

# Average distortion embeddings, nonlinear spectral gaps, and a metric John theorem

(after Assaf Naor)

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# Norms and convex sets

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The main objects of study in this talk are finite-dimensional normed spaces  $X = (\mathbb{R}^d, \|\cdot\|)$ , where  $\|\cdot\| : \mathbb{R}^d \rightarrow \mathbb{R}_+$  satisfies:

- $\|x\| = 0 \Leftrightarrow x = 0$ ,
- $\|\lambda x\| = |\lambda| \|x\|$  and
- $\|x + y\| \leq \|x\| + \|y\|$ ,

where  $x, y \in \mathbb{R}^d$  and  $\lambda \in \mathbb{R}$ .

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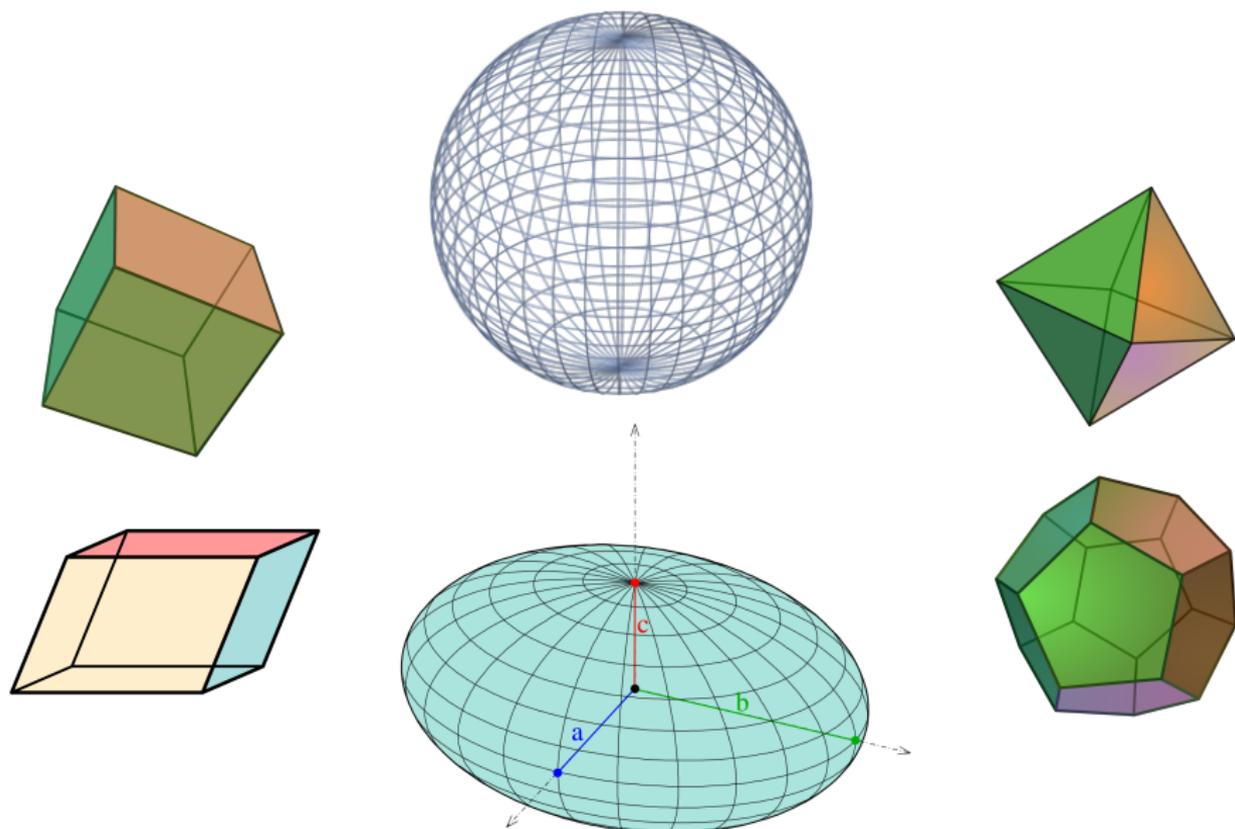
Finite-dimensional normed spaces  $X = (\mathbb{R}^d, \|\cdot\|)$  are in 1-1 correspondance with symmetric compact convex sets  $K$  in  $\mathbb{R}^d$  with non-empty interior via the bijection

$$X \mapsto \mathbf{B}_X \stackrel{\text{def}}{=} \{x \in \mathbb{R}^d : \|x\| \leq 1\},$$

whose inverse is

$$K \mapsto \|x\|_K \stackrel{\text{def}}{=} \min\{t \geq 0 : x \in tK\}.$$

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For any normed space  $X = (\mathbb{R}^d, \|\cdot\|)$ , there exists a linear operator  $T : X \rightarrow \ell_2^d = (\mathbb{R}^d, \|\cdot\|_{\ell_2^d})$  with  $\|T\|_{\text{op}} \|T^{-1}\|_{\text{op}} \leq \sqrt{d}$ , i.e.

