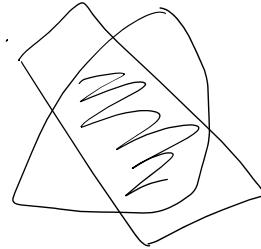


The Localization Method for High-dimensional Inequalities (TUTORIAL)

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

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

$\exists c > 0$



Thm [Klartag-Ledoux; Borell]

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

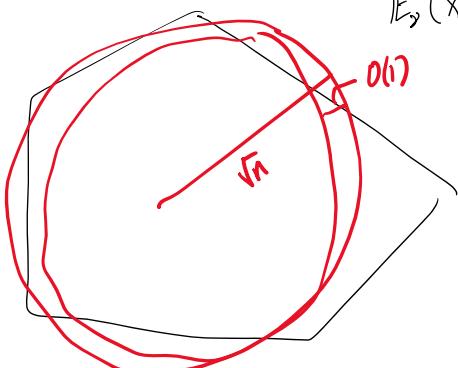
\mathcal{V} in \mathbb{R}^n log-concave f is log-concave
 $f: \mathbb{R}^n \rightarrow \mathbb{R}_+$ $\forall x, y \in \mathbb{R}^n, \lambda \in [0, 1]$
 $f((1-\lambda)x + \lambda y) \geq f(x)^{1-\lambda} f(y)^\lambda$.

\mathcal{V} is isotropic if $E_{\mathcal{V}}(x) = 0$

$$E_{\mathcal{V}}(x x^T) = I$$

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

$$E(\|x\|^2) = n \quad E_{\mathcal{V}}((\|x\| - \sqrt{n})^2) = O(1)$$



③ Large Deviation $t \geq 1$

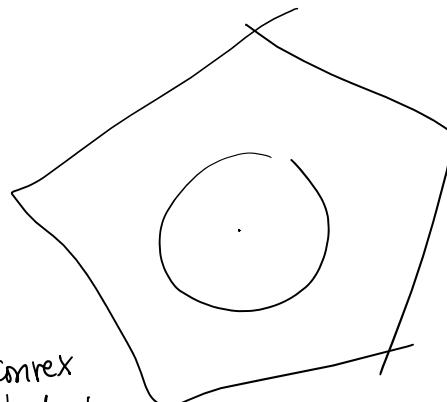
$$\text{Thm} \text{ (Paouris '06)} \quad P_{\mathcal{V}}(\|x\| > t\sqrt{n}) \leq e^{-ct\sqrt{n}}$$

Isotropic, log-concave.

④ orig. small-ball probability

$$\varepsilon < \varepsilon_0 \quad P_{\mathcal{V}}(\|x\| \leq \varepsilon\sqrt{n}) \leq \varepsilon^{cn}$$

↑
Thm (Borell)

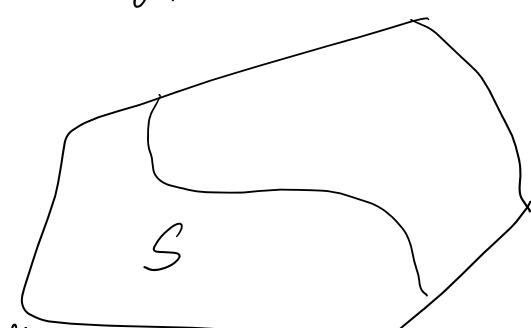


⑤ KLS conjecture. [KLS 95] isotropic, convex body K

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

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

$$\mathcal{V}: \text{log-concave} \quad \mathcal{V}(\partial S) \geq \frac{1}{\sqrt{\|A\|_{op}}} \mathcal{V}(S)$$



$$\mathcal{V}(0) = \sqrt{\|A\|_{op}}$$

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

$$\text{Thm [KLS 95]} \quad \mathcal{V}(\partial S) \geq \frac{c}{\sqrt{\text{tr}(A)}} \min \left\{ \mathcal{V}(S), \mathcal{V}(\mathbb{R}^n \setminus S) \right\}$$

$$\Psi_{KLS}^{(0)} = \sup_{S \in \mathcal{P}} \frac{\mathcal{V}(S)}{\mathcal{V}(\partial S)}$$

