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Article

Degeneracy of the Operator-Valued Poisson Kernel Near the Numerical Range Boundary

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Abstract

Let $A \in \mathbb{C}^{d \times d}$ and let $W(A)$ denote its numerical range. For a bounded convex domain $\Omega \subset \mathbb{C}$ with C^1 boundary containing $\text{spec}(A)$, consider the operator-valued boundary kernel $P_\Omega(\sigma, A) := \text{Re}(n_\Omega(\sigma)(\sigma I - A)^{-1})$, $\sigma \in \partial\Omega$, where $n_\Omega(\sigma)$ is the outward unit normal at σ . For convex Ω with $W(A) \subset \Omega$ this kernel is strictly positive definite on $\partial\Omega$ and underlies boundary-integral functional calculi on convex domains. We analyze the opposite limiting regime $\Omega \downarrow W(A)$. Along any C^1 convex exhaustion $\Omega_\varepsilon \downarrow W(A)$, if $\sigma_\varepsilon \in \partial\Omega_\varepsilon$ approaches $\sigma_0 \in \partial W(A)$ with convergent outward normals and $\sigma_0 \notin \text{spec}(A)$, then $\lambda_{\min}(P_{\Omega_\varepsilon}(\sigma_\varepsilon, A)) \rightarrow 0$ and the corresponding min-eigenvectors converge (up to subsequences and phases) to the canonical subspace $(\sigma_0 I - A)\mathcal{M}(n)$ determined by the maximal eigenspace of $H(n) = \text{Re}(\bar{n}A)$. Quantitatively, we obtain two-sided bounds in terms of an explicit support-gap scalar, yielding a linear degeneracy rate under bounded-resolvent hypotheses and an explicit rate for outer offsets $W(A) + \varepsilon\mathbb{D}$. For normal matrices we compute the eigenvalues of $P_\Omega(\sigma, A)$ explicitly, showing that degeneracy may fail at spectral support points unless the supporting face contains multiple eigenvalues.

Keywords: numerical range; operator-valued Poisson kernel; convex domains; functional calculus; boundary integral methods

1. Introduction

Let $A \in \mathbb{C}^{d \times d}$ and denote its numerical range

$$W(A) := \{x^*Ax : x \in \mathbb{C}^d, \|x\| = 1\}.$$

It is a compact convex subset of $\mathbb{C}$ (Toeplitz–Hausdorff theorem; see, e.g., [1,2]). A central open problem due to Crouzeix asks whether $W(A)$ is a 2-spectral set for A , i.e.,

$$\|p(A)\| \leq 2 \max_{z \in W(A)} |p(z)| \quad \text{for every polynomial } p. \quad (1.1)$$

See [3,4] for the formulation and [5] for the best known universal constant $1 + \sqrt{2}$.

Background and relation to the convex-domain functional calculus. Up to harmless normalization conventions, a recurring tool in the convex-domain approach of Delyon–Delyon and Crouzeix is the operator-valued boundary kernel

$$P_\Omega(\sigma, A) := \text{Re}(n_\Omega(\sigma)(\sigma I - A)^{-1}), \quad \sigma \in \partial\Omega, \quad (1.2)$$

defined for a bounded convex domain $\Omega \subset \mathbb{C}$ with C^1 boundary containing $\text{spec}(A)$, where $n_\Omega(\sigma)$ is the outward unit normal at σ . This kernel appears in double-layer potential representations and boundary integral operators used to obtain functional calculus bounds on convex domains [4–8]. For convex Ω with $W(A) \subset \Omega$, positivity/coercivity of $\sigma \mapsto P_\Omega(\sigma, A)$ on $\partial\Omega$ encodes strict separation of supporting half-planes and serves as a key structural input in such estimates [7–9].

Motivation: loss of coercivity near $\partial W(A)$. In applications and numerical implementations of boundary-integral calculi, one often approximates $W(A)$ by C^1 convex supersets $\Omega_\varepsilon \downarrow W(A)$. It is therefore natural to ask whether coercivity of the pointwise kernel $P_{\Omega_\varepsilon}(\sigma, A)$ can remain uniform as $\varepsilon \rightarrow 0$. The results below show that this is impossible in general: even when the resolvent stays bounded (i.e. at non-spectral boundary points $\sigma_0 \in \partial W(A) \setminus \text{spec}(A)$), the smallest eigenvalue of $P_{\Omega_\varepsilon}(\sigma, A)$ must deteriorate at boundary points $\sigma \in \partial \Omega_\varepsilon$ approaching $\partial W(A)$ in a fixed supporting direction.

What is new in this paper. The existing convex-domain literature primarily exploits positivity of (1.2) for fixed domains $\Omega \supsetneq W(A)$ [4,7–9]. Here we analyze the complementary limiting regime in which Ω shrinks to $W(A)$, and we make explicit the resulting loss of coercivity of the pointwise kernel. The analysis is driven by a congruence identity and by a scalar *support gap* $\delta(\sigma, n) = \text{Re}(\bar{n}\sigma) - \lambda_{\max}(\text{Re}(\bar{n}A))$, which admits a support-function interpretation in standard convex-geometry terminology.

