

Zeros of L -functions near the Central Point and Optimal Test Functions

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Summary

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 - ◇ Riemann zeta function, general L -functions, RMT
 - ◇ nuclear physics, Birch and Swinnerton Dyer conjecture, elliptic curve cryptography.
 - ◇ n -level densities, Katz-Sarnak determinants

Example: Riemann Zeta Function

Riemann Zeta Function

$$\zeta(\mathbf{s}) = \sum_{n=1}^{\infty} \frac{1}{n^{\mathbf{s}}} = \prod_{p \text{ prime}} \frac{1}{1-p^{-\mathbf{s}}} \text{ for } \Re(\mathbf{s}) > 1.$$

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Functional Equation

$$\zeta(s) = 2^s \pi^{s-1} \sin\left(\frac{\pi s}{2}\right) \Gamma(1-s) \zeta(1-s) \text{ for } s \in \mathbb{C} \setminus \{1\}.$$

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Riemann Hypothesis

All nontrivial zeros (not negative even integers) of ζ are of the form $\gamma = \frac{1}{2} + i\sigma$ with $\sigma \in \mathbb{R}$.

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- Euler product:

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- meromorphic continuation to \mathbb{C} , of finite order, at most finitely many poles (all on the line $\Re(s) = 1$).
- Functional equation: $\omega \in \mathbb{R}$, $G(s)$ product of Γ -fns:

$$e^{i\omega} G(s)L(s, f) = e^{-i\omega} \overline{G(1 - \bar{s})L(1 - \bar{s})}.$$

Applications

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- zeros \longleftrightarrow primes.
- Birch and Swinnerton-Dyer conjecture.
 - ◇ Elliptic curve cryptography.

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- Ensembles of matrices (e.g. real symmetric, Hermitian) with entries drawn from probability distribution.
- Study distribution of normalized eigenvalues for given ensemble.

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Applications

- Energy levels of heavy nuclei.
- Elliptic curve cryptography.

1-level Density

Family of Dirichlet L -functions $L(s, f)$ indexed by cuspidal newform $f \in \mathcal{F}$. Riemann hypothesis \implies zeros of $L(s, f)$ are of the form $\rho_f = \frac{1}{2} + i\gamma_f$ with $\gamma_f \in \mathbb{R}$.

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1-level Density

$D(f; \phi) := \sum_{\gamma_f} \phi\left(\frac{\gamma_f}{2\pi} \log(c_f)\right)$ where $\phi \geq 0$ is even, Schwartz, $\hat{\phi}$ compactly supported, $\phi(0) > 0$. $c_f > 1$ is the analytic conductor.

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Idea:

Varying ϕ , $D(f; \phi)$ measures density of zeros of $L(s, f)$ near central point $s = \frac{1}{2}$.

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Then take a limit:

$$\lim_{Q \rightarrow \infty} \mathbb{E}(\mathcal{F}(Q); \phi) = \int_{-\infty}^{\infty} \phi(x) W(\mathcal{F})(x) dx$$

where $W(\mathcal{F})$ is a distribution depending on \mathcal{F} .

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From RMT, $W(\mathcal{F})$ is dependent on a symmetry group $G = G(\mathcal{F})$ of \mathcal{F} . Write $W(\mathcal{F}) = W_{1,G}$. Some examples:

$$W_{1,O}(x) = 1 + \frac{1}{2}\delta(x)$$

$$W_{1,SO(E)}(x) = 1 + \frac{\sin(2\pi x)}{2\pi x}$$

$$W_{1,SO(O)}(x) = 1 - \frac{\sin(2\pi x)}{2\pi x} + \delta(x)$$

1-level Density

Quantity of interest

$\lim_{Q \rightarrow \infty} \text{AveRank}(\mathcal{F}(Q))$, where $\text{AveRank}(\mathcal{F}(Q))$ is average order of vanishing of the L -functions with $f \in \mathcal{F}(Q)$.

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Can show that

$$\lim_{Q \rightarrow \infty} \text{AveRank}(\mathcal{F}(Q)) \leq \frac{\int_{-\infty}^{\infty} \phi(x) W_{1,G}(x) dx}{\phi(0)}$$

n -level Density

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$$D_n(f; \phi) := \sum_{\substack{\gamma_{j,f} \\ |j| \text{ distinct}}} \phi \left(\frac{\gamma_{1,f}}{2\pi} \log(C_f), \frac{\gamma_{2,f}}{2\pi} \log(C_f), \dots, \frac{\gamma_{n,f}}{2\pi} \log(C_f) \right)$$

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Higher Dimensional Bound

$$\lim_{Q \rightarrow \infty} \text{WeightedAveRank}(\mathcal{F}(Q)) \leq \frac{\int_{\mathbb{R}^n} \phi(x) W_{n,G}(x) dx_1 \cdots dx_n}{\phi(\mathbf{0})}$$

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Goal

Higher level densities give stronger bound. Minimize right-hand side over admissible ϕ for n as large as possible.

