Stolarsky Principle: generalizations, extensions, and applications.

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Spherical cap discrepancy

For $x \in \mathbb{S}^d$, $t \in [-1, 1]$ define spherical caps:

$$C(x,t) = \{ y \in \mathbb{S}^d : \langle x, y \rangle \ge t \}.$$

For a finite set $Z = \{z_1, z_2, ..., z_N\} \subset \mathbb{S}^d$ define

$$D_{cap}(Z) = \sup_{x \in \mathbb{S}^d, t \in [-1,1]} \left| \frac{\# (Z \cap C(x,t))}{N} - \sigma (C(x,t)) \right|.$$

Theorem (Beck, '84)

There exists constants c_d , $C_d > 0$ such that

$$c_d N^{-\frac{1}{2} - \frac{1}{2d}} \le \inf_{\#Z = N} D_{cap}(Z) \le C_d N^{-\frac{1}{2} - \frac{1}{2d}} \sqrt{\log N}.$$



Spherical caps: L^2 discrepancy

Define the spherical cap L^2 discrepancy

$$D_{cap,L^2}(Z) = \left(\int_{\mathbb{S}^d} \int_{-1}^1 \left| \frac{\# \big(Z \cap C(x,t) \big)}{N} \right| - \sigma \big(C(x,t) \big) \right|^2 dt \, d\sigma(x) \right)^{\frac{1}{2}}.$$

Theorem (Beck, '84)

There exists constants c_d , $C_d > 0$ such that

$$c_d N^{-\frac{1}{2} - \frac{1}{2d}} \le \inf_{\#Z = N} D_{cap, L^2}(Z) \le C_d N^{-\frac{1}{2} - \frac{1}{2d}}.$$

Theorem (Stolarsky invariance principle)

For any finite set $Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$

$$\frac{1}{N^2} \sum_{i,j=1}^{N} \|z_i - z_j\| + c_d \left[D_{L^2,cap}(Z) \right]^2 = \text{const}$$
$$= \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \|x - y\| \, d\sigma(x) d\sigma(y).$$

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- Stolarsky '73, Brauchart, Dick '12, DB, Dai, Matzke '18
- $C_d = c_d^{-1} = \frac{1}{2} \int_{\mathbb{S}^d} |p \cdot z| d\sigma(z) = \frac{1}{d} \frac{\omega_{d-1}}{\omega_d} \approx \frac{1}{\sqrt{2\pi d}}.$



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- $C_d = c_d^{-1} = \frac{1}{2} \int_{\mathbb{S}^d} |p \cdot z| d\sigma(z) = \frac{1}{d} \frac{\omega_{d-1}}{\omega_d} \approx \frac{1}{\sqrt{2\pi d}}.$
- Easy corollaries:
 - i.i.d. random points: $\mathbb{E}D^2_{L^2,can}(Z) \lesssim N^{-1}$
 - jittered sampling: $\mathbb{E}D^2_{L^2,cap}(Z) \lesssim N^{-1-\frac{1}{d}}$



Proof:

Discrete energy

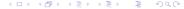
Let $Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$ and let $F : [-1, 1] \to \mathbb{R}$. Discrete energy:

$$E_F(Z) = \frac{1}{N^2} \sum_{i,j=1}^{N} F(z_i \cdot z_j)$$

■ $F(x \cdot y) = -\|x - y\|$: sum of distances (Fejes-Tóth Problem): on \mathbb{S}^2 minimizers known for N = 2, 3, 4, 5, 6 and N = 12.

Questions:

- What are the minimizing configurations?
- Almost minimizers?
- Lower bounds?



Energy integral

Let μ be a Borel probability measure on \mathbb{S}^d . Energy integral

$$I_F(\mu) = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} F(x \cdot y) \, d\mu(x) d\mu(y).$$

i.e.
$$E_F(Z) = I_F\left(\frac{1}{N}\sum \delta_{z_i}\right)$$

Questions:

- What are the minimizers?
- Is σ a minimizer?
- Is it unique?