- We prove a qualitative degeneracy theorem (Theorem 1): along any C^1 convex exhaustion $\Omega_\varepsilon \downarrow W(A)$, if $\sigma_\varepsilon \in \partial \Omega_\varepsilon$ approaches a non-spectral boundary point $\sigma_0 \in \partial W(A) \setminus \text{spec}(A)$ with convergent outward normals $n_{\Omega_\varepsilon}(\sigma_\varepsilon) \rightarrow n$, then $\lambda_{\min}(P_{\Omega_\varepsilon}(\sigma_\varepsilon, A)) \rightarrow 0$ and the limiting min-eigenvector directions lie in $(\sigma_0 I - A)\mathcal{M}(n)$, where $\mathcal{M}(n)$ is the maximal eigenspace of $H(n) = \text{Re}(\bar{n}A)$.
- We establish two-sided bounds for $\lambda_{\min}(P_\Omega(\sigma, A))$ in terms of the support gap $\delta(\sigma, n)$, yielding a linear degeneracy rate under bounded-resolvent hypotheses (Lemma 3 and Corollary 3), and compute δ explicitly for standard outer offsets $W(A) + \varepsilon \mathbb{D}$ (Proposition 2).
- Under a spectral-isolation hypothesis for $\lambda_{\max}(H(n))$, we obtain convergence of the entire near-kernel invariant subspace (spectral projector) along the exhaustion (Proposition 3).
- We analyze the contrasting spectral-support regime $\sigma_0 \in \text{spec}(A) \cap \partial W(A)$ for normal matrices via an explicit eigenvalue formula for $P_\Omega(\sigma, A)$, showing that degeneracy may fail at a spectral support point unless the supporting face contains multiple eigenvalues (Proposition 4 and Examples 1–2).

Organization. Section 2 fixes notation and recalls support-function identities. Section 3 introduces $P_\Omega(\sigma, A)$, proves the key congruence identity, and establishes quantitative support-gap bounds together with a geometric interpretation of δ . Section 4 contains the degeneracy theorem, quantitative corollaries, subspace convergence, and explicit examples, followed by a brief discussion of open problems.

2. Preliminaries

We use the standard notation for disks:

$$\mathbb{D} := \{z \in \mathbb{C} : |z| < 1\}, \quad \bar{\mathbb{D}} := \{z \in \mathbb{C} : |z| \leq 1\}.$$

Throughout, $A \in \mathbb{C}^{d \times d}$ is fixed. For vectors $x \in \mathbb{C}^d$ we use $\|x\| := (x^*x)^{1/2}$. For matrices $B \in \mathbb{C}^{d \times d}$ we use the induced operator norm $\|B\| := \sup_{\|x\|=1} \|Bx\|$. We write B^* for the conjugate transpose and $\text{Re}(B) := (B + B^*)/2$.

For a Hermitian matrix B , we write its eigenvalues in nondecreasing order as

$$\lambda_1^\uparrow(B) \leq \dots \leq \lambda_d^\uparrow(B),$$

and in nonincreasing order as $\lambda_1^\downarrow(B) \geq \dots \geq \lambda_d^\downarrow(B)$. In particular, $\lambda_{\min}(B) = \lambda_1^\uparrow(B)$ and $\lambda_{\max}(B) = \lambda_1^\downarrow(B)$.

Remark 1 (Spectrum is contained in the numerical range). *One has $\text{spec}(A) \subset W(A)$. Indeed, if $Ax = \lambda x$ with $\|x\| = 1$, then $x^*Ax = \lambda \in W(A)$. Consequently, $W(A) \subset \Omega$ implies $\text{spec}(A) \subset \Omega$ for any open set $\Omega \subset \mathbb{C}$.*

2.1. Support Functions and the Hermitian Pencil

For unimodular $\omega \in \mathbb{C}$ (i.e. $|\omega| = 1$), define the Hermitian matrix

$$H(\omega) := \operatorname{Re}(\bar{\omega} A) = \frac{1}{2}(\bar{\omega} A + \omega A^*). \quad (2.1)$$

We will later write $n \in \mathbb{C}$ (with $|n| = 1$) for outward unit normals on $\partial\Omega$; in the support-function identities below and throughout, such an n simply plays the role of the unimodular direction ω .

Let $\lambda_{\max}(H(\omega))$ denote its largest eigenvalue and let

$$\mathcal{M}(\omega) := \operatorname{Ker}(\lambda_{\max}(H(\omega))I - H(\omega))$$

denote the corresponding maximal eigenspace.

Lemma 1 (Support function of the numerical range). *For every unimodular $\omega \in \mathbb{C}$,*

$$\max_{z \in W(A)} \operatorname{Re}(\bar{\omega} z) = \lambda_{\max}(H(\omega)). \quad (2.2)$$

Moreover, if $x \in \mathbb{C}^d$ is a unit eigenvector of $H(\omega)$ associated with $\lambda_{\max}(H(\omega))$, then $x^* A x \in \partial W(A)$ and

$$\operatorname{Re}(\bar{\omega} x^* A x) = \lambda_{\max}(H(\omega)).$$

Proof. For $\|x\| = 1$,

$$\operatorname{Re}(\bar{\omega} x^* A x) = \operatorname{Re}(x^* (\bar{\omega} A) x) = x^* \operatorname{Re}(\bar{\omega} A) x = x^* H(\omega) x.$$

Taking the maximum over $\|x\| = 1$ yields (2.2) by Rayleigh–Ritz. If x is a maximizing unit vector, then $x^* A x \in W(A)$ attains the support functional in direction ω , hence lies on $\partial W(A)$ and satisfies the stated identity. $\square$

2.2. Convex Domains with C^1 Boundary and Normals

We identify $\mathbb{C}$ with $\mathbb{R}^2$ in the usual way. Let $\Omega \subset \mathbb{C}$ be a bounded open convex set with C^1 boundary. Then for each $\sigma \in \partial\Omega$ there is a unique outward unit normal vector. This C^1 assumption is used only to guarantee that the outward unit normal $n_\Omega(\sigma)$ exists and is unique at every boundary point, ensuring that $P_\Omega(\sigma, A)$ is well-defined; no higher regularity (e.g. curvature bounds) is used. We represent the normal as a unimodular complex number $n_\Omega(\sigma) \in \mathbb{C}$ with $|n_\Omega(\sigma)| = 1$ so that the supporting half-plane at σ is

$$\Pi_\Omega(\sigma) = \left\{ z \in \mathbb{C} : \operatorname{Re}(\overline{n_\Omega(\sigma)}(z - \sigma)) \leq 0 \right\}. \quad (2.3)$$

Equivalently, by convexity one has $\overline{\Omega} \subseteq \Pi_\Omega(\sigma)$ and $\Omega \subset \{z \in \mathbb{C} : \operatorname{Re}(\overline{n_\Omega(\sigma)}(z - \sigma)) < 0\}$. Under the identification $\mathbb{C} \simeq \mathbb{R}^2$, the functional $z \mapsto \operatorname{Re}(\bar{n}z)$ is the Euclidean inner product with the unit vector corresponding to n .

