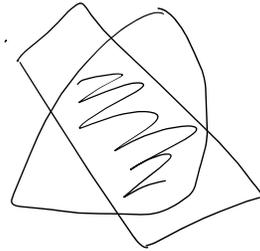


The Localization Method for High-dimensional Inequalities (TUTORIAL)

① Slicing (Bourgain '86) convex body $K \subseteq \mathbb{R}^n$ $x, y \in K$ $[x, y] \subseteq K$

$\text{Vol}(K) = 1$ \exists hyperplane H s.t. $\text{Vol}_{n-1}(K \cap H) \geq c$.
 $\exists c > 0$

Thm [Klartag-Lehec; Bizeul]



② Thin-shell (Attila; Ball, Perissinaki; Bobkov-Koldobsky '03)

ν in \mathbb{R}^n logconcave

f is logconcave

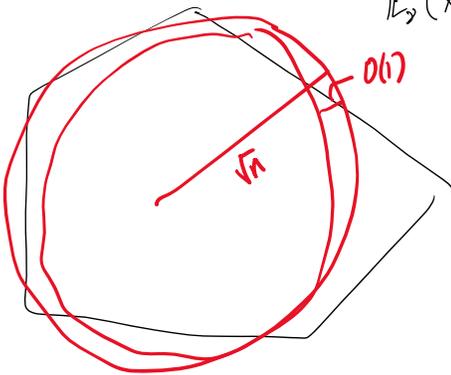
$f: \mathbb{R}^n \rightarrow \mathbb{R}_+$ $\forall x, y \in \mathbb{R}^n, \lambda \in [0, 1]$

$f((1-\lambda)x + \lambda y) \geq f(x)^{1-\lambda} f(y)^\lambda$

ν is isotropic if $E_\nu(x) = 0$

$E_\nu(x x^T) = I$

any dist with bounded second moments can be made isotropic by affine transformation



$E(\|x\|^2) = n$

$E_\nu((\|x\| - \sqrt{n})^2) = \underline{\underline{O(1)}}$

③ Large Deviation $t \geq 1$

Thm (Paouris '06)

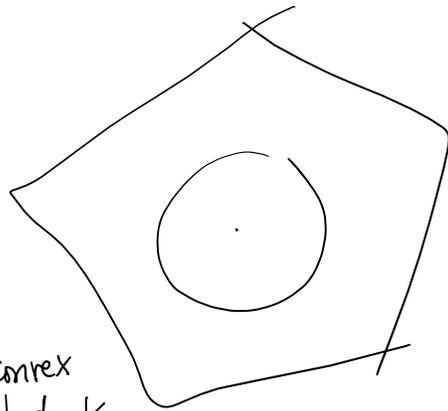
$P_{\nu, 2}(\|x\| > t\sqrt{n}) \leq e^{-ct\sqrt{n}}$

isotropic, logconcave.

④ Log. small-ball probability

$\epsilon < \epsilon_0$ $P_{\nu, 2}(\|x\| \leq \epsilon\sqrt{n}) \leq \epsilon^{cn}$

Thm (Bizeul) \uparrow



⑤ KLS conjecture. [KLS 95]

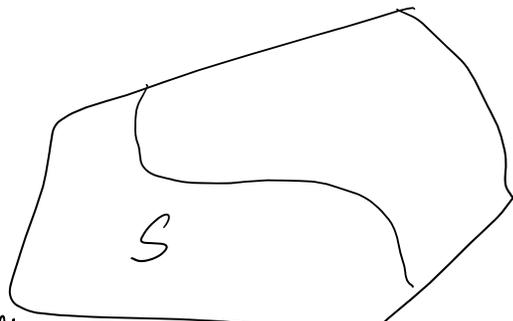
isotropic, convex body K

$\exists c > 0 \forall S \subseteq K$ $\text{Vol}(S) \leq \frac{1}{2} \text{Vol}(K)$