Other versions, extensions, generalizations

$$(L^2 \text{ discrepancy})^2 = \text{ distance integral } - \text{ sum of distances},$$

$$(L^2 \text{ discrepancy})^2 = \text{ discrete energy } - \text{ energy integral.}$$

$$(L^2 \text{ discrepancy of } \mu)^2 = \text{energy of } \mu - \text{optimal energy}$$

Original generalized version (Stolarsky)

For $x, y \in \mathbb{S}^d$ define the distance $\rho(x, y)$ as

$$\rho(x,y) = \int_{\mathbb{S}^d \min\{x \cdot z, y \cdot z\}}^{\max\{x \cdot z, y \cdot z\}} g(t) dt d\sigma(z).$$

Then for any set $Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$ we have

$$\frac{1}{N^2} \sum_{i,j=1}^{N} \rho(z_i, z_j) + 2D_{L^2, cap, g}^2(Z) = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \rho(x, y) \, d\sigma(x) \, d\sigma(y),$$

where $D_{L^2,\text{cap},g}(Z)$ is the g-weighted L^2 spherical cap discrepancy of Z:

$$D_{L^{2},\operatorname{cap},g}^{2}(Z) = \int_{-1}^{1} g(t) \int_{\mathbb{S}^{d}} \left| \frac{1}{N} \sum_{i=1}^{N} \mathbf{1}_{C(z,t)}(z_{i}) - \sigma(C(p,t)) \right|^{2} d\sigma(z) dt.$$

Original generalized version (Stolarsky)

- g(t) = 1:
 the standard spherical cap L^2 -discrepancy, i.e. $D_{L^2,\text{cap},\mathbf{1}} = D_{L^2,\text{cap}},$ $\rho(x,y) = 2C_d ||x-y|| \text{ is a multiple of the Euclidean}$ distance.
- The metric

$$d_1(x,y) = \int_{\mathbb{S}^d} |d(x,z) - d(y,z)| d\sigma(z),$$

where $d(\cdot, \cdot)$ is the geodesic distance, also has this form with $g(t) = (1 - t^2)^{-1/2}$.



Further generalization: Brauchart

Case n=2:

Let

$$\rho(x,y) = \int\limits_{\mathbb{S}^d} \int\limits_{\min\{x \cdot z, y \cdot z\}}^{\max\{x \cdot z, y \cdot z\}} \int\limits_{\min\{x \cdot z, y \cdot z\}}^{g(t,t')} dt \, dt' \, d\sigma(z).$$

Then

$$\frac{1}{N^2} \sum_{i,j} \rho(z_i, z_j) + 2 \int_{-1}^{1} \int_{-1}^{1} g(t, t') D(Z; t, t') dt dt' = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \rho(x, y) d\sigma d\sigma,$$

where

$$D(Z;t,t') = \int_{\mathbb{S}^d} D(Z,C(z,t))D(Z,C(z,t'))d\sigma(z).$$



Application: even powers of Euclidean distances

Taking
$$g(u) = m(2m - 1)u^{2(m-1)}$$

$$\frac{1}{N^2} \sum_{i,j} \|z_i - z_j\|^{2m} + c_{d,m} \int_{-1}^{1} \int_{-1}^{1} g(|t - t'|) D(Z; t, t') dt dt'$$

$$= \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \|x - y\|^{2m} d\sigma(x) d\sigma(y)$$

Euclidean distance energy integrals

Let μ be a Borel probability measure on the sphere \mathbb{S}^d . For $\lambda > 0$ define the energy integral

$$I_{\lambda} = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \|x - y\|^{\lambda} d\mu(x) d\mu(y)$$

Maximizers (Bjorck '56):

- \bullet 0 < λ < 2: unique maximizer is surface measure,
- $\lambda = 2$: any measure with center of mass at 0,
- $\lambda > 2$: mass $\frac{1}{2}$ at two opposite poles.

