Equidistribution of Random Waves on Small Balls

Xiaolong Han¹, Melissa Tacy²

¹ Department of Mathematics, California State University, Northridge, CA 91330, USA. E-mail: xiaolong.han@csun.edu

² Department of Mathematics and Statistics, University of Otago, Dunedin, Otago 9016, New Zealand. E-mail: mtacy@maths.otago.ac.nz

Received: 14 May 2019 / Accepted: 5 October 2019 Published online: 11 November 2019 – © Springer-Verlag GmbH Germany, part of Springer Nature 2019

Abstract: In this paper, we investigate the small scale equidistribution properties of randomised sums of Laplacian eigenfunctions (i.e. random waves) on a compact manifold. We prove small scale expectation and variance results for random waves on all compact manifolds. Here, "small scale" refers to balls of radius $r(\lambda) \rightarrow 0$ such that $r/r_{\text{Planck}} \rightarrow \infty$, where r_{Planck} is the Planck scale. For balls at a larger scale (although still $r(\lambda) \rightarrow 0$) we also obtain estimates showing that the probability that a random wave fails to equidistribute decays exponentially with the eigenvalue.

1. Introduction

Studying the behaviour of random combinations of either plane waves or eigenfunctions has lately proved to be an exciting research area. It is conjectured, by Berry [B] in the 1970s, that eigenfunctions of chaotic systems such as planar domains with chaotic billiard flow behave like random waves. That is, their behaviour is modelled by functions of the form

$$\sum_{j} a_{j} e^{i\lambda \langle x, \xi_{j} \rangle}, \tag{1.1}$$

where the $\{\xi_j\}$ are chosen as a set of equidistributed (at scale λ^{-1}) directions on the (n - 1)-dim unit sphere \mathbb{S}^{n-1} and the coefficients a_j are chosen randomly, e.g. independent Gaussian random variables.

In the setting of manifolds where the underlying geodesic flow displays chaotic properties equidistribution for Laplacian eigenfunctions has been studied by Shnirelman [Sn], Zelditch [Z1], Colin de Verdière [CdV]. In particular, on manifolds with ergodic geodesic flow such as negatively curved manifolds, there is a full density sequence of eigenfunctions in any eigenbasis that equidistribute. Recently a number of small scale equidistribution results were established in various settings and at various scales by Han [Ha1,Ha2], Hezari–Rivière [HR1,HR2], Lester–Rudnick [LR], Chang–Zelditch [CZ].



Communications in

Mathematical Physics Notably on manifolds with negative curvature, equidistribution at logarithmic scales (i.e. $r = (\log \lambda)^{-\alpha}$ for some $\alpha > 0$) is established in [Ha1] and [HR1]. See also the application of small scale equidistribution to other eigenfunction problems in Hezari [He1,He2,He3], Sogge [So1,So2], Zelditch [Z5], etc.

As in the Euclidean case, it is interesting to study the model case of random behaviour on manifolds. On a compact manifold \mathbb{M} , the natural class of objects that replace plane waves $e^{i\lambda(x,\xi_j)}$ are eigenfunctions. That is, we consider sums

$$\sum_{\lambda_j \in \Lambda} a_j e_j(x),$$

where $\Lambda \subset [0, \infty)$, the e_j 's are orthonormal Laplacian eigenfunctions on \mathbb{M} with eigenvalues λ_j^2 , and the coefficients a_j are prescribed in a random fashion.

The obvious first question is how to pick the set Λ . Notice that, because $\xi_j \in \mathbb{S}^{n-1}$, the plane waves $e^{i\lambda(x,\xi_j)}$ are generalised eigenfunctions of the Laplacian with eigenvalues λ^2 . So on compact manfolds, initially it may seem natural to fix an eigenspace E_{λ} with eigenvalue λ^2 and randomise only over the eigenfunctions in E_{λ} , as is done in [Ha2] on manifolds including tori and spheres. See also de Courcy-Ireland [CI1] on spheres. However, the multiplicity of this eigenvalue may be low. In fact, in chaotic cases such as when \mathbb{M} has negative curvature it is conjectured that the eigenvalues have very low multiplicity. Indeed, on manifolds with generic metric, the eigenvalues are simple due to Uhlenbeck [U]. Therefore, to capture the random behaviour, we allow ourselves to randomise over eigenfunctions whose eigenvalues sit in a spectral window. Such randomisations were introduced in Zelditch [Z3]. That is, we set $\Lambda = [\lambda - W, \lambda]$ for $1 \leq W \leq \lambda$ and consider the functions

$$u = \sum_{\lambda_j \in [\lambda - W, \lambda]} a_j e_j(x).$$

We point out that the spectral window width W here is allowed to depend on λ . These randomised functions are commonly referred as "random waves". We adopt this terminology and reserve the term "random eigenfunctions" for those combinations taken over a single eigenspace. In Zelditch [Z3], the two special cases of W = 1 and $W = \lambda$ are called the asymptotically fixed frequency ensembles and the cut-off ensembles, respectively. Both of these ensembles are included in our analysis here.

Having selected a window to randomise over, we must now consider how we pick our random variables a_j in the random wave $u = \sum a_j e_j$. We always normalize such that $\mathbb{E}\left(\|u\|_{L^2(\mathbb{M})}^2\right) = 1$. Some common choices of a_j include independent random variables such as Gaussian or Rademacher random variables with proper normalization ([B,CI1,CI2,Z3]) and uniform probability density on unit spheres ([BL,Ha2,M,Z2, Z4].) The coefficients a_j in the former randomisation procedure are independent so it is straightforward to compute some probabilistic estimates such as the covariance. The latter randomisation process is not independent however in Sect. 2.2 we show, in a rather elementary way, that these key probabilistic estimates are asymptotically the same as the independent case. Where the coefficients a_j are chosen so that $a = (a_1, \ldots, a_d)$ lies on the unit sphere \mathbb{S}^d the randomisation procedure admits a dual interpretation, namely

- we pick coefficients a_j at random from a probability density,
- we pick an L^2 normalised function at random in span $\lambda_j \in [\lambda W, \lambda] \{e_j\}$. (Note that here $||u||^2_{L^2(\mathbb{M})} = \sum |a_j|^2 = 1$ for all random waves, whereas one can only require $\mathbb{E}\left(||u||^2_{L^2(\mathbb{M})}\right) = 1$ in the case when a_j are chosen as Gaussian independent variables).

In addition, the Levy concentration of measure (Theorem 2.7) on the unit sphere serves as an important tool to study uniform equidistribution of random waves on the whole manifold (see Theorem 1.5.) Therefore, in this paper, we use the uniform probability density on unit spheres and ask about the expected behaviour of random waves as well as the variance in behaviour. In particular, we focus on small scale behaviour. We want to understand when random waves equidistribute on small balls.

There are two parts to understanding this equidistribution. The first is to ascertain when

$$\mathbb{E}\left(\int_{B(x,r)} |u|^2 \, d\text{Vol}\right) \to \frac{\text{Vol}(B(x,r))}{\text{Vol}(\mathbb{M})} \quad \text{as } \lambda \to \infty, \tag{1.2}$$

in which $B(x, r) \subset \mathbb{M}$ is a geodesic ball with center x and radius r. However, while the expectation value might equidistribute, it is still possible that the probability of nonequidistribution is high. To that end we also determine, for given $r = r(\lambda) \to 0$ as $\lambda \to \infty$, whether

$$\operatorname{Var}\left(\int_{B(x,r)} |u|^2 d\operatorname{Vol}\right) = o\left(\operatorname{Vol}(x,r)^2\right) \quad \text{as } \lambda \to \infty. \tag{1.3}$$

The variance estimate tells us whether we may expect that a typical sum of eigenfunctions equidistributes at scale *r* around $x \in \mathbb{M}$. If the scale *r* in (1.2) and (1.3) is independent of λ , i.e. equidistribution at fixed scales, the analysis follows from Zelditch [Z2,Z4] and Maples [M] in various random settings.

At small scales $r = r(\lambda)$ such that $r^{-1} = o(\lambda)$ as $\lambda \to \infty$, we are able to obtain sufficient conditions for (1.2) and (1.3). Notice that these scales are just above the Planck scale

$$r_{\text{Planck}} := \lambda^{-1}.$$

Precisely, the scale r satisfies that $r/r_{\text{Planck}} \rightarrow \infty$. See Theorems 1.2 and 1.3 below.

For a *fixed* $x \in \mathbb{M}$, (1.2) and (1.3) tell us that probability such that the random wave u does not equidistribute on B(x, r) decays as $\lambda \to \infty$. Finally, we consider small scale equidistribution of random waves *uniformly* on the manifold, i.e. uniformity across balls B(x, r) of radius r and for all $x \in \mathbb{M}$. See Theorem 1.5 below.

At this point, some details of our setup is in order. Let (\mathbb{M}, g) be an *n*-dim compact, smooth Riemannian manifold without boundary. Denote $\Delta = \Delta_g$ the (positive) Laplace-Beltrami operator. Let $\{e_j\}_{j=0}^{\infty}$ be an orthonormal basis of eigenfunctions (i.e. eigenbasis) of Δ with eigenvalues λ_j^2 (counting multiplicities). That is, $\Delta e_j = \lambda_j^2 e_j$, where λ_j is called the eigenfrequency. Denote Inj \mathbb{M} the injectivity radius of \mathbb{M} . We assume, without loss of generality, that Inj $\mathbb{M} \geq 1$.

We define the probability space for the random waves in a similar fashion as in Zelditch [Z3].

Definition 1.1 (Random waves). Let

$$\mathcal{H}_W(\lambda) = \operatorname{span}_{\lambda_j \in [\lambda - W, \lambda]} \{e_j\} \quad \text{and} \quad N_W(\lambda) = \dim \mathcal{H}_W(\lambda). \tag{1.4}$$

We assume that the eigenfunctions e_j are real-valued. We define the random wave $u_{\lambda,a} \in \mathcal{H}_W(\lambda)$ as

$$u_{\lambda,a} := \sum_{\lambda_j \in [\lambda - W, \lambda]} a_j e_j, \quad \text{for } a \in \mathbb{S}^{N_W(\lambda) - 1}.$$
(1.5)

Here, $\mathbb{S}^{N_W(\lambda)-1}$ is equipped with uniform probability measure $\mu_{N_W(\lambda)}$. That is, $u_{\lambda,a}$ is a sum of eigenfunctions in $\mathcal{H}_W(\lambda)$ with random coefficient $a \in \mathbb{S}^{N_W(\lambda)-1}$ so $||u_{\lambda,a}||_{L^2(\mathbb{M})} = 1$. For brevity, we also write u_{λ} as $u_{\lambda,a}$ with the understanding that a is the random variable.

Remark. One can similarly consider random waves as combinations of complex-valued Laplacian eigenfunctions, in which case the coefficient a in (1.5) is chosen randomly from the complex unit sphere. The analysis is similar so we omit details here.

Our first main theorem states that

Theorem 1.2. Let $1 \leq W \leq \lambda$ and $x \in \mathbb{M}$.

(i) For r > 0, the expected value with respect to the probability measure $\mu_{N_W(\lambda)}$

$$\mathbb{E}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \left[1 + O\left(W^{-1}\right)\right],\tag{1.6}$$

where the term $O(W^{-1})$ is independent of x.