$\text{Vol}_{n-1}(\partial S) \geq c \text{Vol}(S)$

ν : logconcave

$\nu(\partial S) \geq \frac{1}{\|\mathbb{1}_S\|_{0,2}} \nu(S)$



$$\nu(0) = \frac{1}{\sqrt{\|A\|_{op}}}$$

Hypersplane cuts are within $O(1)$ of the optimal isoperimetric subset.



$$\Psi_{KLS}(\nu) = \sup \frac{\nu(S)}{\nu(\partial S)}$$

Thm [KLS95] $\nu(\partial S) \geq \frac{c}{\sqrt{\text{Tr}(A)}} \min \{ \nu(S), \nu(\mathbb{R}^n \setminus S) \}$

$$\sqrt{\text{Tr}(A)} = \sqrt{\mathbb{E}(\|x\|^2)} \leq \text{Diam}$$

ν has a Poincaré Inequality with "constant" C_{PI}

if f smooth f $\text{Var}_\nu(f) \leq C_{PI} \mathbb{E}_\nu(\|\nabla f\|^2)$

For logconcave ν : $\Psi_{KLS}^2 \lesssim C_{PI} \lesssim \Psi_{KLS}^2$

KLS \Rightarrow thin-shell \Rightarrow slicing \Updownarrow small-ball

Thm (Klartag 23) $\Psi_{KLS} = O(\sqrt{\log n})$
 $C_{PI} \lesssim \log n$

⑥ ν is t -strongly logconcave $\nu(x) = e^{-\frac{t\|x\|^2}{2}} \cdot \nu_0(x)$

Thm $\frac{\nu(\partial S)}{\min \nu(S), \nu(\mathbb{R}^n \setminus S)} \gtrsim \sqrt{t} \Leftrightarrow C_{PI} \lesssim t$

[Brascamp-Lieb] $\text{Var}_\nu(f) \leq \mathbb{E}_\nu(\langle \nabla f, (\nabla^2 \log f)^{-1} \nabla f \rangle) \leq t \mathbb{E}(\|\nabla f\|^2)$

Thm $\nu(S) \leq \frac{1}{2}$
 $\frac{\nu(\partial S)}{\nu(S) \sqrt{\log \frac{1}{\nu(S)}}} \gtrsim \sqrt{t}$ "log-Cheeger".

Log-Sobolev inequality (LSI) ν satisfies LSI with "constant" C_{LSI}
 $\text{Ent}_\nu(f) \leq C_{LSI} \mathbb{E}(\|\nabla f\|^2)$

$C_{LSI} = O(1) \Rightarrow$ log-Cheeger holds with constant $\Omega(1)$.

⑥ Certifiable hypercontractivity $f(x) = \langle \langle 1, x \rangle^k, A \rangle$
 $\text{Var}_\nu(f) \lesssim (C, k)^{2k} \|A\|_F^2$ isotropic, logconcave ν

$$\text{Var}_K(f) \leq (C, K) \|A\|_F \quad \text{isomorphic, logconcave}$$

[KS] $C_1 \approx C_{PI}$

(7) Polynomial anti-concentration

[Carbery-Whitney] deg d $p: \mathbb{R}^n \rightarrow \mathbb{R}$ on a convex body $K \subseteq \mathbb{R}^n$.
 $d \leq n$ $\text{Var}_K(p(x)) = 1 \Rightarrow \forall \epsilon > 0 \exists \delta > 0 \forall x \in K \mathbb{P}_x(|p(x) - t| \leq \epsilon) \leq d \epsilon^{1/\delta}$