Definition 1 (C^1 convex exhaustion). *A family $\{\Omega_\varepsilon\}_{\varepsilon>0}$ is called a C^1 convex exhaustion of a compact convex set $K \subset \mathbb{C}$ if:*

- (i) each $\Omega_\varepsilon \subset \mathbb{C}$ is a bounded open convex set with C^1 boundary;
- (ii) $\Omega_{\varepsilon'} \subset \Omega_\varepsilon$ for $0 < \varepsilon' < \varepsilon$;
- (iii) $K \subset \Omega_\varepsilon$ for all $\varepsilon > 0$;
- (iv) $\bigcap_{\varepsilon>0} \overline{\Omega_\varepsilon} = K$.

Remark 2 (Subsequence selection for convergent normals). *Let $\varepsilon_k \downarrow 0$ and $\sigma_k \in \partial\Omega_{\varepsilon_k}$ be any sequence. Since each outward normal $n_k := n_{\Omega_{\varepsilon_k}}(\sigma_k)$ is unimodular, the sequence $\{n_k\} \subset \{z \in \mathbb{C} : |z| = 1\}$ lies in a*

compact set. Hence there is always a subsequence (not relabeled) such that $n_k \rightarrow n$ for some unimodular n . In particular, the normal convergence hypothesis in Theorem 1 can always be arranged by passing to a subsequence.

3. The Operator-Valued Poisson Kernel

Let $\Omega \subset \mathbb{C}$ be a bounded open convex set with C^1 boundary and assume $\text{spec}(A) \subset \Omega$. Then $(\sigma I - A)^{-1}$ exists for all $\sigma \in \partial\Omega$.

Definition 2 (Operator-valued Poisson kernel). For $\sigma \in \partial\Omega$, define

$$P_\Omega(\sigma, A) := \text{Re}\left(n_\Omega(\sigma) (\sigma I - A)^{-1}\right). \quad (3.1)$$

3.1. A Congruence Identity

Lemma 2 (Congruence identity). Let $\sigma \notin \text{spec}(A)$ and let $n \in \mathbb{C}$ be unimodular. Then

$$(\sigma I - A)^* \text{Re}(n(\sigma I - A)^{-1}) (\sigma I - A) = \text{Re}(\bar{n}(\sigma I - A)) = \text{Re}(\bar{n}\sigma) I - \text{Re}(\bar{n}A). \quad (3.2)$$

Proof. Write $R := (\sigma I - A)^{-1}$. Then $R(\sigma I - A) = I$ and $(\sigma I - A)^* R^* = I$. Using $\text{Re}(X) = \frac{1}{2}(X + X^*)$,

$$\begin{aligned} (\sigma I - A)^* \text{Re}(nR)(\sigma I - A) &= \frac{1}{2} \left((\sigma I - A)^* (nR)(\sigma I - A) + (\sigma I - A)^* (\bar{n}R^*)(\sigma I - A) \right) \\ &= \text{Re}(\bar{n}(\sigma I - A)). \end{aligned}$$

Expanding gives (3.2). $\square$

3.2. Support-Gap Bounds

For unimodular $n \in \mathbb{C}$ define the support gap

$$\delta(\sigma, n) := \text{Re}(\bar{n}\sigma) - \lambda_{\max}(H(n)), \quad H(n) = \text{Re}(\bar{n}A).$$

Lemma 3 (Support-gap characterization and quantitative bounds). Let $A \in \mathbb{C}^{d \times d}$, let $\sigma \notin \text{spec}(A)$, and let $n \in \mathbb{C}$ be unimodular. Set

$$P(\sigma, n) := \text{Re}(n(\sigma I - A)^{-1}), \quad \alpha := \text{Re}(\bar{n}\sigma), \quad \delta := \alpha - \lambda_{\max}(H(n)).$$

(This notation emphasizes dependence on the prescribed direction n ; when $n = n_\Omega(\sigma)$ one has $P(\sigma, n) = P_\Omega(\sigma, A)$.) Then:

- (a) $P(\sigma, n) \succeq 0$ if and only if $\delta \geq 0$, and $P(\sigma, n) \succ 0$ if and only if $\delta > 0$.
- (b) If $\delta = 0$, then $P(\sigma, n)$ is singular and

$$\text{Ker}(P(\sigma, n)) = (\sigma I - A) \mathcal{M}(n), \quad \mathcal{M}(n) = \text{Ker}(\lambda_{\max}(H(n))I - H(n)).$$

- (c) If $\delta > 0$, then

$$\frac{\delta}{\|\sigma I - A\|^2} \leq \lambda_{\min}(P(\sigma, n)) \leq \delta \|(\sigma I - A)^{-1}\|^2. \quad (3.3)$$

Proof. Let $B := \sigma I - A$ and $P := P(\sigma, n)$. By Lemma 2,

$$B^*PB = \text{Re}(\bar{n}B) = \alpha I - \text{Re}(\bar{n}A) = \alpha I - H(n) =: Q.$$

Since B is invertible, congruence by B preserves (semi)definiteness, so $P \succeq 0 \iff Q \succeq 0$ and $P \succ 0 \iff Q \succ 0$. As Q is Hermitian with $\lambda_{\min}(Q) = \alpha - \lambda_{\max}(H(n)) = \delta$, this proves (a).