(ii) For $r^{-1} = o(\lambda)$ as $\lambda \to \infty$, i.e. $r/r_{\text{Planck}} \to \infty$, we have that the variance with respect to the probability measure $\mu_{N_W(\lambda)}$

$$\operatorname{Var}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = \operatorname{Vol}(B(x,r))^2 \left[o(1) + O\left(W^{-2}\right)\right] \quad as \ \lambda \to \infty, \ (1.7)$$

where the terms o(1) and $O(W^{-2})$ are independent of x.

In particular, if the spectral window width $W = W(\lambda) \rightarrow \infty$ as $\lambda \rightarrow \infty$, then at all scales r such that $r/r_{\text{Planck}} \rightarrow \infty$,

$$\mathbb{E}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\text{Vol}\right) = \frac{\text{Vol}(B(x,r))}{\text{Vol}(\mathbb{M})} + o\left(r^n\right) \quad as \ \lambda \to \infty$$

and

$$\operatorname{Var}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = o\left(r^{2n}\right) \quad as \ \lambda \to \infty.$$

We remark that more precise estimates of the o(1) and $O(W^{-2})$ terms in (1.7) are given in the proof of Theorem 1.2.

If the spectral window width W is independent of λ , e.g. the case of the asymptotically fixed frequency ensemble when W = 1, then Theorem 1.2 does *not* imply small scale equidistribution results of the random waves u_{λ} . Instead, according to (1.6) and (1.7),

we can only conclude that the L^2 integral of u_{λ} on the ball B(x, r) is proportional to the normalised volume of the ball.

However, with a geometric condition on the manifold \mathbb{M} , we are able to recover small scale equidistribution results. The relevant condition is that, for all $x \in \mathbb{M}$ the set of the geodesic loop directions

$$\mathcal{L}_x := \{\xi \in S^*\mathbb{M} : G_t(x,\xi) = (x,\eta) \text{ for some } t > 0 \text{ and } \eta \in S^*_x\mathbb{M}\}$$

is of measure zero in $S_x^*\mathbb{M}$. Here, $S_x^*\mathbb{M}$ is the cosphere space of \mathbb{M} at x, $S^*\mathbb{M}$ is the cosphere bundle of \mathbb{M} and $G_t(x, \xi)$ is the geodesic flow on \mathbb{M} . This pointwise aperodic condition is called the non self-focal condition. Examples of manifolds satisfying the non self-focal condition include the negatively curved manifolds (i.e. all sectional curvatures are negative everywhere). Since manifolds with negative curvature are a key class of manifolds that we wish to understand using randomisation, making such an assumption is not as restrictive as may first appear.

The above non self-focal condition is a natural dynamical condition to study the precise behavior of eigenfunctions restricted to a fixed-width spectrum window. See Sect. 2.1 for the background.

Concerning the small scale equidistribution of random waves in fixed-width spectral windows, we prove that

Theorem 1.3 (Small scale equidistribution of random waves in fixed-width spectral windows). Suppose that $W \ge 1$ is independent of λ . Assume that, for all $x \in \mathbb{M}$ the set of loop directions \mathcal{L}_x is of measure zero in $S_x^*\mathbb{M}$.

(i) The expectation

$$\mathbb{E}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} + o\left(r^n\right) \quad as \; \lambda \to \infty,$$

where the term $o(r^n)$ is independent of x. (ii) If in addition $r^{-1} = o(\lambda)$, i.e. $r/r_{\text{Planck}} \to \infty$, then the variance

$$\operatorname{Var}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = o\left(r^{2n}\right) \quad as \ \lambda \to \infty,$$

where the term $o(r^{2n})$ is independent of x.

Remark. In particular, if we choose W = 1, then u_{λ} in Theorem 1.3 is the asymptotically fixed frequency ensemble considered in Zelditch [Z3]. In this case, Theorem 1.3 states that such ensembles are equidistributed at all scales r such that $r/r_{\text{Planck}} \rightarrow \infty$, on manifolds that satisfy the non self-focal condition.

Theorems 1.2 and 1.3 give sufficient conditions such that a typical random wave $u_{\lambda,a}$ equidistributes on the ball B(x, r) for $x \in \mathbb{M}$ and $r = r(\lambda)$ such that $r/r_{\text{Planck}} \to \infty$. That is, the probability (i.e. measure in the probability space $\mathbb{S}^{N_W(\lambda)-1}$) that $u_{\lambda,a}$ does not equidistribute on B(x, r) decays in λ .

However, neither theorem provides a quantitative estimate of the decay of such probability, nor do they conclude equidistribution of $u_{\lambda,a}$ on the whole manifold, i.e. on balls B(x, r) for all $x \in \mathbb{M}$. We now address these two problems and provide a quantitative estimate of the probability for small scale equidistribution of random waves $u_{\lambda,a}$ uniformly on \mathbb{M} . We measure the deviation from equidistribution with an o(1) order function. That is, we say $m(\lambda)$ is an o(1) order function if $m(\lambda) : \mathbb{R}^+ \to \mathbb{R}^+$ and $m(\lambda) \to 0$ as $\lambda \to \infty$. For the random variable $a \in \mathbb{S}^{N_W(\lambda)-1}$, we say that $u_{\lambda,a}$ equidistributes at scale $r = r(\lambda)$ to order $m(\lambda)$ uniformly on \mathbb{M} if

$$\sup_{x \in \mathbb{M}} \left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \right| \le r^n m(\lambda). \tag{1.8}$$

Therefore, the introduction of $m(\lambda)$ quantitatively characterizes the reminder term in the equidistribution statements. In general, to make the remainder smaller (and equidistribution better), we need to pick larger scale balls.

We define the exceptional set $\tilde{S}_r(m)$ as the set of point in $\mathbb{S}^{N_W(\lambda)-1}$ where (1.8) fails.

Definition 1.4. The exceptional set $S_r(m)$ is given by

$$S_{r}(m) = \left\{ a \in \mathbb{S}^{N_{W}(\lambda)-1} : \exists x \in \mathbb{M} \text{ such that } \left| \int_{B(x,r)} |u_{\lambda,a}|^{2} d\text{Vol} - \frac{\text{Vol}(B(x,r))}{\text{Vol}(\mathbb{M})} \right| \ge r^{n} m(\lambda) \right\}.$$

In the following theorem, we prove that the exceptional set $S_r(m)$ has exponentially small measure in $\mathbb{S}^{N_W(\lambda)-1}$ at certain scales that are larger than the ones in Theorems 1.2 and 1.3.

Theorem 1.5 (Uniform equidistribution of random waves at small scales). There exist constants $c, \alpha, K > 0$ depending only on \mathbb{M} such that the following statements are true.

(i) Suppose $W = W(\lambda)$ such that $1 \le W \le \lambda$ and $W \to \infty$ as $\lambda \to \infty$. Set

$$r_1=W^{-\frac{1}{2n}}\lambda^{-\frac{n-1}{2n}}.$$

Let $m(\lambda) \ge K W^{-1}$ and $r = r(\lambda)$ such that

$$\max\left\{W^{-1}, \alpha \log(\lambda)^{\frac{1}{2n}} r_1 m(\lambda)^{-\frac{1}{n}}\right\} \le r \le \operatorname{Inj} \mathbb{M}.$$

Then

$$\mu_{N_W(\lambda)}(S_r(m)) \le \exp\left(-\frac{cr^{2n}m(\lambda)^2}{r_1^{2n}}\right).$$

(ii) Suppose $W = W(\lambda)$ such that $1 \le W \le \lambda$ and $W \to \infty$ as $\lambda \to \infty$. Set

$$r_2 = W^{\frac{1}{2(n-1)}} \lambda^{-\frac{1}{2}}.$$

Let $m(\lambda) \ge K W^{-1}$ and $r = r(\lambda)$ such that

$$\alpha r_2 \log(\lambda)^{\frac{1}{2(n-1)}} m(\lambda)^{-\frac{1}{n-1}} \leq r \leq W^{-1}.$$

Then

$$\mu_{N_W(\lambda)}(S_r(m)) \le \exp\left(-\frac{cr^{2(n-1)}m(\lambda)^2}{r_2^{2(n-1)}}\right).$$

(iii) Assume that, for all $x \in \mathbb{M}$ the set of loop directions \mathcal{L}_x is of measure zero in $S_x^*\mathbb{M}$. Suppose that W > 0 is independent of λ . Then there exists an o(1) order function $m(\lambda)$ such that if

$$r \geq \alpha \log(\lambda)^{\frac{1}{2(n-1)}} \lambda^{-1/2} m(\lambda)^{-\frac{1}{n-1}},$$

then

$$\mu_{N_W(\lambda)}(S_r(m)) \le \exp\left(-c\lambda^{n-1}r^{2(n-1)}m(\lambda)^2\right).$$

Remark. The generality of Thoerem 1.5 can make it difficult to parse. In the special cases of the cut-off and asymptotically fixed ensembles, the results can be stated in a simpler fashion. That is, Theorem 1.5 concludes uniform equidistribution of $u_{\lambda,a}$ at scales approaching $\lambda^{-1/2}$ except an exponentially small set $S \subset \mathbb{S}^{N_W(\lambda)-1}$. More precisely,

- (1) If $W = \lambda$, then $u_{\lambda,a}$ is the cut-off ensemble. By Case (i) of Theorem 1.5, with $m(\lambda) = (\log \lambda)^{-\varepsilon}$ for $\varepsilon > 0$ arbitrarily small, we have equidistribution up to scales r such that $r \ge \alpha \log(\lambda)^{\frac{1+2\varepsilon}{2n}} \lambda^{-1/2}$.
- (2) If W = 1, then $u_{\lambda,a}$ is the fixed frequency ensemble. Assume further the loopset condition as in Case (iii) of Theorem 1.5. Then for some o(1) order function $m(\lambda)$, we have equidistribution up to scales r such that $r \ge \alpha \log(\lambda)^{\frac{1}{2n}} m(\lambda)^{-\frac{1}{n-1}} \lambda^{-1/2}$.

Throughout this paper, $A \leq B$ ($A \geq B$) means $A \leq cB$ ($A \geq cB$) for some constant c depending only on the manifold; $A \approx B$ means $A \leq B$ and $B \leq A$; the constants c and C may vary from line to line.

2. Preliminaries

A key technique in the study of randomisations of eigenfunctions is to reduce questions about the expectation or variance of a random variable to problems involving the spectral projection operator (see Proposition 3.1). On a compact manifold \mathbb{M} , let $\{e_j\}_{j=0}^{\infty}$ be a real-valued eigenbasis of the Laplacian Δ with eigenvalues λ_j^2 . Then the spectral projection operator onto the space

span
$$\lambda_i \in [0,\lambda] \{e_j\}$$

has the kernel

$$E_{[0,\lambda]}(x, y) = \sum_{\lambda_j \in [0,\lambda]} e_j(x) e_j(y).$$

The highest order asymptotics of the kernel of $E_{[0,\lambda]}$ are well understood and there are a number of estimates linking the geometry of \mathbb{M} to the behaviour of lower order terms. (See Theorems 2.1 and 2.4.) In Sect. 2.1, we recall these spectral estimates of the Laplacian and their connection to underlying geometry.

In Sect. 2.2, we discuss some probabilistic estimates including the Levy concentration of measure (Theorem 2.7) from probability theory. It is this concentration of measure that we use to prove uniform equidistribution of random waves at small scales in Theorem 1.5.