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

\mathcal{V} has a Poincaré Inequality with "constant" C_{PI}

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

$\leq \text{Diam}$

$$\text{For log-concave } \mathcal{V}: \quad \Psi_{KLS}^2 \lesssim C_{PI} \lesssim \Psi_{KLS}^2$$

$$\underbrace{\text{KLS}}_{\Rightarrow \text{ thin-shell} \Rightarrow \text{ slicing}} \quad \uparrow \quad \text{Thm (Klartag 23)} \quad \Psi_{KLS} = O(\sqrt{\log n})$$

small-ball

$$C_{PI} \lesssim \log n.$$

$$\textcircled{6} \quad \mathcal{V} \text{ is } t\text{-strongly log-concave} \quad \mathcal{V}(x) = e^{-\frac{t\|x\|^2}{2}} \cdot \mathcal{V}_0(x)$$

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

$$\begin{aligned} & \text{[Brascamp-Lieb]} \quad \mathbb{E}_{\mathcal{V}}(f) \leq \mathbb{E}_{\mathcal{V}}(\langle \nabla f, [\nabla^2 \log f]^{-1} \nabla f \rangle) \\ & \leq t \mathbb{E}(\|\nabla f\|^2) \end{aligned}$$

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

Log-Sobolev inequality (LSI) \mathcal{V} satisfies LSI with "constant" C_{LSI}

$$\mathbb{E}_{\mathcal{V}}(\nabla f) \leq C_{LSI} \mathbb{E}(\|\nabla f\|^2)$$

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

$$\textcircled{6} \quad \text{Certifiable hypercontractivity} \quad f(x) = \langle \langle 1, x \rangle^{\otimes K}, A \rangle$$

$$\mathbb{E}_{\mathcal{V}}(f) \leq (C, K)^{2K} \|A\|_F^2 \quad \text{isotropic, log-concave } \mathcal{V}$$

cont

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

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

(7) Polynomial anticoncentration

[Carbery-Wright] $\deg d$ $p: \mathbb{R}^n \rightarrow \mathbb{R}$ on a convex body $K \subseteq \mathbb{R}^n$.
 $d \leq n$ $\text{Var}_K(p(x)) = 1 \Rightarrow \forall \varepsilon \geq 0 \quad \Pr_K(\|p(x) - t\| \leq \varepsilon) \lesssim d \varepsilon^{\frac{1}{d}}$

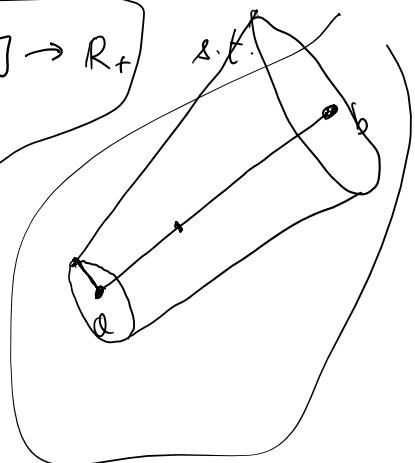
Lemma [LS93] $f, g: \mathbb{R}^n \rightarrow \mathbb{R}$ lower semi-continuous

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

Then \exists needle: $a, b \in \mathbb{R}^n$ linear fraction $\ell: [0, 1] \rightarrow \mathbb{R}_+$

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

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



Lemma [KLS95] $f_1, f_2, f_3, f_4: \mathbb{R}^n \rightarrow \mathbb{R}$ continuous.

