Measures on spaces of Riemannian metrics, IAS, Princeton, July 2, 2014

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[CJW]: Y. Canzani, I. Wigman, DJ: arXiv:1002.0030, Jour. of Geometric Analysis, 2013
[CCKJST]: B. Clarke, N. Kamran, L. Silberman, J. Taylor,

Y. Canzani, DJ: arXiv:1309.1348

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Level spacings conjectures (Bohigas, Gianoni, Schmit, 1984): spacings between eigenvalues of the Dirichlet Laplacian for ergodic (mixing) 2-dimensional billiards follow the GOE statistics.

Manifold version: *M* - compact connected negatively curved manifold of dimension *n*; $0 = \lambda_0 < \lambda_1 \leq \lambda_2 \leq \ldots$ - the eigenvalues of the Laplacian Δ . Let $\mu_i = \lambda_i^{n/2}$ ("unfolding"). Then

$$\lim_{n\to\infty}\sum_{k=1}^n\delta(\mu_k-\mu_{k-1})\to d\mu_{GOE}.$$

Conjecture *fails* for arithmetic hyperbolic manifolds (Luo-Sarnak), because of high multiplicity in the length spectrum; but was confirmed numerically for *generic* negatively curved manifolds, and for certain *manifolds with boundary*. *Few* rigorous results. One such result (Jakobson, Zelditch; idea suggested by Sarnak): On a negatively curved compact surface *S*, level spacings distribution for eigenvalues of the Laplacian Δ (if they exist); and the level spacings distribution for eigenvalues of the Schrödinger operator $\Delta + V$ do not depend on V.

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The question seems *difficult* to answer for *every* Riemannian metric;

Idea: can we prove some results in this direction by *averaging* over different Riemannian metrics?

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Previous related results due to Sarnak and Vanderkam, averaging over spaces of flat tori.

Fix a compact smooth Riemannian manifold M^n . We shall discuss several measures on manifolds of metrics on M.

• Measures on conformal classes of metrics: concentrated near a reference metric g_0 , supported on regular (e.g. Sobolev, real-analytic) metrics a.s. Applications to the study of Gauss curvature.

• Measures on manifolds of metrics with the fixed *volume form*, applications to the study of L^2 (Ebin) distance function, and to integrability of the diameter, eigenvalue and volume entropy functionals.

• **Remark:** All measures are invariant by the action of diffeomorphisms.

Conformal class: g_0 - reference metric on M. Conformal class of g_0 : { $g_1 = e^f \cdot g_0$ }; f is a random (suitably regular) function on M.

 Δ_0 - Laplacian of g_0 . Spectrum: $\Delta_0\phi_j + \lambda_j\phi_j = 0, \ 0 = \lambda_0 < \lambda_1 \le \lambda_2 \le \dots$ Define *f* by

$$f(\mathbf{x}) = -\sum_{j=1}^{\infty} a_j c_j \phi_j(\mathbf{x}),$$

 $a_j \sim \mathcal{N}(0, 1)$ are i.i.d standard Gaussians, $c_j = F(\lambda_j) \rightarrow 0$ (*damping*):

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The covariance function $r_f(x, y) := \mathbb{E}[f(x)f(y)] = \sum_{j=1}^{\infty} c_j^2 \phi_j(x) \phi_j(y)$, for $x, y \in M$. For $x \in M$, f(x) is mean zero Gaussian of variance $r_f(x, x) = \sum_{j=1}^{\infty} c_j^2 \phi_j(x)^2$.

Examples:

• Random *Sobolev* metric: $c_j = \lambda_j^{-s}$, \Longrightarrow

 $r_f(x,y) = \sum_j rac{\phi_j(x)\phi_j(y)}{\lambda_i^{2s}}$, spectral zeta function.

• Random *real-analytic* metric $c_j = e^{-\lambda_j t}$, \implies $r_f(x, y) = \sum_j \phi_j(x)\phi_j(y)e^{-2\lambda_j t}$, heat kernel.