2.1. Spectral estimates. Let $T^*\mathbb{M} = \{(x,\xi) : x \in \mathbb{M}, \xi \in T^*_x\mathbb{M}\}$ be the cotangent bundle of \mathbb{M} and $|\cdot|_x$ be the induced metric on the cotangent space $T^*_x\mathbb{M}$. We denote \exp_y the exponential map at $y \in \mathbb{M}$. Since $\operatorname{Inj} \mathbb{M} \ge 1$, $\exp_y(x)$ is diffeomophism if d(x, y) is small enough. Here, d denotes the Riemannian distance on \mathbb{M} . The next theorem from Hörmander [Ho, Theorem 4.4] provides the estimates of the kernel $E_{[0,\lambda]}$.

Theorem 2.1 (Spectral projection kernel estimates). *There is constant* d_0 *depending only on* \mathbb{M} *such that if* $d(x, y) < d_0$, *then*

$$E_{[0,\lambda]}(x, y) = \frac{1}{(2\pi)^n} \int_{|\xi|_{g_y} \le \lambda} e^{i \langle \exp_y^{-1}(x), \xi \rangle} \frac{d\xi}{\sqrt{|g_y|}} + R(x, y, \lambda),$$

where $R(x, y, \lambda) = O(\lambda^{n-1})$ as $\lambda \to \infty$ uniformly for $x, y \in \mathbb{M}$ such that $d(x, y) < d_0$.

Letting x = y in the above theorem, we immediately get the pointwise Weyl asymptotics as well as the Weyl asymptotics for the distribution of eigenvalues.

Corollary 2.2. (Pointwise Weyl asymptotics) We have that

$$E_{[0,\lambda]}(x,x) = \sum_{\lambda_j \le \lambda} |e_j(x)|^2 = c_n \lambda^n + R(\lambda, x), \quad \text{where } R(\lambda, x) = O(\lambda^{n-1}) \text{ as } \lambda \to \infty$$
(2.1)

uniformly for all $x \in \mathbb{M}$. Here, c_n is the volume of the unit ball in \mathbb{R}^n . Moreover, let $N(\lambda) := \#\{j : \lambda_j \leq \lambda\}$. Then

$$N(\lambda) = c_n \operatorname{Vol}(\mathbb{M})\lambda^n + R(\lambda), \quad \text{where } R(\lambda) = O(\lambda^{n-1}) \text{ as } \lambda \to \infty.$$
(2.2)

The remainder term estimate $R(\lambda, x) = O(\lambda^{n-1})$ in (2.1) is sharp on the sphere \mathbb{S}^n . The sharp growth rate λ^{n-1} is achieved at the poles of zonal harmonics on \mathbb{S}^n . See Hörmander [Ho, Section 6].

However, on some manifolds other than the sphere, the above estimates of $R(\lambda, x)$ and $R(\lambda)$ may be improved. Such improvements are related to the dynamical properties of the geodesic flow on \mathbb{M} . The geodesic flow G_t is the Hamiltonian flow with Hamiltonian defined on $T^*\mathbb{M}$ as $H(x, \xi) = |\xi|_x^2$. The geodesic flow G_t preserves the Liouville measure on $T^*\mathbb{M}$. Denote the cosphere bundle $S^*\mathbb{M} = \{(x, \xi) \in T^*\mathbb{M} : |\xi|_x = 1\}$. Then G_t acts on $S^*\mathbb{M}$ by homogeneity and leaves the induced Liouville measure on $S^*\mathbb{M}$ invariant.

Denote the set of periodic geodesics on $S^*\mathbb{M}$ as

$$\Pi = \{ (x, \xi) \in S^* \mathbb{M} : G_t(x, \xi) = (x, \xi) \text{ for some } t > 0 \}.$$

Duistermaat-Guillemin [DG, Theorem 3.5] proved that

Theorem 2.3 (Improved Weyl asymptotics). Assume that the set of periodic geodesics Π is of Liouville measure zero in S^{*}M. Then

$$N(\lambda) = c_n \operatorname{Vol}(\mathbb{M})\lambda^n + R(\lambda), \quad \text{where } R(\lambda) = o(\lambda^{n-1}) \text{ as } \lambda \to \infty.$$
(2.3)

To improve the pointwise Weyl law, we need a pointwise dynamical condition on the geodesics that is similar to the one in Theorem 2.3. A geodesic loop through x is a geodesic L(t) parametrized by arclength so that for some $t_0 > 0$ such that $L(0) = L(t_0) = x$. Define the loop directions at x as

$$\mathcal{L}_x := \{ \xi \in S^* \mathbb{M} : G_t(x, \xi) = (x, \eta) \text{ for some } t > 0 \text{ and } \eta \in S^*_x \mathbb{M} \}.$$

Canzani-Hanin [CH, Theorem 2] proved that

Theorem 2.4 (Improved spectral projection estimate). Assume that \mathcal{L}_x is of measure zero in $S_x^* \mathbb{M}$ for all $x \in \mathbb{M}$. Then

$$E_{[0,\lambda]}(x, y) = \frac{1}{(2\pi)^n} \int_{|\xi|_{g_y} < \lambda} e^{i \langle \exp_y^{-1}(x), \xi \rangle} \frac{d\xi}{\sqrt{|g_y|}} + R(x, y, \lambda),$$

where $R(x, y, \lambda) = o(\lambda^{n-1})$ uniformly for all $x, y \in \mathbb{M}$. In particular, the pointwise Weyl asymptotic asserts that

$$E_{[0,\lambda]}(x,x) = \sum_{\lambda_j \le \lambda} |e_j(x)|^2 = c_n \lambda^n + R(\lambda, x), \quad \text{where } R(\lambda, x) = o(\lambda^{n-1}) \text{ as } \lambda \to \infty$$
(2.4)

uniformly for all $x \in \mathbb{M}$.

- *Remark* .(1). If \mathcal{L}_x is of measure zero on $S_x^*\mathbb{M}$ for all $x \in \mathbb{M}$, then the set of periodic geodesics Π is of Liouville measure zero on $S^*\mathbb{M}$. Hence, one has that $R(\lambda) = o(\lambda^{n-1})$ as $\lambda \to \infty$ as an immediate corollary of Theorem 2.4 (one can also instead integrate (2.4) on \mathbb{M} directly).
- (2). There is a long history of works investigating the relationship between the geometric condition of the manifold and the improved pointwise Weyl asymptotic (2.4) over (2.2). See Safarov [Sa], Sogge-Zelditch [SZ], Sogge-Toth-Zelditch [STZ], and Canzani-Hanin [CH] for more details.

2.2. Probabilistic estimates. We define in (1.5) the random waves $u = \sum a_j e_j$ such that the random coefficients are chosen from the unit sphere with uniform probability measure. In this section, we gather some standard estimates of this probabilistic density.

We mention that there are different randomisation precedures ([B,CI1,CI2,Z3]) where a_j are chosen as identical and independent variables with proper normalization, e.g. Gaussian random variables or Rademacher random variables. As noted by Zelditch [Z3, Section 0.1], choosing the random variables from the unit sphere is more intuitive. In addition, on the spheres, the Levy concentration of measure (Theorem 2.7) is crucial to establish the uniform equidistribution of random waves in Theorem 1.5.

Let $\mathbb{S}^{d-1} \subset \mathbb{R}^d$ be the (d-1)-dim unit sphere endowed with the uniform probability measure μ_d . Write

$$u = \sum_{j=1}^{d} a_j s_j$$
, where $a = (a_1, ..., a_d) \in \mathbb{S}^{d-1}$ and $s = (s_1, ..., s_d) \in \mathbb{R}^d$.

Notice that

|u| > t if and only if $|\langle (a_1, ..., a_d), (s_1(x), ..., s_d(x)) \rangle_{\mathbb{R}^d} | > t$.

We then have the following fact. See e.g. Burq–Lebeau [BL, Appendix A] for an elementary proof.

Lemma 2.5.

$$\mu_d(|u| > t) = \begin{cases} \left(1 - \frac{t^2}{|s|^2}\right)^{\frac{d}{2} - 1} & \text{if } 0 \le t < |s|, \\ 0 & \text{if } t \ge |s|, \end{cases}$$

where |s| is the length of $s = (s_1, ..., s_d) \in \mathbb{R}^d$.

Because $|a_j|$, j = 1, ..., d, has identical distribution for $a \in \mathbb{S}^{d-1}$, it is obvious (since $\sum_{j=1}^{d} |a_j|^2 = |a|^2 = 1$ for $a \in \mathbb{S}^{d-1}$) that $\mathbb{E}(|a_j|^2) = d^{-1}$. Using Lemma 2.5, one can directly compute the *p*-moment $\mathbb{E}(|a_j|^p)$ for $p \ge 0$. Taking *s* as the unit vector in the *j*-th axis of \mathbb{R}^d , we have that $\langle a, s \rangle = a_j$, therefore,

$$\mathbb{E}(|a_{j}|^{p}) = \int_{\mathbb{S}^{d-1}} |a_{j}|^{p} da.$$

$$= \int_{0}^{\infty} t^{p-1} \mu_{d}(|a_{j}| > t) dt$$

$$= \int_{0}^{1} t^{p-1} \left(1 - t^{2}\right)^{\frac{d}{2} - 1} dt$$

$$= \frac{1}{2} \beta \left(\frac{d}{2}, \frac{p}{2}\right)$$

$$= c_{p} d^{-\frac{p}{2}} + O_{p} \left(d^{-\frac{p}{2} - 1}\right). \qquad (2.5)$$

Here, $\beta(\cdot, \cdot)$ is the beta function and c_p is a constant that depends only on p. The asymptotic follows from the fact that if y is fixed and x large, $\beta(x, y) \sim \Gamma(y)x^{-y}$.

In the estimation of the variance in Theorem 1.2, we also need the expectation of $\mathbb{E}(|a_i|^2|a_j|^2)$ for $i \neq j$ (see Sect. 3). If a_j are chosen as independent random variables such as in the Gaussian ensemble $\sum_{j=1}^{d} a_j e_j$, then we normalize $\mathbb{E}\left(||u||_{L^2}^2\right) = 1$ by setting $\mathbb{E}(|a_j|^2) = d^{-1}$. Hence, it is straightforward to see that $\mathbb{E}(|a_i|^2|a_j|^2) = d^{-2}$ for $i \neq j$ from the independence of a_i and a_j . More generally, for $1 \leq m \leq d$ and any $j_1, ..., j_m \in \{1, ..., d\}$ distinct, we have that

$$\mathbb{E}\left(|a_{j_1}|^2\cdots|a_{j_m}|^2\right)=d^{-m}.$$

In Lemma 2.6 below, we show that that when a_j are chosen from the unit sphere \mathbb{S}^{d-1} , $\mathbb{E}\left(|a_{j_1}|^2 \cdots |a_{j_m}|^2\right) = d^{-m} + O(d^{-m-1})$, which agrees with the independent variable case modulo lower order term.

First, let $p = (p_1, ..., p_d)$ for $p_j \ge 0$, j = 1, ..., d. Denote $|p| = p_1 + \cdots + p_d$. Then by (2.5) and Hölder inequality, we have that

$$\mathbb{E}\left(|a_{1}|^{p_{1}}\cdots|a_{d}|^{p_{d}}\right) = \int_{\mathbb{S}^{d-1}} |a_{1}|^{p_{1}}\cdots|a_{d}|^{p_{d}} da$$

$$\leq \left(\int_{\mathbb{S}^{d-1}} |a_{1}|^{|p|} da\right)^{\frac{p_{1}}{|p|}}\cdots\left(\int_{\mathbb{S}^{d-1}} |a_{d}|^{|p|} da\right)^{\frac{p_{d}}{|p|}}$$

$$= O\left(d^{-\frac{|p|}{2}}\right).$$
(2.6)

We now prove the following fact.