The following are equivalent:

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

\int_E F f_1 \quad \int_E F f_2 \leq \int_E F f_3 \quad \int_E F f_4

(2) If exp. needle $a, b \in \mathbb{R}^n$ $x \in \mathbb{R}$ $\int_E f = \int_0^{(b-a)} f(a + t(b-a)) e^{xt} dt$

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

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

$P_f = \left\{ \mu \text{ logconcave} : \int_K f d\mu \geq 0 \right\}$. Then Extreme measures are

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

(1) Dirac $x^* \in K$ s.t. $f(x) \geq 0$

(2) ν s.t. $\nu(x) \propto e^{\ell(x)} \mathbb{1}_{[a,b]}$ $[a,b] \subseteq K$

$$\int f d\nu = 0 \quad \int_a^x f d\nu > 0 \quad \text{for } x \in (a, b) \text{ and} \\ \int_x^b f d\nu > 0$$

Convex

ϕ

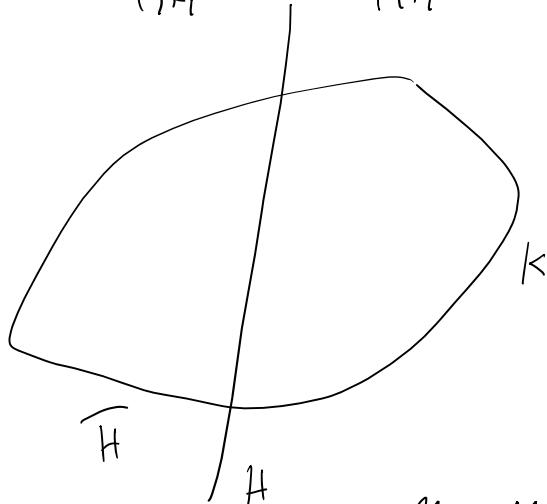
$\arg\max \phi$

$\mu \in P_f$

$$\int f > 0 \quad \int g > 0 \quad K_0: \text{huge ball.}$$

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

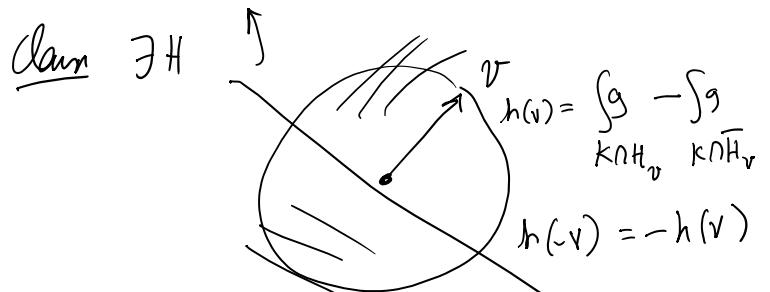
$$\int_{K_i} f > 0 \quad \int_{K_i} g > 0 \quad \lim_{i \rightarrow \infty} \bigcap K_i \rightarrow$$



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

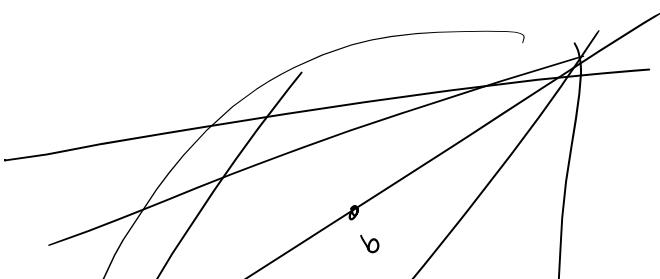
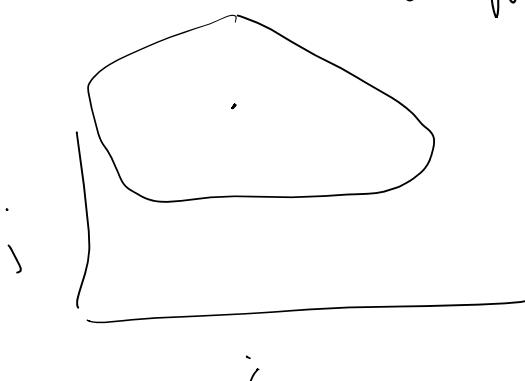
$v \perp A_{n-2}$

