

Hypercontractivity of spherical averages in Hamming space

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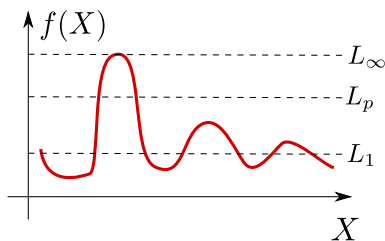
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Hypercontractivity: definition

- L_p -norms of random variables:

$$\|f(x)\|_{L_p(\mu)} \triangleq \left(\int |f(x)|^p \mu(dx) \right)^{\frac{1}{p}}$$
$$\|U\|_p \triangleq (\mathbb{E}[|U|^p])^{\frac{1}{p}}$$



- Let P_{XY} – joint distribution.
Define $T_{Y|X}$ – stochastic (Markov) averaging operator

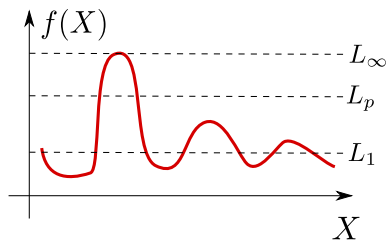
$$T_{Y|X} f(x) \triangleq \mathbb{E}[f(Y)|X = x]$$

$$\Rightarrow \|Tf\|_p \leq \|f\|_p \text{ by Jensen's.}$$

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$$\implies \|Tf\|_p \leq \|f\|_p \text{ by Jensen's.}$$

- T is hypercontractive if $\|T\|_{p \rightarrow q} = 1$ for $p < q$:

$$\|Tf\|_q \leq \|f\|_p, \quad \|T\|_{p \rightarrow q} \triangleq \sup_f \frac{\|Tf\|_q}{\|f\|_p}$$

Tf gains integrability! (by smoothing peaks)

- T is HC for some $p \iff T$ is HC for all p .

- Nelson-Gross: (X, Y) – **jointly Gaussian**:

$$\|\mathbb{E}[f(Y)|X]\|_4 \leq \|f(Y)\|_2 \iff \rho(X, Y) \leq \frac{1}{\sqrt{3}}.$$

- Bonami-Gross: $(X, Y) \sim \frac{1}{2} \begin{bmatrix} 1 - \delta & \delta \\ \delta & 1 - \delta \end{bmatrix} \triangleq \text{BSS}(\delta)$:

$$\|\mathbb{E}[f(Y)|X]\|_q \leq \|f(Y)\|_p \iff \frac{p-1}{q-1} \geq (1-2\delta)^2$$

- Bonami-Beckner:

$$\|T\|_{p \rightarrow q} = 1 \iff \|T^{\otimes n}\|_{p \rightarrow q} = 1$$

- Ahslwede-Gács: for finite X, Y pretty much **any $T_{Y|X}$ is HC**

What is HC good for?

Canonical example:

- Estimate probability of a rectangle $\mathbb{P}[X \in A, Y \in B]$, where

$$\mathbb{P}[X \in A] \approx \mathbb{P}[Y \in B] = \theta \ll 1.$$

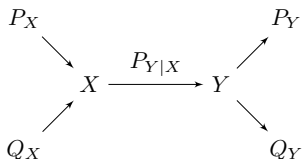
- Spectral gap (expanders, etc): $1 = \sigma_1(T) > \sigma_2(T) > \dots$

$$\mathbb{P}[X \in A, Y \in B] \leq \theta^2 + \sigma_2 \theta (1 - \theta) \approx \sigma_2 \cdot \theta$$

- Hypercontractivity:

$$\begin{aligned} \mathbb{P}[X \in A, Y \in B] &= (1_A, T1_B) \\ &\leq \|1_A\|_{q'} \|T1_B\|_q \leq \|1_A\|_{q'} \|1_B\|_p \\ &= \theta^{1 - \frac{1}{q} + \frac{1}{p}} = \theta^{1+\epsilon} \ll \sigma_2 \theta \quad (!) \end{aligned}$$

Note: Used 0/1 nature of indicator functions.