Lemma 2.6. Let $1 \le m \le d$ and $j_k \in \{1, ..., d\}$, k = 1, ..., m, be distinct. Then

$$A_m := \mathbb{E}\left(a_{j_1}^2 \cdots a_{j_m}^2\right) = \frac{1}{d^m}\left(1 + O_m\left(d^{-1}\right)\right).$$

Proof. Since $a \in \mathbb{S}^{d-1}$,

$$1 = \left(\sum_{j=1}^{d} a_j^2\right)^m = \sum_{j_k \text{ are distinct}} a_{j_1}^2 \cdots a_{j_m}^2 + \sum_{\text{others}} a_{j_1}^2 \cdots a_{j_m}^2$$

The first summation on the right-hand-side has $d(d-1)\cdots(d-m+1)$ terms and the second summation has $O_m(d^{m-1})$ terms. Note that the expectations of all the terms in the first summation are identical. Hence, taking the expectation of both sides in the above equation, we have that

$$1 = d(d-1)\cdots(d-m+1)A_m + \sum_{\text{others}} \mathbb{E}\left(a_{j_1}^2\cdots a_{j_m}^2\right) \\ = d(d-1)\cdots(d-m+1)A_m + O_m(d^{m-1})O(d^{-m})$$

where we use (2.6) to estimates the expectations in the second sum. The lemma therefore follows. \Box

To establish the uniform equidistribution of random waves in Theorem 1.5, we need to control the probability that a function deviates from the expectation. To this end, we use the principle of concentration of measure. It is here that the high dimensionality of the probability spaces we consider comes into play. Concentration of measure requires that a random variable F(a) cannot take values away from its median too often. Exactly how close to the median depends on regularity properties of F. Let

$$||F||_{\text{Lip}} := \sup_{a \neq b} \frac{|F(a) - F(b)|}{\operatorname{dist}(a, b)},$$

where dist(\cdot , \cdot) is the geodesic distance on \mathbb{S}^{d-1} . A number $\mathcal{M}(F)$ is said to be a median value of *F* if

$$\mu_d(F \ge \mathcal{M}(F)) \ge \frac{1}{2}$$
 and $\mu_d(F \le \mathcal{M}(F)) \ge \frac{1}{2}$.

Levy concentration of measures [Le, Theorem 2.3, (1.10), and (1.12)] then asserts that a Lipschitz function on \mathbb{S}^{d-1} is highly concentrated around its median value when its dimension is large.

Theorem 2.7 (Levy concentration of measures). *Consider a Lipschitz function* F on \mathbb{S}^{d-1} . *Then for any* t > 0, we have that

$$\mu_d(|F - \mathcal{M}(F)| > t) \le \exp\left(-\frac{(d-2)t^2}{2\|F\|_{\operatorname{Lip}}^2}\right).$$

3. Proofs of Theorems 1.2 and 1.3

In this section, we prove the small scale equidistribution results in Theorems 1.2 and 1.3. Recall that

$$\mathcal{H}_W(\lambda) = \operatorname{span}_{\lambda_j \in [\lambda - W, \lambda]} \{e_j\} \text{ and } N_W(\lambda) = \dim \mathcal{H}_W(\lambda).$$

We write the kernel of the spectral projection operator onto $\mathcal{H}_W(\lambda)$ as

$$E_{[\lambda-W,\lambda]}(x, y) = \sum_{\lambda_j \in [\lambda-W,\lambda]} e_j(x) e_j(y) \text{ for } x, y \in \mathbb{M}.$$

For $a \in \mathbb{S}^{N_W(\lambda)-1}$, let

$$u_{\lambda,a} = \sum_{\lambda_j \in [\lambda - W, \lambda]} a_j e_j \in \mathcal{H}_W(\lambda)$$

be a random wave. In the following proposition, we reduce the estimates of expectation and variance in Theorems 1.2 and 1.3 to integrals involving the spectral projection kernel $E_{[\lambda-W,\lambda]}$.

Proposition 3.1 (Expectation and variance of random waves via the spectral kernel). Let $\Omega \subset \mathbb{M}$ be measurable. Write

$$F_{\Omega}(a) = \int_{\Omega} |u_{\lambda,a}(x)|^2 dx.$$
(3.1)

Then

(*i*).
$$\mathbb{E}(F_{\Omega}) = \frac{1}{N_W(\lambda)} \int_{\Omega} E_{[\lambda - W, \lambda]}(x, x) dx,$$

and

(*ii*).
$$\operatorname{Var}(F_{\Omega}) = \frac{2}{N_W(\lambda)^2} \left(1 + O\left(\frac{1}{N_W(\lambda)}\right) \right) \int_{\Omega} \int_{\Omega} \int_{\Omega} E^2_{[\lambda - W, \lambda]}(x, y) \, dx \, dy$$

 $+ O\left(s_{\Omega} \frac{\mathbb{E}(F_{\Omega})}{N_W(\lambda)} \right) + O\left(\frac{\mathbb{E}(F_{\Omega})^2}{N_W(\lambda)} \right),$

where

$$s_{\Omega} = \sup_{\lambda_j \in [\lambda - W, \lambda]} \int_{\Omega} e_j^2(x) \, dx.$$
(3.2)

Proof. For simplicity of notation, we renumber the eigenbasis of $\mathcal{H}_W(\lambda)$ as $\{e_1, ..., e_{N_W(\lambda)}\}$.

(i) To prove the expectation result, denote

$$e_{W,\lambda}(x) = |(e_1(x), ..., e_{N_W(\lambda)}(x))|$$
 for $x \in \mathbb{M}$,

that is, the length of the vector $(e_1(x), ..., e_{N_W(\lambda)}(x)) \in \mathbb{R}^{N_W(\lambda)}$. Then

$$\mathbb{E}(F_{\Omega}) = \int_{\mathbb{S}^{N_W(\lambda)-1}} \int_{\Omega} \sum_{i,j=1}^{N_W(\lambda)} a_i a_j e_i(x) e_j(x) \, dx \, d\mu_{N_W(\lambda)}.$$

Recall that the μ_d is the uniform probability measure on the sphere \mathbb{S}^{d-1} . The symmetry of the sphere ensures that

$$\int_{\mathbb{S}^{N_W(\lambda)-1}} a_i a_j \, d\mu_{N_W(\lambda)} = 0, \quad \text{if } i \neq j,$$

and $\mathbb{E}(a_i^2) = 1/N_W(\lambda)$. Therefore

$$\mathbb{E}(F_{\Omega}) = \frac{1}{N_W(\lambda)} \int_{\Omega} \sum_{i=1}^{N_W(\lambda)} |e_i(x)|^2 dx = \frac{1}{N_W(\lambda)} \int_{\Omega} E_{[\lambda - W, \lambda]}(x, x) dx$$

(ii) To prove the variance result, we directly compute that

$$\begin{aligned} \operatorname{Var}(F_{\Omega}) &= \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \left| F_{\Omega}(a) - \mathbb{E}(F_{\Omega}) \right|^{2} d\mu_{N_{W}(\lambda)} \\ &= \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \left| \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} a_{i}a_{j}e_{i}(x)e_{j}(x) \, dx - \mathbb{E}(F_{\Omega}) \right|^{2} d\mu_{N_{W}(\lambda)} \\ &= \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \int_{\Omega} \int_{\Omega} \sum_{i,j,k,l=1}^{N_{W}(\lambda)} a_{i}a_{j}a_{k}a_{l}e_{i}(x)e_{j}(x)e_{k}(y)e_{l}(y) \, dxdyd\mu_{N_{W}(\lambda)} \\ &- 2\mathbb{E}(F_{\Omega}) \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} a_{i}a_{j}e_{i}(x)e_{j}(x) \, dxd\mu_{N_{W}(\lambda)} + \mathbb{E}(F_{\Omega})^{2}. \end{aligned}$$

Note that the symmetry of the sphere gives ensures that any term containing odd powers of the a_i is zero in expectation. So the terms with even powers remains only. Hence, in the notation of Lemma 2.6

$$\begin{aligned} \operatorname{Var}(F_{\Omega}) &= \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \int_{\Omega} \int_{\Omega} \sum_{\substack{i,j,k,l=1\\i,j,k,l=1}}^{N_{W}(\lambda)} a_{i}a_{j}a_{k}a_{l}e_{i}(x)e_{j}(x)e_{k}(y)e_{l}(y) \\ &dxdyd\mu_{N_{W}(\lambda)} - \mathbb{E}(F_{\Omega})^{2} \\ &= 2 \int_{\Omega} \int_{\Omega} \sum_{\substack{i,j=1\\i\neq j}}^{N_{W}(\lambda)} A_{2}e_{i}(x)e_{i}(y)e_{j}(x)e_{j}(y) dxdy \\ &+ \int_{\Omega} \int_{\Omega} \sum_{\substack{i,j=1\\i\neq j}}^{N_{W}(\lambda)} A_{2}e_{i}^{2}(x)e_{j}^{2}(y) dxdy \\ &+ \int_{\Omega} \int_{\Omega} \sum_{\substack{i,j=1\\i\neq j}}^{N_{W}(\lambda)} \mathbb{E}(a_{i}^{4})e_{i}^{2}(x)e_{i}^{2}(y)dxdy - \mathbb{E}(F_{\Omega})^{2}. \end{aligned}$$

Applying Lemma 2.6 with m = 2 and $d = N_W(\lambda)$, we have that for $i \neq j$,

$$A_2 = \mathbb{E}\left(a_i^2 a_j^2\right) = \frac{1}{N_W(\lambda)^2} \left(1 + O\left(N_W(\lambda)^{-1}\right)\right)$$

By (2.6), we have that $\mathbb{E}(a_i^4) = O(N_W(\lambda)^{-2})$. We will see that the diagonal contribution

$$\sum_{i=1}^{N_W(\lambda)} \mathbb{E}[a_i^4] e_i(x) e_y(y)$$

makes only a small contribution to the overall computation. Therefore we can add and subtract it from the main terms as necessary. So

$$\begin{aligned} \operatorname{Var}(F_{\Omega}) &= \frac{2}{N_{W}(\lambda)^{2}} \left(1 + O\left(\frac{1}{N_{W}(\lambda)}\right) \right) \int_{\Omega} \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} e_{i}(x)e_{i}(y)e_{j}(x)e_{j}(y) \, dx \, dy \\ &+ \frac{1}{N_{W}(\lambda)^{2}} \left(1 + O\left(\frac{1}{N_{W}(\lambda)}\right) \right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} e_{i}^{2}(x)e_{j}^{2}(y) \, dx \, dy - \mathbb{E}(F_{\Omega})^{2} \\ &+ O\left(\frac{1}{N_{W}(\lambda)^{2}}\right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i=1}^{N_{W}(\lambda)} e_{i}^{2}(x)e_{i}^{2}(y) \, dx \, dy \end{aligned}$$
$$&= \frac{2}{N_{W}(\lambda)^{2}} \left(1 + O\left(\frac{1}{N_{W}(\lambda)}\right) \right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i=1}^{E_{1}^{2}-W,\lambda]} (x, y) \, dx \, dy \\ &+ O\left(\frac{1}{N_{W}(\lambda)^{2}}\right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i=1}^{N_{W}(\lambda)} e_{i}^{2}(x)e_{i}^{2}(y) \, dx \, dy \end{aligned}$$
$$&+ O\left(\frac{1}{N_{W}(\lambda)^{3}}\right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} e_{i}^{2}(x)e_{j}^{2}(y) \, dx \, dy \end{aligned}$$
$$&+ O\left(\frac{1}{N_{W}(\lambda)^{3}}\right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i,j=1}^{N_{W}(\lambda)} e_{i}^{2}(x)e_{j}^{2}(y) \, dx \, dy \end{aligned}$$
$$&\leq \frac{2}{N_{W}(\lambda)^{2}} \left(1 + O\left(\frac{1}{N_{W}(\lambda)}\right) \right) \int_{\Omega} \int_{\Omega} \int_{\Omega} \sum_{i,j=1}^{2} e_{i}^{2}(x)e_{j}^{2}(y) \, dx \, dy \end{aligned}$$
$$&+ O\left(s_{\Omega} \frac{\mathbb{E}(F_{\Omega})}{N_{W}(\lambda)}\right) + O\left(\frac{\mathbb{E}(F_{\Omega})^{2}}{N_{W}(\lambda)}\right).$$