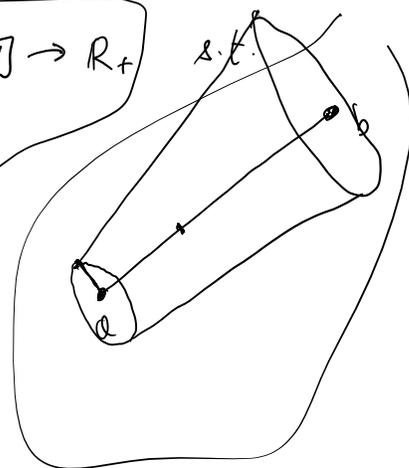
Lemma [LS93] $f, g: \mathbb{R}^n \rightarrow \mathbb{R}$ lower semi-continuous

$$\int_{\mathbb{R}^n} f > 0 \quad \int_{\mathbb{R}^n} g > 0$$

Then \exists needle: $a, b \in \mathbb{R}^n$ linear function $l: [0, 1] \rightarrow \mathbb{R}_+$ s.t.

$$\int_N f = \int_0^1 f((1-t)a + tb) l(t)^{n-1} dt$$

s.t. $\int_N f > 0 \quad \int_N g > 0$



Lemma [KLS98] $f_1, f_2, f_3, f_4: \mathbb{R}^n \rightarrow \mathbb{R}$ continuous.

The following are equivalent:

(1) \exists logconcave $F: \mathbb{R}^n \rightarrow \mathbb{R}_+$ with compact support

$$\int F f_1 \int F f_2 \leq \int F f_3 \int F f_4$$

(2) \exists exp. needle $a, b \in \mathbb{R}^n \quad \gamma \in \mathbb{R} \quad \int_E f = \int_0^{|b-a|} f(a+t(b-a)) e^{\gamma t} dt$

$$\int_E f_1 \int_E f_2 \leq \int_E f_3 \int_E f_4$$

Lemma [Fradelizi-Guedon] $K: \text{convex body in } \mathbb{R}^n \quad f: \text{upper semi-continuous}$

$\mathcal{P}_f = \{ \sum \mu \text{ logconcave} : \int_K f d\mu \geq 0 \}$. Then extreme measures are

$P_f \equiv \{ \mu \text{ logconcave} : \int_K f d\mu \geq 0 \}$. Then extreme measures are

(1) Dirac $\mu \ll f(x) \geq 0$

(2) ν s.t. $\nu(x) \propto e^{-\ell(x)} \uparrow_{[a,b]}$ $[a,b] \subseteq K$

$\int f d\nu = 0$ $\int_a^x f d\nu > 0$ $\int_x^b f d\nu > 0$ $x \in (a,b)$

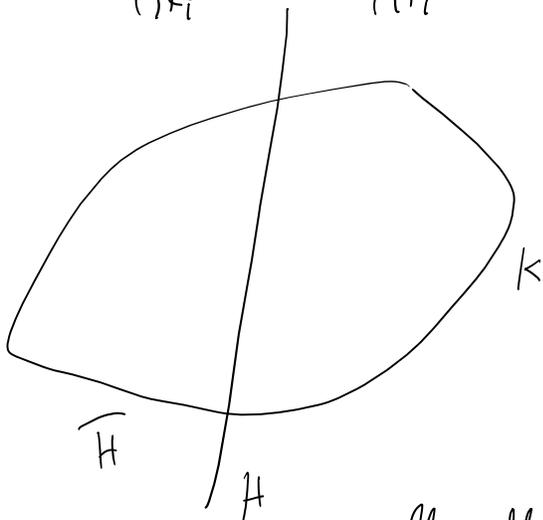
convex ϕ

argmax ϕ
 $\mu \in P_f$

$\int f > 0$ $\int g > 0$ K_0 : huge ball.

