

The Kadison-Singer Problem

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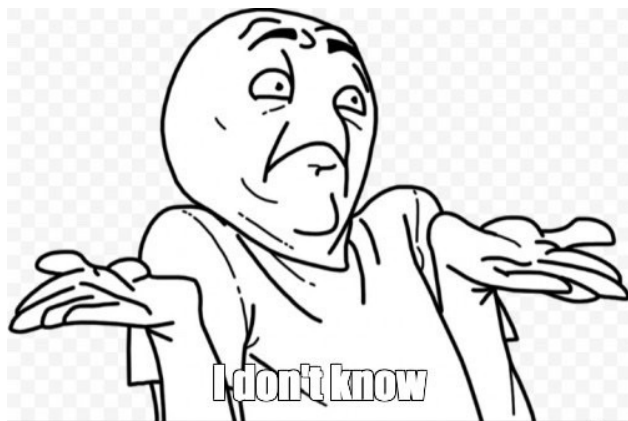
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KADISON-SINGER PROBLEM

Does every pure state on the (abelian) von Neumann algebra \mathbb{D} of bounded diagonal operators on ℓ_2 have a unique extension to a pure state on $B(\ell_2)$, the von Neumann algebra of all bounded operators on ℓ_2 ?

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Does every pure state on the (abelian) von Neumann algebra \mathbb{D} of bounded diagonal operators on ℓ_2 have a unique extension to a pure state on $B(\ell_2)$, the von Neumann algebra of all bounded operators on ℓ_2 ?



EQUIVALENTLY, ...

* Thm: If $\varepsilon > 0$ and v_1, \dots, v_m are independent random vectors in \mathbb{C}^n with finite support s.t.

$$\mathbb{E} \sum_i v_i v_i^* = I_n, \quad \text{and}$$

$$\mathbb{E} \|v_i\|^2 \leq \varepsilon, \quad \forall i, \text{ then}$$

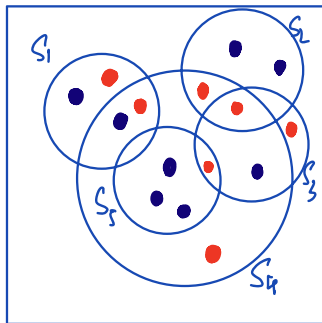
$$\mathbb{P} \left[\left\| \sum_{i=1}^m \hat{v}_i \hat{v}_i^* \right\| \leq \underbrace{(1 + \sqrt{\varepsilon})^2}_{=} \right] > 0.$$

• Rmk: Concentration Ineq $\Rightarrow \underbrace{\| \sum v_i v_i^* \|}_{=}$ $\leq (C\varepsilon) \cdot \log(n)$ whp.

DISCREPANCY THEORY

- (SPENCER): Given sets $S_1, \dots, S_n \subset [n]$, colour elements of $[n]$ Red = Blue s.t.

$$\forall S_i : \quad \left| |S_i \cap R| - |S_i \cap B| \right| \leq 6\sqrt{n}$$



UNIFORMLY PARTITIONING VECTORS

Thm: Given vectors $v_1, \dots, v_m \in \mathbb{R}^n$, satisfying

$$\|v_i\|^2 \leq \alpha, \quad \text{and}$$

$$\sum_{i=1}^m \langle v_i, x \rangle^2 = 1, \quad \forall \|x\| = 1 \quad \sum v_i v_i^T = I$$

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there exists partition $T_1 \cup T_2 = [m]$:

$$\left| \sum_{i \in T_j} \langle v_i, x \rangle^2 - \frac{1}{2} \right| \leq \underbrace{5\sqrt{\alpha}}_{\hookrightarrow \text{Discrepancy}}, \quad \forall \|x\| = 1$$

- non trivial guarantee only if $5\sqrt{\alpha} < \frac{1}{2}$

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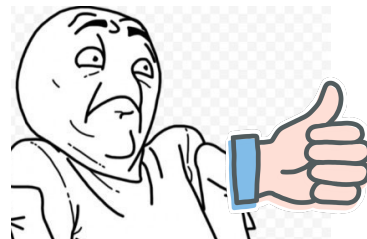
$$\sum_{i=1}^m \langle v_i, x \rangle^2 = 1, \quad \forall \|x\| = 1$$

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Does every pure state on the (abelian) von Neumann algebra \mathbb{D} of bounded diagonal operators on ℓ_2 have a unique extension to a pure state on $B(\ell_2)$, the von Neumann algebra of all bounded operators on ℓ_2 ?



