

Dvoretzky-Milman theorem

Note Title

2015-11-25

3 Some key ideas from geometric functional analysis.

DMT: "Any norm $\|\cdot\|$ on \mathbb{R}^n is almost Euclidean on a random subspace E ".
I.e.

$$\|x\|_n \approx M \|x\|_2 \quad \forall x \in E$$

proportionality coefficient

Value of M ? $\|x\| \approx M \quad \forall x \in S^{n-1} \cap E$

$$M := E\|x\| = \int_{S^{n-1}} \|x\| d\sigma(x)$$

\uparrow
 $\sim \text{Unif}(S^{n-1})$ $\underbrace{\text{normalized Lebesgue measure}}$

Dvoretzky-Milman Thm '73

Let $\|\cdot\|$ be a norm on \mathbb{R}^n s.t. $\|x\| \leq b \|x\|_2 \quad \forall x \in \mathbb{R}^n$.
Set

$$M := E\|x\|, \quad x \sim \text{Unif}(S^{n-1}).$$

Then, $\forall \epsilon > 0$, a random subspace $E \in G_{n,k}$ of dimension $k = C_\epsilon (M/b)^2 n$ satisfies w.p. $\geq 1 - 2e^{-k}$ that

$$(1-\epsilon) M \|x\|_2 \leq \|x\| \leq (1+\epsilon) M \|x\|_2 \quad \forall x \in E.$$

Remark $b := \max_{x \in S^{n-1}} \|x\|$. $M = \text{avg}_{x \in S^{n-1}} \|x\|$

$\frac{M}{b} < 1$ measures how far $\|\cdot\|$ is from Euclidean.

$$\text{Ex) If } \frac{1}{2}\|X\|_2 \leq \|X\| \leq 2\|X\|_2 \Rightarrow \left(\frac{m}{b}\right) \geq \frac{1}{4}$$

\Leftarrow

$$\text{Ex) } \|\cdot\|_1 = \|\cdot\|_1,$$

$$\|\cdot\|_1 \leq \sqrt{n} \|\cdot\|_2$$

$\|$
 b

Calculate M :

$$\mathbb{E} \|X\|_1 = ?$$

\uparrow
unif(\mathbb{S}^{n-1})

Replace w/ Gaussian expectation. Let $g \sim \mathcal{N}(0, I_n)$

$$g = X \cdot \|g\|_2 \quad X = \frac{g}{\|g\|_2}$$

\uparrow
unif(\mathbb{S}^{n-1}) \nwarrow index of X

$$\begin{aligned} \mathbb{E} \|g\|_1 &= \mathbb{E} \|X\|_2 \|g\|_1 = \mathbb{E} \|X\|_1 \cdot \|g\|_2 = \mathbb{E} \|X\|_1 \cdot \mathbb{E} \|g\|_2 \\ &= \mathbb{E} \|X\|_1 \cdot (\sqrt{n} + o(1)) \end{aligned}$$

$$\Rightarrow \mathbb{E} \|X\|_1 = \frac{\mathbb{E} \|g\|_1}{(\sqrt{n} + o(1))} = \frac{\sqrt{\frac{2}{\pi}} \cdot n}{\sqrt{n} + o(1)} \approx \boxed{\sqrt{\frac{2}{\pi}} \cdot \sqrt{n} \approx m}$$

$$\Rightarrow \frac{m}{b} \geq c.$$

Thus, $\|\cdot\|_1$ is almost Euclidean on a $c \cdot n$ dimensional subspace!

$$\text{Ex) } \|\cdot\| = \|\cdot\|_\infty \quad b = 1$$

$$\mathbb{E} \|X\|_\infty \approx \frac{\mathbb{E} \|g\|_\infty}{\sqrt{n}} \approx c \frac{\sqrt{\log n}}{\sqrt{n}} = m$$

$$\left(\frac{m}{b}\right)^2 \geq \frac{\log n}{n}$$

$\|\cdot\|_\infty$ is almost Euclidean on a $c \log n$ dimensional subspace 

Assume

Proof of DMT: (Simple covering argument) ✓

- ① Cover
- ② Deviation ≠
- ③ Union bound
- ④ Continuation argument

Sample E as follows: Fix $\mathbb{R}^k = \text{span}\{e_1, \dots, e_k\} \subseteq \mathbb{R}^n$.
Let $U \in \mathcal{O}(k)$ be random uniform. Take
 $E = U(\mathbb{R}^k)$.