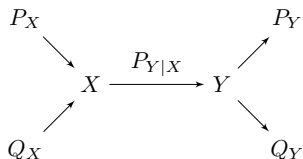


- Apply to typical sets:

$$\begin{aligned} 2^{-nD(Q_X \| P_X)} &\approx \mathbb{P}[X^n \in T_{[Q_X]}^n, Y^n \in T_{[Q_Y]}^n] \\ &\lesssim 2^{-n \frac{1}{q} D(Q_X \| P_X) - n \frac{1}{p} D(Q_Y \| P_Y)} \end{aligned}$$

- So we get:

$$\|T_{Y|X}\|_{p \rightarrow q} = 1 \implies \boxed{D(Q_Y \| P_Y) \leq \frac{p}{q} D(Q_X \| P_X) \quad \forall Q_X}$$

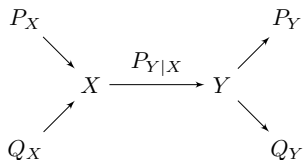


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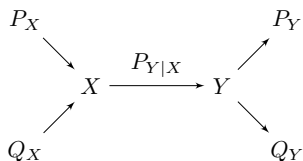
- So we get:

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- For fixed P_{XY} , **KL-contraction ratio**:

$$\eta_{\text{KL}}(P_{XY}) \triangleq \sup_{Q_X} \frac{D(Q_Y || P_Y)}{D(Q_X || P_X)} = \sup_{U \rightarrow X \rightarrow Y} \frac{I(U; Y)}{I(U; X)}$$



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Theorem (Ahlsvede-Gács)

$$\eta_{\text{KL}}(P_{XY}) = \inf \left\{ \frac{p}{q} : \|T_{Y|X}\|_{p \rightarrow q} = 1 \right\}$$

- Roughly:

$$\forall Q_X : D(Q_Y \| P_Y) \leq r D(Q_X \| P_X) \implies \boxed{\|T_{Y|X}\|_{p \rightarrow \frac{p}{r}} = 1, p \gg 1}$$

[Nair'2014]:

- Apply HC to typical sets of *arbitrary* Q_{XY} to get

$$\|T_{Y|X}\|_{p \rightarrow q} = 1 \implies$$

$$\boxed{\frac{1}{p}D(Q_Y \| P_Y) \leq \frac{1}{q}D(Q_X \| P_X) + D(Q_{Y|X} \| P_{Y|X} | Q_X) \quad \forall Q_{XY}}$$

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- Proof: Take $\|g_1\|_1 = 1$ and let $f = (g_1)^{\frac{1}{p}} \in L_p$
- STP: $\|Tf\|_q \leq 1$.

Ahlsvede-Gács:

$$Q_X(x) = P_X(x) \frac{Tf(x)^q}{\|Tf\|_q^q}$$

$$Q_{Y|X}(y|x) = P_{Y|X}(y|x)$$

Nair:

$$Q_X(x) = \dots \text{ same } \dots$$

$$Q_{Y|X}(y|x) = P_{Y|X}(y|x) \cdot \frac{f(y)}{Tf(x)}$$

$$\log \|Tf\|_q \leq \frac{1}{p}D(Q_Y \| P_Y) - \frac{1}{q}D(Q_X \| P_X) - D(Q_{Y|X} \| P_{Y|X} | Q_X) \leq 0$$

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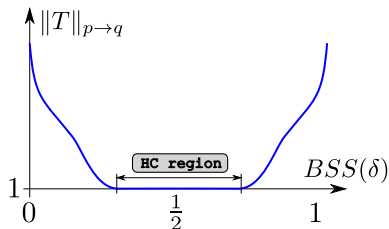
$$\log \|Tf\|_q \leq \frac{1}{p}D(Q_Y \| P_Y) - \frac{1}{q}D(Q_X \| P_X) - D(Q_{Y|X} \| P_{Y|X} | Q_X) \leq 0$$

- Note: For $p \gg 1$ both have $Q_{Y|X} \approx P_{Y|X}$.

