

Best conditioned bases and minimal projections

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Preliminaries

where the concepts are introduced and the problems raised.

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A **minimal projection** from X onto V is a projection $P : X \twoheadrightarrow V$ such that

$$\|P\| = p(V, X).$$

Best conditioned basis

The ℓ_p -condition number of a basis $\underline{v} = (v_1, \dots, v_n)$ of a normed space $(V, \|\bullet\|)$ is

$$\kappa_p(\underline{v}) := \sup_{a \in \ell_p^n \setminus \{0\}} \frac{\|\sum_{i=1}^n a_i v_i\|}{\|a\|_p} \times \sup_{a \in \ell_p^n \setminus \{0\}} \frac{\|a\|_p}{\|\sum_{i=1}^n a_i v_i\|}$$

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and a **best ℓ_p -conditioned basis** \underline{v} of V is a basis such that $\kappa_p(\underline{v}) = \kappa_p(V)$.

In the terminology of Banach space geometry, the **Banach-Mazur** distance between X and Y is defined by

$$d(X, Y) := \inf\{\|T\| \times \|T^{-1}\|, T : X \rightarrow Y \text{ isomorphism}\}.$$

Optimal normalization of a basis

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Therefore, $\kappa_\infty(\underline{v}^N) \leq \kappa_\infty(\underline{v})$.

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Proposition. Let V be a subspace of a normed space X , and consider an embedding $i : X \hookrightarrow C(K)$, where K is a compact Hausdorff space, then

$$p(V, X) \leq p(i(V), C(K)).$$

Interlude

where our liking for extreme points is revealed.

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$$\text{Ex}(B_{V^*}) \subseteq \text{Ex}(B_{X^*})|_V := \{\lambda|_V, \lambda \in \text{Ex}(B_{X^*})\}.$$

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Lemma (Auerbach). V be a n -dimensional space, there exist (v_1, \dots, v_n) basis of V and $(\lambda_1, \dots, \lambda_n)$ basis of V^* such that

$$\forall i, j \in \{1, \dots, n\}, \|v_i\| = 1, \lambda_i \in \text{cl}(\text{Ex}(B_{V^*})) \text{ (hence } \|\lambda_i\| = 1) \text{ and } \lambda_i(v_j) = \delta_{i,j}.$$

Transition

where we return to our original aim.

The case $p(V, C(K)) = \kappa_\infty(V) = 1$

For V n -dimensional subspace of $C(K)$, there exist $t_1, \dots, t_n \in K$ (the **Fekete points**) and a basis (v_1, \dots, v_n) of V such that

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Proposition (Wulbert). If V is a n -dimensional WS-subspace of $C(K)$ such that $p(V, C(K)) = 1$, then $\kappa_\infty(V) = 1$ (and also $p_{\text{int}}(V, C(K)) = 1$).

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The case $p(V, C(K)) = \kappa_\infty(V) = 1$

For V n -dimensional subspace of $C(K)$, there exist $t_1, \dots, t_n \in K$ (the **Fekete points**) and a basis (v_1, \dots, v_n) of V such that

$$\forall i, j \in \{1, \dots, n\}, \|v_i\| = 1, \quad \text{and} \quad v_j(t_i) = \delta_{i,j}.$$

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It is possible to find projections of norm 1 onto WS-subspaces, and those must be unique.

V and X	minimal projection	best ℓ_∞ -conditioned basis
	$p(V, X)$	$\kappa_\infty(V)$
	uniqueness	uniqueness
ℓ_2^n and ℓ_2	orthogonal	orthonormal
	1	\sqrt{n}
	yes	yes
\mathcal{T}_n and $L_p(\mathbb{T})$	Fourier projection	?
	$\frac{1}{\pi} \int_{\mathbb{T}} D_n $?
	yes	?
\mathcal{T}_1 and $C(\mathbb{T})$	Fourier projection	equidistant Lagrange
	1.435991...	$\frac{5}{3}$
	yes	yes
\mathcal{P}_1 and $L_1([-1, 1])$...	$(x \mapsto \frac{1}{2}, x \mapsto x)$
	1.220404...	$\frac{5}{4}$
	yes	yes
\mathcal{P}_2 and $C([-1, 1])$
	1.220173...	1.248394...?
	?	yes?

The heart of the matter

where the problems of existence, uniqueness and characterization are approached.

Existence and Uniqueness

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No contradiction: if V is a 1-dimensional subspace of X , then $p(X, V) = 1$.

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when C and K are convex, K compact, the function $f : C \times K \rightarrow \mathbb{R}$ is convex with respect to both variables, $f(x, \bullet)$ is continuous for any $x \in C$ and $(f(\bullet, \lambda))_{\lambda \in D}$ is equicontinuous at x^* .

Applications

Proposition (Cheney). If V is a Haar subspace of X , and if $P : X \rightarrow V$ is a minimal projection with $\|P\| > 1$, then there exists $\lambda_1, \dots, \lambda_{n+1} \in \text{Ex}(B_{X^*})$ ($\lambda_i \neq \pm \lambda_j$, for $i \neq j$) such that $\|P^*(\lambda_i)\| = \|P\|$.