Katz-Sarnak Determinants

Set $K_\epsilon(x, y) := \frac{\sin(\pi(x-y))}{\pi(x-y)} + \epsilon \frac{\sin(\pi(x+y))}{\pi(x+y)}$, $\epsilon \in \{0, \pm 1\}$. n -level weights for classical compact groups are:

$$W_{n, \text{SO}(\mathbb{E})}(x) = \det (K_1(x_i, x_j))_{i, j \leq n}$$

$$W_{n, \text{SO}(\mathbb{O})}(x) = \det (K_{-1}(x_i, x_j))_{i, j \leq n} + \sum_{k=1}^n \delta(x_k) \det (K_{-1}(x_i, x_j))_{i, j \neq k}$$

$$W_{n, \mathbb{O}}(x) = \frac{1}{2} W_{n, \text{SO}(\mathbb{E})}(x) + \frac{1}{2} W_{n, \text{SO}(\mathbb{O})}(x)$$

$$W_{n, \mathbb{U}}(x) = \det (K_0(x_i, x_j))_{i, j \leq n}$$

$$W_{n, \text{Sp}}(x) = \det (K_{-1}(x_i, x_j))_{i, j \leq n}$$

Main Results

Main Idea

Restrict domain to only those ϕ which are linear combinations of single variable test functions:

$$\phi(\mathbf{x}) = \sum_{j=1}^m \phi_{j,1}(x_1) \cdots \phi_{j,n}(x_n).$$

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Main Result 1

- 1 Choosing first $n - 1$ terms $\phi_{j,1}, \dots, \phi_{j,n-1}$ carefully, can integrate first $n - 1$ variables to obtain new weight function of same form as 1-dimensional weights.
- 2 1-level case already solved, so choose ϕ_n optimally for new weight.

Main Results

Main Idea

Restrict domain to only those ϕ which are linear combinations of single variable test functions:

$$\phi(\mathbf{x}) = \sum_{j=1}^m \phi_{j,1}(x_1) \cdots \phi_{j,n}(x_n).$$

Main Result 2

- 1 Using functional analysis, reduce to a problem similar to the 1-level case.
- 2 Deal with the added difficulties of poorly behaved Fourier transforms of weight functions.
- 3 Obtain stronger bound than Main Result 1.

1-level Case

2 Steps

- 1 Reduce problem to different functional analysis optimization problem.

1-level Case

2 Steps

- 1 Reduce problem to different functional analysis optimization problem.
- 2 Use functional analysis tools to solve reduced problem.

Step 1: Reduce Problem

Assume $\text{supp}(\hat{\phi}) \subset [-1, 1]$. Plancherel on numerator, taking then inverting Fourier transform in denominator:

$$\frac{\int_{-\infty}^{\infty} \phi(x) W_{1,G}(x) dx}{\phi(\mathbf{0})} = \frac{\int_{-1}^1 \hat{\phi}(\xi) \widehat{W}_{1,G}(\xi) d\xi}{\int_{-1}^1 \hat{\phi}(\xi) d\xi}.$$

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Ahiezer's Theorem and the Paley-Wiener Theorem show ϕ admissible $\iff \hat{\phi}(\xi) = (g * \check{g})(\xi)$ for some $g \in L^2[-\frac{1}{2}, \frac{1}{2}]$, where $\check{g}(\xi) = \overline{g(-\xi)}$.

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Step 1: Reduce Problem

Some functional analysis: define compact, self-adjoint linear operator $K : L^2[-\frac{1}{2}, \frac{1}{2}] \rightarrow L^2[-\frac{1}{2}, \frac{1}{2}]$

$$(Kg)(x) = \int_{-\frac{1}{2}}^{\frac{1}{2}} m(x-y)g(y) dy.$$

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Some manipulations:

$$\frac{\int_{-1}^1 \hat{\phi}(\xi) \widehat{W_{1,G}}(\xi) d\xi}{\int_{-1}^1 \hat{\phi}(\xi) d\xi} = \frac{\int_{-1}^1 (g * \check{g})(\xi)(\delta(\xi) + m(\xi)) d\xi}{\int_{-1}^1 (g * \check{g})(\xi) d\xi}$$