"Higher smoothness" (Brauchart and Dick, 2013)

$$K_{\beta}(x,y) = \int_{-1}^{1} \int_{\mathbb{S}^d} (x \cdot z - t)_{+}^{\beta - 1} (y \cdot z - t)_{+}^{\beta - 1} d\sigma(z) dt$$

Notice $K_1(x, y) = 1 - C_d ||x - y||$

$$\begin{split} D^2_{L^2,\beta}(Z) := \int\limits_{-1}^1 \int\limits_{\mathbb{S}^d} \left| \frac{1}{N} \sum_{j=1}^N (z_j \cdot z - t)_+^{\beta - 1} - \int\limits_{\mathbb{S}^d} (x \cdot z - t)_+^{\beta - 1} d\sigma(x) \right|^2 \! dz dt \\ &= \frac{1}{N^2} \sum_{i,j} K_\beta(z_i, z_j) - \int\limits_{\mathbb{S}^d} \int\limits_{\mathbb{S}^d} K_\beta(x, y) d\sigma(x) d\sigma(y). \end{split}$$

Application: odd powers of Euclidean distances

- Notice $K_1(x,y) = 1 C_d ||x y||$
- If $\beta = M$ is a positive integer

$$K_M(x,y) = Q_{M-1}(x,y) + (-1)^M c_{d,M} ||x-y||^{2M-1},$$

where Q_{M-1} is a polynomial of degree M-1.

• For any $Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$

$$D_{L^2,\beta}(Z) \ge c_{\beta,d} N^{-\frac{1}{2} - \frac{\beta - 1/2}{d}}.$$

• For any spherical t-design Z:

$$D_{L^2,\beta}(Z) \le C_{\beta,d} (t^d)^{-\frac{1}{2} - \frac{\beta - 1/2}{d}}.$$



L^1 invariance principle (Skriganov)

Let M be a distance-invariant metric space of diameter π with a fixed probability measure μ .

Consider the "symmetric difference" metric

$$\theta^{\Delta}(\eta, x, y) = \int_0^{\pi} \theta_r^{\Delta}(x, y) \eta(r) dr,$$

where

$$\theta_r^{\Delta}(x,y) = \frac{1}{2}\mu\bigg(B_r(x)\Delta B_r(y)\bigg)$$

Once can also write

$$\theta_r^{\Delta}(x,y) = \frac{1}{2} \int_M \left| \mathbf{1}_{B_r(x)}(z) - \mathbf{1}_{B_r(y)}(z) \right| d\mu(z).$$

L^1 invariance principle (Skriganov)

 L^2 -discrepancy of an N-point set $Z \subset M$:

$$D_{\eta}^{2}(Z) := \int_{0}^{\pi} \int_{M} \left(\frac{1}{N} \cdot \#Z \cap B_{r}(z) - \mu(B_{r}(y)) \right)^{2} d\mu(z) \ \eta(r) dr.$$

L_1 invariance principle

$$D^2_{\eta}(Z) + \frac{1}{N^2} \sum_{x,y,\in Z} \theta^{\Delta}(\eta,x,y) = \int\limits_{M} \int\limits_{M} \theta^{\Delta}(\eta,x,y) d\mu(x) d\mu(y)$$

L^2 invariance principle (Skriganov)

Let Q be a connected compact two-point homogeneous space, i.e. \mathbb{S}^d , \mathbb{RP}^n , \mathbb{CP}^n , \mathbb{HP}^n , or \mathbb{OP}^2 , with geodesic distance θ , normalized so that $\operatorname{diam}(Q) = \pi$. Define the *chordal distance*:

$$\tau(x,y) = \sin \frac{\theta(x,y)}{2}$$
.