$$\forall x \in \mathbb{R}^d, \quad \|x\| \leq \|Tx\|_{\ell_2^d} \leq \sqrt{d}\|x\|.$$

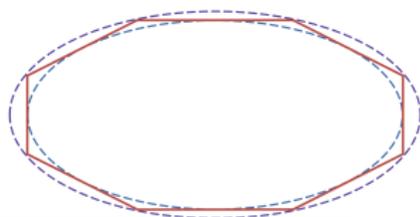
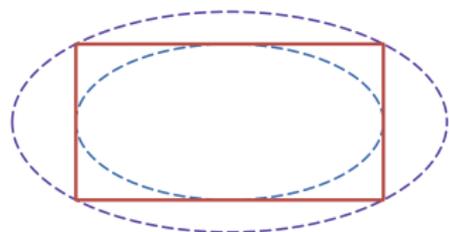
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Equivalently, for any compact symmetric convex set  $K$  in  $\mathbb{R}^d$ , there exists an ellipsoid  $\mathcal{E}$  satisfying  $\mathcal{E} \subseteq K \subseteq \sqrt{d}\mathcal{E}$ .



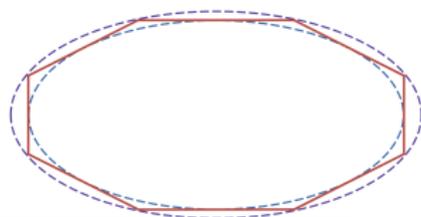
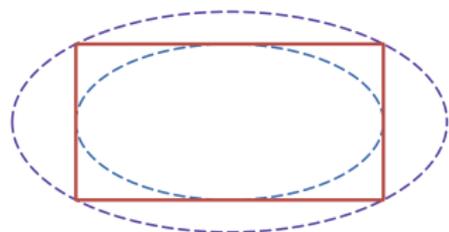
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Moreover, the factor  $\sqrt{d}$  is optimal (e.g. for the cube  $[-1, 1]^d$  which corresponds to the supremum norm on  $\mathbb{R}^d$ ).

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## Theorem (F. John, 1948)

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By classical differentiation principles, this is an **equivalent** reformulation of John's theorem.

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$$\iint_{M \times M} \|f(x) - f(y)\|_Y^q d\mu(x) d\mu(y) \geq \iint_{M \times M} d_M(x, y)^q d\mu(x) d\mu(y).$$

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$$\iint_{\mathcal{M} \times \mathcal{M}} \|f(x) - f(y)\|_Y^q d\mu(x) d\mu(y) \leq D^q \iint_{\mathcal{M} \times \mathcal{M}} d_{\mathcal{M}}(x, y)^q d\mu(x) d\mu(y).$$

- If  $\theta \in (0, 1]$ , the  $\theta$ -**snowflake** of a metric space  $(\mathcal{M}, d_{\mathcal{M}})$  is the metric space  $(\mathcal{M}, d_{\mathcal{M}}^{\theta})$ .

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## Theorem (A. Naor, 2019)

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- $\frac{1}{2}$  is the least amount of snowflaking for which the resulting distortion depends subpolynomially on  $d$ .
- The bound  $O(\sqrt{\log d})$  is optimal for the quadratic average distortion of a  $d$ -dimensional normed space.

# Application: expanders

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Let  $G = (V, E)$  be an  $r$ -regular graph and let  $A(G)$  be its normalized adjacency matrix, whose entries are given by

$$\forall u, v \in V, \quad A(G)_{u,v} = \frac{\mathbf{1}_{\{u,v\} \in E}}{r}.$$

$A(G)$  is a symmetric stochastic matrix with real eigenvalues

$$1 = \lambda_1(G) \geq \lambda_2(G) \geq \dots \geq \lambda_{|V|}(G) \geq -1.$$

Denote by  $\gamma(G) = \frac{1}{1-\lambda_2(G)}$  the **reciprocal spectral gap** of  $G$ .

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A sequence of  $r$ -regular graphs  $\{G_n = (V_n, E_n)\}_{n=1}^{\infty}$  such that  $|V_n| \rightarrow \infty$  as  $n \rightarrow \infty$  is an **expander graph sequence** if there exists  $C \in (0, \infty)$  such that  $\gamma(G_n) \leq C$  for all  $n \in \mathbb{N}$ .