Step 1: Reduce Problem

$$\begin{aligned}
 &= \frac{\int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-1}^1 \left(\delta(\xi) g(\xi + y) \overline{g(y)} + m(\xi) g(\xi + y) \overline{g(y)} \right) d\xi dy}{\int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-1}^1 g(\xi + y) \overline{g(y)} d\xi dy} \\
 &= \frac{\langle g, g \rangle_{L^2} + \int_{-1}^1 \int_{-\frac{1}{2}+\xi}^{\frac{1}{2}+\xi} m(\xi) g(y) \overline{g(-\xi + y)} dy d\xi}{\int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-1}^1 g(\xi + y) d\xi \overline{g(y)} dy} \\
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Step 1: Reduce Problem

$$\begin{aligned}
 &= \frac{\langle g, g \rangle_{L^2} + \int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-\frac{1}{2}}^{\frac{1}{2}} m(\xi - y) g(y) dy \overline{g(\xi)} d\xi}{\int_{-\frac{1}{2}}^{\frac{1}{2}} \int_{-1}^1 g(\xi + y) d\xi \overline{g(y)} dy} \\
 &= \frac{\langle g, g \rangle_{L^2} + \langle Kg, g \rangle_{L^2}}{\langle g, \mathbf{1} \rangle_{L^2} \langle \mathbf{1}, g \rangle_{L^2}} \\
 &= \frac{\langle (I + K)g, g \rangle_{L^2}}{|\langle \mathbf{1}, g \rangle_{L^2}|^2}
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 \end{aligned}$$

New Problem

Defining $R : L^2[-\frac{1}{2}, \frac{1}{2}] \rightarrow L^2[-\frac{1}{2}, \frac{1}{2}]$ by $R(g) := \frac{\langle (I+K)g, g \rangle_{L^2}}{|\langle \mathbf{1}, g \rangle_{L^2}|^2}$,
 minimize R over subset of $L^2[-\frac{1}{2}, \frac{1}{2}]$ with denominator $\neq 0$.

Step 2: Minimization

Some observations:

- $R(g) \geq \lim_{Q \rightarrow \infty} \text{AveRank}(\mathcal{F}(Q)) \geq 0.$
- Spectral Theorem \implies orthonormal basis of eigenvectors of K , eigenvalues λ_j .
- $\lambda_j \geq -1.$

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Case 1: Eigenvalue (-1)

If \exists a (-1) -eigenvector $f_0 \in L^2[-\frac{1}{2}, \frac{1}{2}]$ not orthogonal to 1, then $R(f_0) = \frac{\langle (I+K)f_0, f_0 \rangle_{L^2}}{|\langle 1, f_0 \rangle_{L^2}|^2} = \frac{\langle f_0, f_0 \rangle_{L^2} - \langle f_0, f_0 \rangle_{L^2}}{|\langle 1, f_0 \rangle_{L^2}|^2} = 0.$

Step 2: Minimization

Case 2: $\lambda_j > -1$ for all j . More functional analysis!

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- Fredholm Theory $\implies \exists f_0 \in L^2[-\frac{1}{2}, \frac{1}{2}]$ satisfying $(I + K)f_0 = 1$.
- $A := \langle 1, f_0 \rangle = \langle (I + K)f_0, f_0 \rangle_{L^2} > 0$.

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For $g = f_0 + h \in L^2[-\frac{1}{2}, \frac{1}{2}]$ with $\langle 1, g \rangle_{L^2} \neq 0$, WLOG $\langle 1, g \rangle_{L^2} = A$. Then $\langle 1, h \rangle_{L^2} = 0$, so

$$\begin{aligned} R(g) &= \frac{\langle 1, f_0 \rangle_{L^2} + \langle (I + K)h, h \rangle_{L^2} + \langle 1, h \rangle_{L^2} + \langle h, 1 \rangle_{L^2}}{|A|^2} \\ &= \frac{A + \langle (I + K)h, h \rangle_{L^2} + 0 + 0}{|A|^2} \geq \frac{1}{A} = R(f_0) \end{aligned}$$

$n \geq 2$

Challenges

- 1 Don't have Ahiezer/Paley-Wiener to write $\hat{\phi} = g * \check{g}$.
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- 1 Don't have Ahiezer/Paley-Wiener to write $\hat{\phi} = g * \check{g}$.
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A Solution

Restrict to minimizing over $\phi(x) = \sum_{j=1}^m \phi_{j,1}(x_1) \cdots \phi_{j,n}(x_n)$ with $\phi_{j,k}$ as in 1-level case.