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Sobolev regularity: If

$$\mathbb{E}||f||_{H^s}^2 = \sum_j c_j^2 (1+\lambda_j)^s < \infty$$

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then $f \in H^{s}(M)$ a.s. Weyl's law + Sobolev embedding imply **Proposition:** If $c_{j} = O(\lambda_{j}^{-s})$, s > n/2, then $f \in C^{0}$ a.s; if $c_{j} = O(\lambda_{j}^{-s})$, s > n/2 + 1, then $f \in C^{2}$ a.s.

- [CJW]: Let n = 2, and let g_0 have non-vanishing Gauss curvature ($M \neq T^2$). Can estimate the probability that after a random conformal perturbation, the Gauss curvature will change sign somewhere on M.
- Techniques: curvature transformation in 2*d* under conformal changes of metrics, large deviation estimates (Borell, Tsirelson-Ibragimov-Sudakov, Adler-Taylor).
- $n \ge 3$: related results for scalar curvature and *Q*-curvature.

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- ► Random (Sobolev) embeddings into R^k: 1-dimensional i.i.d. Gaussians → k-dimensional i.i.d. Gaussians.
- ► F. Morgan (1979): $M = S^1$, k = 3: a.s. results about minimal surfaces spanned by random "knots."
- ► F. Morgan (1982): general compact *M*, a.s. Whitney embedding theorems + applications.
- We shall use similar ideas to define Gaussian measures on manifolds of metrics with *fixed volume form*; transverse to conformal classes.

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Metrics = sections of Pos(M) ⊂ Sym(M) ⊂ Hom(TM, T*M) (positive-definite, symmetric maps); symmetric matrices in local coordinates. GL(T_xM) acts on Pos_x(M) with stabilizer isomorphic to O(n).

Fix a volume form v, consider $Met_v(M)$. $SL(T_xM)$ acts on the fibre $Pos_x^v(M)$ by

$$h.g_x = h^T \circ g_x \circ h;$$

stabilizer isomorphic to SO(n). We have

 $\operatorname{Pos}_{X}^{V}(M) \cong \operatorname{SL}_{n}(\mathbf{R})/\operatorname{SO}_{n}$

Fix $g^0 \in \operatorname{Met}_V$; $dv(x) = \sqrt{|\det g^0(x)|} dx_1 \wedge \ldots \wedge dx_n$. Let $G_x = \operatorname{SL}(T_x M), K_x = \operatorname{SO}(g_x^0)$.

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► Fix $g^0 \in \operatorname{Met}_V$; $dV(x) = \sqrt{|\det g^0(x)|} dx_1 \wedge \ldots \wedge dx_n$. Let $G_x = \operatorname{SL}(T_x M), K_x = \operatorname{SO}(g_x^0)$.

f_x frame in *T_xM* orthonormal w.r.to *g⁰_x*, *A_x* ⊂ *G_x* positive diagonal matrices of determinant 1 (w.r. to *f_x*).
 Every *g¹_x* ∈ Pos^v_x(*M*) can be represented as

$$g_x^1=(k_xa_x)g_x^0,\qquad k_x\in K_x,a_x\in A_x;$$

unique up to S_n acting on f_x .

- ► Assumption: *M* is *parallelizable* (∃ global section of the frame bundle). Examples:
 - All 3-manifolds;
 - All Lie groups;
 - The frame bundle of any manifold;

• The sphere S^n iff $n \in \{1, 3, 7\}$.

Necessary condition: vanishing of the 2nd Stiefel-Whitney class. For orientable \iff spin.

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M parallelizable. Choose a global section f^0 of the frame bundle orthonormal w.r. to g^0 .

To define g_x^1 , we apply to f^0 a composition of a rotation $k_x \in SO(T_xM)$ and a diagonal unimodular transformation $a_x \in SL(T_xM)$ which will define an orthonormal basis f_x^1 for g_x^1 . By construction, g^0 and g^1 will have the same volume form. We let $a_x = \exp(H(x))$, where $H : M \to \mathfrak{a} \cong \mathbb{R}^{n-1}$ is the Lie algebra of $\operatorname{Diag}_0(n) \subset SL_n$. Similarly, let $k_x = \exp h(x)$, where $h : M \to \mathfrak{so}_n$, the Lie algebra of SO_n . Choice of $g^0 +$ parallelizability makes the above construction well-defined. We define Gaussian measures on $\{H : M \to \mathfrak{a}\}$ and $\{h : M \to \mathfrak{so}_n\}$ as in Morgan. In the sequel, only need *H*; constructions are analogous.