$L_p \rightarrow L_q$ norm as measure of dependence

$\|T_{Y|X}\|_{p \rightarrow q}$ measures dependence:

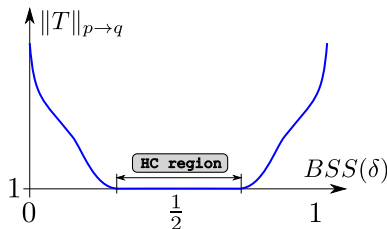
- Invariant to relabeling
- Satisfies data-processing
- Unusual: “sticky at 1”.



$L_p \rightarrow L_q$ norm as measure of dependence

$\|T_{Y|X}\|_{p \rightarrow q}$ measures dependence:

- Invariant to relabeling
- Satisfies data-processing
- Unusual: “sticky at 1”.
- Turns out the last one is a general phenomenon:



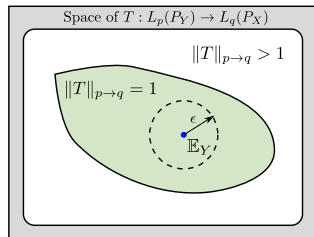
All operators T sufficiently noisy are $p \rightarrow q$ HC.

Theorem

For any $q \leq 2 \leq p$ there is $\epsilon = \epsilon(p, q)$ s.t.
simultaneously for all P_{XY} :

$$\|T_{Y|X} - \mathbb{E}_Y\|_{p \rightarrow q} \leq \epsilon \quad \Rightarrow \quad \|T_{Y|X}\|_{p \rightarrow q} = 1$$

where $\mathbb{E}_Y f \triangleq 1_X \cdot \mathbb{E}[f(Y)]$.



Remarks:

- Also for general $q < p$ with extra condition: $\|T - \mathbb{E}\|_{2 \rightarrow 2} \leq \epsilon$
- Also for other operators of the type:

$$T1_Y = 1_X, \quad (1_X, T \cdot) = (1_X, \mathbb{E} \cdot)$$

- Orig. result is for $P_X = P_Y$, but proof easily generalizes.
- For finite (X, Y) , suff. to check

$$\|P_{XY} - P_X P_Y\|_{TV} \leq \epsilon' = \epsilon'(p, q, P_X, P_Y)$$

Proof for $q \leq 2 \leq p$:

- WLOG can take $f = 1 + Z$ with $Z \perp 1$. Decompose $T = \mathbb{E} + \Delta$. STP:

$$(*) \quad \|1 + \Delta Z\|_q \leq \|1 + Z\|_p$$

- **Cool fact:** \exists universal τ, a, b s.t.

$$\|1 + Z\|_q \leq 1 + a\|Z\|_q^2 \quad \text{if } \|Z\|_q \leq \tau$$

$$\|1 + Z\|_p \geq 1 + b\|Z\|_p^2 \quad \text{if } \|Z\|_p \leq \tau$$

- Whenever $\|\Delta\|_{p \rightarrow q}^2 \leq \frac{b}{a}$ and $\|Z\|_p \leq \tau$:

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- Whenever $\|\Delta\|_{p \rightarrow q}^2 \leq \frac{b}{a}$ and $\|Z\|_p \leq \tau$:

$$\|1 + \Delta Z\|_q \leq 1 + a\|\Delta Z\|_q^2 \leq \|1 + Z\|_p$$

- Large Z : Another **cool fact:** \exists universal $\beta_p(\cdot)$

$$\|1 + Z\|_p \geq 1 + \beta_p(\|Z\|_p), \quad \text{and } \frac{\beta_p(u)}{u} \text{--increasing}$$

So $(*)$ holds if $\|\Delta\|_{p \rightarrow q} \leq \frac{\beta_p(\tau)}{\tau}$.

Spherical averages in Hamming space

- Hamming space \mathbb{F}_2^n with uniform probability
- Bonami-Gross operator

$$N_\delta f(x) \triangleq \mathbb{E}[f(x + Z)], \quad Z \sim \text{Bern}(\delta)^n$$

- Spherical average:

$$T_\delta f(x) \triangleq \mathbb{E}[f(x + Z)], \quad Z \sim \text{Unif}(\mathbb{S}_{\delta n})$$

where $\mathbb{S}_j \triangleq \{y : \text{wt}(y) = j\}$.