INTERPRETING: Uniformly Partitioning Vectors

• Norm Bounds $\|v_i\|^2 \leq \alpha$ is necessary:

• Suppose v_1, \dots, v_m s.t. $\sum v_i v_i^T = I$, but
 $\|v_1\|^2 = \frac{3}{4}$, $\forall \|v_i\|^2 \leq \alpha$.

\Rightarrow Partition T_1 contains $v_1 \Rightarrow$

$$\sum_{i \in T_1} \langle v_i, x \rangle^2 \geq \|v_1\|^2 = \frac{3}{4}.$$

\therefore This partition has discrepancy at least $\frac{1}{4}$.

\Rightarrow No way to get closer to $\frac{1}{2}$ with splitting v_1 .

UNIFORMLY PARTITIONING VECTORS

Thm: Given vectors $v_1, \dots, v_m \in \mathbb{R}^n$, satisfying

$$\|v_i\|^2 \leq \alpha, \quad \text{and}$$

$$\sum_{i=1}^m \langle v_i, x \rangle^2 = 1, \quad \forall \|x\| = 1$$

there exists partition $T_1 \cup T_2 = [m]$:

$$\left| \sum_{i \in T_j} \langle v_i, x \rangle^2 - \frac{1}{2} \right| \leq 5\sqrt{\alpha}, \quad \forall \|x\| = 1$$

• Theorem says that the ONLY obstacle to obtaining low disc. solution is large vectors.

• Rank : Disc. $O(\sqrt{\alpha})$ is tight.

$$\sum_{i=1}^m v_i v_i^T = I \quad \Rightarrow \quad \text{ISOTROPY CONDITION}$$

(a normalization)

$w_1, \dots, w_m \in \mathbb{R}^n$ not isotropic.

s.t. $\text{Span}(w_1, \dots, w_m) = \mathbb{R}^n$.

$$\Rightarrow W = \sum_{i=1}^m w_i w_i^T \quad \Rightarrow \text{invertible}$$

$$\forall v_i = W^{-\frac{1}{2}} w_i$$

\Downarrow

$$\sum v_i v_i^T = W^{-\frac{1}{2}} \left(\sum w_i w_i^T \right) W^{-\frac{1}{2}} = I.$$

$$\forall \|v_i\|^2 = \|W^{-\frac{1}{2}} w_i\|^2.$$

$$\underline{\|v_i\|^2 = \|w^{\frac{1}{2}} w_i\|^2} \Rightarrow \text{Interpretation}$$

$$\begin{aligned}\|v_i\|^2 &= \|w^{\frac{1}{2}} w_i\|^2 = \sup_{x \neq 0} \frac{\langle x, w^{\frac{1}{2}} w_i \rangle^2}{x^T x} \\ &= \sup_{y = w^{\frac{1}{2}} x \neq 0} \frac{\langle w^{\frac{1}{2}} y, w^{\frac{1}{2}} w_i \rangle^2}{y^T w y} \\ &= \sup_{y \neq 0} \frac{\langle y, w_i \rangle^2}{\sum_i \langle y, w_i \rangle^2}.\end{aligned}$$

$\dots \|v_i\|^2$ measures max fraction of quadratic form of w that a single vector w_i can be responsible for.

Thm: Given vectors $v_1, \dots, v_m \in \mathbb{R}^n$, satisfying

$$\|v_i\|^2 \leq \alpha, \quad \text{and}$$

$$\sum_{i=1}^m \langle v_i, x \rangle^2 = 1, \quad \forall \|x\| = 1$$

there exists partition $T_1 \cup T_2 = [m]$:

$$\left| \sum_{i \in T_j} \langle v_i, x \rangle^2 - \frac{1}{2} \right| \leq 5\sqrt{\alpha}, \quad \forall \|x\| = 1.$$

Idea 1: Randomly partitioning the vectors.
+
Matrix-Chernoff. } $\Rightarrow O(\sqrt{\alpha \log n})$
whp

REMOVING log FACTOR

To remove the log factor, we use the following theorem:

* Thm: If $\alpha > 0$ and v_1, \dots, v_m are independent random vectors in \mathbb{R}^n with finite support s.t.

$$\sum_{i=1}^m \mathbb{E} \hat{v}_i \hat{v}_i^{\top} = \frac{\mathbf{I}_n}{2} \quad \text{and}$$
$$\mathbb{E} \|\hat{v}_i\|^2 \leq \alpha, \quad \forall i, \text{ then}$$

$$\mathbb{P} \left[\left\| \sum_{i=1}^m \hat{v}_i \hat{v}_i^{\top} \right\| \leq (1 + \sqrt{\alpha})^2 \right] > 0$$