Let

$$H(x) = \sum_{j=1}^{\infty} \pi_n(A_j) \beta_j \psi_j(x), \qquad (1)$$

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where

- $\Delta \psi_j + \lambda_j \psi_j = 0;$
- *A_j* are i.i.d standard Gaussians in **R**^{*n*};

• $\pi_n : \mathbf{R}^n \to \{x \in \mathbf{R}^n : x \cdot (1, \dots, 1) = 0\} \simeq \mathbf{R}^{n-1}$ - projection into the hyperplane $\sum_{j=1}^n x_j = 0$; • $\beta_j = F_2(\lambda_j) > 0$, where $F_2(t)$ is (eventually) monotone decreasing function of $t, F(t) \to 0$ as $t \to \infty$.

- ▶ Smoothness: Morgan showed **Proposition 1:** If $\beta_j = O(j^{-r})$ where $r > (q + \alpha)/n + 1/2$, then *H* converges a.s. in $C^{q,\alpha}(M, \mathbb{R}^{n-1})$.
- ▶ Proposition 1 + Weyl's law \implies **Proposition 2:** If $\beta_j = O(\lambda_j^{-s})$ where s > q/2 + n/4, then *H* converges a.s. in $C^q(M, \mathbb{R}^{n-1})$.

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Lipschitz distance ρ :

$$\rho(g_0, g_1) = \sup_{x \in M} \sup_{0 \neq \xi \in T_x M} \left| \ln \frac{g_1(\xi, \xi)}{g_0(\xi, \xi)} \right|$$
(2)

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A related expression appeared in the paper by Bando-Urakawa. If $g_1(x) = (k(x)d(x))g_0(x)$ then ρ only depends on the diagonal part d(x).

Tail estimate for ρ :

One can show that for large R,

 $\operatorname{Prob}\{\rho(g_0,g_1)>R\} \le 2^n(n+\epsilon) \cdot \operatorname{Prob}\{\sup_{x\in M} d_1(x)>R/2\} \quad (3)$

Definition of d_1 :

Recall from (1): $H(x) = \sum_{j=1}^{\infty} \pi_n(A_j)\beta_j\psi_j(x)$. Define $D(x) = (d_1(x), \dots, d_n(x))$ by

$$D(x) = \sum_{j=1}^{\infty} A_j \beta_j \psi_j(x)$$

("don't project A_j "). The covariance function for $d_1(x)$:

$$r_{d_1}(x,y) = \sum_{k=1}^{\infty} \beta_k^2 \psi_k(x) \psi_k(y),$$

Define σ^2 by $\sigma^2 := \sigma(d_1)^2 := \sup_{x \in M} r_{d_1}(x, x).$ (4)

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Borell-TIS theorem applied to the random field d_1 implies **Proposition 3.** Let σ be as in (4). Then

$$\lim_{R\to\infty}\frac{\ln\operatorname{Prob}\{\rho(g_0,g_1)>R\}}{R^2}\leq\frac{-1}{8\sigma^2}.$$
(5)

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More precise result:

Proposition 4. There exists $\alpha > 0$ such that for a fixed $\epsilon > 0$ and for large enough *R*, we have

$$\operatorname{Prob}\{\rho(g_1,g_0)>R\}\leq 2^n(n+\epsilon)\exp\left(\frac{\alpha R}{2}-\frac{R^2}{8\sigma^2}\right).$$

ρ controls diameter and eigenvalues: Proposition 5.