If $\delta = 0$, then $Q \succeq 0$ is singular with $\text{Ker}(Q) = \mathcal{M}(n)$, and $P \succeq 0$ by (a). For $P \succeq 0$, $x \in \text{Ker}(P) \iff x^*Px = 0$. Writing $x = By$,

$$x^*Px = y^*Qy,$$

so $x \in \text{Ker}(P) \iff y \in \text{Ker}(Q) = \mathcal{M}(n)$, proving (b).

If $\delta > 0$, then $Q \succ 0$ and $P = B^{-*}QB^{-1} \succ 0$. For $\|x\| = 1$ and $y = B^{-1}x$, one has $x = By$ and hence $\|y\| \geq 1/\|B\|$; thus

$$x^*Px = y^*Qy \geq \lambda_{\min}(Q)\|y\|^2 = \delta\|y\|^2 \geq \delta/\|B\|^2,$$

giving the lower bound in (3.3). For the upper bound, take y a unit eigenvector of Q for $\lambda_{\min}(Q) = \delta$ and set $x = By/\|By\|$; then

$$x^*Px = \frac{y^*Qy}{\|By\|^2} = \frac{\delta}{\|By\|^2} \leq \delta\|B^{-1}\|^2.$$

□

Remark 3 (Connection with the convex-domain Poisson kernel literature). *Up to normalization conventions, $P_{\Omega}(\sigma, A)$ is the operator-valued boundary kernel appearing in the Carl Neumann double-layer potential framework for convex domains; see, e.g., [7–9]. Lemma 3 isolates the dependence of $\lambda_{\min}(P(\sigma, n))$ on the scalar support gap $\delta(\sigma, n)$.*

3.3. Strict positivity when $W(A) \subset \Omega$

Lemma 4 (Strict separation at a supporting line). *Let $\Omega \subset \mathbb{C}$ be a bounded open convex set with C^1 boundary and let $K \subset \Omega$ be compact. Fix $\sigma \in \partial\Omega$ and let $n = n_{\Omega}(\sigma)$ be the outward unit normal. Then*

$$\max_{z \in K} \text{Re}(\bar{n}z) < \text{Re}(\bar{n}\sigma).$$

Proof. By (2.3), $\Omega \subset \{z : \text{Re}(\bar{n}(z - \sigma)) < 0\}$, hence $K \subset \{z : \text{Re}(\bar{n}(z - \sigma)) < 0\}$. The continuous function $z \mapsto \text{Re}(\bar{n}(z - \sigma))$ attains its maximum on compact K , and this maximum is strictly negative. Rearranging yields the claim. □

Proposition 1 (Positivity of the Poisson kernel). *Assume $W(A) \subset \Omega$. Then for every $\sigma \in \partial\Omega$,*

$$P_{\Omega}(\sigma, A) \succ 0.$$

Proof. Fix $\sigma \in \partial\Omega$ and set $n := n_{\Omega}(\sigma)$ and $\alpha := \text{Re}(\bar{n}\sigma)$. By Lemma 4 with $K = W(A)$ and Lemma 1,

$$\lambda_{\max}(H(n)) = \max_{z \in W(A)} \text{Re}(\bar{n}z) < \alpha,$$

so $\delta(\sigma, n) = \alpha - \lambda_{\max}(H(n)) > 0$. Now apply Lemma 3(a). □

3.4. Geometric Meaning of the Support Gap and Offset Exhaustions

For a compact convex set $K \subset \mathbb{C}$ and unimodular $n \in \mathbb{C}$, define its *support function*

$$h_K(n) := \max_{z \in K} \text{Re}(\bar{n}z).$$

If $\Omega \subset \mathbb{C}$ is a bounded open convex set with C^1 boundary and $\sigma \in \partial\Omega$ has outward normal $n = n_{\Omega}(\sigma)$, then necessarily $\text{Re}(\bar{n}\sigma) = h_{\bar{\Omega}}(n)$, i.e. the boundary point lies on the supporting line in direction n .

Lemma 5 (Support gap as a support-function difference). *Let $\Omega \subset \mathbb{C}$ be a bounded open convex set with C^1 boundary and $\sigma \in \partial\Omega$. Let $n := n_\Omega(\sigma)$. Then*

$$\delta(\sigma, n) = \operatorname{Re}(\bar{n}\sigma) - \lambda_{\max}(H(n)) = h_{\bar{\Omega}}(n) - h_{W(A)}(n).$$

In particular, $\delta(\sigma, n)$ measures the separation between the supporting line of $\bar{\Omega}$ in direction n and the corresponding supporting line of $W(A)$.