For
$$Q = \mathbb{S}^d$$
, $\tau(x,y) = \frac{1}{2} ||x - y||$. For $Q = \mathbb{FP}^n$,

$$\tau(x,y) = \frac{1}{\sqrt{2}} \|\Pi(x) - \Pi(y)\|_F.$$

L^2 invariance principle (Skriganov)

Let $\eta^{\natural}(r) = \sin r$. Then

$$\tau(x,y) = \gamma(Q)\theta^{\Delta}(\eta^{\natural},x,y).$$

■ Therefore, for any N-point set $Z \subset Q$:

$$\gamma(Q) D_{\eta^{\natural}}^{2}(Z) + \frac{1}{N^{2}} \sum_{x,y,\in Z} \tau(x,y) = \int_{M} \int_{M} \tau(x,y) d\mu(x) d\mu(y)$$

• If $Q = \mathbb{S}^d$, then $D_{\eta^{\natural}}(Z) = D_{L^2,cap}(Z)$.



Positive definite functions on the sphere

Lemma

For a function $F \in C[-1,1]$ the following are equivalent:

- i F is positive definite on \mathbb{S}^d .
- ii Gegenbauer coefficients of F are non-negative, i.e.

$$\widehat{F}(n,(d-1)/2) \ge 0$$
 for all $n \ge 0$.

- iii For any signed measure $\mu \in \mathcal{B}$ the energy integral is non-negative: $I_F(\mu) \geq 0$.
- iv There exists a function $f \in L^2_{w_{(d-1)/2}}[-1,1]$ such that

$$F(x \cdot y) = \int_{\mathbb{S}^d} f(x \cdot z) f(z \cdot y) \, d\sigma(z), \quad x, y \in \mathbb{S}^d.$$

v $I_F(\mu) \ge I_F(\sigma) \ge 0$ for any Borel probability measure μ .

Define the L^2 discrepancy of a Borel probability measure μ w.r.t. the function $f: [-1,1] \to \mathbb{R}$ as

$$D_{L^2,f}^2(\mu) = \int\limits_{\mathbb{S}^d} \left| \int\limits_{\mathbb{S}^d} f(x \cdot y) d\mu(y) - \int\limits_{\mathbb{S}^d} f(x \cdot y) d\sigma(y) \right|^2 d\sigma(x).$$

In particular, if $\mu = \frac{1}{N} \sum \delta_{z_i}$

$$D_{L^2,f}^2(Z) = \int_{\mathbb{S}^d} \left(\frac{1}{N} \sum_{i=1}^N f(x \cdot z_i) - \int_{\mathbb{S}^d} f(x \cdot y) d\sigma(y) \right)^2 d\sigma(x).$$

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$$D_{L^2,f}^2(\mu) = \int_{\mathbb{S}^d} \left| \int_{\mathbb{S}^d} f(x \cdot y) d(\mu - \sigma)(y) \right|^2 d\sigma(x).$$

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Theorem (DB, R. Matzke, F. Dai, '18)

Generalized Stolarsky principle:

Let F be positive definite and f as in (iv), then

$$I_F(\mu) - I_F(\sigma) = D_{L^2, f}^2(\mu).$$



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$$I_F(\mu) - I_F(\sigma) = I_F(\mu - \sigma).$$



Discrepancy/energy bounds

Theorem (DB, F. Dai, '19)

Assume that F is positive definite and f as in (iv).

Let
$$Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d \text{ and } \mu = \frac{1}{N} \sum_{i=1}^N \delta_{z_i}$$
.

■ Upper bound:

$$\inf_{\#Z=N} D_{L^2,f}^2(\mu) \lesssim \frac{1}{N} \max_{0 < \theta \lesssim N^{-\frac{1}{d}}} \left(F(1) - F(\cos \theta) \right).$$

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■ Lower bound:

$$\inf_{\#Z=N} D^2_{L^2,f}(\mu) \gtrsim \min_{1 \le k \lesssim N^{1/d}} \widehat{F}(k,\lambda).$$



Stolarsky principle on metric spaces

- lacksquare Ω compact metric space
- $K: \Omega \times \Omega \to \mathbb{R}$ continuous, symmetric, and positive definite, i.e. $\sum_{i,j=1}^{n} K(x_i, x_j) c_i c_j \geq 0$.
- Define $T_K f(x) = \int_{\Omega} K(x, y) f(y) d\mu(y)$
- $T_K \phi_i = \lambda_i \phi_i, \ \lambda_i \ge 0.$

Theorem (Mercer)

$$K(x,y) = \sum_{i=1}^{\infty} \lambda_i \phi_i(x) \phi_i(y),$$

where the sum above converges absolutely and uniformly.