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$$p_{\text{g.int}}(V, C(K)) := \inf \{ \|P\|, P : C(K) \rightarrow V \text{ generalized interpolating projection} \}.$$

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Proposition (Odyniec-Lewicki). P a discrete projection from $C(K)$ onto V ,

$$[P \text{ is minimal}] \iff$$

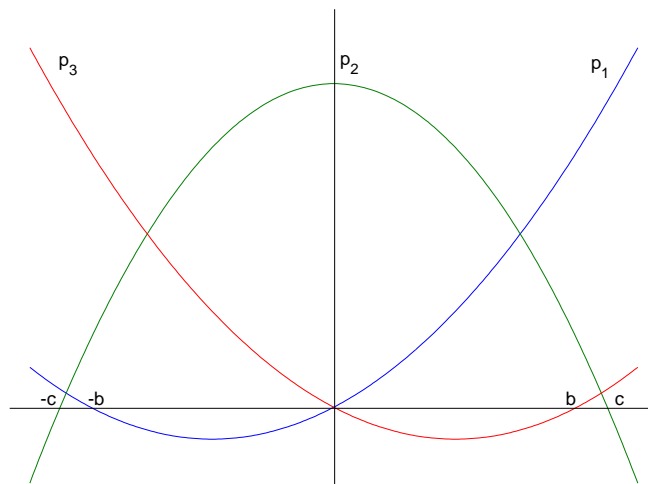
$$\left[\forall Q : C(K) \rightarrow V \text{ projection}, \exists x \in S_{C(K)}, \exists t \in K : |Q(x)(t)| \geq P(x)(t) = \|P\| \right].$$

Tailpiece

where we briefly explain how to obtain $\kappa_\infty(\mathcal{P}_2)$.

Let $\underline{p} := \underline{p}(b, c, d)$ be the following symmetric basis of \mathcal{P}_2 , where $b, c, d \in (0, +\infty)$:

$$p_1(x) := \frac{x(x+b)}{2d}, \quad p_2(x) := c^2 - x^2, \quad p_3(x) := \frac{x(x-b)}{2d}, \quad x \in [-1, 1].$$



Theorem. $\min_{b,c,d>0} \kappa_{\infty}(\underline{p(b,c,d)}) \approx 1.248394563 < \frac{5}{4} = p_{\text{int}}(\mathcal{P}_2, C([-1, 1]))$.

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The dual basis of \underline{p} has the expression, for $f \in \mathcal{P}_2$, with $C := c^2$ and $t^* := \frac{C(1+b)}{b+C}$,

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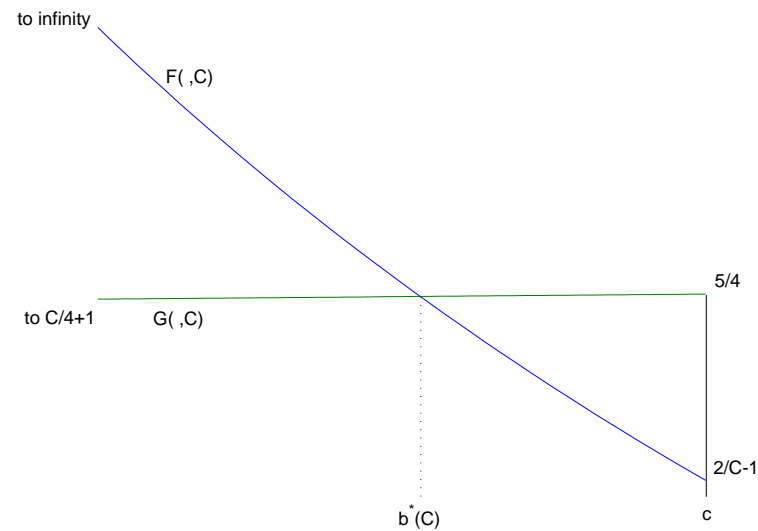
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Obtaining $\kappa_{\infty}(\underline{p}) < \frac{5}{4}$ is only possible if $b \leq c \leq 1$. In this case, the optimal normalization of \underline{p} leads to the choice

$$d = \frac{1}{\lambda} = \frac{(b+C)^2 - C(C-b^2)}{(b+C)^2 + C(C-b^2)}.$$

We then have to minimize $\max(F(b, C), G(b, C))$, where

$$F(b, C) := \frac{\lambda + 1}{C} - 1 \quad \text{and} \quad G(b, C) := \frac{b^2 \lambda^2}{4C} + 1.$$



Therefore, we have to minimize $F(b^*(C), C) = G(b^*(C), C) =: H(C)$ for $C \in \left[\frac{8}{9}, 1\right]$.

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The optimal (b, C) must be solution of the following (polynomial) system:

$$\begin{cases} F(b, C) - G(b, C) &= 0, \\ \left[\frac{\partial F}{\partial b} \frac{\partial G}{\partial C} - \frac{\partial F}{\partial C} \frac{\partial G}{\partial b} \right] (b, C) &= 0. \end{cases}$$

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Using the Groebner package from Maple, this system is equivalent to

$$\begin{aligned} 144C^8 + 6498C^7 + 25839C^6 - 25108C^5 + 9827C^4 - 17192C^3 + 2336C^2 + 1088C - 192 &= 0, \\ 60b^8 - 906b^7 - 1452b^6 + 2261b^5 + 6451b^4 + 568b^3 - 3704b^2 - 1408b - 192 &= 0. \end{aligned}$$

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In the prescribed domain, this has the unique solution

$$C \approx 0.9402938300 \quad \text{and} \quad b \approx 0.8675381234.$$

THE END