Assume that $d \operatorname{vol}(g_0) = d \operatorname{vol}(g_1)$ and $\rho(g_0, g_1) < R$. Then

$$e^{-R} \le rac{\operatorname{diam}(M, g_1)}{\operatorname{diam}(M, g_0)} \le e^R$$
 (6)

and

$$e^{-2R} \leq rac{\lambda_k(\Delta(g_1))}{\lambda_k(\Delta(g_0))} \leq e^{2R}.$$
 (7)

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Propositions 4 and 5 imply **Theorem 6.**

Let $h : \mathbb{R}^+ \to \mathbb{R}^+$ be a monotonically increasing function such that for some $\delta > 0$

$$h(e^{y}) = O\left(\exp\left[y^2(1/(8\sigma^2) - \delta)\right]\right).$$

Then $h(\text{diam}(g_1))$ is integrable with respect to the probability measure $d\omega(g_1)$ constructed earlier.

and

Theorem 7. Let $h : \mathbb{R}^+ \to \mathbb{R}^+$ be a monotonically increasing function such that for some $\delta > 0$

$$h(e^{2y}) = O\left(\exp\left[y^2(1/(8\sigma^2) - \delta)\right]\right).$$

Then $h(\lambda_k(\Delta(g_1)))$ is integrable with respect to the probability measure $d\omega(g_1)$ constructed earlier. etc

Similar results can be established for volume entropy,

$$h_{vol} = \lim_{s \to \infty} \frac{\ln \operatorname{vol} B(x,s)}{s}$$

 L² or *Ebin* distance between can be computed as follows [Ebin, Clarke]:

$$\Omega_2^2(g^0,g^1) := \int_M d_{2,x}(g^0(x),g^1(x))^2 dv(x)$$

where $d_{2,x}(g^0(x), g^1(x))$ is the distance in SL_n/SO_n ;

$$= \int_M \langle H(x), H(x) \rangle_{g^0(x)} dv(x).$$

▶ For us Ω_2^2 is a *random variable* whose distribution we shall compute. Note: only depends on *H*, hence it suffices to consider the Gaussian measure on $\{H : M \rightarrow a\}$.

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In local coordinates, let

$$a_x = ext{diag}(ext{exp}(b_1(x)), ext{exp}(b_2(x)), \dots, ext{exp}(b_n(x))),$$

where $\sum_{j=1}^n b_j(x) = 0, orall x \in M$. Then

$$d_x(g_x^0,g_x^1)^2 = \sum_{j=1}^n b_j(x)^2.$$

Accordingly,

$$\Omega_2(g^0,g^1)^2 = \int_M \left(\sum_{j=1}^n b_j(x)^2\right) dv(x).$$

 $\pi_n : \mathbf{R}^n \to \{x : \sum x_j = 0\}$ standard projection. P_n - matrix of π_n (in the usual basis of \mathbf{R}^n) with singular values

$$(1,\ldots,1,0):=\mu_{i,n}, 1\leq i\leq n.$$

Then in distribution

$$\Omega_2^2 \stackrel{D}{=} \sum_j \beta_j^2 \sum_{i=1}^n \mu_{i,n}^2 W_{i,j}$$

where $W_{i,j} \sim \chi_1^2$ are i.i.d. We get $\Omega_2^2 \stackrel{D}{=} \sum_j \beta_j^2 V_j$ where $V_j \sim \chi_{n-1}^2$ are i.i.d.

Theorem 8. Moment generating function of Ω_2^2 :

$$M_{\Omega_2^2}(t) = E(\exp(t\Omega_2^2)) = \prod_j \prod_{i=1}^n M_{\chi_1^2}(t\mu_{i,n}^2\beta_j^2)$$

$$=\prod_{j}\prod_{i=1}^{n}(1-2t\mu_{i,n}^{2}\beta_{j}^{2})^{-1/2}=\prod_{j}(1-2t\beta_{j}^{2})^{-(n-1)/2}$$

Characteristic function of Ω_2^2 :

$$\prod_{j}\prod_{i=1}^{n}(1-2it\mu_{i,n}^{2}\beta_{j}^{2})^{-1/2}=\prod_{j}(1-2it\beta_{j}^{2})^{-(n-1)/2}$$

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Corollary 9. Tail estimates for Ω_2^2 : applying results of Laurent-Massart, one can show that

$$\operatorname{Prob}\{\Omega_2^2 \ge s^2\} \le \exp(-s^2/(2\beta_1^2)).$$

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