Proof. Since n is the outward unit normal at $\sigma \in \partial\Omega$, the supporting half-plane characterization implies $\operatorname{Re}(\bar{n}z) \leq \operatorname{Re}(\bar{n}\sigma)$ for all $z \in \bar{\Omega}$, hence $h_{\bar{\Omega}}(n) = \operatorname{Re}(\bar{n}\sigma)$. By Lemma 1, $h_{W(A)}(n) = \lambda_{\max}(H(n))$. Combining gives the claim. $\square$

Proposition 2 (Outer offsets: δ is explicit). *Let $K \subset \mathbb{C}$ be compact and convex and fix $\varepsilon > 0$. Define the outer offset (outer parallel set)*

$$K_\varepsilon := K + \varepsilon\mathbb{D} = \{z + w : z \in K, w \in \mathbb{C}, |w| < \varepsilon\}.$$

Then for every unimodular $n \in \mathbb{C}$,

$$h_{\bar{K}_\varepsilon}(n) = h_K(n) + \varepsilon.$$

In particular, taking $K = W(A)$ and $\Omega_\varepsilon := W(A) + \varepsilon\mathbb{D}$, for any boundary point $\sigma \in \partial\Omega_\varepsilon$ with outward normal $n = n_{\Omega_\varepsilon}(\sigma)$ (whenever defined) one has

$$\delta(\sigma, n) = \varepsilon.$$

Consequently, since $\sigma \in \partial\Omega_\varepsilon$ implies $\sigma \notin W(A)$ and hence $\sigma \notin \operatorname{spec}(A)$ (Remark 1), Lemma 3(c) yields

$$\frac{\varepsilon}{\|\sigma I - A\|^2} \leq \lambda_{\min}(P_{\Omega_\varepsilon}(\sigma, A)) \leq \varepsilon \|(\sigma I - A)^{-1}\|^2.$$

Proof. Fix unimodular n . For any $z \in K$ and $w \in \mathbb{C}$ with $|w| \leq \varepsilon$,

$$\operatorname{Re}(\bar{n}(z + w)) = \operatorname{Re}(\bar{n}z) + \operatorname{Re}(\bar{n}w) \leq h_K(n) + |w| \leq h_K(n) + \varepsilon,$$

so $h_{\bar{K}_\varepsilon}(n) \leq h_K(n) + \varepsilon$. On the other hand, choosing $z_\star \in K$ with $\operatorname{Re}(\bar{n}z_\star) = h_K(n)$ and $w_\star = \varepsilon n$ gives $|w_\star| = \varepsilon$ and

$$\operatorname{Re}(\bar{n}(z_\star + w_\star)) = h_K(n) + \varepsilon,$$

so $h_{\bar{K}_\varepsilon}(n) \geq h_K(n) + \varepsilon$. This proves the support-function identity and hence the displayed formula for δ follows from Lemma 5.

The final eigenvalue bounds are an immediate substitution of $\delta = \varepsilon$ into (3.3). $\square$

Remark 4 (Smoothness versus offsets). *If K has flat faces, then $\partial(K + \varepsilon\mathbb{D})$ is typically only $C^{1,1}$ (curvature may jump at transitions between translated faces and rounded arcs). Proposition 2 is therefore best viewed as a geometric model illustrating how the support gap scales with the outer distance parameter ε . For the purposes of Definition 1, one may replace $K + \varepsilon\mathbb{D}$ by any convex domain with C^1 boundary whose support function differs from h_K by a quantity comparable to ε ; the same interpretation of δ then applies. For example, one may take Minkowski sums with a fixed smooth strictly convex unit ball (instead of $\mathbb{D}$) or smooth the support function to obtain a genuine C^1 (indeed smooth) convex exhaustion with the same first-order support-gap scaling.*

3.5. Hausdorff distance and support-function control of the support gap

For a nonempty compact set $K \subset \mathbb{C}$ and $z \in \mathbb{C}$, write

$$1.7em(z, K) := \inf_{w \in K} |z - w|.$$

For nonempty compact sets $K, L \subset \mathbb{C}$, define the (Euclidean) Hausdorff distance

$$d_H(K, L) := \max \left\{ \sup_{z \in K} 1.7em(z, L), \sup_{w \in L} 1.7em(w, K) \right\}.$$

Lemma 6 (Hausdorff distance via support functions). *Let $K, L \subset \mathbb{C}$ be nonempty compact convex sets and let $\mathbb{D} := \{z \in \mathbb{C} : |z| \leq 1\}$. Then*

$$d_H(K, L) = \sup_{|n|=1} |h_K(n) - h_L(n)|.$$

If moreover $K \subseteq L$, then $h_K(n) \leq h_L(n)$ for all $|n| = 1$ and hence

$$d_H(L, K) = \sup_{|n|=1} (h_L(n) - h_K(n)).$$

Proof. For $t \geq 0$ and a nonempty compact set K , the Minkowski sum

$$K + t\mathbb{D} = \{z + w : z \in K, |w| \leq t\}$$

is the closed t -neighborhood of K , i.e. $K + t\mathbb{D} = \{u \in \mathbb{C} : 1.7em(u, K) \leq t\}$. Consequently,

$$d_H(K, L) = \inf \left\{ t \geq 0 : K \subseteq L + t\mathbb{D} \text{ and } L \subseteq K + t\mathbb{D} \right\}.$$

For compact convex sets $M, N \subset \mathbb{C}$ one has $M \subseteq N$ if and only if $h_M(n) \leq h_N(n)$ for all $|n| = 1$. (Indeed, the forward direction is immediate; conversely, if $x \in M \setminus N$, a separating supporting line for the convex compact set N yields a unimodular n with $\operatorname{Re}(\bar{n}x) > \max_{z \in N} \operatorname{Re}(\bar{n}z) = h_N(n)$, hence $h_M(n) \geq \operatorname{Re}(\bar{n}x) > h_N(n)$.)