\exists sequence of convex bodies $K_0 \supseteq K_1 \supseteq K_2 \dots$

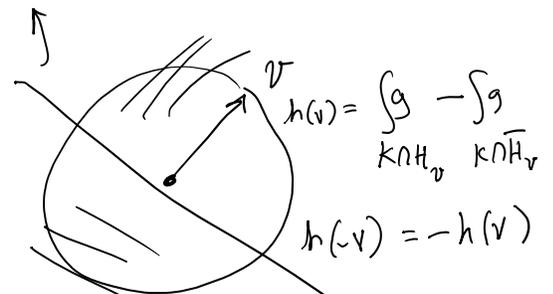
$\int_{K_i} f > 0$ $\int_{K_i} g > 0$ $\text{Lim}_{i \rightarrow \infty} \bigcap K_i \rightarrow \bullet$



$\int_{K \cap H} g = \int_{K \cap \bar{H}} g > 0$

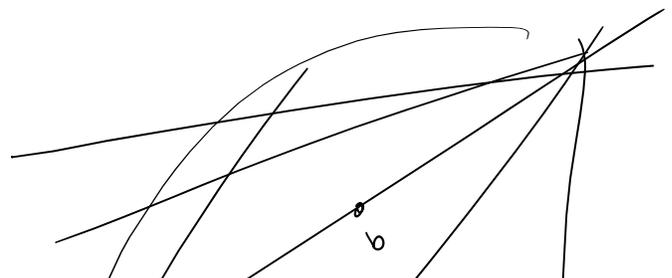
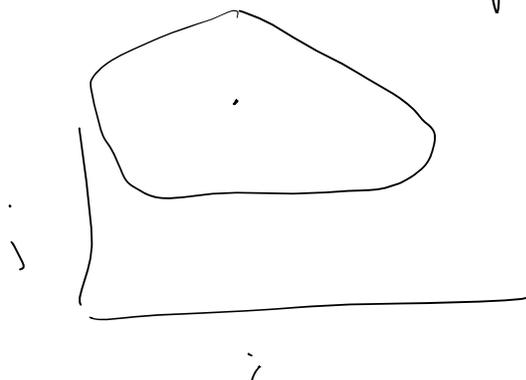
$\nu \perp A_{n-2}$

claim $\exists H$

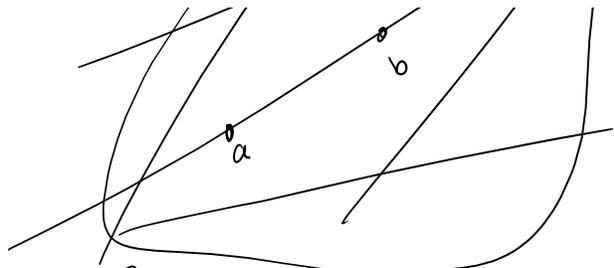


all affine subspaces with rational coordinates

$\{ x_i = r_i, y_j = r_j \}$



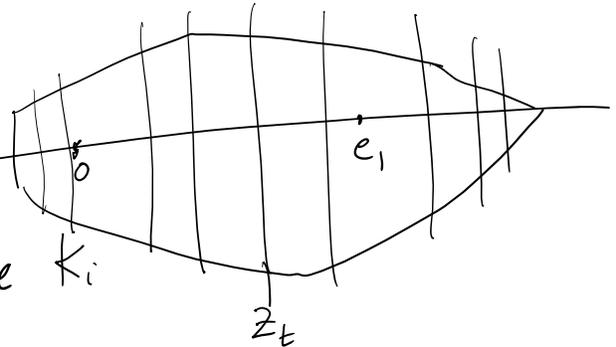
i



Claim \exists concave $\psi: [0,1] \rightarrow \mathbb{R}_+$ s.t.
 $\int_0^1 f \psi(t)^{n-1} \geq 0 \quad \int_0^1 g \psi(t)^{n-1} \geq 0$