Approach 1

Outline

- 1 Choose $\phi_{j,1}, \dots, \phi_{j,n-1}$ nicely to obtain new weight function similar to 1-level weight after integrating first $n - 1$ variables.

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- 1 Choose $\phi_{j,1}, \dots, \phi_{j,n-1}$ nicely to obtain new weight function similar to 1-level weight after integrating first $n - 1$ variables.
- 2 Use 1-level approach to minimize over $\phi_{j,n}$.

Approach 1 Example: $W_{2,U}$

Problem

Minimize

$$\frac{\int_{\mathbb{R}^2} \phi_1(x_1)\phi_2(x_2)W_{2,U}(x) dx_1 dx_2}{\phi_1(0)\phi_2(0)} = \frac{\int_{[-1,1]^2} \hat{\phi}_1(\xi_1)\hat{\phi}_2(\xi_2)\widehat{W_{2,U}}(\xi) d\xi_1 d\xi_2}{\phi_1(0)\phi_2(0)} \text{ over}$$

ϕ_1, ϕ_2 even, Schwartz, with $\text{supp}(\phi_1), \text{supp}(\phi_2) \subset [-1, 1]$.

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$\chi(x) = \chi_{[-1,1]}(x)$. A short computation:

$$W_{2,U}(x) = 1 - \frac{\sin^2(\pi(x_1 - x_2))}{\pi^2(x_1 - x_2)^2}$$

$$\widehat{W_{2,U}}(\xi) = \delta(\xi_1)\delta(\xi_2) + \delta(\xi_1 + \xi_2)(|\xi_1| - 1)\chi(\xi_1)$$

Approach 1 Example: $W_{2,U}$

For ϕ_2 arbitrary,

$$\begin{aligned} \frac{1}{\phi_2(\mathbf{0})} \int_{\xi_2 \in \mathbb{R}} \hat{\phi}_2(\xi_2) \widehat{W_{2,U}}(\xi) d\xi_2 &= \delta(\xi_1) + \frac{\hat{\phi}_2(-\xi_1)}{\phi_2(\mathbf{0})} (|\xi_1| - 1) \chi(\xi_1) \\ &= \delta(\xi_1) + m(\xi_1) \end{aligned}$$

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Choose (for example)

$$\phi_2(\mathbf{y}) = \sqrt{\frac{2}{\pi}} \frac{\sin(\mathbf{y})}{\mathbf{y}} \implies \hat{\phi}_2(\omega) = \eta(\omega) := \begin{cases} 0 & |\omega| > 1 \\ \frac{1}{2} & |\omega| = 1 \\ 1 & |\omega| < 1 \end{cases}$$

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Then

$$m(\xi_1) = (|\xi_1| - 1)\eta(\xi_1)$$

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1-level case \implies optimal ϕ_1 is

$$\phi_1(x_1) = \int_{-1}^1 (g * \check{g})(\xi_1) e^{2\pi i x_1 \xi_1} d\xi_1$$

where $g \in L^2[-\frac{1}{2}, \frac{1}{2}]$ satisfies one of

$$0 = g(x) + \int_{y=-\frac{1}{2}}^{\frac{1}{2}} (|x - y| - 1)g(y) dy$$

$$1 = g(x) + \int_{y=-\frac{1}{2}}^{\frac{1}{2}} (|x - y| - 1)g(y) dy$$

Approach 2

Goal

Get better bound by minimizing over all factors of $\phi(\mathbf{x}) = \sum_{j=1}^m \phi_{j,1}(\mathbf{x}_1) \cdots \phi_{j,n}(\mathbf{x}_n)$ simultaneously.

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- 1 Use similar techniques to 1-level to obtain messier version of R from 1-level case.

Approach 2

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- 3 Minimize.

Approach 2 Step 1

 $\widehat{W}_{n,G}$

Convolution Theorem + induction \implies

$\widehat{W}_{n,G}(\xi) = \delta(x) + \sum_{f,l} f(x) \prod_{(i,j,\epsilon) \in l} \delta(x_i + \epsilon x_j)$ with $\epsilon = \pm 1$;
 $(i, i, -1) \notin l$; and f bounded, real-valued, even.

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Self-adjoint operator $K : L^2([-\frac{1}{2}, \frac{1}{2}]^n) \rightarrow L^2([-\frac{1}{2}, \frac{1}{2}]^n)$

$$(Kg)(x) = \int_{[-\frac{1}{2}, \frac{1}{2}]^n} m(x-y)g(y) dy_1 \cdots dy_n$$

with $m(x) = \sum_{f,l} f(x) \prod_{(i,j,\epsilon) \in l} \delta(x_i + \epsilon x_j)$.