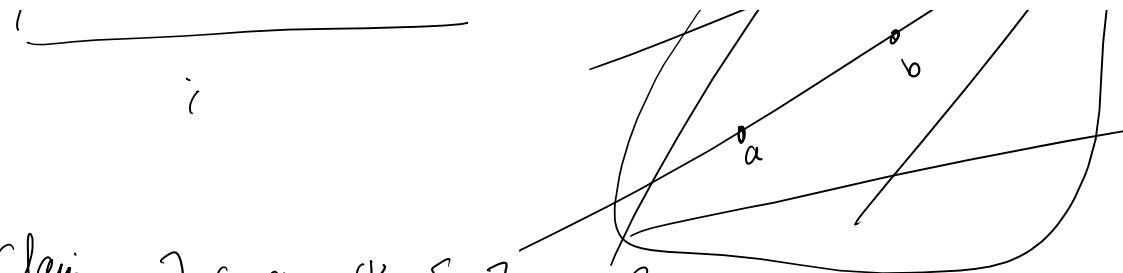
claim $\exists H$



all affine subspaces with rational coordinates

$$\{x_i = r_1, y_i = r_2\}$$





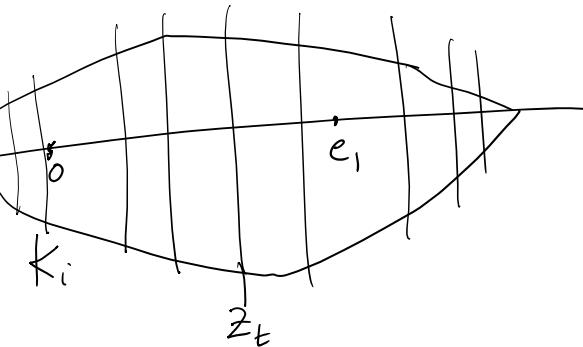
Claim \exists concave $\psi: [0, 1] \rightarrow \mathbb{R}_+$ st.

$$\int_0^1 f \psi(t)^{n-1} dt \geq 0 \quad \int_0^1 g \psi(t)^{n-1} dt \geq 0$$

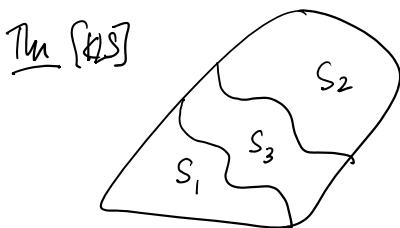
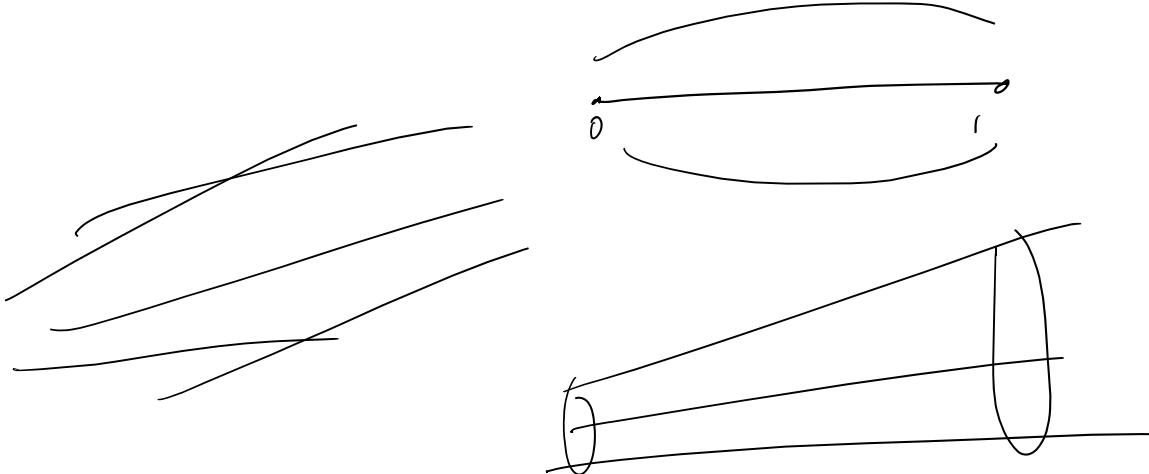
$$[a, b] = [0, e_1]$$