Here, s_{Ω} is defined in (3.2) and we use the fact that

$$\int_{\Omega} \int_{\Omega} \sum_{i=1}^{N_W(\lambda)} e_i^2(x) e_i^2(y) \, dx \, dy \leq s_{\Omega} \int_{\Omega} \sum_{i=1}^{N_W(\lambda)} e_i^2(x) \, dx = s_{\Omega} N_W(\lambda) \mathbb{E}(F_{\Omega}).$$

Now we prove the small scale equidistribution results of random waves in Theorem 1.2.

Proof of Theorem 1.2. Since the spectral projection kernel

$$E_{[\lambda-W,\lambda]} = E_{[0,\lambda]} - E_{[0,\lambda-W)},$$

the spectral estimates of the Laplacian in Sect. 2.1 apply.

(i) To prove the expectation result we use the pointwise Weyl asymptotics in Corollary 2.2 to see that

$$E_{[\lambda-W,\lambda]}(x,x) = E_{[0,\lambda]}(x,x) - E_{[0,\lambda-W)}(x,x)$$

= $c_n \lambda^n + R(\lambda,x) - c_n (\lambda-W)^n - R(\lambda-W,x)$
= $nc_n W \lambda^{n-1} + O(\lambda^{n-1}).$

Also from the Weyl asmptotics of the eigenvalues

$$N_W(\lambda) = nc_n W \lambda^{n-1} \operatorname{Vol}(\mathbb{M}) + O(\lambda^{n-1}).$$
(3.3)

Setting $\Omega = B(x, r)$ in (i) of Proposition 3.1, we conclude (1.6) in Theorem 1.2.

(ii) To prove the variance result, we need the off-diagonal description of $E_{[\lambda-W,\lambda]}(x, y)$ of Theorem 2.1. For $x, y \in \mathbb{M}$ close enough, say $d(x, y) < d_0$ as in Theorem 2.1,

$$E_{[\lambda-W,\lambda]}(x,y) = \frac{1}{(2\pi)^n} \int_{\lambda-W \le |\xi|_{gy} \le \lambda} e^{i\langle \exp_y^{-1}(x),\xi \rangle} \frac{d\xi}{\sqrt{|g_y|}} + R(x,y,\lambda) - R(x,y,\lambda-W).$$
(3.4)

Here, for notational simplicity, we write

$$R(\lambda, W) = \sup_{d(x,y) < d_0} |R(x, y, \lambda) - R(x, y, \lambda - W)|$$

Hence, by Theorem 2.1

$$R(\lambda, W) = O(\lambda^{n-1}).$$
(3.5)

Since x and y are close, we can assume that they are in the same coordinate patch. Indeed, we may assume that y is the centre of that patch and $g_y = \text{Id}$. Therefore, the integral in (3.4) becomes

$$\frac{1}{(2\pi)^n}\int_{\lambda-W<|\xi|\leq\lambda}e^{i\langle\exp_y^{-1}(x),\xi\rangle}\,d\xi.$$

Note that the inner product here is understood by associating $\exp_y^{-1}(x)$ with an element of \mathbb{R}^n so effectively what we need to calculate is

$$\frac{1}{(2\pi)^n} \int_{\lambda - W \le |\xi| \le \lambda} e^{i\langle z, \xi \rangle} d\xi = \frac{1}{(2\pi)^n} \int_{\lambda - W}^{\lambda} \int_{|\xi| = \rho} e^{i\langle z, \xi \rangle} d\xi d\rho.$$

That is, we need to take the inverse Fourier transform of the surface measure of the sphere of radius ρ . This is a classical problem from harmonic analysis and can be computed by stationary phase to give

$$\left| \int_{|\xi|=\rho} e^{i\langle z,\xi\rangle} \, d\xi \right| \le c\rho^{n-1} (1+\rho|z|)^{-\frac{n-1}{2}},$$

in which c depends only on n. See e.g. Sogge [So3, Section 1.2]. Therefore,

$$\frac{1}{(2\pi)^n} \int_{\lambda - W < |\xi| \le \lambda} e^{i \langle \exp_y^{-1}(x), \xi \rangle} d\xi \le \begin{cases} cW\lambda^{n-1}, & \text{if } 0 \le |x-y| \le \lambda^{-1}; \\ cW\lambda^{\frac{n-1}{2}}|x-y|^{-\frac{n-1}{2}}, & \text{if } \lambda^{-1} \le |x-y| \le 1. \end{cases}$$

Taking $\Omega = B(x_0, r)$ in (ii) of Proposition 3.1, we compute that

$$\begin{split} &\int_{B(x_0,r)} \int_{B(x_0,r)} E^2_{[\lambda-W,\lambda]}(x,y) \, dx dy \\ &\leq c \int_{B(x_0,r)} \int_{B(x_0,r) \cap B(x,\lambda^{-1})} W^2 \lambda^{2(n-1)} \, dy dx \\ &+ c \int_{B(x_0,r)} \int_{B(x_0,r) \setminus B(x,\lambda^{-1})} W^2 \lambda^{n-1} |x-y|^{-(n-1)} \, dy dx \\ &+ \int_{B(x_0,r)} \int_{B(x_0,r)} R(\lambda,W)^2 \, dx dy \\ &\leq c W^2 \lambda^{n-2} \operatorname{Vol}(B(x_0,r)) + c W^2 \lambda^{n-1} r \operatorname{Vol}(B(x_0,r)) \\ &+ R(\lambda,W)^2 \operatorname{Vol}(B(x_0,r))^2. \end{split}$$

Combining with (3.3),

$$\operatorname{Var}(F_{B(x_0,r)}) = \frac{2}{N_W(\lambda)^2} \left(1 + O\left(\frac{1}{N_W(\lambda)}\right) \right) \int_{B(x_0,r)} \int_{B(x_0,r)} E_{[\lambda-W,\lambda]}^2(x, y) \, dx \, dy + O\left(s_{B(x_0,r)} \frac{\mathbb{E}(F_{B(x_0,r)})}{N_W(\lambda)} \right) + O\left(\frac{\mathbb{E}(F_{B(x_0,r)})^2}{N_W(\lambda)}\right) \leq c \lambda^{-n} \operatorname{Vol}(B(x_0, r)) + c \lambda^{-(n-1)} r \operatorname{Vol}(B(x_0, r)) + c W^{-2} \lambda^{-2(n-1)} R(\lambda, W)^2 \operatorname{Vol}(B(x_0, r))^2 + O\left(s_{B(x_0,r)} \frac{\mathbb{E}(F_{B(x_0,r)})}{N_W(\lambda)} \right) + O\left(\frac{\mathbb{E}(F_{B(x_0,r)})^2}{N_W(\lambda)} \right).$$
(3.6)

If $r^{-1} = o(\lambda)$, then $\lambda^{-n} = o(r^n)$ and so

$$\lambda^{-n} \operatorname{Vol}(B(x_0, r)) = o\left(r^{2n}\right) \text{ and } \lambda^{-(n-1)} r \operatorname{Vol}(B(x_0, r)) = o\left(r^{2n}\right).$$

In Section 4 we will see that, Lemma 4.2,

$$s_{B(x_0,r)} \leq \begin{cases} CWr & \text{if } \lambda^{-1} \leq r \leq W^{-1} \\ 1 & \text{if } W^{-1} \leq r \leq \text{Inj } (\mathbb{M}). \end{cases}$$

So in either of the above cases,

$$s_{B(x_0,r)}\frac{\mathbb{E}(F_{B(x_0,r)})}{N_W(\lambda)} = o(r^{2n}).$$

Therefore, (ii) in Theorem 1.2 follows from the fact that $R(\lambda, W) = O(\lambda^{n-1})$ in (3.5).

We next provide a short proof of Theorem 1.3 in the case of fixed window length with the help of Theorem 2.4.

Proof of Theorem 1.3. (i). To prove the expectation result, we use the improved pointwise Weyl asymptotics in Theorem 2.4 to see that

$$E_{[\lambda-W,\lambda]}(x,x) = E_{[0,\lambda]}(x,x) - E_{[0,\lambda-W]}(x,x)$$

= $c_n\lambda^n + R(\lambda,x) - c_n(\lambda-W)^n - R(\lambda-W,x)$
= $nc_nW\lambda^{n-1} + o(\lambda^{n-1}).$

We also have that

$$N_W(\lambda) = nc_n W \lambda^{n-1} \operatorname{Vol}(\mathbb{M}) + o(\lambda^{n-1})$$

Taking $\Omega = B(x, r)$ in (i) of Proposition 3.1, we conclude (i) in Theorem 1.3.

(ii) To prove the variance result, we only need to notice that in the last term of (3.6), $R(\lambda, W) = o(\lambda^{n-1})$ from the improvement of remainder terms provided in Theorem 2.4. \Box

4. Uniform Equidistribution

For a ball B(x, r) with a *fixed* $x \in \mathbb{M}$, the expectation and variance results in Theorems 1.2 and 1.3 tell us that for scales $r = r(\lambda)$ such that $r/r_{\text{Planck}} \to \infty$, the probability that the random wave $u_{\lambda,a}$ does not equidistribute on B(x, r) decays in λ .

In this section, we establish a quantitative estimate of the probability for small scale equidistribution of random waves *uniformly* everywhere on the manifold, i.e. on balls all $B(x, r), x \in \mathbb{M}$.

First for a fixed $x \in \mathbb{M}$, we derive a quantitative estimate of the probability such that $u_{\lambda,a}$ does not equidistribute on B(x, r). The main tool is from the Levy concentration of measure in Theorem 2.7. We then use a covering argument to estimate the probability for equidistribution on all balls in \mathbb{M} .

Levy concentration of measure in Theorem 2.7 requires a Lipschitz norm estimate. To this end, we need a result from Sogge [So1, Section 4] that limits the maximum L^2 growth on a small ball. (It is actually proved for spectral clusters in [So1], which applies to combination of eigenfunctions in the same spectral window.)