Moreover, support functions add under Minkowski sums, and $h_{t\mathbb{D}}(n) = t$ for $|n| = 1$; hence

$$h_{K+t\mathbb{D}}(n) = h_K(n) + t.$$

Therefore, $K \subseteq L + t\mathbb{D}$ is equivalent to $h_K(n) \leq h_L(n) + t$ for all $|n| = 1$, and similarly $L \subseteq K + t\mathbb{D}$ is equivalent to $h_L(n) \leq h_K(n) + t$ for all $|n| = 1$. Thus $d_H(K, L)$ is the smallest t such that $|h_K(n) - h_L(n)| \leq t$ for all $|n| = 1$, i.e.

$$d_H(K, L) = \sup_{|n|=1} |h_K(n) - h_L(n)|.$$

If $K \subseteq L$, then $h_K \leq h_L$, so the absolute value may be dropped, giving the second identity. $\square$

Corollary 1 (Support gap bounded by the Hausdorff approximation error). *Assume $W(A) \subset \Omega$, and set*

$$\Delta(\Omega) := d_H(\overline{\Omega}, W(A)) = \sup_{|n|=1} (h_{\overline{\Omega}}(n) - h_{W(A)}(n)).$$

Then for every $\sigma \in \partial\Omega$ with outward normal $n = n_\Omega(\sigma)$,

$$\delta(\sigma, n) = \operatorname{Re}(\bar{n}\sigma) - \lambda_{\max}(H(n)) = h_{\overline{\Omega}}(n) - h_{W(A)}(n) \leq \Delta(\Omega).$$

Consequently, since $\sigma \notin \operatorname{spec}(A)$ for $\sigma \in \partial\Omega$, Lemma 3(c) yields

$$\lambda_{\min}(P_\Omega(\sigma, A)) \leq \delta(\sigma, n) \|(\sigma I - A)^{-1}\|^2 \leq \Delta(\Omega) \|(\sigma I - A)^{-1}\|^2.$$

Moreover, there exists $\sigma_\star \in \partial\Omega$ such that

$$\delta(\sigma_\star, n_\Omega(\sigma_\star)) = \Delta(\Omega),$$

and for this point one has the two-sided estimate

$$\frac{\Delta(\Omega)}{\|\sigma_* I - A\|^2} \leq \lambda_{\min}(P_{\Omega}(\sigma_*, A)) \leq \Delta(\Omega) \|(\sigma_* I - A)^{-1}\|^2.$$

Proof. The identity $\delta(\sigma, n) = h_{\overline{\Omega}}(n) - h_{W(A)}(n)$ is Lemma 5, and the bound $\delta(\sigma, n) \leq \Delta(\Omega)$ follows from the definition of $\Delta(\Omega)$. The eigenvalue bounds are then immediate from Lemma 3(c).

Finally, the function $n \mapsto h_{\overline{\Omega}}(n) - h_{W(A)}(n)$ is continuous on the unit circle, so it attains its maximum at some unimodular n_* . Choose $\sigma_* \in \overline{\Omega}$ such that $\operatorname{Re}(\overline{n_*} \sigma_*) = h_{\overline{\Omega}}(n_*)$; then $\sigma_* \in \partial\Omega$ and the supporting line $\{z : \operatorname{Re}(\overline{n_*} z) = \operatorname{Re}(\overline{n_*} \sigma_*)\}$ is a supporting line for $\overline{\Omega}$ at σ_* . Since $\partial\Omega$ is C^1 , the outward unit normal at σ_* is uniquely defined and equals n_* , and hence

$$\delta(\sigma_*, n_{\Omega}(\sigma_*)) = h_{\overline{\Omega}}(n_*) - h_{W(A)}(n_*) = \Delta(\Omega).$$

□

4. Degeneracy Along a C^1 Convex Exhaustion

4.1. Qualitative Degeneracy and Limiting Kernel Directions

Theorem 1 (Degeneracy of the Operator-Valued Poisson Kernel). *Let $A \in \mathbb{C}^{d \times d}$ and let $\{\Omega_{\varepsilon}\}_{\varepsilon>0}$ be a C^1 convex exhaustion of $W(A)$ (Definition 1). For $\sigma \in \partial\Omega_{\varepsilon}$, set*

$$P_{\Omega_{\varepsilon}}(\sigma, A) := \operatorname{Re}\left(n_{\Omega_{\varepsilon}}(\sigma) (\sigma I - A)^{-1}\right).$$

Fix any sequence $\varepsilon_k \downarrow 0$ and points $\sigma_k \in \partial\Omega_{\varepsilon_k}$ such that

$$\sigma_k \rightarrow \sigma_0 \in \partial W(A), \quad n_k := n_{\Omega_{\varepsilon_k}}(\sigma_k) \rightarrow n \in \mathbb{C}, \quad |n| = 1.$$

(After passing to a subsequence, the convergence $n_k \rightarrow n$ is automatic; see Remark 2.) Assume $\sigma_0 \notin \operatorname{spec}(A)$. Let $H(n) = \operatorname{Re}(\overline{n}A)$ and $\mathcal{M}(n) = \operatorname{Ker}(\lambda_{\max}(H(n))I - H(n))$.

Then:

- (1) (Vanishing) $\lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A)) \rightarrow 0$ as $k \rightarrow \infty$.
- (2) (Limiting directions) *If u_k is any unit eigenvector of $P_{\Omega_{\varepsilon_k}}(\sigma_k, A)$ for $\lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A))$, then every accumulation point u_0 of $\{u_k\}$ satisfies*

$$u_0 \in (\sigma_0 I - A) \mathcal{M}(n).$$

- (3) (One-dimensional case) *If $\dim \mathcal{M}(n) = 1$, then there exist phases $\theta_k \in \mathbb{R}$ such that*

$$e^{i\theta_k} u_k \rightarrow \frac{(\sigma_0 I - A)v}{\|(\sigma_0 I - A)v\|} \quad (k \rightarrow \infty),$$

where v is any unit vector spanning $\mathcal{M}(n)$.

Proof. Set $B_k := \sigma_k I - A$ and $R_k := B_k^{-1}$, and define

$$P_k := \operatorname{Re}(n_k R_k), \quad \alpha_k := \operatorname{Re}(\overline{n_k} \sigma_k).$$

Define also $B_0 := \sigma_0 I - A$, $R_0 := B_0^{-1}$, $P_0 := \operatorname{Re}(n R_0)$, $\alpha_0 := \operatorname{Re}(\overline{n} \sigma_0)$.