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

Brun-Minkowski $\Rightarrow \psi_i$ is concave



$$\lim_{i \rightarrow \infty} \psi_i \rightarrow \psi$$



The $[S_i]$ partition of $K \subseteq \mathbb{R}^n$

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

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

Spherical note. Then

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

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

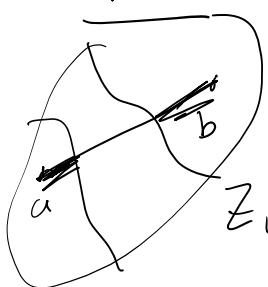
$$\begin{aligned} \alpha & f = \alpha \mathbf{1}_{S_1} - \mathbf{1}_{S_3} \\ & g = \alpha \mathbf{1}_{S_2} - \mathbf{1}_{S_3} \end{aligned}$$

$$\alpha \text{Vol}(S_2) > \text{Vol}(S_3) \quad | \rightarrow \quad g = \alpha \mathbf{1}_{S_2} - \mathbf{1}_{S_3}$$

$$\text{sf}, \int g > 0.$$

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

Appls LL. \exists needle $a, b \in \mathbb{R}^n$ $\ell(t)$ s.t.



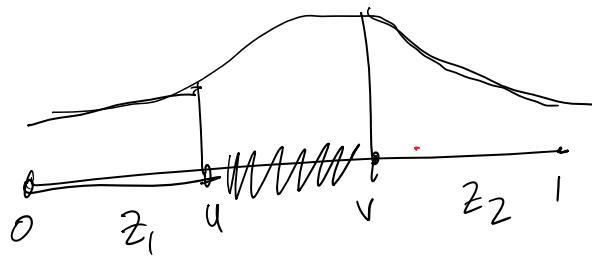
$$\int_0^1 f((1-t)a+tb) \ell(t)^{n-1} dt > 0$$

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

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

$$\int_{Z_3} \ell(t)^{n-1} dt < \alpha \int_{Z_1} \ell(t)^{n-1}, \quad \alpha \int_{Z_2} \ell(t)^{n-1}$$

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



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

$\ell(t)^{n-1}$ is log-convex unifdral $f \geq |u-v|$

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

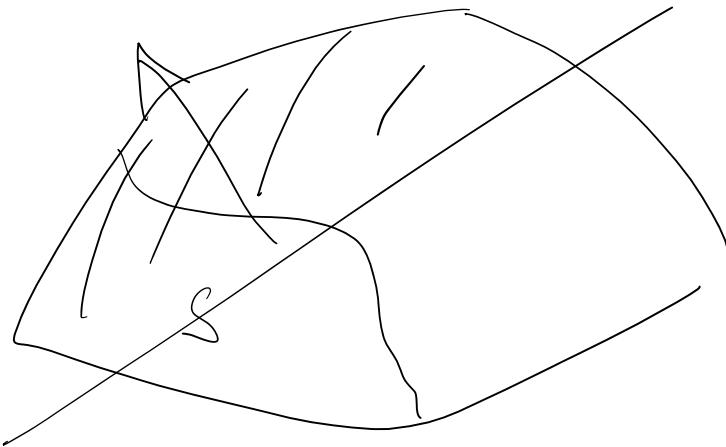
$$\int_0^u f \leq u f(u)$$

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

Part II: Stochastic Localization.