Lemma 4.1. On a compact manifold \mathbb{M} , let $u = \sum_{\lambda_j \in [\lambda - 1, \lambda]} a_j e_j$. Then for all $x \in \mathbb{M}$ and $\lambda^{-1} \leq r \leq \operatorname{Inj} \mathbb{M}$, we have that

$$\int_{B(x,r)} |u|^2 \, d\text{Vol} \le cr \, \|u\|_{L^2(\mathbb{M})}^2 \,, \tag{4.1}$$

where c > 0 depending only on \mathbb{M} .

Lemma 4.1 is of course an improvement on the trivial estimate $\int_{B(x,r)} |u|^2 \leq \int_{\mathbb{M}} |u|^2$. In fact, it is already sharp on \mathbb{S}^n , as the estimate is saturated by the zonal harmonics on balls centred at one of the poles. See Sogge [So1, Section 4] for more discussion.

Since we consider window widths $W = W(\lambda)$ such that $1 \le W \le \lambda$, we need to get an analogous estimate to (4.1) for

$$\sum_{\lambda_j\in[\lambda-W,\lambda]}a_je_j.$$

We are able to use Lemma 4.1 to obtain the necessary estimates.

Lemma 4.2. Let $1 \leq W \leq \lambda$. Suppose that for $a_j \in \mathbb{R}$,

$$u = \sum_{\lambda_j \in [\lambda - W, \lambda]} a_j e_j.$$

Then there exists a positive constant c depending only on \mathbb{M} such that

$$\int_{B(x,r)} |u|^2 \, d\text{Vol} \le \begin{cases} c Wr \, \|u\|_{L^2(\mathbb{M})}^2, & \text{if } \lambda^{-1} \le r \le W^{-1}, \\ \|u\|_{L^2(\mathbb{M})}^2, & \text{if } W^{-1} \le r \le \text{Inj } \mathbb{M}. \end{cases}$$
(4.2)

Proof. The inequalities in (4.2) are trivially true if $r \ge W^{-1}$ so we assume that Wr > 1. That is, $\lambda^{-1} \le r \le W^{-1}$. Also if $W \in [\lambda/2, \lambda]$, then

$$Wr \ge \frac{\lambda}{2} \cdot \lambda^{-1} \ge \frac{1}{2}.$$

Therefore (4.2) holds with c = 2.

Now we assume that $W \leq \lambda/2$. Write

$$u = \sum_{0 \le k \le W-1} u_k$$
, where $u_k = \sum_{\lambda_j \in [\lambda - W + k, \lambda - W + k+1]} a_j e_j$.

Note that each u_k is a fixed window spectral cluster at frequency $\mu_k = \lambda - W + k + 1 > \lambda/2$ so we may apply Lemma 4.1 to each of the u_k separately. Thus,

$$\int_{B(x_0,r)} |u|^2 \, d\text{Vol} = \sum_{0 \le m, k \le W-1} \int_{B(x_0,r)} u_k(x) u_{(k+m)_{W-1}}(x) \, d\text{Vol},$$

where

$$(k+m)_{W-1} = k+m \mod (W-1).$$

Applying Lemma 4.1 and the Cauchy-Schwarz inequality, we have that

$$\begin{split} \int_{B(x_0,r)} |u|^2 \, d\text{Vol} &\lesssim r \sum_{0 \le m, k \le W-1} \|u_k\|_{L^2(\mathbb{M})} \|u_{(k+m)_{W-1}}\|_{L^2(\mathbb{M})} \\ &\lesssim r \sum_{0 \le m \le W-1} \left(\sum_{0 \le k \le W-1} \|u_k\|_{L^2(\mathbb{M})}^2 \right)^{1/2} \\ &\left(\sum_{0 \le k \le W-1} \|u_{(k+m)_{W-1}}\|_{L^2(\mathbb{M})}^2 \right)^{1/2} \\ &\lesssim r \, W \, \|u\|_{L^2(\mathbb{M})}^2 \, . \end{split}$$

Remark. It turns out that the above simple estimates are sharp. Spectral clusters u of window width W in Lemma 4.2 are special cases of approximate eigenfunctions with L^2 error no greater than $W\lambda$. That is,

$$\left\| (\Delta - \lambda^2) u \right\|_{L^2(\mathbb{M})} \lesssim W \lambda \, \| u \|_{L^2(\mathbb{M})} \, .$$

Such functions can localize in one ball with radius $r = W^{-1}$. See e.g. Tacy [T].

We now prove the following estimate on the Lipschitz norm of $F_{B(x,r)}$ defined in (3.1).

Proposition 4.3. There exists a positive constant c depending only on \mathbb{M} such that

$$\|F_{B(x,r)}\|_{\operatorname{Lip}} \leq \begin{cases} crW, & \text{if } \lambda^{-1} \leq r \leq W^{-1}; \\ c, & \text{if } W^{-1} \leq r \leq \operatorname{Inj} \mathbb{M}. \end{cases}$$

Proof. Given $u, v \in \mathcal{H}_W(\lambda)$, let

$$u = \sum_{j=1}^{N_W(\lambda)} a_j e_j$$
 and $v = \sum_{j=1}^{N_W(\lambda)} b_j e_j$,

where $a = (a_1, ..., a_{N_W(\lambda)})$ and $b = (b_1, ..., b_{N_W(\lambda)})$ are in $\mathbb{S}^{N_W(\lambda)-1}$. We have that

$$\begin{aligned} \left| F_{B(x,r)}(a) - F_{B(x,r)}(b) \right| &= \left| \int_{B(x,r)} |u(y)|^2 \, dx - \int_{B(x,r)} |v(y)|^2 \, dy \right| \\ &\leq \int_{B(x,r)} \left| u(y)^2 - v(y)^2 \right| \, dx \\ &= \int_{B(x,r)} |u(y) - v(y)| |u(y) + v(y)| \, dy \\ &\leq \left(\int_{B(x,r)} |u(y) - v(y)|^2 \, dy \right)^{\frac{1}{2}} \\ &\qquad \left(\int_{B(x,r)} |u(y) + v(y)|^2 \, dy \right)^{\frac{1}{2}} \\ &\leq \left\| u - v \right\|_{L^2(B(x,r))} \| u + v \|_{L^2(B(x,r))} \, . \end{aligned}$$

If $\lambda^{-1} \leq r \leq W^{-1}$, then by applying Lemma 4.2, we obtain that

$$||u - v||_{L^2(B(x,r))} \le cr^{1/2}W^{1/2} ||u - v||_{L^2(\mathbb{M})},$$

and

$$||u + v||_{L^2(B(x,r))} \le cr^{1/2}W^{1/2} ||u + v||_{L^2(\mathbb{M})}$$

It then follows that

$$\left| F_{B(x,r)}(a) - F_{B(x,r)}(b) \right| \le cr W \| u - v \|_{L^2(\mathbb{M})} \| u + v \|_{L^2(\mathbb{M})} \le cr W \| u - v \|_{L^2(\mathbb{M})},$$
(4.3)

since $||u + v||_{L^2(\mathbb{M})} \le ||u||_{L^2(\mathbb{M})} + ||v||_{L^2(\mathbb{M})} = 2$. From

$$u(x) - v(x) = \sum_{j=1}^{N_W(\lambda)} (a_j - b_j) e_j(x),$$

we also have that

$$||u - v||_{L^2(\mathbb{M})} = |a - b| \approx \operatorname{dist}(a, b).$$
 (4.4)

Here, |a - b| is the distance of a and b in $\mathbb{R}^{N_W(\lambda)}$ while dist(a, b) is the distance of a and b on $\mathbb{S}^{N_W(\lambda)-1}$. Putting (4.4) together with (4.3),

$$\|F_{B(x,r)}\|_{\operatorname{Lip}} = \sup_{a,b\in\mathbb{S}^{N_{W}(\lambda)-1},a\neq b} \frac{\left|F_{B(x,r)}(a) - F_{B(x,r)}(b)\right|}{\operatorname{dist}(a,b)} \le crW.$$

If $W^{-1} \leq r \leq \text{Inj } \mathbb{M}$, we use the trivial estimates that

$$||u - v||_{L^2(B(x,r))} \le ||u - v||_{L^2(\mathbb{M})}$$
 and $||u + v||_{L^2(B(x,r))} \le ||u + v||_{L^2(\mathbb{M})}$.

Thus,

$$\left|F_{B(x,r)}(a) - F_{B(x,r)}(b)\right| \le \|u - v\|_{L^{2}(\mathbb{M})} \|u + v\|_{L^{2}(\mathbb{M})} \le c \|u - v\|_{L^{2}(\mathbb{M})}.$$

In the view of (4.4) again, the Lipschitz norm of $F_{B(x,r)}$ when $W^{-1} \leq r \leq \text{Inj} \mathbb{M}$ follows. \Box

We can now use Levy concentration of measure to control the probability that for a fixed *x*, $F_{B(x,r)}$ deviates from Vol(*B*)/Vol(\mathbb{M}).

Proposition 4.4. Let $m(\lambda)$ be an o(1) order function (i.e. $m : \mathbb{R}^+ \to \mathbb{R}^+$ and $m(\lambda) \to 0$ as $\lambda \to \infty$). For $x \in \mathbb{M}$, denote

$$S_{r,x}(m) = \left\{ a \in \mathbb{S}^{N_W(\lambda) - 1} : \left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \right| \ge r^n m(\lambda) \right\}.$$
(4.5)

For some c and K depending only on \mathbb{M} , the following statements are true.

(i) Suppose $W = W(\lambda)$ such that $1 \le W \le \lambda$ and $W \to \infty$ as $\lambda \to \infty$. Set

$$r_1 = W^{-\frac{1}{2n}} \lambda^{-\frac{n-1}{2n}}.$$

Then for all $m(\lambda) \ge K W^{-1}$ and $r = r(\lambda)$ such that

$$\max\left\{W^{-1}, r_1 m(\lambda)^{-\frac{1}{n}}\right\} \le r \le \operatorname{Inj}(\mathbb{M}),$$

we have that

$$\mu_{N_W(\lambda)}\left(S_{r,x}(m)\right) \le \exp\left(-\frac{cr^{2n}m(\lambda)^2}{r_1^{2n}}\right)$$

(ii) Suppose $W = W(\lambda)$ such that $1 \le W \le \lambda$ and $W \to \infty$ as $\lambda \to \infty$. Set

$$r_2 = W^{\frac{1}{2(n-1)}} \lambda^{-\frac{1}{2}}.$$

Then for all $m(\lambda) \ge K W^{-1}$ and $r = r(\lambda)$ such that

$$r_2 m(\lambda)^{-\frac{1}{n-1}} \le r \le W^{-1},$$

we have that

$$\mu_{N_W(\lambda)}\left(S_{r,x}(m)\right) \le \exp\left(-\frac{cr^{2(n-1)}m(\lambda)^2}{r_2^{2(n-1)}}\right)$$

(iii) Assume that the set of loop directions \mathcal{L}_x is of measure zero in $S_x^* \mathbb{M}$ for all $x \in \mathbb{M}$. Suppose that W > 0 is independent of λ . Then there exists some $m(\lambda)$ that is o(1) as $\lambda \to \infty$ such that for any r satisfying

$$\lambda^{-1/2}m(\lambda)^{-\frac{1}{n-1}} \leq r \leq \operatorname{Inj}(\mathbb{M}),$$

we have

$$\mu_{N_W(\lambda)}\left(S_{r,x}(m)\right) \leq \exp\left(-C\lambda^{n-1}r^{2(n-1)}m(\lambda)^2\right).$$

Proof. First we recall from the expectation estimate (1.6) in Theorem 1.2 that

$$\mathbb{E}\left(F_{B(x,r)}\right) = \mathbb{E}\left(\int_{B(x,r)} |u_{\lambda}|^2 \, d\operatorname{Vol}\right) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \left[1 + O\left(W^{-1}\right)\right],$$

in which the term $O(W^{-1})$ is independent of $x \in \mathbb{M}$.

