

# Eigenvalue bounds for sets avoiding norm 1 in $\mathbb{R}^d$

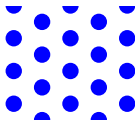
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# Sets avoiding norm 1

- ▶ A subset  $A$  of  $\mathbb{R}^d$  **avoids norm 1** if  $\|x - y\| \neq 1$  for all  $x, y \in A$ .
- ▶ Example in dimension 2, Euclidean norm:



Disks of diameter 1, centers at distance at least 2 apart.

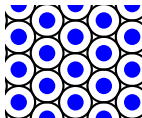
- ▶ The **density**  $\delta(A)$  of a measurable subset  $A$  is defined as usual:

$$\delta(A) = \limsup_{r \rightarrow +\infty} \frac{\text{vol}(A \cap B(r))}{\text{vol}(B(r))}.$$

Question: **How large can be  $\delta(A)$  if  $A$  avoids norm 1 ?**

# Sets avoiding norm 1

- ▶  $\delta(A) = \pi/8\sqrt{3} \approx 0.226$



- ▶ In general (arbitrary dimension and norm), a similar construction achieves

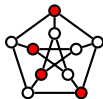
$$\delta(A) = (\text{density of an optimal packing of unit balls})/2^d.$$

- ▶ In dimension 2 for the Euclidean norm the best known construction is an hexagonal arrangement of *tortoises*, giving  $\delta \approx 0.229$ .

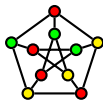


# Finite graphs $G = (V, E)$

- ▶ A **stable set** or **independent set** is a subset  $S$  of  $V$  such that  $S^2 \cap E = \emptyset$ . The **independence number**  $\alpha(G)$  is the maximal number of elements of an independent set.



- ▶ The **chromatic number**  $\chi(G)$  is the least number of colors needed to color the vertices of  $G$  so that vertices connected by an edge receive different colors.



- ▶ Because the color classes are independent sets, we have

$$\chi(G) \geq \frac{|V|}{\alpha(G)}$$

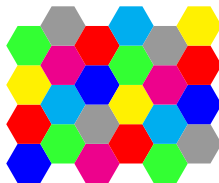
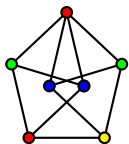
# The unit distance graph

- ▶ It is the graph with **vertex set**  $\mathbb{R}^d$  and **edge set**  $\{xy : \|x - y\| = 1\}$ .
- ▶ A set  $A$  avoiding norm 1 is **an independent set** of the unit distance graph. Its **independence number (ratio)** is

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|) := \sup_{A \text{ avoids } 1} \delta(A)$$

- ▶ The determination of its **chromatic number**  $\chi(\mathbb{R}^d)$  (Euclidean norm) is a widely open famous problem (introduced by Nelson 1950 for the plane).

# The chromatic number of the plane



$$4 \leq \chi(\mathbb{R}^2) \leq 7 \quad (\text{Nelson and Isbell, 1950})$$

# The chromatic number of $\mathbb{R}^d$

- ▶ Lower bounds based on

$$\chi(\mathbb{R}^d) \geq \chi(G)$$

for all finite induced subgraph of the unit distance graph  $G \hookrightarrow \mathbb{R}^d$ .

- ▶ De Bruijn and Erdős (1951):

$$\chi(\mathbb{R}^d) = \max_{\substack{G \text{ finite} \\ G \hookrightarrow \mathbb{R}^d}} \chi(G)$$

- ▶ Good sequences of graphs: Raiski (1970), Larman and Rogers (1972), Frankl and Wilson (1981), Székely and Wormald (1989).

# $\chi(\mathbb{R}^d)$ for large $d$

$$(1.2 + o(1))^d \leq \chi(\mathbb{R}^d) \leq (3 + o(1))^d$$

- ▶ Lower bound : Frankl and Wilson (1981).
- ▶ FW  $1.207^d$  is improved to  $1.239^d$  by Raigorodskii (2000).
- ▶ Upper bound: Larman and Rogers (1972). They use Voronoï decomposition of lattice packings.



# Frankl and Wilson graphs

- ▶  $p < d/4$  is a prime number.
- ▶  $\text{FW}(d, p)$  is the graph with:

$$V = \{x \in \{0, 1\}^d : \text{wt}(x) = 2p - 1\} \quad E = \{xy : |x \cap y| = p - 1\}.$$

- ▶ Then

$$\alpha(\text{FW}(d, p)) \leq \binom{d}{p-1}.$$

- ▶ Follows from Frankl and Wilson intersection theorems (1981).