Theorem (P'13)

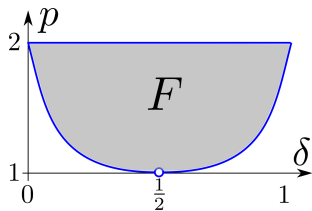
$\forall \delta \neq \frac{1}{2}$ and $p = 1 + (1 - 2\delta)^2$ there is a *dim-independent* $C_{p,\delta}$ s.t.

$$\|T_\delta f\|_2 \leq C_{p,\delta} \|f\|_p$$

Remark: Bonami-Gross $\|N_\delta f\|_2 \leq 1 \cdot \|f\|_p$

Hypercontractivity of T_δ : exact version

$$F = \left\{ (\delta, p) : \begin{array}{l} 1 + (1 - 2\delta)^2 \leq p \leq 2 \\ 0 \leq \delta \leq 1 \\ p \neq 1 \end{array} \right\}$$



Theorem (P'13)

For any compact subset $K \subset F$ there is a constant C_K s.t.

$$\|T_\delta\|_{p \rightarrow 2} \leq C_K \quad \forall (\delta, p) \in K,$$

where $T_\delta f = f * P_Z$, $P_Z = \text{Unif}(\mathbb{S}_{\delta n})$.

For $p = 1$, $\delta = 1/2$:

$$\|T_{1/2}\|_{p \rightarrow 2} = \theta(n^{1/4}).$$

Note: for Bernoulli noise $\|N_\delta\|_{p \rightarrow 2} = 1$ everywhere on **closure** of F .

Why is constant not 1?

- Let $1_{\text{even}} = 1\{x : \text{wt}(x)\text{-even}\}$
- Then:

$$T_\delta 1_{\text{even}} = 1_{\text{even}} \quad (\text{or } 1_{\text{odd}})$$

- So:

$$\|T_\delta\|_{p \rightarrow q} \geq 2^{\frac{1}{p} - \frac{1}{q}} > 1.$$

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Beautiful methods that failed:

- Tensorization (T_δ not a tensor power)
- Semigroup method, aka $\frac{d}{d\delta}$ (T_δ not a semigroup)
- Comparison of Dirichlet forms (not useful for $\delta \gg \frac{1}{n}$)

Ugly method that worked:

- Direct computation of eigenvalues of T_δ

- T_δ and N_δ are S_n -equivariant \Rightarrow

$$f(x) = \text{poly}((-1)^{x_1}, \dots, (-1)^{x_n}) = \sum_{a=0}^n f_a(x),$$

where $f_a(x) =$ degree a part of the poly.

- By odd/even trick can assume $f_a = 0$ for $a \geq n/2$.
- Have:

$$N_\delta f_a = (1 - 2\delta)^a \cdot f_a$$

$$T_\delta f_a = (\quad ?? \quad) \cdot f_a$$

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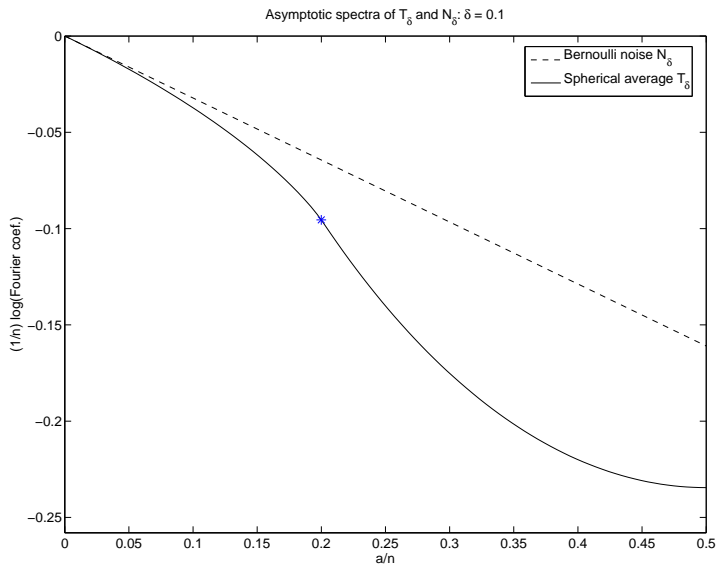
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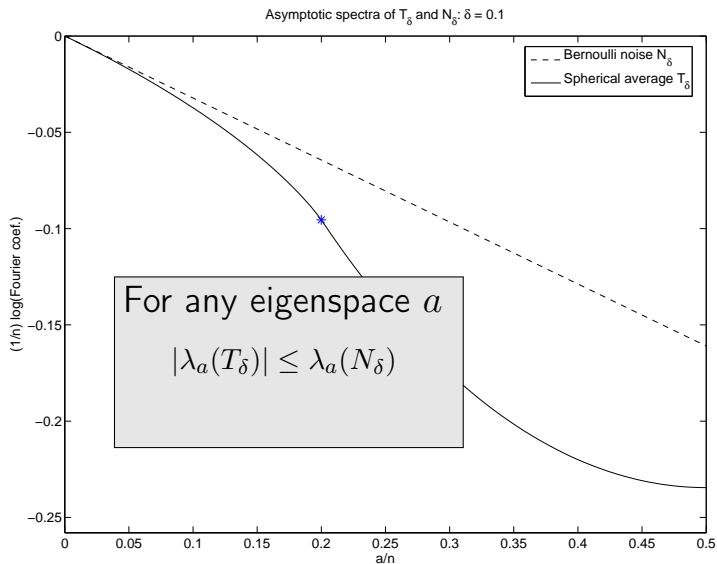
$$T_\delta f_a = \kappa_{\delta n}(a) \cdot f_a$$