By Levy concentration of measures in Theorem 2.7, we estimate the difference between the expectation and the median.

$$\begin{split} |\mathbb{E}(F_{B(x,r)}) - \mathcal{M}(F_{B(x,r)})| &= \left| \int_{\mathbb{S}^{N_{W}(\lambda)-1}} F_{B(x,r)}(a) \, da - \mathcal{M}(F_{B(x,r)}) \right| \\ &\leq \int_{\mathbb{S}^{N_{W}(\lambda)-1}} \left| F_{B(x,r)}(a) - \mathcal{M}(F_{B(x,r)}) \right| \, da \\ &= \int_{0}^{\infty} \mu_{N_{W}(\lambda)} \left(|F_{B(x,r)}(a) - \mathcal{M}(F_{B(x,r)})| > t \right) \, dt \\ &\leq \int_{0}^{\infty} \exp\left(-\frac{(N_{W}(\lambda) - 2)t^{2}}{\|F_{B(x,r)}\|_{\mathrm{Lip}}^{2}} \right) \, dt \\ &\leq \frac{c \|F_{B(x,r)}\|_{\mathrm{Lip}}}{N_{W}(\lambda)^{\frac{1}{2}}} \end{split}$$

for some absolute constant c > 0. Putting this together with the expectation (1.6), we then have that

$$\mathcal{M}(F_{B(x,r)}) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} + R_1 + R_2,$$

where

$$R_1 = O(W^{-1})r^n$$
 and $R_2 = O\left(\frac{\|F_{B(x,r)}\|_{\text{Lip}}}{N_W(\lambda)^{\frac{1}{2}}}\right).$

Now we divide into the three cases listed in the proposition.

Case (i). Since $r \ge W^{-1}$, the second inequality of Lipschitz norm estimate in Proposition 4.3 applies. That is,

$$\|F_{B(x,r)}\|_{\mathrm{Lip}} \leq c.$$

Hence as $r \ge r_1 m(\lambda)^{-\frac{1}{n}}$,

$$R_{2} = O\left(\frac{\|F_{B(x,r)}\|_{\text{Lip}}}{N_{W}(\lambda)^{\frac{1}{2}}}\right)$$

= $O\left(N_{W}(\lambda)^{-\frac{1}{2}}\right) = O\left(W^{-\frac{1}{2}}\lambda^{-\frac{n-1}{2}}\right) = O(r_{1}^{n}) = O(r^{n}m(\lambda)).$

Therefore,

$$\mathcal{M}(F_{B(x,r)}) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} + O\left(r^n(W^{-1} + m(\lambda))\right).$$
(4.6)

Now we use the Levy concentration of measure to control deviance from the median. Define

$$\widetilde{S}_{r,x}(m) = \left\{ a \in \mathbb{S}^{N_W(\lambda) - 1} : \left| \int_{B(x,r)} |u_{\lambda,a}|^2 d\operatorname{Vol} - \mathcal{M}(F_{B(x,r)}) \right| \ge r^n m(\lambda) \right\}.$$
(4.7)

By the Levy concentration of measure in Theorem 2.7,

$$\mu_{N_{W}(\lambda)}\left(\widetilde{S}_{r,x}\left(\frac{m}{2}\right)\right) \leq \exp\left(-\frac{(N_{W}(\lambda)-2)r^{2n}m(\lambda)^{2}}{8\|F_{B(x,r)}\|_{\operatorname{Lip}}^{2}}\right)$$
$$\leq \exp\left(-cW\lambda^{n-1}r^{2n}m(\lambda)^{2}\right)$$
$$= \exp\left(-\frac{cr^{2n}m(\lambda)^{2}}{r_{1}^{2n}}\right),$$

as $r_1 = W^{-\frac{1}{2n}} \lambda^{-\frac{n-1}{2n}}$.

Now suppose that $a \in S_{r,x}(m)$. Then

$$\left|\int |u_{\lambda,a}|^2 d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})}\right| \ge r^n m(\lambda).$$

So combining with (4.6),

$$\left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \mathcal{M}(F_{B(x,r)}) \right| \ge \left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \right| - \left| \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} - \mathcal{M}(F_{B(x,r)}) \right| \\\ge r^n m(\lambda) + O\left(r^n (W^{-1} + m(\lambda))\right).$$

If $m \ge K W^{-1}$ for some sufficiently large K, then

$$\left|\int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \mathcal{M}(F_{B(x,r)})\right| \geq \frac{3r^n m(\lambda)}{4} + O\left(r^n m(\lambda)\right).$$

Since $m(\lambda) \to 0+$ as $\lambda \to \infty$, we can conclude that

$$\left|\int_{B(x,r)} |u_{\lambda,a}|^2 \, d\text{Vol} - \mathcal{M}(F_{B(x,r)})\right| \ge \frac{r^n m(\lambda)}{2}$$

for sufficiently large λ . That is, $a \in \widetilde{S}_{r,x}(m/2)$. Therefore,

$$\mu_{N_W(\lambda)}(S_{r,x}(m)) \le \mu_{\mathbb{S}^{N_W(\lambda)-1}}\left(\widetilde{S}_{r,x}\left(\frac{m}{2}\right)\right) \le \exp\left(-\frac{cr^{2n}m(\lambda)^2}{r_1^{2n}}\right).$$

Case (ii). The reasoning follows as Case (i) however the Lipschitz norm is different. Since $r \le W^{-1}$, the first inequality of Lipschitz norm estimate in Proposition 4.3 applies. That is,

$$||F_{B(x,r)}||_{\operatorname{Lip}} \leq crW.$$

Hence as $r \ge r_2 m(\lambda)^{-\frac{1}{n-1}}$, $R_2 = O\left(\frac{\|F_{B(x,r)}\|_{\text{Lip}}}{N_W(\lambda)^{\frac{1}{2}}}\right)$ $= O\left(\frac{crW}{N_W(\lambda)^{\frac{1}{2}}}\right) = O\left(rW^{\frac{1}{2}}\lambda^{-\frac{n-1}{2}}\right) = O(rr_2^{n-1}) = O(r^n m(\lambda)).$

Note that the above equation is also independent of $x \in M$. So again,

$$\mathcal{M}(F_{B(x,r)}) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} + O(r^n(W^{-1} + m(\lambda))).$$

Then with $\widetilde{S}_{r,w}(m)$ defined as in (4.7), we have that

$$\mu_{N_{W}(\lambda)}\left(\widetilde{S}_{r,x}\left(\frac{m}{2}\right)\right) \leq \exp\left(-\frac{(N_{W}(\lambda)-2)r^{2n}m(\lambda)^{2}}{8\|F_{B(x,r)}\|_{\operatorname{Lip}}^{2}}\right)$$
$$\leq \exp\left(-cW^{-1}\lambda^{n-1}r^{2(n-1)}m(\lambda)^{2}\right)$$
$$= \exp\left(-\frac{cr^{2(n-1)}m(\lambda)^{2}}{r_{2}^{2(n-1)}}\right),$$

as $r_2 = W^{\frac{1}{2(n-1)}} \lambda^{-\frac{1}{2}}$. As in Case (i), $a \in S_{r,x}(m)$ implies that $a \in \widetilde{S}_{r,x}(m/2)$ so we also have

$$\mu_{N_W(\lambda)}(S_{r,x}(m)) \le \exp\left(-\frac{cr^{2(n-1)}m(\lambda)^2}{r_2^{2(n-1)}}\right).$$

Case (iii). Now we address the case where the window width is allowed to be fixed but we assume that \mathbb{M} satisfies the loop set conditions. By Theorem 1.3, there exists some o(1) order function $m(\lambda)$ so that

$$\mathbb{E}\left(F_{B(x,r)}\right) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} + r^{n}m(\lambda).$$

In this case as *W* is fixed the first Lipschitz norm estimate in Proposition 4.3 applies. So as $r \ge \lambda^{-1/2} m(\lambda)^{-\frac{1}{n-1}}$,

$$\left|\mathbb{E}\left(F_{B(x,r)}\right)-\mathcal{M}\left(F_{B(x,r)}\right)\right|\leq r\lambda^{-\frac{n-1}{2}}\leq r^{n}m(\lambda).$$

Therefore,

$$\mathcal{M}\left(F_{B(x,r)}\right) = \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(M)} + r^n m(\lambda) + O(r^n m(\lambda)).$$

As in Case (ii), we have that for this specific $m(\lambda)$,

$$\mu_{N_W(\lambda)}\left(\widetilde{S}_{r,x}\left(\frac{m}{2}\right)\right) \leq \exp\left(-c\lambda^{n-1}r^{2(n-1)}m(\lambda)^2\right).$$

Also by the reasoning of Case (i), if $a \in S_{r,x}(m)$ then $a \in \widetilde{S}_{r,x}(m/2)$ so

$$\mu_{N_W(\lambda)}(S_{r,x}(m)) \le \exp\left(-c\lambda^{n-1}r^{2(n-1)}m(\lambda)^2\right).$$

To prove uniform equidistribution in Theorem 1.5, we use a covering lemma that is similar to the one in Han [Ha2, Section 3.2].

Lemma 4.5. For any d > 0 there exists a family of geodesic balls that covers \mathbb{M} :

$$\bigcup_{p=1}^{N_d} B(x_p, d) \supset \mathbb{M} \quad with \ N_d \le cd^{-n},$$

where c > 0 depends only on \mathbb{M} .

Given a cover $\{B(x_p, d)\}$, note that for any r > d, $\{B(x_p, r)\}$ remains a cover. Moreover, in the cover $\{B(x_p, r)\}$, the centers x_p of the balls are separated by distances d < r. This enables us to efficiently approximate the L^2 mass of u on B(x, r) by the L^2 mass on one of the $B(x_p, r)$. Then the set $S_r(m)$ of all $a \in \mathbb{S}^{N_W(\lambda)-1}$ for which equidistribution fails at some point is contained in the union of the $S_{r,x_p}(m)$, for which we have estimates from Proposition 4.4. See below for the details of the proof.