# Frankl and Wilson graphs

- ▶ If  $p \sim ad$ ,

$$\chi(\text{FW}(d, p)) \geq \frac{|V_d|}{\alpha(\text{FW}(d, p))} \geq \frac{\binom{d}{2p-1}}{\binom{d}{p-1}} \approx e^{(H(2a)-H(a))d}$$

- ▶ Optimizing on  $a$  leads to  $(1.207)^d$ .
- ▶ Raigorodski uses vertices in  $\{0, 1, -1\}^d$  and a similar proof.

# The measurable chromatic number of $\mathbb{R}^d$

- ▶ The **measurable chromatic number**  $\chi_m(\mathbb{R}^d)$ : the color classes are required to be measurable.
- ▶ Obviously  $\chi_m(\mathbb{R}^d) \geq \chi(\mathbb{R}^d)$ .
- ▶ Falconer (1981):  $\chi_m(\mathbb{R}^d) \geq d + 3$ . In particular

$$\chi_m(\mathbb{R}^2) \geq 5$$

- ▶ The color classes avoid norm 1, thus are **independent sets of the unit distance graph**, so:

$$\chi_m(\mathbb{R}^d) \geq \frac{1}{\bar{\alpha}(\mathbb{R}^d)}.$$

## Upper bounds for $\bar{\alpha}(\mathbb{R}^d, \|\cdot\|)$

- ▶ Larman and Rogers (1972): if  $G = (V, E)$  is a finite induced subgraph of the unit distance graph, and if  $\alpha(G)$  denotes its independence number,

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|) \leq \bar{\alpha}(G) := \frac{\alpha(G)}{|V|}.$$

- ▶ Proof is easy: if  $A$  avoids norm 1,

$$(\mathbf{1}_A * \delta_V)(x) \leq \alpha(G).$$

Indeed, if  $\mathbf{1}_A * \delta_V$  reaches a value  $m > \alpha(G)$ , there exists  $x$  s.t.  $x = a_1 - v_1 = \dots = a_m - v_m$ ; then  $\|v_i - v_j\| = \|a_i - a_j\| \neq 1$  so  $\{v_1, \dots, v_m\}$  is an independent set of  $G$ , a contradiction.  
Taking densities,

$$|V|\delta(A) \leq \alpha(G).$$

# Upper bounds for $\bar{\alpha}(\mathbb{R}^d, \|\cdot\|)$

- ▶ Example: for  $\|\cdot\|_\infty$ ,  $V = \{0, 1\}^d$  leads to the complete graph so  $\bar{\alpha}(G) = 1/2^d$ . It shows

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|_\infty) = \frac{1}{2^d}.$$

- ▶ For  $\|\cdot\|_p$ ,  $1 \leq p < \infty$ , the Frankl-Wilson graphs lead to the asymptotic

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|_p) \lesssim \frac{1}{1.207^d}.$$

- ▶ For small dimensions and  $p = 2$  Szekely and Wormald (1989) give better bounds.

# An upper bounds for $\bar{\alpha}(\mathbb{R}^d, \|\cdot\|)$ from Fourier analysis

**Theorem** [B., E. de Corte, F.M. de Oliveira Filho, F. Vallentin (2013)]

Let  $\mu$  be a signed Borel measure centrally symmetric and supported on  $S_{\|\cdot\|}^{d-1} := \{x \in \mathbb{R}^d : \|x\| = 1\}$ , let

$$m_\mu := \min_{\xi \in \mathbb{R}^d} \widehat{\mu}(\xi).$$

Then,

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|) \leq \frac{-m_\mu}{\widehat{\mu}(0^d) - m_\mu}.$$

B., E. de Corte, F.M. de Oliveira Filho, F. Vallentin, *Spectral bounds for the independence ratio and the chromatic number of an operator*, arxiv:1301.1054, to appear in Israel J. Math.

## Sketch of proof:

We will pretend  $\mathbb{R}^d$  is a probability space (!!!).

- ▶ Let  $A$  avoids norm 1. Because  $\mu$  is supported on  $S_{\|\cdot\|}^{d-1}$ ,

$$(\mathbf{1}_A * \mu, \mathbf{1}_A) = 0.$$

Indeed:

$$(\mathbf{1}_A * \mu, \mathbf{1}_A) = \int \int \mathbf{1}_A(x+y) \mathbf{1}_A(x) d\mu(y) dx = 0$$

- ▶ We decompose  $\mathbf{1}_A$  orthogonally:

$$\mathbf{1}_A = \beta \mathbf{1} + g, \quad (\mathbf{1}, g) = 0$$

then replace in  $(\mathbf{1}_A * \mu, \mathbf{1}_A) = 0$ .