Step 1: Congruence identities. By Lemma 2,

$$B_k^* P_k B_k = \alpha_k I - H(n_k), \quad B_0^* P_0 B_0 = \alpha_0 I - H(n), \quad (4.1)$$

where $H(n_k) = \operatorname{Re}(\overline{n_k} A)$.

Step 2: $\alpha_0 = \lambda_{\max}(H(n))$. Since n_k is the outward normal at $\sigma_k \in \partial\Omega_{\varepsilon_k}$, the supporting half-plane property gives $\operatorname{Re}(\overline{n_k}z) \leq \alpha_k$ for all $z \in \Omega_{\varepsilon_k}$ and hence for all $z \in W(A)$. Passing to the limit yields $\operatorname{Re}(\overline{n}z) \leq \alpha_0$ for all $z \in W(A)$. Because $\sigma_0 \in W(A)$, equality holds at $z = \sigma_0$, so $\alpha_0 = \max_{z \in W(A)} \operatorname{Re}(\overline{n}z)$. Lemma 1 now gives

$$\alpha_0 = \lambda_{\max}(H(n)), \quad \operatorname{Ker}(\alpha_0 I - H(n)) = \mathcal{M}(n),$$

so $\alpha_0 I - H(n) \succeq 0$ is singular.

Step 3: $P_k \rightarrow P_0$ in operator norm. Since $\sigma_0 \notin \operatorname{spec}(A)$, B_0 is invertible. Write

$$B_k = B_0 + (\sigma_k - \sigma_0)I = B_0(I + E_k), \quad E_k := (\sigma_k - \sigma_0)R_0.$$

Then $\|E_k\| \rightarrow 0$, so for large k , $I + E_k$ is invertible and

$$R_k = B_k^{-1} = (I + E_k)^{-1}R_0, \quad \|R_k - R_0\| \rightarrow 0.$$

Therefore,

$$\|P_k - P_0\| = \|\operatorname{Re}(n_k R_k - n R_0)\| \leq |n_k - n| \|R_k\| + \|R_k - R_0\| \rightarrow 0.$$

Step 4: $\lambda_{\min}(P_0) = 0$ and $\lambda_{\min}(P_k) \rightarrow 0$. Since P_k, P_0 are Hermitian, Weyl's inequality yields

$$|\lambda_{\min}(P_k) - \lambda_{\min}(P_0)| \leq \|P_k - P_0\| \rightarrow 0,$$

so $\lambda_{\min}(P_k) \rightarrow \lambda_{\min}(P_0)$. By (4.1) and Step 2,

$$B_0^* P_0 B_0 = \alpha_0 I - H(n) \succeq 0 \text{ is singular.}$$

Since B_0 is invertible, $P_0 \succeq 0$ is singular, hence $\lambda_{\min}(P_0) = 0$, proving (1). Moreover,

$$\operatorname{Ker}(P_0) = B_0 \operatorname{Ker}(\alpha_0 I - H(n)) = (\sigma_0 I - A)\mathcal{M}(n)$$

by Lemma 3(b) (with $\delta = 0$).

Step 5: Limiting eigenvectors. Let u_k be unit min-eigenvectors: $P_k u_k = \lambda_{\min}(P_k) u_k$. Along a convergent subsequence, $u_k \rightarrow u_0$. Then

$$u_0^* P_0 u_0 = \lim_{k \rightarrow \infty} u_k^* P_0 u_k = \lim_{k \rightarrow \infty} (u_k^* P_k u_k + u_k^* (P_0 - P_k) u_k) = \lim_{k \rightarrow \infty} (\lambda_{\min}(P_k) + o(1)) = 0.$$

Since $P_0 \succeq 0$, this implies $u_0 \in \operatorname{Ker}(P_0) = (\sigma_0 I - A)\mathcal{M}(n)$, proving (2).

Step 6: One-dimensional case. If $\dim \mathcal{M}(n) = 1$, then $\dim \operatorname{Ker}(P_0) = 1$, so the smallest eigenvalue of P_0 is simple. By the Davis–Kahan sin Θ theorem for invariant subspaces (see [10]), the corresponding one-dimensional eigenspaces of P_k converge to $\operatorname{Ker}(P_0)$ in gap metric, hence there exist phases θ_k such that $e^{i\theta_k} u_k \rightarrow u_*$, where u_* spans $\operatorname{Ker}(P_0) = (\sigma_0 I - A)\mathcal{M}(n)$. This gives (3). $\square$

Remark 5 (Why $\sigma_0 \notin \operatorname{spec}(A)$ is essential). The hypothesis $\sigma_0 \notin \operatorname{spec}(A)$ ensures that $(\sigma I - A)^{-1}$ remains bounded near σ_0 , so P_0 is a finite Hermitian matrix. When $\sigma_0 \in \operatorname{spec}(A)$, the resolvent diverges and the behavior of $\lambda_{\min}(P_{\Omega_\varepsilon}(\sigma_\varepsilon, A))$ depends on the spectral geometry; see Proposition 4 below and Section 4.7.