Proof of Theorem 1.5. We begin by choosing a cover as in Lemma 4.5 with

$$d = r\lambda^{-(n-1)}m(\lambda)^2.$$

Then for any $x \in \mathbb{M}$ there exists an x_p so that $x \in B(x_p, d)$. We now approximate $\operatorname{Vol}(B(x, r))$ by $\operatorname{Vol}(B(x_p, r))$ and the L^2 mass of $u_{\lambda,a}$ in B(x, r) by the one in $B(x_p, r)$, respectively. First,

$$\frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} - \frac{\operatorname{Vol}(B(x_p,r))}{\operatorname{Vol}(\mathbb{M})} \le c\operatorname{Vol}\left(B(x_p,r+d) \setminus B(x_p,r-d)\right) \\ \le cdr^{n-1} \\ \le cr^n\lambda^{-(n-1)}m(\lambda)^2.$$
(4.8)

Second, since $\|u_{\lambda,a}\|_{L^{\infty}} \le c\lambda^{\frac{n-1}{2}}$ (see e.g. [So3, Section 4.2]), we have that

$$\left| \int_{B(x,r)} |u_{\lambda,a}|^2 d\operatorname{Vol} - \int_{B(x_p,r)} |u_{\lambda,a}|^2 d\operatorname{Vol} \right|$$

$$\leq c \left\| u_{\lambda,a} \right\|_{L^{\infty}}^2 \operatorname{Vol} \left(B(x_p, r+d) \setminus B(x_p, r-d) \right)$$

$$\leq c r^n m(\lambda)^2.$$
(4.9)

We recall that

$$S_{r}(m) = \left\{ a \in \mathbb{S}^{N_{W}(\lambda)-1} : \exists x \in \mathbb{M} \text{ such that } \left| \int_{B(x,r)} |u_{\lambda,a}|^{2} d\text{Vol} - \frac{\text{Vol}(B(x,r))}{\text{Vol}(\mathbb{M})} \right| \ge r^{n} m(\lambda) \right\}.$$

Now suppose that $a \in S_r(m)$. Then by (4.8) and (4.9), there exists x_p such that

$$\begin{aligned} \left| \int_{B(x_p,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x_p,r)}{\operatorname{Vol}(\mathbb{M})} \right| &\geq \left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} \right| \\ &- \left| \frac{\operatorname{Vol}(B(x,r))}{\operatorname{Vol}(\mathbb{M})} - \frac{\operatorname{Vol}(B(x_p,r))}{\operatorname{Vol}(\mathbb{M})} \right| \\ &- \left| \int_{B(x,r)} |u_{\lambda,a}|^2 \, d\operatorname{Vol} - \int_{B(x_p,r)} |u_{\lambda,a}|^2 d\operatorname{Vol} \right| \\ &\geq r^n m(\lambda) - \left(cr^n \lambda^{-(n-1)} m(\lambda)^2 + cr^n m(\lambda)^2 \right) \\ &\geq r^n m(\lambda) + O(r^n m(\lambda)^2) \\ &\geq r^n m(\lambda)/2, \end{aligned}$$

for sufficiently large λ (since $m(\lambda) \rightarrow 0+$ as $\lambda \rightarrow \infty$). It is then immediate that

$$S_r(m) \in \bigcup_{p=1}^{N_d} S_{r,x_p}(m/2).$$
 (4.10)

Let us consider each case in the theorem separately.

Case (i). All the conditions in Proposition 4.4 are satisfied. In fact, we have a stronger condition that $r \ge \alpha r_1 m(\lambda)^{-\frac{1}{n}} \log(\lambda)^{\frac{1}{2n}}$. Therefore,

$$\begin{split} \mu_{N_W(\lambda)}\left(S_{r,x_p}\left(\frac{m}{2}\right)\right) &\leq \exp\left(-\frac{cr^{2n}m(\lambda)^2}{8r_1^{2n}}\right) \\ &= \exp\left(-\frac{cr^{2n}m(\lambda)^2}{16r_1^{2n}}\right)\exp\left(-\frac{cr^{2n}m(\lambda)^2}{16r_1^{2n}}\right) \\ &\leq \exp\left(-\frac{c\alpha^{2n}\log\lambda}{16}\right)\exp\left(-\frac{cr^{2n}m(\lambda)^2}{16r_1^{2n}}\right) \\ &\leq \lambda^{-\frac{c\alpha^{2n}}{16}}\exp\left(-\frac{cr^{2n}m(\lambda)^2}{16r_1^{2n}}\right). \end{split}$$

Now by (4.10) we have that

$$\mu_{N_W(\lambda)}\left(S_r(m)\right) \le N_d \lambda^{-\frac{c\alpha^{2n}}{16}} \exp\left(-\frac{cr^{2n}m(\lambda)^2}{8r_1^{2n}}\right).$$

Further, notice that from our initial choice of d,

$$N_d = cd^{-n} = cr^{-n}\lambda^{n(n-1)}m(\lambda)^{-2n} \le c\lambda^{3n+n(n-1)}$$

Here, we used the fact that $r, m(\lambda) \gtrsim W^{-1} \geq \lambda^{-1}$. Since N_d grows as a power of λ we can choose α large enough (depending only on the dimension of \mathbb{M}) so

$$N_d \lambda^{-\frac{c\alpha^{2n}}{16}} \le 1.$$

We arrive at the desired estimate that

$$\mu_{N_W(\lambda)}(S_r(m)) \le \exp\left(-\frac{cr^{2n}m(\lambda)^2}{16r_1^{2n}}\right).$$

Case (ii) follows the same reasoning with suitable adjustments (comparing r to r_2) so we omit the proof.

Case (iii). Proposition 4.4 tells us that there exists an o(1) order function $m(\lambda)$ such that for any x,

$$\mu_{N_W(\lambda)}\left(S_{r,x}(m)\right) \le \exp\left(-c\lambda^{n-1}r^{2(n-1)}m(\lambda)^2\right).$$

Provided $m(\lambda) = O(\lambda^{-\beta})$ for some $\beta > 0$, the argument remains the same as in Case (i).¹ If the $m(\lambda)$ extracted from Proposition 4.4 decays faster, we simply pick some $\widetilde{m}(\lambda) \le m(\lambda)$ so that $\widetilde{m}(\lambda) = O(\lambda^{-\beta})$ for some $\beta > 0$. Note that the results of Proposition 4.4 will also hold for $\widetilde{m}(\lambda)$ and we complete the proof. \Box

Publisher's Note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

References

- [B] Berry, M.: Regular and irregular semiclassical wavefunctions. J. Phys. A 10(12), 2083–2091 (1977)
- [BL] Burq, N., Lebeau, G.: Injections de Sobolev probabilistes et applications. Ann. Sci. Éc. Norm. Supér. (4) 46(6), 917–962 (2013)
- [CH] Canzani, Y., Hanin, B.: Scaling limit for the kernel of the spectral projector and remainder estimates in the pointwise Weyl law. Anal. PDE **8**(7), 1707–1731 (2015)
- [CdV] Colin de Verdière, Y.: Ergodicité et fonctions propres du laplacien. Commun. Math. Phys. 102(3), 497–502 (1985)
- [CII] de Courcy-Ireland, M.: Small-scale equidistribution for random spherical harmonics. arXiv:1711.01317
- [CI2] de Courcy-Ireland, M.: Shrinking scale equidistribution for monochromatic random waves on compact manifolds. arXiv:1902.05271
- [CZ] Chang, R., Zelditch, S.: Log-scale equidistribution of zeros of quantum ergodic eigensections. Ann. Henri Poincaré 19(12), 3783–3814 (2018)
- [DG] Duistermaat, J.J., Guillemin, V.W.: The spectrum of positive elliptic operators and periodic bicharacteristics. Invent. Math. 29, 39–79 (1975)
- [Ha1] Han, X.: Small scale quantum ergodicity in negatively curved manifolds. Nonlinearity 28(9), 3263– 3288 (2015)
- [Ha2] Han, X.: Small scale equidistribution of random eigenbases. Commun. Math. Phys. 349(1), 425–440 (2017)
- [He1] Hezari, H.: Applications of small scale quantum ergodicity in nodal sets. Anal. PDE 11(4), 855–871 (2018)
- [He2] Hezari, H.: Inner radius of nodal domains of quantum ergodic eigenfunctions. Proc. Am. Math. Soc. 146(11), 4661–4666 (2018)
- [He3] Hezari, H.: Quantum ergodicity and L^p norms of restrictions of eigenfunctions. Commun. Math. Phys. **357**(3), 1157–1177 (2018)
- [HR1] Hezari, H., Rivière, G.: L^p norms, nodal sets, and quantum ergodicity. Adv. Math. **290**, 938–966 (2016)

¹ This is indeed true for all the cases such that the improvement in the pointwise Weyl law in Theorem 2.4 is known. See e.g. [CH].

- [HR2] Hezari, H., Rivière, G.: Quantitative equidistribution properties of toral eigenfunctions. J. Spectr. Theory 7(2), 471–485 (2017)
- [Ho] Hörmander, L.: The spectral function of an elliptic operator. Acta Math. 121, 193–218 (1968)
- [Le] Ledoux, M.: The Concentration of Measure Phenomenon. American Mathematical Society, Providence, RI (2001)
- [LR] Lester, S., Rudnick, Z.: Small scale equidistribution of eigenfunctions on the torus. Commun. Math. Phys. 350(1), 279–300 (2017)
- [M] Maples, K.: Quantum unique ergodicity for random bases of spectral projections. Math. Res. Lett. 20(6), 1115–1124 (2013)
- [Sa] Safarov, Y.: Asymptotics of a spectral function of a positive elliptic operator without a nontrapping condition, Funktsional. Anal. i Prilozhen. 22:3 (1988), 53–65, 96. In: Russian: translated in Funct. Anal. Appl. 22:3 (1988), 213–223
- [Sn] Shnirelman, A.: The asymptotic multiplicity of the spectrum of the Laplace operator. Uspehi Mat. Nauk 30(4 (184)), 265–266 (1975)
- [So1] Sogge, C.: Localized L^p -estimates of eigenfunctions: a note on an article of Hezari and Rivière. Adv. Math. **289**, 384–396 (2016)
- [So2] Sogge, C.: Problems related to the concentration of eigenfunctions. Journées équations aux dérivées partielles, Volume (2015), p. 1–11
- [So3] Sogge, C.: Fourier Integrals in Classical Analysis, 2nd edn. Cambridge University Press, Cambridge (2017)
- [STZ] Sogge, C., Toth, J., Zelditch, S.: About the blowup of quasimodes on Riemannian manifolds. J. Geom. Anal. 21(1), 150–173 (2011)
- [SZ] Sogge, C., Zelditch, S.: Riemannian manifolds with maximal eigenfunction growth. Duke Math. J. 114(3), 387–437 (2002)
- [SZ2] Sogge, C., Zelditch, S.: A note on *L^p*-norms of quasi-modes. Some topics in harmonic analysis and applications, 385–397, Adv. Lect. Math. (ALM), 34, Int. Press, Somerville, MA (2016)
- [T] Tacy, M.: A note on constructing sharp examples for L^p norms of eigenfunctions and quasimodes near submanifolds. Proc. Am. Math. Soc. 146, 2909–2924 (2018)
- [U] Uhlenbeck, K.: Generic properties of eigenfunctions. Am. J. Math. 98(4), 1059–1078 (1976)
- [Z1] Zelditch, S.: Uniform distribution of eigenfunctions on compact hyperbolic surfaces. Duke Math. J. 55(4), 919–941 (1987)
- [Z2] Zelditch, S.: A random matrix model for quantum mixing. Int. Math. Res. Not. 3, 115–137 (1996)
- [Z3] Zelditch, S.: Real and complex zeros of Riemannian random waves. Spectral analysis in geometry and number theory, 321–342, Contemp. Math., 484, Am. Math. Soc., Providence, RI (2009)
- [Z4] Zelditch, S.: Quantum ergodicity of random orthonormal bases of spaces of high dimension. Philos. Trans. R. Soc. Lond. Ser. A Math. Phys. Eng. Sci. 372 (2014), no. 2007
- [Z5] Zelditch, S.: Logarithmic lower bound on the number of nodal domains. J. Spectr. Theory 6(4), 1047–1086 (2016)

Communicated by J. Marklof