Part II : Stochastic Localization

Yours



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

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

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

$$t \geq 0 \quad p_t \quad \mathbb{E}_t(x) = \int x p_t(x) dx$$

$$(x) \quad d\mathbb{P}_t(x) = \langle x - \mathbb{E}_t, dw_t \rangle p_t(x)$$

dw_t Brownian Motion
 $w_t = N(0, tI)$

$$\mathbb{E}(d\mathbb{P}_t(x)) = 0 \quad \mathbb{E}(p_t(x)) = p_0(x) \quad \text{Martingale}$$

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

$$\mathbb{E}(w_t^2) = t \quad dw_t = \frac{w_t}{\sqrt{dt}} dt$$

$$\mathbb{E}(dw_t)^2 = (\frac{w_t}{\sqrt{dt}})^2 dt = dt$$

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

$$(*) \quad d\mathbb{P}_t(x) = p_t(x)(x - \mathbb{E}_t)^T dw_t$$

$$d \log \mathbb{P}_t(x) = \frac{1}{\mathbb{P}_t(x)} d\mathbb{P}_t(x) - \frac{1}{2} \frac{\|x - \mathbb{E}_t\|^2}{\mathbb{P}_t(x)^2} p_t(x)^T dt$$

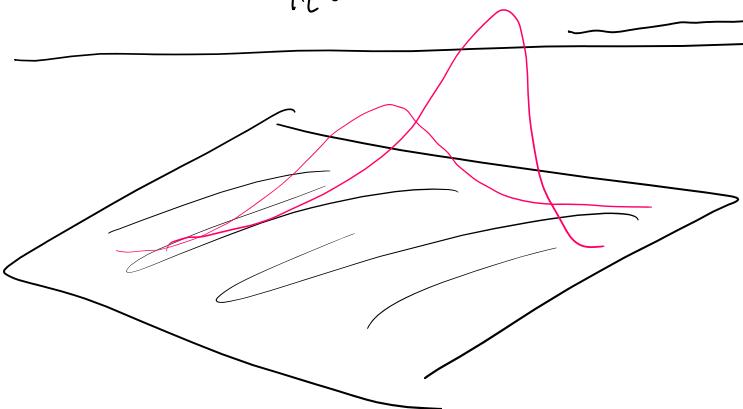
$$= (x - \mathbb{E}_t)^T dw_t - \frac{1}{2} \|x - \mathbb{E}_t\|^2 dt$$

$$= -\frac{1}{2} \|x\|^2 dt + x^T (\mathbb{E}_t dt + dw_t) - \underbrace{\mathbb{E}_t^T dw_t}_{-\frac{1}{2} \|p_t\|^2 dt} - \underbrace{\frac{1}{2} \|p_t\|^2 dt}_{-\frac{1}{2} \|x - \mathbb{E}_t\|^2 dt}$$

$$\begin{aligned}
 &= -\frac{1}{2} \|x\|^2 dt + x^T (\mu_t dt + d\omega_t) - \underbrace{\left(\mu_t^T d\omega_t - \frac{1}{2} \|\mu_t\|^2 dt \right)}_{g(t)} \\
 &= -\frac{1}{2} \|x\|^2 dt + x^T (\mu_t dt + d\omega_t) - g(t) dt
 \end{aligned}$$

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

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



p_t is t -strongly log-concave.

$$\begin{aligned}
 C_{PI} &\leq t \\
 \text{cov}(p_t) &\asymp \frac{1}{t} I
 \end{aligned}$$

Example ISoperimetry Goal : bound C_{PI} , Ψ_{KL} of $p_0 = p$

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

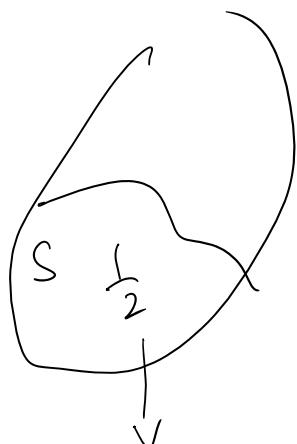
Apply (SL) $p_0 \rightarrow p_t$ $\frac{p_t(\partial S)}{p_t(S)} \gtrsim \sqrt{t}$