PROOF SKETCH

Theorem says: \exists pt. $\omega \in \Omega$ (Prob. sp.)
c.t.

$$\left\| \sum_{i \in M} \hat{v}_i(\omega) \hat{v}_i(\omega)^T \right\| \leq \underline{(1 + \sqrt{K})^2}$$

For every $\omega \in \Omega$, consider polynomial

$$P[\omega](x) := \det \left(xI - \sum_{i \in M} \hat{v}_i(\omega) \hat{v}_i(\omega)^T \right).$$

Note: $\sum \hat{v}_i \hat{v}_i^T$ is a symmetric

• $\left\| \sum \hat{v}_i \hat{v}_i^T \right\|$ is the largest root of the characteristic polynomial.

• characteristic polynomial has real roots.

PROOF SKETCH

STEP 1:

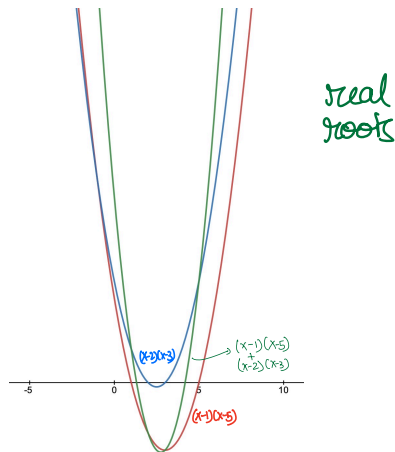
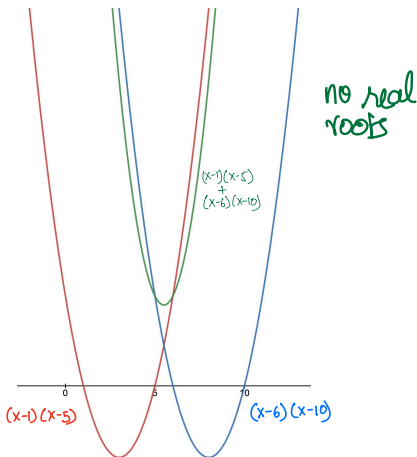
$\exists w \in \Omega$ s.t.

$$P[w](x) := \det\left(xI - \sum_{i \in M} v_i(w) v_i(w)^T\right)$$

$$\lambda_{\max}(P[w]) \leq \lambda_{\max}(EP) \rightarrow \text{probabilistic method vibes}$$

- Roots of sum of polynomials don't have much to do with roots of individual polynomials.

Eg:



STEP 2: Upper bound roots of expected polynomials

$$\mu(x) := \mathbb{E}P(x)$$

• $\mu(x)$:= linear transform of m -variate polynomial
 $Q(z_1, \dots, z_m)$

• Q does not have any roots in certain region of $\mathbb{R}^m \rightarrow$ and use barrier functions.



uses theory of "real stable"
polynomial.


• We'll focus on STEP 1 & show it is sufficient to bound roots of expected characteristic polynomial.

INTERLACING

f : degree n polynomial with real roots $\{\alpha_i\}$.

g : degree n or $n-1$ with all real roots $\{\beta_i\}$

g interlaces f if their roots alternate


$$\beta_n \leq \alpha_n \leq \beta_{n-1} \leq \dots \leq \beta_1 \leq \alpha_1.$$

NOTATION: $g \longrightarrow f \Rightarrow$ largest root belongs to f .

- If a single g interlaces a family f_1, \dots, f_m
 \Rightarrow have a common interlacing.

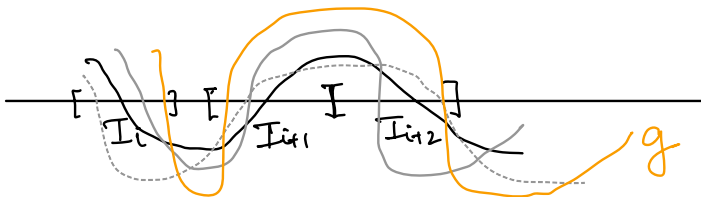
- f_1, \dots, f_n has common interlacing \Leftrightarrow Every pair has common interlacing

\Updownarrow
 $I_n \leq I_{n-1} \leq \dots \leq I_1$ closed intervals
 i -th root of f_j is contained in I_i .

Thm: Suppose f_1, \dots, f_m are real-rooted of degree n with positive leading coeff. $\lambda_k(f_j)$: k -th largest root of f_j
 μ be any dist on $[m]$.