and $\kappa_{\delta n}(a)$ – Krawtchouk polynomial of degree δn .

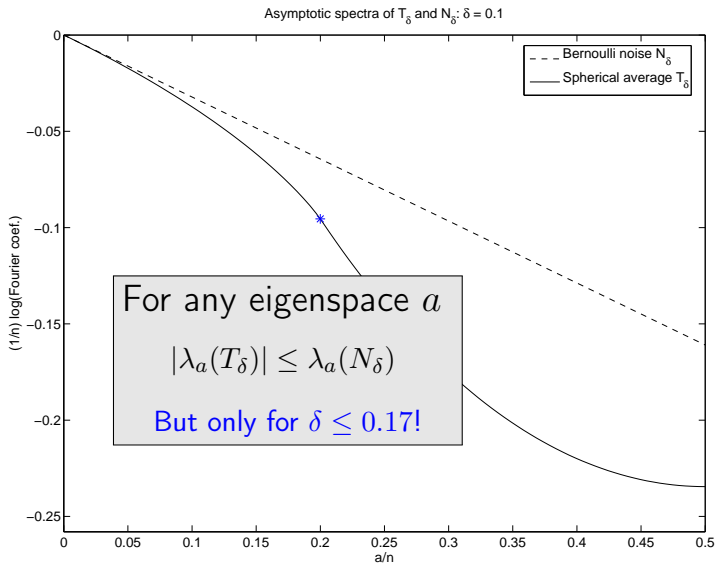
Proof (cont'd): Eigenvalues of N_δ vs T_δ