Stolarsky principle on metric spaces

Theorem (DB, O. Vlasiuk, '19?)

Let $K: \Omega \times \Omega \to \mathbb{R}$ continuous, symmetric, and positive definite. Let $\tilde{\mu}$ be the equilibrium measure of I_K and $\operatorname{supp} \mu \subset \operatorname{supp} \tilde{\mu}$.

Then there exists $f: \widetilde{\Omega} \times \widetilde{\Omega} \to \mathbb{R}$ such that

$$I_K(\mu) - I_K(\tilde{\mu}) = D_{L^2, f, \tilde{\mu}}^2(\mu),$$

where

$$D_{L^2,f,\tilde{\mu}}^2(\mu) = \int_{\Omega} \left| \int_{\Omega} f(x,y) d\mu(y) - \int_{\Omega} f(x,y) d\tilde{\mu}(y) \right|^2 d\tilde{\mu}(x).$$

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$$f(x,y) = \sum_{i=1}^{\infty} \sqrt{\lambda_i} \, \phi_i(x) \phi_i(y).$$

Stolarsky principle for hemispheres

Theorem (DB, Dai, Matzke '18; Skriganov '18)

$$D_{L^2,\text{hem}}^2(Z) = \frac{1}{2\pi} \left(\int_{\mathbb{S}^d} \int_{\mathbb{S}^d} d(x, y) \, d\sigma(x) \, d\sigma(y) - \frac{1}{N^2} \sum_{i,j=1}^N d(z_i, z_j) \right)$$

Stolarsky principle for hemispheres

Theorem (DB, Dai, Matzke '18; Skriganov '18)

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Corollary (DB, Dai, Matzke '18)

For any
$$Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$$

$$\frac{1}{N^2} \sum_{i,j=1}^N d(z_i, z_j) \leq \frac{\pi}{2}$$

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$$\frac{1}{N^2} \sum_{i,j=1}^{N} d(z_i, z_j) \le \frac{\pi}{2}$$

For even N:

$$\frac{1}{N^2} \sum_{i,j=1}^{N} d(z_i, z_j) = \frac{\pi}{2} \iff Z - \text{symmetric.}$$



Simple corollaries: a conjecture of Fejes Tóth

Corollary (DB, Dai, Matzke '18)

For odd N:

$$\frac{1}{N^2} \sum_{i,j=1}^{N} d(z_i, z_j) \le \frac{\pi}{2} - \frac{\pi}{2N^2}.$$

Maximum is achieved if and only if $Z = Z_1 \cup Z_2$, where

- \blacksquare Z_1 is symmetric,
- Z_2 lies on a two-dimensional hyperplane (a great circle) and is a maximizer for \mathbb{S}^1 .

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- Fejes-Toth '59: d = 1 and conjectured for $d \ge 2$.
- Sperling, '60 (d = 2, even N)
- Larcher, '61 (d = 2, odd N, ?)
- Nielsen, '65 (d=2)
- Kelly, '69 $(d \ge 2)$



Let H(x) = C(x, 0) denote the hemisphere with center at x and let μ be a Borel probability measure on \mathbb{S}^d .

$$\int\limits_{\mathbb{S}^d} \bigg(\mu\big(H(x)\big) - \frac{1}{2}\bigg)^2 d\sigma(x) = \frac{1}{2\pi} \cdot \bigg(\frac{\pi}{2} - \int\limits_{\mathbb{S}^d} \int\limits_{\mathbb{S}^d} d(x,y) \, d\mu(x) d\mu(y)\bigg).$$

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- For any probability measure μ : $I_{\text{geod}}(\mu) \leq \frac{1}{2}$.
- $I_{\text{geod}}(\mu) = \frac{1}{2}$ (i.e. μ is a maximizer) iff $\mu(H(x)) = \frac{1}{2}$ for σ -a.e. $x \in \mathbb{S}^d$