# Sketch of proof:

- ▶ We obtain:

$$0 = (\mathbf{1}_A * \mu, \mathbf{1}_A) = ((\beta \mathbf{1} + g) * \mu, \beta \mathbf{1} + g) = \beta^2 + (g * \mu, g)$$

- ▶ Applying Parseval:

$$(g * \mu, g) = (\widehat{g\mu}, \widehat{g}) \geq m_\mu(g, g)$$

- ▶ Thus:

$$\beta^2 = -(g * \mu, g) \leq -m_\mu(g, g)$$

- ▶ To conclude we notice:

$$\beta = (\mathbf{1}_A, \mathbf{1}) = \delta(A) \quad \text{et} \quad (g, g) = (\mathbf{1}_A, \mathbf{1}_A) - \beta^2 = \delta(A) - \delta(A)^2$$



# An analogy with finite graphs

- ▶ This upper bound is the analog of the so-called **Delsarte bound for graphs**:
- ▶  $G = (V, E)$  a finite graph. For all symmetric matrix  $B \in \mathbb{R}^{V \times V}$  s.t.  $B\mathbf{1} = d\mathbf{1}$  and  $B_{x,y} = 0$  if  $xy \notin E$ ,

$$\bar{\alpha}(G) = \frac{\alpha(G)}{|V|} \leq \frac{-\lambda_{\min}(B)}{d - \lambda_{\min}(B)}$$

where  $\lambda_{\min}(B)$  is the minimal eigenvalue of  $B$ .

- ▶ If  $G$  is regular of degree  $d$ , one can take for  $B$  the adjacency matrix of  $G$ . It leads to the **Hoffman bound**.
- ▶ If  $\text{Aut}(G)$  is transitive on  $E$ , the adjacency matrix is an optimal choice for  $B$ .

# An analogy with finite graphs

- ▶  $G = (V, E)$  a finite graph,  $B \in \mathbb{R}^{V \times V}$  s.t.  $B_{x,y} = 0$  if  $xy \notin E$  defines an operator:

$$B : \mathbb{R}^V \rightarrow \mathbb{R}^V$$

$$f \mapsto Bf, \quad (Bf)_x = \sum_{y \in V(x)} B_{x,y} f_y$$

- ▶ For the unit distance graph, the measure  $\mu$  also defines an operator:

$$L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$$

$$f \mapsto f * \mu, \quad (f * \mu)(x) = \int_{\|y\|=1} f(x+y) d\mu(y)$$

whose spectrum is  $\{\widehat{\mu}(\xi), \xi \in \mathbb{R}^d\}$ . So  $m_\mu = \min \widehat{\mu}(\xi)$  replaces  $\lambda_{\min}(B)$ .

# How to optimize over $\mu$

- ▶ Recall the bound: if  $\mu$  is supported on  $S_{\|\cdot\|}^{d-1}$ , let  $m_\mu := \min \widehat{\mu}(\xi)$ ,

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|) \leq \frac{-m_\mu}{\widehat{\mu}(0^d) - m_\mu}.$$

- ▶ Problem: how should we choose  $\mu$  so that this bound is good ?
- ▶ Because the RHS is convex,  $\mu$  can be assumed to be invariant under  $\text{Aut}(S_{\|\cdot\|}^{d-1})$ .
- ▶ For the Euclidean norm, it means  $\mu$  is invariant under  $O(\mathbb{R}^d)$  so there is essentially one choice: the surface measure of the unit sphere.

# The Fourier bound for the Euclidean norm

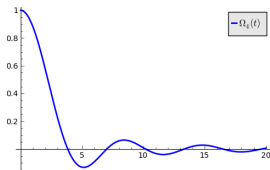
- ▶ Taking  $\mu = \omega_d$  the normalized surface measure on  $S^{d-1}$ ,

$$\widehat{\omega}_d(\xi) = \int_{S^{d-1}} e^{2i\pi(x \cdot \xi)} d\omega_d(x) = \Omega_d(\|\xi\|)$$

where

$$\Omega_d(t) = \Gamma(d/2)(2/t)^{d/2-1} J_{d/2-1}(t)$$

$J_{d/2-1}(t)$  is the Bessel function of the first kind with parameter  $d/2 - 1$ .



$\min \Omega_d(t) = \Omega_d(j_{d/2,1})$  where  $j_{d/2,1}$  is the first zero of  $J_{d/2}$ .