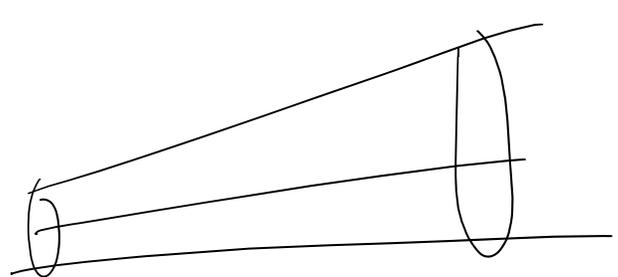
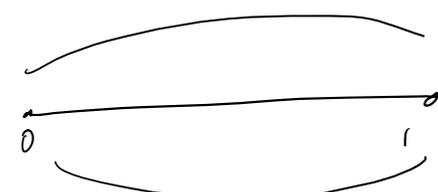
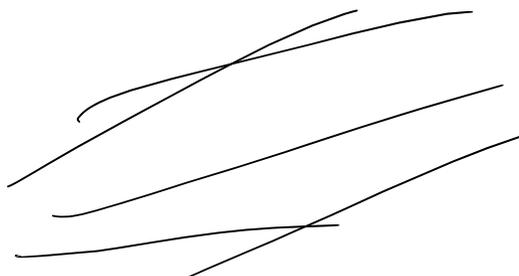
$[a,b] = [0, e_1]$

$\psi_i(t) = \left(\frac{\text{Vol}_n(K \cap Z_t)}{\text{Vol}(K)} \right)^{\frac{1}{n-1}}$

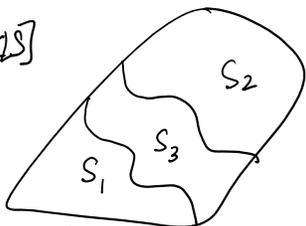


Blum-Minkowski $\Rightarrow \psi_i$ is concave K_i

Lim $\psi_i \rightarrow \psi$
 $i \rightarrow \infty$



Thm [K2]



S_1, S_2, S_3 partition of $K \in \mathbb{R}^n$

$\text{Diam}(K) = D$

$\text{Vol}(S_3) \geq \left(\frac{2}{D} d(S_1, S_2) \right) \min \{ \text{Vol}(S_1), \text{Vol}(S_2) \}$

Suppose not. Then

$\alpha \text{Vol}(S_1) > \text{Vol}(S_2)$

$\alpha \text{Vol}(S_2) > \text{Vol}(S_3)$

| \rightarrow

$f = \alpha \mathbb{1}_{S_1} - \mathbb{1}_{S_2}$

$g = \alpha \mathbb{1}_{S_2} - \mathbb{1}_{S_3}$

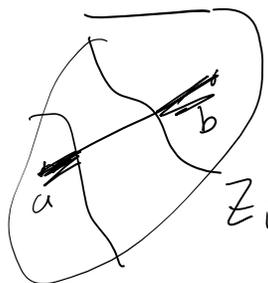
$$\alpha \text{Vol}(S_2) > \text{Vol}(S_1) \quad | \rightarrow$$

$$g = \alpha \mathbb{1}_{S_2} - \mathbb{1}_{S_1}$$

$$\int f, \int g > 0$$

$$\frac{2d(S_1, S_2)}{|a-b|} \geq \frac{2}{D} \underline{d(S_1, S_2)}$$

Apply LL. \exists needle $a, b \in \mathbb{R}^n$ s.t.



$$\int_0^1 f((1-t)a + tb) l(t)^{n-1} dt > 0$$

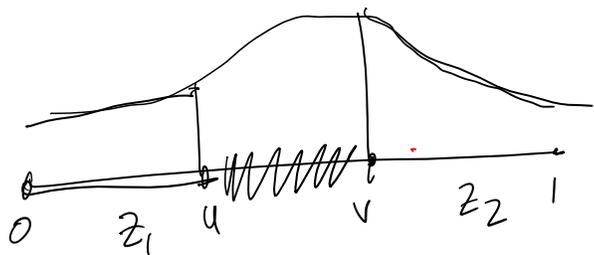
$$\int g l(t)^{n-1} > 0$$

$$Z_i = \{t \in [0, 1] : (1-t)a + tb \in S_i\}$$

$$\int_{Z_1} l(t)^{n-1} dt < \alpha \int_{Z_2} l(t)^{n-1} dt, \quad \alpha \int_{Z_2} l(t)^{n-1} dt$$



Claim suffices to prove the contrary for Z_1, Z_2, Z_3 single intervals.