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## Theorem (A. Naor, 2017)

*Suppose that  $G = (V, E)$  is an  $n$ -vertex connected 4-regular expander which admits a bi-Lipschitz embedding into a  $d$ -dimensional normed space  $X = (\mathbb{R}^d, \|\cdot\|)$  with distortion at most  $D$ . Then  $d \geq n^{c/D}$  for some universal constant  $c \in (0, \infty)$ .*

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The sharpness of the JLS theorem had previously been established in important work of **J. Matoušek** (1996) via an ingenious construction of random metric spaces which relied on input from real algebraic geometry.

# Proof using the average John theorem

Let  $G = (V, E)$  be an  $n$ -vertex 4-regular connected graph and denote  $\gamma(G) = \gamma$ . Suppose that there exists a  $d$ -dimensional normed space  $X = (\mathbb{R}^d, \|\cdot\|)$  and a map  $f : V \rightarrow X$  satisfying

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By the average John theorem for the measure  $\frac{1}{n} \sum_{u \in V} \delta_{f(u)}$ , there exists a  $O(\sqrt{\log d})$ -Lipschitz map  $g : (X, \|\cdot\|^{1/2}) \rightarrow \ell_2$  with

$$\frac{1}{n^2} \sum_{u, v \in V} \|g(f(u)) - g(f(v))\|_{\ell_2}^2 \geq \frac{1}{n^2} \sum_{u, v \in V} \|f(u) - f(v)\|.$$

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**Remark.**  $\gamma$  is the best constant such that any  $h : V \rightarrow \ell_2$  satisfies

$$\frac{1}{n^2} \sum_{u, v \in V} \|h(u) - h(v)\|_{\ell_2}^2 \leq \frac{\gamma}{|E|} \sum_{\{a, b\} \in E} \|h(a) - h(b)\|_{\ell_2}^2.$$

# Proof using the average John theorem

Applying this to  $g \circ f$  along with the upper and lower bounds,

$$\begin{aligned} \frac{1}{n^2} \sum_{u,v \in V} d_G(u,v) &\leq \frac{1}{n^2} \sum_{u,v \in V} \|g(f(u)) - g(f(v))\|_{\ell_2}^2 \\ &\leq \frac{\gamma}{|E|} \sum_{\{a,b\} \in E} \|g(f(a)) - g(f(b))\|_{\ell_2}^2 \lesssim \gamma D \log d. \end{aligned}$$

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Finally, as the graph  $G$  is 4-regular, for any fixed  $u \in V$ , at least  $\log_4(\lfloor n/2 \rfloor)$  satisfy  $d_G(u,v) \geq \frac{n}{2}$ . Therefore

$$\frac{\log n}{100} \leq \frac{1}{n^2} \sum_{u,v \in V} d_G(u,v) \lesssim \gamma D \log d,$$

which completes the proof. □

# Nonlinear spectral gap inequalities

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Let  $\pi = (\pi_1, \dots, \pi_n)$  be a probability measure on  $\{1, \dots, n\}$ . A stochastic matrix  $A \in M_n(\mathbb{R})$  is  $\pi$ -**reversible** if  $\pi_i a_{ij} = \pi_j a_{ji}$ . We think of  $A$  as the transition matrix of a Markov chain  $\{X_t\}_{t \geq 0}$ , i.e.

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If  $A$  is  $\pi$ -reversible, then  $\pi$  is a stationary measure for  $\{X_t\}_{t \geq 0}$ ,

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Moreover,  $A$  defines a self-adjoint operator on  $L_2(\pi)$  whose norm is

$$\|x\|_{L_2(\pi)} \stackrel{\text{def}}{=} \left( \sum_{i=1}^n \pi_i x_i^2 \right)^{\frac{1}{2}}$$

and therefore has real eigenvalues

$$1 = \lambda_1(A) \geq \lambda_2(A) \geq \dots \geq \lambda_n(A) \geq -1.$$

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As in the case of regular graphs,  $\gamma(A) \stackrel{\text{def}}{=} \frac{1}{1-\lambda_2(A)}$  is the least constant such that for any  $x_1, \dots, x_n \in \ell_2$ , we have

$$\sum_{i,j=1}^n \pi_i \pi_j \|x_i - x_j\|_{\ell_2}^2 \leq \gamma(A) \sum_{i,j=1}^n \pi_i a_{ij} \|x_i - x_j\|_{\ell_2}^2.$$