Let H(x) = C(x, 0) denote the hemisphere with center at x and let μ be a Borel probability measure on \mathbb{S}^d .

$$\int\limits_{\mathbb{S}^d} \bigg(\mu\big(H(x)\big) - \frac{1}{2}\bigg)^2 d\sigma(x) = \frac{1}{2\pi} \cdot \bigg(\frac{\pi}{2} - \int\limits_{\mathbb{S}^d} \int\limits_{\mathbb{S}^d} d(x,y) \, d\mu(x) d\mu(y)\bigg).$$

- For any probability measure μ : $I_{\text{geod}}(\mu) \leq \frac{1}{2}$.
- $I_{\text{geod}}(\mu) = \frac{1}{2}$ (i.e. μ is a maximizer) iff $\mu(H(x)) = \frac{1}{2}$ for σ -a.e. $x \in \mathbb{S}^d$ iff μ is symmetric, i.e. $\mu(E) = \mu(-E)$.



Geodesic distance energy integrals

Let μ be a Borel probability measure on the sphere \mathbb{S}^d . For $\lambda > 0$ define the energy integral

$$I_{\lambda}(\mu) = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \left(d(x, y) \right)^{\lambda} d\mu(x) d\mu(y)$$

Maximizers (DB, Dai, Matzke '18):

- $0 < \lambda < 1$: unique maximizer is σ ,
- $\lambda = 1$: any symmetric measure,
- $\lambda > 1$: mass $\frac{1}{2}$ at two opposite poles.

Euclidean distance energy integrals

Let μ be a Borel probability measure on the sphere \mathbb{S}^d . For $\lambda > 0$ define the energy integral

$$I_{\lambda}(\mu) = \int_{\mathbb{S}^d} \int_{\mathbb{S}^d} \|x - y\|^{\lambda} d\mu(x) d\mu(y)$$

Maximizers (Björck '56):

- $0 < \lambda < 2$: unique maximizer is surface measure,
- $\lambda = 2$: any measure with center of mass at 0,
- $\lambda > 2$: mass $\frac{1}{2}$ at two opposite poles.



Another conjecture of Fejes Tóth

Conjecture (Sum of acute angles)

Let
$$Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$$
 and define

$$F(x \cdot y) = \min \{\arccos(x \cdot y), \pi - \arccos(x \cdot y)\} = \arccos|x \cdot y|,$$

i.e. the acute angle between the lines through $x, y \in \mathbb{S}^d$.

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■ The discrete energy $E_F(Z) = \frac{1}{N^2} \sum_{i,j=1}^N F(z_i \cdot z_j)$ is maximized by the set $Z = \{z_1, ..., z_N\} \subset \mathbb{S}^d$ with $z_i = e_{i \mod (d+1)}$.

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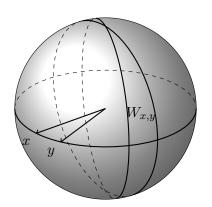
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- $\max I_F(\mu) = I_F(\nu_{ONB}) = \frac{\pi}{2} \cdot \frac{d}{d+1}$, where

$$\nu_{ONB} = \frac{1}{d+1} \sum_{i=1}^{d+1} \delta_{e_i}$$

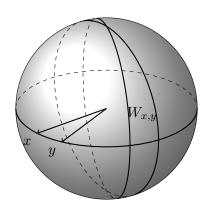
Acute angles: known results

- Partial results for $d \ge 2$ Fodor, Vigh, Zarnocz, '16; DB, R. Matzke, '18
- Known on \mathbb{S}^1
- Stolarsky-type proof: Quadrant discrepancy

$$\sigma(Q(x) \cap Q(y)) = \frac{1}{2} - \frac{1}{\pi} \arccos|x \cdot y|$$
$$D_{L^2,quad}^2(Z) = \frac{1}{4} - \frac{1}{\pi} E_F(Z)$$