# The Fourier bound for the Euclidean norm

**Theorem** [F.M. de Oliveira Filho, F. Vallentin 2010]

$$\bar{\alpha}(\mathbb{R}^d) \leq \frac{-\Omega_d(j_{d/2,1})}{1 - \Omega_d(j_{d/2,1})}$$

- ▶ Asymptotically,

$$\frac{-\Omega_d(j_{d/2,1})}{1 - \Omega_d(j_{d/2,1})} \approx (\sqrt{e/2})^{-n} \approx 1.165^{-d}$$

So it is not as good as the Frankl-Wilson  $1.207^{-d}$  and Raigorodskii  $1.239^{-d}$  bounds (although better for small dimensions).

- ▶ It is possible to improve the Fourier bound by additional graphical constraint.

# The improved Fourier bound for the Euclidean norm

Let  $G \hookrightarrow \mathbb{R}^d$ , for  $x_i \in V$ , let  $r_i := \|x_i\|$ .

$$\vartheta_G(\mathbb{R}^d) := \inf \left\{ z_0 + z_2 \frac{\alpha(G)}{|V|} : \begin{aligned} & z_2 \geq 0 \\ & z_0 + z_1 + z_2 = 1 \\ & z_0 + z_1 \Omega_d(t) + z_2 \left( \frac{1}{|V|} \sum_{i=1}^{|V|} \Omega_d(r_i t) \right) \geq 0 \\ & \text{for all } t > 0 \end{aligned} \right\}.$$

**Theorem** [Filho, Vallentin 2010 for simplices]

$$\bar{\alpha}(\mathbb{R}^d) \leq \vartheta_G(\mathbb{R}^d)$$

**Theorem** [B., A. Thiery 2012]

$$\vartheta_R(\mathbb{R}^d) \lesssim (1.268)^{-d} \quad d \rightarrow +\infty$$

# Numerical results: upper bounds for $\bar{\alpha}(\mathbb{R}^d)$

d	previous	Fourier bound	improved F. bound	G
2	0.279069	0.287120	0.2623	Moser Spindles
3	0.187500	0.178466	0.165609	simplices [OV 2010]
4	0.128000	0.116826	0.10006	600-cell
5	0.0953947	0.0793346	0.0752845	simplex [OV 2010]
6	0.0708129	0.0553734	0.04870	Schäfli/kissing
7	0.0531136	0.0394820	0.02764	kissing of $E_8$
8	0.0346096	0.0286356	0.01959	$E_8$
9	0.0288215	0.0210611	0.01678	J(10,5,2)
10	0.0223483	0.0156717	0.01269	J(11,5,2)
11	0.0178932	0.0117771	0.0088775	J(12,6,2)*
12	0.0143759	0.00892554	0.006111	J(13,6,2)*
13	0.0120332	0.00681436	0.00394332	J(14,7,3)*
14	0.00981770	0.00523614	0.00300286	J(15,7,3)*
15	0.00841374	0.00404638	0.00242256	J(16,8,3)*
16	0.00677838	0.00314283	0.00161645	J(17,8,3)*
17	0.00577854	0.00245212	0.00110487	J(18,9,4)*
18	0.00518111	0.00192105	0.00084949	J(19,9,4)*
19	0.00380311	0.00151057	0.00074601	J(20,9,3)*
20	0.00318213	0.001191806	0.00046909	J(21,10,4)*
21	0.00267706	0.000943209	0.00031431	J(22,11,5)*
22	0.00190205	0.000748582	0.00024621	J(23,11,5)*
23	0.00132755	0.000595665	0.0002122678	J(24,12,5)
24	0.00107286	0.000475128	0.00018437	orth. graph [KP 2008]

# Numerical results : lower bounds for $\chi_m(\mathbb{R}^d)$

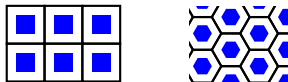
d	previous	$\chi_m(\mathbb{R}^d)$	G
2	5		
3	6	7	simplices
4	8	10	600-cells
5	11	14	simplex
6	15	21	Schläfli/kissing
7	19	37	kissing of $E_8$
8	30	52	$E_8$
9	35	60	J(10,5,2)
10	48	79	J(11,5,2)
11	64	113	J(12,6,2)*
12	85	164	J(13,6,20)*
13	113	254	J(14,7,3)*
14	147	334	J(15,7,3)*
15	191	413	J(16,8,3)*
16	248	619	J(17,8,3)*
17	319	906	J(18,9,4)*
18	408	1178	J(19,9,4)*
19	521	1341	J(20,9,3)*
20	662	2132	J(21,10,4)*
21	839	3182	J(22,11,5)*
22	1060	4062	J(23,11,5)*
23	1336	4712	J(24,12,5)*
24	1679	5424	orth. graph



# Polytopal norms

- ▶ Joint ongoing work with D. Henrion, J.-B. Lasserre, S. Robins, F. Vallentin. Many open questions!
- ▶ For the hypercube ( $\|\cdot\|_\infty$ ) we have seen  $\bar{\alpha}(\mathbb{R}^d, \|\cdot\|_\infty) = 1/2^d$ .