$$\int_u^v f \geq 2|u-v| \min \left\{ \int_0^u f, \int_v^1 f \right\}$$

$l(t)^{n-1}$ is logconcave unimodal $f \geq |u-v|$

$$\int_u^v f \geq |u-v| \min \{f(u), f(v)\} \quad f(u) \leq f(v)$$

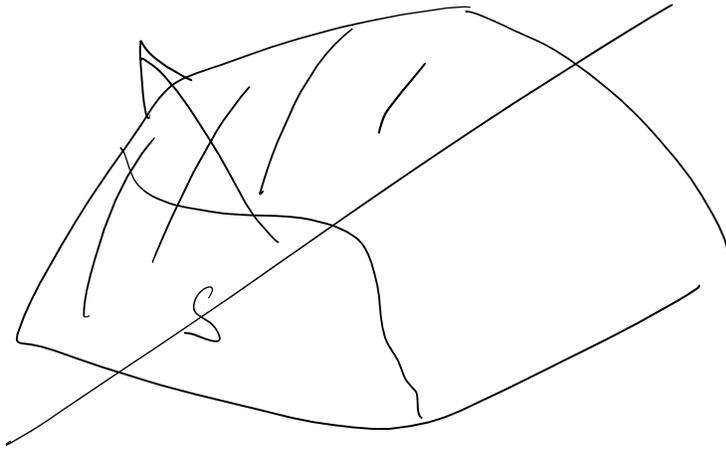
$$\int_0^u f \leq u f(u)$$

$$\begin{aligned} \int_u^v f &\geq \frac{|u-v|}{u} \cdot u f(u) \\ &\geq \frac{|u-v|}{u} \int_0^u f \\ &\geq |u-v| \end{aligned}$$

Part II: Stochastic localization.

Part II: Stochastic Localization

4/17



$$\frac{\text{Vol}(S)}{\text{Vol}(K)} = C$$

$$\text{Vol}(\partial S) \geq c \text{Vol}(S)$$

Start with an isotropic logconcave measure $p_0 = p$ on \mathbb{R}^n

$$t \geq 0 \quad p_t \quad m_t = \int x p_t(x) dx$$

$$(*) \quad dp_t(x) = \langle x - m_t, dW_t \rangle p_t(x)$$

dW_t Brownian Motion

$$W_t = N(0, tI)$$

$$E(dp_t(x)) = 0 \quad E(p_t(x)) = p_0(x)$$

Martingale

$$dx_t = b_t dt + \sigma(t) dW_t$$

$$E(W_t^2) = t \quad dW_t = \frac{W_{dt}}{\sqrt{dt}} \quad E(dW_t)^2 = dt$$

Ito's

$$df(x_t) = \langle \nabla f(x_t), dx_t \rangle + \frac{1}{2} \langle \nabla^2 f(x_t), \sigma(t) \sigma(t)^T \rangle dt$$

$$(*) \quad dp_t(x) = p_t(x) (x - m_t)^T dW_t$$

$$d \log p_t(x) = \frac{1}{p_t(x)} dp_t(x) - \frac{1}{2} \frac{\|x - m_t\|^2}{p_t(x)^2} p_t(x)^2 dt$$

$$= (x - m_t)^T dW_t - \frac{1}{2} \|x - m_t\|^2 dt$$

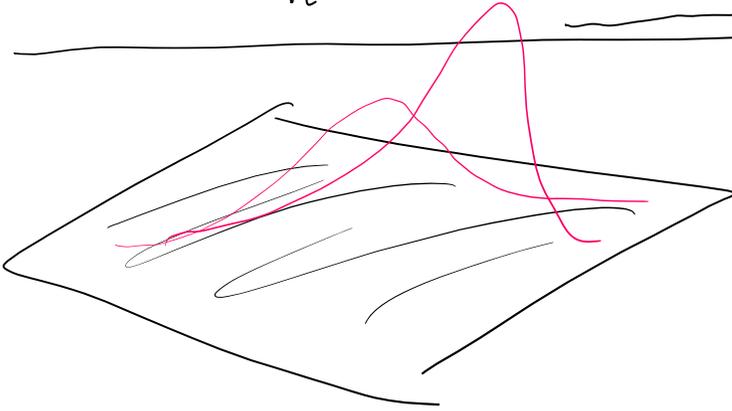
$$= -\frac{1}{2} \|x\|^2 dt + x^T (m_t dt + dW_t) - \left(m_t^T dW_t - \frac{1}{2} \|m_t\|^2 dt \right)$$