Using Euclidean balls

① Cover: Let N_1 be an ϵ -cover of $\mathbb{R}^{kn} S^{n-1}$ w/ $|N_1| \leq (\frac{3}{\epsilon})^k$. Then

$N_2 := U(N_1)$ is an ϵ -cover of $E \cap S^{n-1}$
w/ $|N_2| \leq (\frac{3}{\epsilon})^k$.

② Deviation ≠: Fix $x \sim N_1$. Let $y = Ux \in N_2$.
Note $y \sim \text{Unif}(S^{n-1})$.

Let $f(z) := \|z\|$.

By assumption, f is b -Lipschitz.
Assume $b=1$ for simplicity.

Then, by spherical concentration

$$\|f(y) - \mathbb{E} f(y)\|_{\chi_2} = \|\|y\| - M\|_{\chi_2} \leq \frac{\epsilon}{\sqrt{n}}$$

i.e. $P(\|y\| - M > t) \leq 2 \exp(-ct^2 n)$

$$\Rightarrow P(\|y\| - M > M\epsilon) \leq 2 \exp(-c M^2 \epsilon^2 n)$$

③ Union bound:

$$\begin{aligned} P\left(\max_{y \in N_2} \|\|y\| - M\| > M\epsilon\right) &\leq 2\left(\frac{3}{\epsilon}\right)^k \exp(-c M^2 \epsilon^2 n) \\ &= 2 \exp(k \lg(\frac{3}{\epsilon}) - c M^2 \epsilon^2 n) \end{aligned}$$

$$(\text{By def of } k) \leq 2 \exp(-k)$$

$$\Rightarrow \text{w.p. } \geq 1 - 2 \exp(-k),$$

$$M(1-\epsilon) \leq \|y\| \leq M(1+\epsilon) \quad \text{on the set } N_\epsilon.$$

④ Continuation argument.

Let $\bar{z} \in S^{n-1} \cap E$. Then $\exists \bar{z} \in N_\epsilon$ s.t.

$$\|\bar{z} - z\|_2 \leq \epsilon \quad \text{Set } w = \frac{\bar{z} - z}{\|\bar{z} - z\|_2}$$

$$\Rightarrow \|z\| \leq \|\bar{z}\| + \|\bar{z} - z\| \leq \|\bar{z}\| + \epsilon \|w\| \quad (*)$$

$$\text{Set } a := \sup_{\bar{z} \in S^{n-1} \cap E} \|\bar{z}\|$$

Take supremums on both sides of (*) :

$$a \leq \max_{\bar{z} \in N_\epsilon} \|\bar{z}\| + \epsilon a$$

$$\Rightarrow (1-\epsilon)a \leq M(1+\epsilon) \quad \text{w.h.p.}$$

$$\Rightarrow a \leq M \frac{(1+\epsilon)}{(1-\epsilon)} \leq M(1+C\epsilon) \quad \text{for } \epsilon \ll 1.$$

Similar lower bound

$$\Rightarrow \text{w.p. } \geq 1 - 2 \exp(-k)$$

$$(1-C\epsilon)M \leq \|z\| \leq M(1+C\epsilon) \quad \forall z \in E \cap S^{n-1}$$

Absorb C into def of k to finish proof

QED

Geometric form of DMT

Recall: One-to-one correspondence between norms and symmetric convex bodies (in R^n).

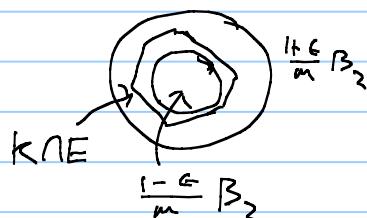
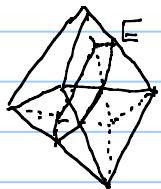
$$\| \cdot \|_X \rightarrow B_X \xleftarrow[\text{convex body}]{\text{unit ball}} \quad K \rightarrow \| \cdot \|_K := \inf \{ \lambda > 0 : x \in \lambda K \}$$

- Given convex body K , apply DMT for $\| \cdot \|_K$.