$$\begin{split} H_x &= \{z \ : \ \langle z, x \rangle > 0\} \\ W_{xy} &= H_x \triangle H_y \\ &= \{z \in \mathbb{S}^d : \mathrm{sign} \langle z, x \rangle \neq \mathrm{sign} \langle z, y \rangle \} \end{split}$$



$$H_x = \{z : \langle z, x \rangle > 0\}$$

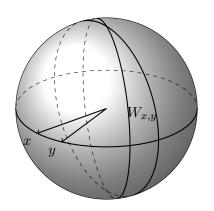
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$$\mathbb{P}(z^{\perp} \text{ separates } x \text{ and } y)$$

$$= \mathbb{P}(\operatorname{sign}\langle z, x \rangle \neq \operatorname{sign}\langle z, y \rangle)$$

$$= \sigma(W_{xy}) = d(x, y)$$



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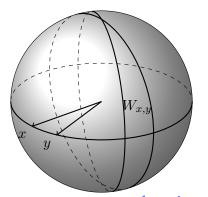
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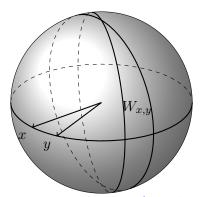
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Define $d_H(x,y) := \frac{1}{N} \cdot \#\{z_k \in Z : \operatorname{sgn}(x \cdot z_k) \neq \operatorname{sgn}(y \cdot z_k)\}$, i.e. the proportion of hyperplanes z_k^{\perp} that separate x and y.



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$$\Delta_Z(x,y) = d_H(x,y) - d(x,y) = \frac{\#(Z \cap W_{xy})}{N} - \sigma(W_{xy})$$

Stolarsky principle for wedge discrepancy (DB, Lacey)

Define the L^2 discrepancy for wedges

$$\left\|\Delta_Z(x,y)\right\|_2^2 = \int\limits_{\mathbb{S}^d} \int\limits_{\mathbb{S}^d} \left(\frac{1}{N} \sum_{k=1}^N \mathbf{1}_{W_{xy}}(z_k) - \sigma(W_{xy})\right)^2 d\sigma(x) d\sigma(y)$$

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Theorem (Stolarsky principle for the tessellation of the sphere)

For any finite set $Z = \{z_1, \ldots, z_N\} \subset \mathbb{S}^d$

$$\|\Delta_{Z}(x,y)\|_{2}^{2} = \frac{1}{N^{2}} \sum_{i,j=1}^{N} \left(\frac{1}{2} - d(z_{i}, z_{j})\right)^{2} - \int_{\mathbb{S}^{d}} \int_{\mathbb{S}^{d}} \left(\frac{1}{2} - d(x,y)\right)^{2} d\sigma(x) d\sigma(y).$$

Finite metric spaces (Barg)

- X finite metric space, $Z = \{z_1, ..., z_N\} \subset X$
- $D_{L^2}^2(Z) = \sum_{t=0}^n \sum_{x \in X} \left(\frac{1}{N} \sum \mathbf{1}_{B(x,t)}(z_j) \frac{\#B(x,t)}{\#X} \right)^2$
- Define the distance

$$\lambda(x,y) = \frac{1}{2} \sum_{u \in X} |d(x,u) - d(y,u)|.$$

Stolarsky principle:

$$D_{L^2}^2(Z) = \frac{1}{(\#X)^2} \sum_{x,y \in X} \lambda(x,y) - \frac{1}{N^2} \sum_{i,j=1}^N \lambda(z_i, z_j).$$



Hamming cube (Barg)

$$D_{L^2}^2(Z) = \frac{1}{(\#X)^2} \sum_{x,y \in X} \lambda(x,y) - \frac{1}{N^2} \sum_{i,j=1}^N \lambda(z_i, z_j).$$

Lemma

Let $X = \{0,1\}^n$ be the Hamming cube. Assume that the Hamming distance d(x,y) = w. Then

$$\lambda(x,y) = 2^{n-w} w \binom{w-1}{\lceil \frac{w}{2} \rceil - 1}.$$