$$= -\frac{1}{2} \|x\|^2 dt + x^T (\mu_t^T dt + dw_t) - \underbrace{\mu_t^T dw_t - \frac{1}{2} \|\mu_t\|^2 dt}_{g(t)}$$

$$= -\frac{1}{2} \|x\|^2 dt + x^T (\mu_t^T dt + dw_t) - g(t) dt \quad | \quad g(t)$$

$$\log \frac{p_t(x)}{p_0(x)} = -\frac{t}{2} \|x\|^2 + c_t^T x - \bar{g}(t)$$

$$p_t(x) \propto e^{-\frac{t}{2} \|x\|^2 + c_t^T x} \cdot p_0(x)$$



p_t is t -strongly logconcave.

$$C_{PI} \leq t$$

$$\text{cov}(p_t) \preceq \frac{1}{t} I$$

Example Isoperimetry Goal: bound C_{PI} , Ψ_{KLS} of $p_0 = p$

Fix subset S . $p_0(S) = \frac{1}{2}$ Goal: bound $\frac{p_0(\partial S)}{p_0(S)}$

Apply (SL) $p_0 \rightarrow p_t$ $\frac{p_t(\partial S)}{p_t(S)} \gtrsim \sqrt{t}$

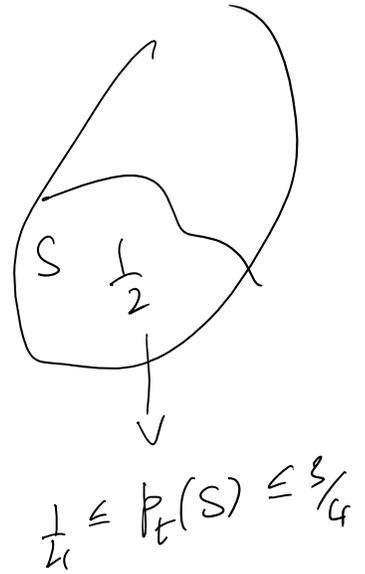
$$\rightarrow p_0(\partial S) = \mathbb{E}(p_t(\partial S)) \quad | \quad p_0(S) = \mathbb{E}(p_t(S))$$

(by C_{PI}) $(\gtrsim \sqrt{t} \mathbb{E}(p_t(S)))$

$$\rightarrow \gtrsim \sqrt{t} P_2 \left(\frac{1}{4} \leq p_t(S) \leq \frac{3}{4} \right)$$

$$\underline{d(p_t(S))} = d\left(\int_S p_t(x) dx\right) = \int_S dp_t(x) dx$$

$$= \int_S \langle x - \mu_t, dw_t \rangle p_t(x) dx$$



$$\frac{1}{4} \leq p_t(S) \leq \frac{3}{4}$$

$$d p_t(s) = \underbrace{\left(\int_S (x - \mu_t) p_t(x) dx \right)^T}_{\sigma(t)} dW_t$$

$$\sigma(t) \sigma(t)^T dt$$

$$\begin{aligned} d [p_t(s)]_t &= \left\| \int_S (x - \mu_t) p_t(x) dx \right\|^2 dt \\ &\leq \sup_{\|v\|=1} \left(\int_S v^T (x - \mu_t) p_t(x) dx \right)^2 dt \leq \sup_v \int_S (v^T (x - \mu_t))^2 p_t(x) dx dt \\ &\leq \sup_{\|v\|=1} v^T \left(\int_{\mathbb{R}^n} (x - \mu_t) (x - \mu_t)^T p_t(x) dx \right) v dt \\ &\leq \|A_t\|_{op} dt \end{aligned}$$