Approach 2 Step 1

For $\phi(\mathbf{x}) = \sum_{j=1}^{\ell} \phi_{j,1}(x_1) \cdots \phi_{j,n}(x_n)$,

$$\frac{\int_{\mathbb{R}^n} \phi(\mathbf{x}) W_{n,G}(\mathbf{x})}{\phi(\mathbf{0})} = \frac{\sum_{j=1}^{\ell} \langle (I+K)g_j, g_j \rangle_{L^2}}{\sum_{j=1}^{\ell} |\langle \mathbf{1}, g_j \rangle_{L^2}|^2}$$

where $g_j(x) = g_{j,1}(x_1) \cdots g_{j,n}(x_n)$.

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Goal of Step 2

Get rid of the sums...

Approach 2 Step 2

Assume $f(x) = f_1(x_1) \cdots f_n(x_n)$, $g(x) = g_1(x_1) \cdots g_n(x_n)$,
 $\langle 1, f \rangle_{L^2} \neq 0$, $\langle 1, g \rangle_{L^2} \neq 0$.

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$$\frac{\langle (I+K)f, f \rangle_{L^2}}{|\langle \mathbf{1}, f \rangle_{L^2}|^2} \leq \frac{\langle (I+K)g, g \rangle_{L^2}}{|\langle \mathbf{1}, g \rangle_{L^2}|^2}$$

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$$\langle (I+K)f, f \rangle_{L^2} (|\langle \mathbf{1}, g \rangle_{L^2}|^2 + |\langle \mathbf{1}, f \rangle_{L^2}|^2) \leq (\langle (I+K)g, g \rangle_{L^2} + \langle (I+K)f, f \rangle_{L^2}) |\langle \mathbf{1}, f \rangle_{L^2}|^2$$

$$\frac{\langle (I+K)f, f \rangle_{L^2}}{|\langle \mathbf{1}, f \rangle_{L^2}|^2} \leq \frac{\langle (I+K)g, g \rangle_{L^2} + \langle (I+K)f, f \rangle_{L^2}}{|\langle \mathbf{1}, g \rangle_{L^2}|^2 + |\langle \mathbf{1}, f \rangle_{L^2}|^2}$$

Approach 2 Step 2

More manipulations...

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$$(\langle (I+K)f, f \rangle_{L^2} + \langle (I+K)g, g \rangle_{L^2}) |\langle \mathbf{1}, g \rangle_{L^2}|^2 \leq \langle (I+K)g, g \rangle_{L^2} (|\langle \mathbf{1}, f \rangle_{L^2}|^2 + |\langle \mathbf{1}, g \rangle_{L^2}|^2)$$

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For $\langle \mathbf{1}, f \rangle_{L^2} \neq 0$ and $\langle \mathbf{1}, g \rangle_{L^2} \neq 0$:

$$\frac{\langle (I+K)f, f \rangle_{L^2}}{|\langle \mathbf{1}, f \rangle_{L^2}|^2} \leq \frac{\langle (I+K)f, f \rangle_{L^2} + \langle (I+K)g, g \rangle_{L^2}}{|\langle \mathbf{1}, f \rangle_{L^2}|^2 + |\langle \mathbf{1}, g \rangle_{L^2}|^2} \leq \frac{\langle (I+K)g, g \rangle_{L^2}}{|\langle \mathbf{1}, g \rangle_{L^2}|^2}$$

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Continuity argument:

- Take $(g_\alpha)_{\alpha=1}^\infty$ sequence of the form $g_\alpha(x) = g_{\alpha,1}(x_1) \cdots g_{\alpha,n}(x_n)$ with $\langle 1, g_\alpha \rangle_{L^2} \neq 0$ and $g_\alpha \rightarrow g$ in L^2 .

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Reduced Problem

Suffices to minimize $R(g) := \frac{\langle (I+K)g, g \rangle_{L^2}}{|\langle 1, g \rangle_{L^2}|^2}$ over $g(x) = g_1(x_1) \cdots g_n(x_n)$ with $\langle 1, g \rangle_{L^2} \neq 0$.

Approach 2 Step 3

Currently in progress!

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More careful treatment of K :

- Working termwise in $L^2([-\frac{1}{2}, \frac{1}{2}]^k)$ for $k < n$.

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- More explicit calculation of $\widehat{W}_{n,G}$?

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