2021). On 3-dimensional Berry’s model. *Lat. Am. J. Probab. Math. Stat.*, 18, 379–399. MR4213863
- [17] Dierickx, G., Nourdin, I., Peccati, G. and Rossi, M. (2023). Small scale CLTs for the nodal length of monochromatic waves. *Comm. Math. Phys.*, 397, 1–36. MR4538280
- [18] Döbler, C. and Peccati, G. (2018). The fourth moment theorem on the Poisson space. *Ann. Probab.*, 46, no. 4, 1878–1916. MR3813981
- [19] Döbler, C., Vidotto, A. and Zheng, G. (2018). Fourth moment theorems on the Poisson space in any dimension. *Electron. J. Probab.*, 23, 1–27. MR3798246
- [20] Durastanti, C., Marinucci, D. and Peccati, G. (2014). Normal approximations for wavelet coefficients on spherical Poisson fields. *J. Math. Anal. Appl.*, 409, no. 1, 212–227. MR3095032
- [21] Eichelsbacher, P. and Thäle, C. (2014). New Berry-Essen bounds for non-linear functionals of Poisson random measures. *Electron. J. Probab.*, 19, article 102, 1–25. MR3275854
- [22] Krishnapur, M., Kurlberg, P. and Wigman, I. (2013). Nodal length fluctuations for arithmetic random waves. *Ann. Math.*, 177, no. 2, 699–737. MR3010810
- [23] Lachièze-Rey, R., Schulte, M. and Yukich, J. E. (2019). Normal approximation for stabilizing functionals, *Ann. Appl. Probab.*, 29, no. 2, 931–993. MR3910021
- [24] Last, G., Peccati, G. and Schulte, M. (2016). Normal approximation on Poisson spaces: Mehler’s formula, second order Poincaré inequalities and stabilization, *Probab. Theory Relat. Fields* 165, no. 3-4, 667–723. MR3520016
- [25] Last, G. and Penrose, M. (2017). *Lectures on the Poisson Process*, Cambridge University Press. MR3791470
- [26] Maffucci, R. W. (2019). Nodal intersections for arithmetic random waves against a surface. *Ann. Henri Poincaré*, 20, no. 11, 3651–3691. MR4019200
- [27] Marinucci, D. and Peccati, G. (2013). Mean-square continuity on homogeneous spaces of compact groups. *Electron. Commun. Probab.*, 18, no. 37, 10. MR3064996
- [28] Marinucci, D. and Peccati, G. (2011). *Random Fields on the Sphere: Representations, Limit Theorems and Cosmological Applications*, London Mathematical Society Lecture Notes, Cambridge University Press. MR2840154
- [29] Marinucci, D., Peccati, G., Rossi, M. and Wigman, I. (2016). Non-universality of nodal length distribution for arithmetic random waves. *Geom. Funct. Anal.*, 26, 926–960. MR3540457

- [30] Marinucci, D., Rossi, M. and Wigman, I. (2020). The asymptotic equivalence of the sample trispectrum and the nodal length for random spherical harmonics. *Ann. Inst. Henri Poincaré Probab. Stat.*, 56, no. 1, 374–390. MR4058991
- [31] Marinucci, D. and Wigman, I. (2014). On nonlinear functionals of random spherical eigenfunctions. *Commun. Math. Phys.*, 327, no. 3, 849–872. MR3192051
- [32] Marinucci, D. and Wigman, I. (2011). On the excursion sets of spherical Gaussian eigenfunctions. *J. Math. Phys.*, 52, 093301. MR2867816
- [33] Narcowich, F. J., Petrushev, P. and Ward, J. D. (2006). Localized tight frames on spheres. *SIAM J. Math. Anal.*, 38, no. 2, 574–594. MR2237162
- [34] Nazarov, F. and Sodin, M. (2009). On the number of nodal domains of random spherical harmonics. *Am. J. Math.*, 131, no. 5, 1337–1357. MR2555843
- [35] Nourdin, I. and Peccati, G. (2012). *Normal Approximations Using Malliavin Calculus: From Stein’s Method to Universality*, Cambridge University Press. MR2962301
- [36] Nourdin, I., Peccati, G. and Rossi, M. (2019). Nodal statistics of planar random waves, *Comm. Math. Phys.*, 369, no. 1, 99–151. MR3959555
- [37] Peccati, G. and Zheng, C. (2010). Multi-dimensional Gaussian fluctuations on the Poisson space. *Electron. J. Probab.*, 15, 1487–1527. MR2727319
- [38] Rudnick, Z. and Wigman, I. (2008). On the volume of nodal sets for eigenfunctions of the Laplacian on the torus. *Ann. Henri Poincaré*, 9, no. 1, 109–130. MR2389892
- [39] Sarnak, P. and Wigman, I. (2019). Topologies of nodal sets of random band-limited functions. *Comm. Pure Appl. Math.* 72, no. 2, 275–342. MR3896022
- [40] Szegő, G. (1975). *Orthogonal Polynomials*, American Mathematical Society, Colloquium Publications, Vol. XXIII. MR0372517
- [41] Todino, A. P. (2020). Nodal lengths in shrinking domains for random eigenfunctions on S^2 . *Bernoulli* 26, no. 4, 3081–3110. MR4140538
- [42] Varshalovich, D. A., Moskalev, A. N. and Khersonskii, V. K. (1988). *Quantum Theory of Angular Momentum. Irreducible Tensors, Spherical Harmonics, Vector Coupling Coefficients, 3nj Symbols*. Translated from the Russian. World Scientific Publishing Co., Inc., Teaneck, NJ. MR1022665
- [43] Vidotto, A. (2021). A note on the reduction principle for the nodal length of planar random waves. *Statist. Probab. Lett.*, 174, 109090. MR4237481
- [44] Wigman, I. (2010). Fluctuations of the nodal length of random spherical harmonics. *Comm. Math. Phys.*, 298, no. 3, 787–831. MR2670928
- [45] Wigman, I. (2012). On the nodal lines of random and deterministic Laplace eigenfunctions, in: *Spectral Geometry*, Proc. Sympos. Pure Math., 84, Amer. Math. Soc., Providence, RI, 285–297. MR2985322

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