$$\left[p_t(s) \right]_t \leq \sup_{s \leq t} \|A_s\|_{op} \cdot t$$

Lemma For $t \leq t_0 = \frac{c}{(\log n)^2}$ $P_1(\exists s \leq t_0: \|A_s\|_{op} > 2) \leq e^{-\frac{1}{ct_0}}$

M_t Martingale
 $\forall u > 0, \sigma^2$

$$P_1(\exists t > 0, M_t > u) \leq e^{-\frac{u^2}{2\sigma^2}}$$

β

$$\frac{1}{\beta} \ln t_0 e^{\beta A_t}$$

$$|p_t(s) - p_0(s)| \leq 4 t_0$$

$$\frac{p_0(\partial S)}{p_0(S)} \geq \sqrt{t_0} = \frac{1}{\log n}$$

$$\Psi_{KLS} = \frac{O(\log n)}{\sqrt{\log n}} \quad P_I = \frac{O(\log^2 n)}{\log n}$$

$$t_0 = \frac{c}{\log n}$$

(1) Classical via SL (2) Steins \Leftrightarrow Small-ball prop

① Classical via SL

② Strong \Leftrightarrow Small-ball prop c_n
 (SB) $\underline{\underline{P_1(\|X\| \leq \varepsilon \sqrt{n}) \geq \varepsilon}}}$

Prop [Bizenl'25] For a $\Omega(1)$ -strongly logconcave p_t
 $\tau_2(A_t) = \Omega(n) \Rightarrow$ (SB) holds.
 $t=1 \rightarrow \Omega(1)$ strongly logconcave p_t .

Thm (Guan'24) WHP $\tau_2(A_t^2) = O(n)$ for $t \rightarrow \underline{\underline{c}}$.

Lemma for any $\lambda > 1$, w.p. $1 - \frac{1}{\lambda}$ $\forall S$ $p_0(S) \leq e^{cnt} \underbrace{(p_t(S))^{1/\lambda}}_{\varepsilon^{cn}}$
 $S = \varepsilon \sqrt{n} B_n$

PF. $g_t = p_t(S)$ $dg_t = \int_S \langle x - \mu_t, \underline{dw_t} \rangle p_t(x) dx$
 $d[g_t]_t = \mathbb{E} \left\| \int_S (x - \mu_t) p_t(x) dx \right\|^2 dt$
 $\leq D^2 g_t^2 dt$ $D = d(\sqrt{n})$
 $\leq cn g_t^2 dt$

$$\mathbb{E} d \log g_t \geq \mathbb{E} \frac{1}{g_t} dg_t - \frac{1}{g_t^2} cn g_t^2 dt$$

$$\mathbb{E} \log g_t \geq \log g_0 - cnt$$

$$\log g_0 \leq \frac{1}{\lambda} \log g_t + cnt$$

$$g_0 \leq e^{cnt} (g_t)^{1/\lambda}$$

Markov to $\log \frac{1}{g_t}$
 w.p. $1 - \frac{1}{\lambda}$
 $\log \frac{1}{g_t} \leq \lambda \mathbb{E}(\log \frac{1}{g_t})$
 $\frac{1}{\lambda} \log g_t \geq \mathbb{E}(\log g_t)$

$$g_0 \leq e^{c \cdot t} (g_t)^{\frac{1}{\lambda}}$$

$$\frac{1}{\lambda} \Rightarrow g_t \leq \dots$$

$$dP_t(x) = \langle x - P_t, dW_t \rangle P_t(x)$$

$$hI$$

$$P_{t+1}(x) = P_t(x) (1 + \sqrt{h} \langle x - P_t, Z \rangle)$$

$$Z \sim N(0, I)$$