$\rightarrow p_0(\partial S) = \mathbb{E}(p_t(\partial S)) \quad | \quad p_0(S) = \mathbb{E}(p_t(S))$
 $(\text{by } C_{PI}) \quad (\gtrsim \sqrt{t} \mathbb{E}(p_t(S)))$

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

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

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



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

$\therefore \int_T \dots$

$$d\hat{p}_t(s) = \int_s^t \left(\int_s^x (x - \mu_t) p_t(x) dx \right)^T d\omega_t$$

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

$$\begin{aligned}
 d[\hat{p}_t(s)]_t &= \left\| \int_s^t (x - \mu_t) \hat{p}_t(x) dx \right\|^2 dt \\
 &\leq \sup_{\|v\|=1} \left(\int_s^t v^T (x - \mu_t) \hat{p}_t(x) dx \right)_t^2 \leq \sup_n \int_s^t (v^T (x - \mu_t))^2 \hat{p}_t(x) dx dt \\
 &\leq \sup_{\|v\|=1} v^T \left(\int_{\mathbb{R}^n} (x - \mu_t) (x - \mu_t)^T \hat{p}_t(x) dx \right) v dt \\
 &\leq \|A_t\|_{op} dt \\
 \int \hat{p}_t(s) ds &\leq \sup_{s \leq t} \|A_s\|_{op} \cdot t
 \end{aligned}$$

M_t Martingale
$\forall u > 0, \sigma^2$
$P_1(\exists t > 0, M_t > u)$
$\left[M_t \right] \leq \sigma^2$
$\leq e^{-\frac{u^2}{2\sigma^2}}$

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

$$|\hat{p}_t(s) - \hat{p}_0(s)| \leq 4t_0$$

$$\frac{\hat{p}_0(\partial S)}{\hat{p}_0(S)} \geq \sqrt{t_0} = \frac{1}{\log n}$$

$$\Psi_{KLS} = O(\log n) \quad PI = O(\log^2 n)$$

$$\sqrt{\log n}$$

$$\log n$$

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

(1) Classical via SL

(2) Glauis \Leftrightarrow small-ball prob

① Classical via SL

② Shows \Leftrightarrow small-ball prop
 $P_1(\|X\| \leq \varepsilon \sqrt{n}) \leq \varepsilon^c n$

Prop [Bizen'25] For a $\mathcal{Q}(1)$ -strongly log-concave p_t
 $\text{tr}(A_t) = Q(n)$ \Rightarrow (SB) holds.
 $t=1 \rightarrow \mathcal{Q}(1)$ strongly log-concave p_t .

The (Guarn '24) WHP $\text{tr}(A_t^2) = O(n)$ for $t \rightarrow c$.

Lemma For any $\lambda > 1$, w.p. $1 - \frac{1}{\lambda}$ $\forall s$ $p_s(s) \leq e^{cnt} (\bar{p}_t(s))^{\frac{1}{\lambda}}$
 $S = \varepsilon \sqrt{n} B_n$

PF.

$$g_t = p_t(s) \quad dg_t = \int_S \langle x - \bar{p}_t, dw_t \rangle p_t(x) dx$$

$$d[g_t]_t = \left\| \int_S (x - \bar{p}_t) p_t(x) dx \right\|^2 dt$$

$$\leq D^2 g_t^2 dt \quad D = O(\sqrt{n})$$

$$\leq cn g_t^2 dt.$$

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

$$|\mathbb{E} \log g_t| \geq \log g_0 + cnt$$

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

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

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

$$\frac{1}{\lambda} \log g_t \geq \mathbb{E}(\log g_t)$$

$$g_0 \leq e^{\text{cnt}} \left(g_t \right)^{\frac{1}{\lambda}}$$

$$dp_t(x) = \langle x - p_t, d\omega_t \rangle p_t(x) \quad hI$$

$$p_{t+1}(x) = p_t(x) \left(1 + \sqrt{h} \langle x - p_t, z \rangle \right) \quad z \sim N(0, I)$$