If f_1, \dots, f_m have common interlacing then $\forall k=1, \dots, n$

$$\min_j \lambda_k(f_j) \leq \lambda_k\left(\prod_{j \in \mu} f_j\right) \leq \max_j \lambda_k(f_j).$$



Proof:

- At some point, it is all positive in I_i
- At some point, it is all negative in I_i

\therefore This bound holds \square

Interlacing helps us achieve step 4:

STEP 4: $\exists w \in \Sigma$ s.t. $P[w](x) := \det(xI - \sum_{i \in M} v_i(w)v_i(w)^T)$

$\lambda_{\max}(P[w]) \leq \lambda_{\max}(EP) \rightarrow$ Probabilistic method
vibes

FINDING COMMON INTERLACER (Annoying)

Translate "interlacing" to "real-rootedness".

* Thm: Let $\{f_i\}$ be degree n monic polynomials.

The following are equivalent:

[A.] All convex combinations $\sum \mu_i f_i$ has d -real roots.

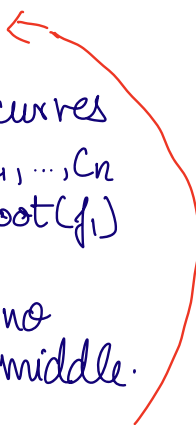
[B.] The collection $\{f_i\}$ has common interlacer.

The condition [A.] is easy to work with.

Pf: Common interlacing is pairwise phenomenon.
consider polynomials f_0 & f_1

(can skip)

$$f_t := (1-t)f_0 + tf_1, \quad t \in [0, 1]$$

- f_0 & f_1 \Rightarrow no common roots, wlog. 
- t varies, roots of f_t : n continuous curves in \mathbb{C}^n plane G_1, \dots, G_n beginning at root of f_0 & ending at root of f_1
- Curves must lie in real line, and no curve can cross roots of f_0 or f_1 in middle.

i.e., $f_t(x) = 0$ & $f_0(x) = 0 \Rightarrow f_1(x) = 0$ ~~\Rightarrow~~

- \therefore Each curve is an interval (non-overlapping)
 \Rightarrow Common interlacing.

RECALL THE THEOREM

* Thm: If $\alpha > 0$ and $\hat{v}_1, \dots, \hat{v}_m$ are independent random vectors in \mathbb{R}^n with finite support s.t.

$$\sum_{i=1}^m \mathbb{E} \hat{v}_i \hat{v}_i^{\top} = \mathbf{I}_n, \quad \text{and}$$
$$\mathbb{E} \|\hat{v}_i\|^2 \leq \alpha, \quad \forall i, \text{ then}$$

$$\mathbb{P} \left[\left\| \sum_{i=1}^m \hat{v}_i \hat{v}_i^{\top} \right\| \leq (1 + \sqrt{\alpha})^2 \right] > 0. \quad (*)$$

- Each \hat{v}_i is a random vector, and we need to show that there exists non-0 prob. of (*) happening.

Let

$$\Theta(\hat{v}_1, \dots, \hat{v}_n) = \min_{v_i \in \text{supp}(\hat{v}_i)} \lambda_{\max} \left(\sum_i v_i v_i^\top \right)$$

$$\Theta(\hat{v}_1, \dots, \hat{v}_n) \leq (1 + \sqrt{\alpha})^2$$

Ideally: If all resulting characteristic polynomial

had common interlacer, we could give this

guarantee. \Rightarrow $\{f_i\}$ all possible characteristic polynomials

· Common interlacer \Rightarrow all cvx. combination real roots

· $\mathbb{E} P$ is cvx. combination & has real roots.

Too much to hope for

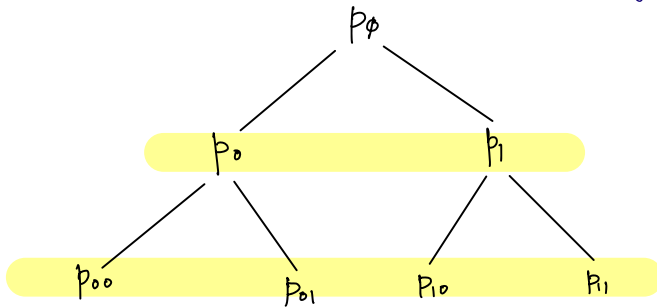
• Thm: Suppose f_1, \dots, f_m are real-rooted of degree n with positive leading coeff. $\lambda_k(f_i)$: k -th largest root of f_i .
 μ be any dist on $[m]$.