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**Definition.** Let  $(\mathcal{M}, d_{\mathcal{M}})$  be a metric space and  $p \in (0, \infty)$ . If  $A$  is a  $\pi$ -reversible stochastic matrix, denote by  $\gamma(A, d_{\mathcal{M}}^p)$  the least constant such that for any  $n \in \mathbb{N}$  and  $x_1, \dots, x_n \in \mathcal{M}$ , we have

$$\sum_{i,j=1}^n \pi_i \pi_j d_{\mathcal{M}}(x_i, x_j)^p \leq \gamma(A, d_{\mathcal{M}}^p) \sum_{i,j=1}^n \pi_i a_{ij} d_{\mathcal{M}}(x_i, x_j)^p.$$

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$$\sum_{i,j=1}^n \pi_i \pi_j d_m(x_i, x_j)^p \leq \gamma(A, d_m^p) \sum_{i,j=1}^n \pi_i a_{ij} d_m(x_i, x_j)^p.$$

**Major open problem.** For which metric spaces  $\mathcal{M}$  do we have

$$\gamma(A; d_m^p) \leq \Psi\left(\frac{1}{1-\lambda_2(A)}\right)$$

for some function  $\Psi$  and all reversible stochastic matrices  $A$ ?

# Extrapolation

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We shall need the following vector-valued version of Matoušek's extrapolation principle for Poincaré inequalities (1997) due to T. de Laat and M. de la Salle (2017).

## Proposition

For every normed space  $(X, \|\cdot\|)$ , every  $\pi$ -reversible matrix  $B$  and every  $1 \leq p \leq q$ ,

$$\gamma(B, \|\cdot\|^q)^{\frac{p}{q}} \lesssim_{p,q} \gamma(B, \|\cdot\|^p) \lesssim_{p,q} \gamma(B, \|\cdot\|^q).$$

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Suppose that  $(\mathcal{M}, d_{\mathcal{M}})$  embeds into  $(Y, \|\cdot\|_Y)$  with  $q$ -average distortion  $D$  and let  $A$  be a  $\pi$ -reversible stochastic matrix. Then, given  $x_1, \dots, x_n \in \mathcal{M}$  there exist  $y_1, \dots, y_n \in Y$  satisfying  $\|y_i - y_j\|_Y \leq D d_{\mathcal{M}}(x_i, x_j)$  for all  $i, j \in \{1, \dots, n\}$  and

$$\sum_{i,j=1}^n \pi_i \pi_j \|y_i - y_j\|^q \geq \sum_{i,j=1}^n \pi_i \pi_j d_{\mathcal{M}}(x_i, x_j)^q.$$

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$$\sum_{i,j=1}^n \pi_i \pi_j \|y_i - y_j\|^q \geq \sum_{i,j=1}^n \pi_i \pi_j d_{\mathcal{M}}(x_i, x_j)^q.$$

Therefore,

$$\begin{aligned} \sum_{i,j=1}^n \pi_i \pi_j d_{\mathcal{M}}(x_i, x_j)^q &\leq \gamma(A, \|\cdot\|_Y^q) \sum_{i,j=1}^n \pi_i a_{ij} \|y_i - y_j\|_Y^q \\ &\leq D^q \gamma(A, \|\cdot\|_Y^q) \sum_{i,j=1}^n \pi_i a_{ij} d_{\mathcal{M}}(x_i, x_j)^q, \end{aligned}$$

which implies that  $\gamma(A, d_{\mathcal{M}}^q) \leq D^q \gamma(A, \|\cdot\|_Y^q)$ .

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## Theorem (A. Naor, 2014)

*Suppose that  $q, D \in [1, \infty)$ . Let  $(\mathcal{M}, d_{\mathcal{M}})$  be a metric space and  $(Y, \|\cdot\|_Y)$  be a Banach space such that for every  $n \in \mathbb{N}$ , every reversible stochastic matrix  $A \in M_n(\mathbb{R})$  satisfies*

$$\gamma(A, d_{\mathcal{M}}^q) \leq D^q \gamma(A, \|\cdot\|_Y^q).$$

*Then, for any  $\varepsilon > 0$ ,  $\mathcal{M}$  embeds into some ultrapower of  $\ell_q(Y)$  with  $q$ -average distortion at most  $D + \varepsilon$ .*

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The proof is a clever Hahn–Banach separation argument.

# Average John as a spectral gap estimate

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## Theorem

Let  $(X, \|\cdot\|)$  be a finite-dimensional normed space. Then, for every  $n \in \mathbb{N}$ , every  $\pi$ -reversible stochastic matrix  $A \in M_n(\mathbb{R})$  satisfies

$$\gamma(A, \|\cdot\|_X) \leq \frac{C \log(\dim(X) + 1)}{1 - \lambda_2(A)},$$

where  $C \in (0, \infty)$  is a universal constant.

Thank you!