*Question: does it hold for any polytope that tiles ?*



It is open for the hexagon. The best we can prove: 0.28.

# Polytopal norms

- ▶ In the Fourier bound,

$$\bar{\alpha}(\mathbb{R}^d, \|\cdot\|) \leq \frac{-m_\mu}{\hat{\mu}(0^d) - m_\mu}$$

*Question: what is the optimal measure  $\mu$  ?*

Symmetrization does not lead to a single measure.

- ▶ For the hypercube, a weighted sum of point measures at the center of all faces gives back the  $1/2^d$  bound. Not the surface measure!



- ▶ In general, point measures lead to **polynomial optimization problems**.

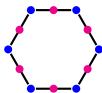
# Point measures and polynomial optimization

- ▶ Let  $\mathcal{Q} = \{Q_1, \dots, Q_m\}$  rational points in  $\mathbb{R}^d$ . Let  $\mu = \sum_{j=1}^m w_j \delta_{Q_j}$ . We assume (for simplicity) that  $\mathcal{Q}$  is invariant by flipping the signs of coordinates. Then

$$\hat{\mu}(\xi) = \sum_{j=1}^m w_j \cos(2\pi Q_{j1}\xi_1) \dots \cos(2\pi Q_{jd}\xi_d).$$

- ▶ If  $k$  is the lcm of denominators of coordinates of  $Q_j$ , the above is a polynomial in the variables  $x_1 := \cos(2\pi\xi_1/k), \dots, x_d := \cos(2\pi\xi_d/k)$  (using Chebyshev polynomials) with  $\deg_{x_j} \leq k$ .

► Example: the hexagon



$$\widehat{\mu}_{\text{vertices}}(\xi) = P_v(x_1, x_2) := (8x_1^4 + 8x_1^2x_2^2 - 12x_1^2 - 4x_2^2 + 3)/3$$

$$\widehat{\mu}_{\text{edges}}(\xi) = P_e(x_1, x_2) := (8x_1^3x_2 - 6x_1x_2 + 2x_2^2 - 1)/3.$$

We have

$$6(P_v(x_1, x_2) + 2P_e(x_1, x_2)) = -7 + (4x_1^2 + 4x_1x_2 - 3)^2$$

showing that

$$(\widehat{\mu}_{\text{vertices}} + 2\widehat{\mu}_{\text{edges}})/3$$

has minimum  $-7/18$ , leading to the bound  $7/25 = 0.28$ .

# Point measures, polynomial optimization, and SOS

- ▶ For fixed set of rational points  $\mathcal{Q}$ , optimizing over the weights  $w_j$  amounts to solve:

$$\max\left\{ m : \sum_{j=1}^m w_j P_j(x_1, \dots, x_d) \geq m, \sum_{j=1}^m w_j = 1 \right\}$$

where  $P_j(x_1, \dots, x_d) = \widehat{\delta}_{\mathcal{Q}_j}(\xi)$  and  $x_j \in [-1, 1]$ .

- ▶ **sums of squares relaxations** allow to approximate the above by **semidefinite programming**:

$$P_j(\underline{x}) = m + S_0(\underline{x}) + (1 - x_1^2)S_1(\underline{x}) + \dots + (1 - x_d^2)S_d(\underline{x})$$

where  $S_0, \dots, S_d$  are **SOS**.

# Numerical results

- ▶ Optimizing the weights for the middle of all faces:
  - ▶ For the hexagon, the weights  $1/3, 2/3$  are optimal leading to bound 0.28
  - ▶ For the crosspolytopes it seems the bound is  $1/2d$  (verified for  $d = 2, 3, 4$ ).
- ▶ Optimizing the weights for more points:
  - ▶ For the hexagon it does not improve.
  - ▶ For the crosspolytopes, it does and we obtain:

$d$	$\mathcal{Q}$	$m$	bound
3	$\mathbb{Z}/8$	-0.186	0.155
4	$\mathbb{Z}/6$	-0.104	0.095