If f_1, \dots, f_m have common interlacer

$$\min_j \lambda_k(f_j) \leq \lambda_k(\mathbb{E}_{\mu} f_j) \leq \max_j \lambda_k(f_j).$$

INTERLACING FAMILY

$$\theta(\hat{v}_1, \dots, \hat{v}_n) = \min_{v_i \in \text{supp}(\hat{v}_i)} \lambda_{\max}(\sum v_i v_i^T)$$



- Def: Interlacing family:

Connected tree, where each node is the common interlacer with its children

\Rightarrow Every interlacing family contains leaf nodes $p_{\text{leaf}_1}, p_{\text{leaf}_2}$ s.t.

$$\lambda_k(p_{\text{leaf}_1}) \leq \lambda_k(p_{\text{top}}) \leq \lambda_k(p_{\text{leaf}_2})$$

PUTTING IT TOGETHER

If there exists interlacing family with

$$\{ \chi_{\Sigma_{v_i, v_i^T}}(x) \}_{v_i \in \text{supp}(\hat{v}_i)} \quad \text{as leaf nodes}$$

AND

$$\mathbb{E} \{ \chi_{\Sigma_{v_i, v_i^T}}(x) \} \quad \text{as top node}$$

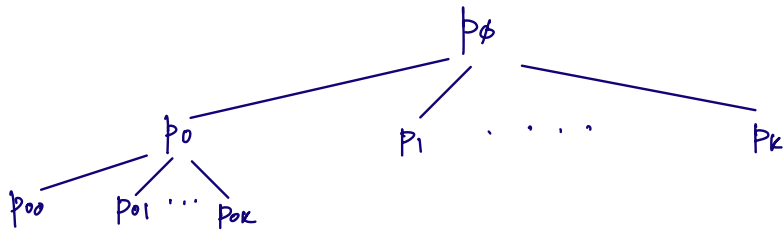
Then

$$\Theta(\hat{v}_1, \dots, \hat{v}_n) \leq \max_{\text{root}} \{ \mathbb{E} \{ \chi_{\Sigma_{v_i, v_i^T}}(x) \} \}.$$

BUILDING SUCH TREES

Let $\hat{V} = \sum v_i v_i^T$

going down tree: Revealing value of each \hat{v}_i



$$\sum v_i v_i^T$$

where

$$P_{s_1 s_2 \dots s_r} = \mathbb{E} \left\{ X_{\hat{V}} \mid \hat{v}_1 = s_1, \dots, \hat{v}_r = s_r \right\}$$

\Rightarrow Siblings at depth r , differ at \hat{v}_r .

* Theorem: Let $\hat{v}_1, \dots, \hat{v}_m$ be indep rand vec s.t.

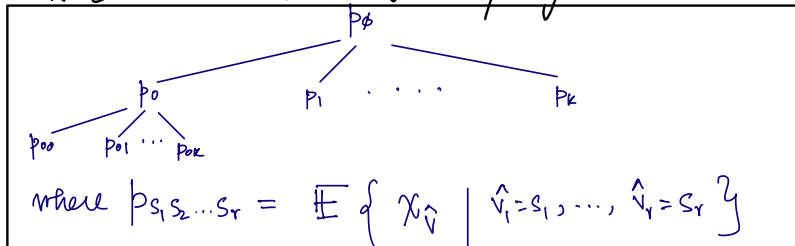
$$\mathbb{E}[\hat{v}_i \hat{v}_i^T] = A_i \quad . \quad \text{Then}$$

$$\mathbb{E}\left\{\chi_{\hat{v}}(x)\right\} = \prod_{i=1}^m \left(1 - \frac{\partial}{\partial z_i}\right) \det \left[x\mathbb{I} + \sum_{i=1}^m z_i A_i \right] \Big|_{z_1 = \dots = z_m = 0}$$

→ Expectation depends only on expected outer prod of random vec.

Call this mixed characteristic polynomial denoted by $\mu[A_1, \dots, A_n](x)$

- Claim: Every polynomial we saw in a tree is a mixed characteristic polynomial.



- Leaf polynomials: ($\sigma_i = v_i$ for $i \in [m]$)

$$p_\phi(x) = \chi_{\sum v_i v_i^T}(x) = \mu[v_1 v_1^T, \dots, v_m v_m^T](x)$$

- Top polynomials: $\mathbb{E}[\chi_{\hat{v}}(x)] = \mu[A_1, \dots, A_m](x)$

- Middle polynomials: ($\sigma_i = v_i$ for $i \in [k]$)

$$p_{s'} = \mathbb{E}[\chi_{\hat{v}}(x) \mid \hat{v}_i = \sigma_i, i \in [k]]$$

TBC.