Assumption: $\|x\|_E \leq b \|x\|_2 \Leftrightarrow K \subseteq \frac{1}{b} B_2^n$

Conclusion: $(1-\epsilon)M\|x\|_2 \leq \|x\|_E = (1+\epsilon)M\|x\|_2$

$$\Leftrightarrow \frac{1-\epsilon}{M} B_2^n \cap E \subseteq K \cap E \subseteq \frac{1+\epsilon}{M} B_2^n \cap E$$



Thm (DMT, random sections) Let K be a symmetric convex body s.t. $K \supseteq \frac{1}{b} B_2^n$. Let

$$M = \mathbb{E} \|X\|_E \quad X \sim \nu_{n-1}(S^{n-1}).$$

Then, $\forall \epsilon > 0$, a random subspace $E \subseteq G_{n,K}$ of dimension $k = C_\epsilon \left(\frac{M}{b}\right)^2 n$ satisfies w.p. $\geq 1 - 2e^{-k\epsilon}$ that

$$\frac{(1-\epsilon)}{M} B_2^n \cap E \subseteq K \subseteq \frac{1+\epsilon}{M} B_2^n \cap E$$

Random Projection Version of DMT

Theme: projections & intersections onto subspaces are in dual relationship.

Recall Given a convex body K , the polar body is

$$K^\circ := \{x \in \mathbb{R}^n : \langle x, y \rangle \leq 1 \quad \forall y \in K\}$$

$$\cdot \|z\|_{K^\circ} := \inf \{\lambda > 0 : z \in \lambda K^\circ\} = \sup_{y \in K^\circ} \langle z, y \rangle$$

• $\|\cdot\|_K$ & $\|\cdot\|_{K^\circ}$ are dual norms

Properties:

- $K \subseteq L \Rightarrow K^\circ \supseteq L^\circ$
- $(qK)^\circ = \frac{1}{q} K^\circ$
- K convex, closed $\Rightarrow K^{\circ\circ} = K$ (bipolar thm)

$$\boxed{(K \cap E)^\circ = P_E K^\circ}$$

↑
subspace

Proof: Exercise

Note: Polar is taken in E not \mathbb{R}^n

Examine DMT for K° :

Assumptions

$$- K^\circ \subseteq \frac{1}{b} B_2^n \Leftrightarrow K \subseteq b B_2^n$$

$$\bullet M := \mathbb{E} \|X\|_{K^\circ} = \mathbb{E} \sup_{Y \in K} \langle X, Y \rangle =: w_s(K) \leftarrow \begin{matrix} \text{spherical} \\ \text{mean width} \end{matrix}$$

Conclusion:

$$\frac{1-\epsilon}{m} B_2^n \cap E \subseteq K^\circ \cap E \subseteq \frac{1+\epsilon}{m} B_2^n \cap E$$

Take polars of all sides (in E !):

$$\Rightarrow \underbrace{\frac{M}{1-\epsilon} P_E B_2^n}_{= E \cap B_2^n} \geq P_E K \geq \underbrace{\frac{M}{1+\epsilon} P_E B_2^n}_{= E \cap B_2^n}$$

\Rightarrow

Thm (DMT random projections)

Let $K \subseteq \mathbb{R}^n$ be a symmetric convex body
s.t. $K \subseteq b B_2^n$. Let

$$M = w_s(K).$$

Then, $\forall \epsilon > 0$, a random projection P_E onto a subspace E uniformly distributed in the Grassmannian satisfies: Suppose $\dim(E) = k = C \left(\frac{m}{b}\right)^2 n$, then $w_P \geq 1 - 2e^{-k}$,

$$(1-\epsilon) M B_2^n \cap E \subseteq P_E K \subseteq (1+\epsilon) M B_2^n \cap E$$

Literature: [Vershynin, "Lectures in geometric..."]