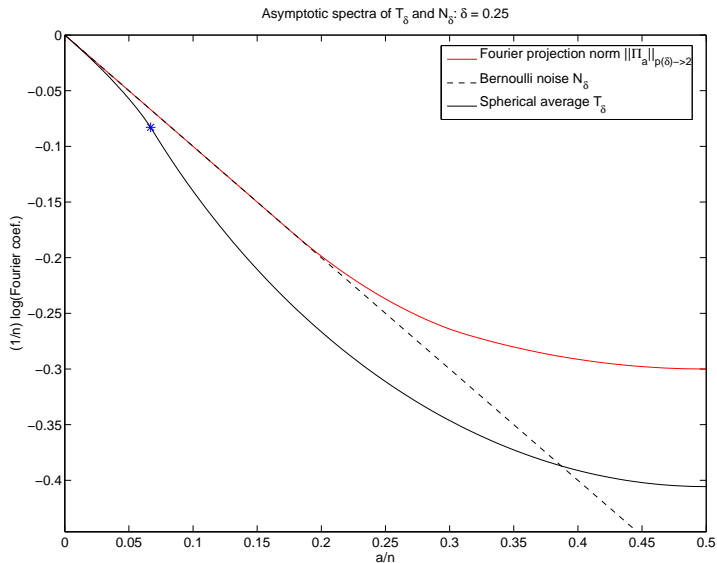
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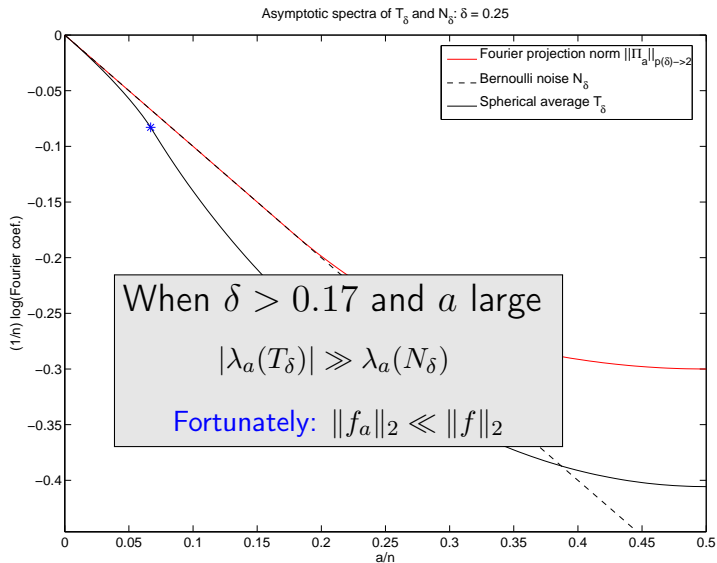
Proof (cont'd): Eigenvalues of N_δ vs T_δ



Proof (cont'd): Failure for $\delta \gtrsim 0.17$



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Proof (cont'd): Overall argument

- Know:

$$\|N_\delta f\|_2^2 = \sum_a (1 - 2\delta)^{2a} \|f_a\|_2^2 \leq \|f\|_p^2 \quad \text{by Bonami!}$$

- Need to show:

$$\|T_\delta f\|_2^2 = \sum_a \kappa_{\delta n}(a)^2 \|f_a\|_2^2 \leq C \|f\|_p^2$$

- $a \leq a_{crit}$:

$$\begin{aligned} \sum_{a \leq a_{crit}} \kappa_{\delta n}(a)^2 \|f_a\|_2^2 &\leq C \sum_a (1 - 2\delta)^{2a} \|f_a\|_2^2 \\ &= C \|N_\delta f\|_2^2 \leq C \|f\|_p^2 \end{aligned}$$

- $a > a_{crit}$: Need extra estimate:

$$\kappa_{\delta n}(a) \|f_a\|_2 \lesssim e^{-Cn} \|f\|_p$$

$$\implies \sum_{a \geq \xi_{crit} n} \text{contributes} \leq C \|f\|_p^2.$$

Proof (cont'd): Overall argument

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- $a \leq a_{crit}$: For $\delta < 0.17$: $a_{crit} = n/2$!

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Known fact:

Subspace $L \subset \mathbb{F}_2^n$ s.t. $|L \cap \mathbb{S}_{\delta n}| \sim |\mathbb{S}_{\delta n}|$ implies $\dim L \approx n$.

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Corollary

Let $A \subset \mathbb{F}_2^n$ be a subset s.t. sumset $A + A$ intersects some $\mathbb{S}_{\delta n}$ with average multiplicity $> \lambda|A|$. Then

$$|A| \geq \lambda^\epsilon \cdot 2^n.$$

Proof:

- Statement means

$$2^n(1_A * 1_A, 1_{\mathbb{S}_{\delta n}}) \geq \lambda \cdot |A| \cdot |\mathbb{S}_{\delta n}|$$

- Rearrange $(1_A * 1_A, 1_{\mathbb{S}}) = (T_\delta 1_A, 1_A)$
- Apply Hölder and HC.

Thank you!

References

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- total variation

$$\|P - Q\|_{\text{TV}} = \frac{1}{2} \int |dP - dQ|$$



- Dobrushin coefficient

$$\eta_{\text{TV}} = \sup_{P_X \neq Q_X} \frac{\|P_Y - Q_Y\|_{\text{TV}}}{\|P_X - Q_X\|_{\text{TV}}} = \sup_{x, x'} \|P_{Y|X=x} - P_{Y|X=x'}\|_{\text{TV}}$$

- Original motivation: mixing of Markov chains

Consider a M.C. with invariant dist. P^* :

$$X_0 \rightarrow X_1 \rightarrow X_2 \rightarrow \dots$$

Contraction:

$$\|P_{X_n} - P^*\|_{\text{TV}} \leq (\eta_{\text{TV}})^n \quad (\text{Dobrushin})$$

$$\chi^2(P_{X_n} \| P^*) \leq (\eta_{\chi^2})^n \cdot \chi^2(P_{X_0} \| P^*) \quad (\text{spectral gap})$$

$$D(P_{X_n} \| P^*) \leq (\eta_{\text{KL}})^n \cdot D(P_{X_0} \| P^*) \quad (\text{log-Sobolev})$$

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TODO: add mixing estimate via HC

From TV to KL ... and further

Theorem (Cohen-Iwasa-Rautu-Ruskai-Seneta-Zbăganu '93)

$$\eta_{\text{KL}} \leq \eta_{\text{TV}}$$

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Fun application:

- Let T – kernel of a **finite-state M.C.** with invariant distribution P^* .
- Let \mathbb{E} be indep. M.C.: $\mathbb{E}f \triangleq \int f dP^*$
- If $\|T(x, \cdot) - \mathbb{E}(x, \cdot)\|_{\text{TV}} < 1/2$ then

$$\eta_{\text{TV}} = \max_{x, x'} \|T(x, \cdot) - T(x', \cdot)\|_{\text{TV}} < 1$$

- Then $\eta_{\text{KL}} < 1$. By Ahlswede-Gács $\exists q \gg 1$

$$\|Tf\|_q \leq \|f\|_p, \quad p = \eta_{\text{KL}} \cdot q < q \quad (\text{hypercont})$$

- \Rightarrow **There is a ball around \mathbb{E} s.t. for all T : $\|T\|_{L_p \rightarrow L_q} = 1$.**

Theorem (Cohen-Iwasa-Rautu-Ruskai-Seneta-Zbăganu '93)

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- Let T – kernel of a **finite-state M.C.** with invariant distribution P^* .
- Let \mathbb{E} be indep. M.C.: $\mathbb{E}f \triangleq \int f dP^*$
- If $\|T(x, \cdot) - \mathbb{E}(x, \cdot)\|_{\text{TV}} < 1/2$ then

$$\eta_{\text{TV}} = \max_{x, x'} \|T(x, \cdot) - T(x', \cdot)\|_{\text{TV}} < 1$$

- Then $\eta_{\text{KL}} < 1$. By Ahlswede-Gács $\exists q \gg 1$

$$\|Tf\|_q \leq \|f\|_p, \quad p = \eta_{\text{KL}} \cdot q < q \quad (\text{hypercont})$$

- \Rightarrow **There is a ball around \mathbb{E} s.t. for all T : $\|T\|_{L_p \rightarrow L_q} = 1$.**
- \Rightarrow Every (mixing) M.C. eventually (hyper-)contracts $L_p \rightarrow L_q$.

Theorem (Cohen-Iwasa-Rautu-Ruskai-Seneta-Zbăganu '93)

$$\eta_{\text{KL}} \leq \eta_{\text{TV}}$$

Fun application:

- Let T – kernel of a **finite-state M.C.** with invariant distribution P^* .
- Let \mathbb{E} be indep. M.C.: $\mathbb{E}f \triangleq \int f dP^*$
- If $\|T(x, \cdot) - \mathbb{E}(x, \cdot)\|_{\text{TV}} < 1/2$ then

$$\eta_{\text{TV}} = \max_{x, x'} \|T(x, \cdot) - T(x', \cdot)\|_{\text{TV}} < 1$$

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- \Rightarrow Every (mixing) M.C. eventually (hyper-)contracts $L_p \rightarrow L_q$.
- **Euclidean space:** planar face of the ball of Fourier multipliers: [Segal'70], [Fefferman-Shapiro'72], [Semenov-Shneiberg'88]