Corollary 2 (Global coercivity collapse along a C^1 convex exhaustion). Assume that A is not a scalar multiple of the identity (equivalently, $W(A)$ is not a singleton). Let $\{\Omega_\varepsilon\}_{\varepsilon>0}$ be a C^1 convex exhaustion of $W(A)$ and define the global coercivity constant

$$c(\varepsilon) := \inf_{\sigma \in \partial\Omega_\varepsilon} \lambda_{\min}(P_{\Omega_\varepsilon}(\sigma, A)), \quad P_{\Omega_\varepsilon}(\sigma, A) = \operatorname{Re}(n_{\Omega_\varepsilon}(\sigma) (\sigma I - A)^{-1}).$$

Then

$$\liminf_{\varepsilon \downarrow 0} c(\varepsilon) = 0.$$

In particular, there do not exist $\varepsilon_0 > 0$ and $c_0 > 0$ such that $P_{\Omega_\varepsilon}(\sigma, A) \succeq c_0 I$ for all $0 < \varepsilon < \varepsilon_0$ and all $\sigma \in \partial\Omega_\varepsilon$.

Proof. Since A is not scalar, the compact convex set $W(A)$ contains more than one point, hence $\partial W(A)$ is infinite, whereas $\text{spec}(A)$ is finite. Choose $\sigma_0 \in \partial W(A) \setminus \text{spec}(A)$.

Fix any sequence $\varepsilon_k \downarrow 0$. We claim that $1.7em(\sigma_0, \partial\Omega_{\varepsilon_k}) \rightarrow 0$. Indeed, if not, then there exist $\delta > 0$ and a subsequence (not relabeled) such that $1.7em(\sigma_0, \partial\Omega_{\varepsilon_k}) \geq \delta$ for all k , hence the open ball $B(\sigma_0, \delta)$ is contained in Ω_{ε_k} for all k . Taking closures and intersecting over k yields $B(\sigma_0, \delta) \subset \bigcap_k \overline{\Omega_{\varepsilon_k}} = W(A)$, contradicting $\sigma_0 \in \partial W(A)$.

Therefore we may choose $\sigma_k \in \partial\Omega_{\varepsilon_k}$ with $\sigma_k \rightarrow \sigma_0$. By compactness of the unit circle, after passing to a subsequence we have $n_{\Omega_{\varepsilon_k}}(\sigma_k) \rightarrow n$ for some unimodular n . Theorem 1 then gives

$$\lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A)) \rightarrow 0.$$

Since $c(\varepsilon_k) \leq \lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A))$, it follows that $\liminf_{\varepsilon \downarrow 0} c(\varepsilon) = 0$. $\square$

4.2. Quantitative Degeneracy Rate

Corollary 3 (Linear rate in terms of the support gap). *In the setting of Theorem 1, define*

$$\delta_k := \text{Re}(\overline{n_k} \sigma_k) - \lambda_{\max}(H(n_k)), \quad H(n_k) = \text{Re}(\overline{n_k} A).$$

Then $\delta_k > 0$ for each k and $\delta_k \rightarrow 0$. Moreover, for all sufficiently large k ,

$$\frac{\delta_k}{4\|\sigma_0 I - A\|^2} \leq \lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A)) \leq 4\delta_k \|(\sigma_0 I - A)^{-1}\|^2. \quad (4.2)$$

In particular, $\lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A)) = \Theta(\delta_k)$.

Proof. Since $W(A) \subset \Omega_{\varepsilon_k}$ and $\sigma_k \in \partial\Omega_{\varepsilon_k}$ with normal n_k , Lemma 4 and Lemma 1 imply $\lambda_{\max}(H(n_k)) < \text{Re}(\overline{n_k} \sigma_k)$, so $\delta_k > 0$.

As $n_k \rightarrow n$ and $\sigma_k \rightarrow \sigma_0$, $\text{Re}(\overline{n_k} \sigma_k) \rightarrow \text{Re}(\overline{n} \sigma_0)$. Also $H(n_k) \rightarrow H(n)$ in operator norm, hence $\lambda_{\max}(H(n_k)) \rightarrow \lambda_{\max}(H(n))$. By Step 2 in the proof of Theorem 1, $\text{Re}(\overline{n} \sigma_0) = \lambda_{\max}(H(n))$, so $\delta_k \rightarrow 0$.

Set $B_k = \sigma_k I - A$ and $B_0 = \sigma_0 I - A$. Since $B_k \rightarrow B_0$ and B_0 is invertible, for large k one has $\|B_k\| \leq 2\|B_0\|$ and $\|B_k^{-1}\| \leq 2\|B_0^{-1}\|$. Applying Lemma 3(c) to $(\sigma, n) = (\sigma_k, n_k)$ gives

$$\frac{\delta_k}{\|B_k\|^2} \leq \lambda_{\min}(P_{\Omega_{\varepsilon_k}}(\sigma_k, A)) \leq \delta_k \|B_k^{-1}\|^2,$$

and the stated constants follow. $\square$

4.3. Convergence of the Near-Kernel Subspace

Proposition 3 (Convergence of the near-kernel spectral projector). *Assume the setting of Theorem 1 and set $m := \dim \mathcal{M}(n)$. Assume that $\lambda_{\max}(H(n))$ is isolated with multiplicity m , i.e.*

$$\gamma_H := \lambda_{\max}(H(n)) - \lambda_{m+1}^\downarrow(H(n)) > 0. \quad (4.3)$$

Let

$$P_0 := \text{Re}(n(\sigma_0 I - A)^{-1}), \quad \mathcal{K}_0 := \text{Ker}(P_0) = (\sigma_0 I - A)\mathcal{M}(n), \quad \Pi_0 : \mathbb{C}^d \rightarrow \mathcal{K}_0$$

be the orthogonal projector onto $\mathcal{K}_0$.

For each k , let $P_k := P_{\Omega_{\varepsilon_k}}(\sigma_k, A)$ and let Π_k be the orthogonal projector onto the direct sum of the eigenspaces of P_k corresponding to its m smallest eigenvalues. Then $\|\Pi_k - \Pi_0\| \rightarrow 0$ as $k \rightarrow \infty$.

Moreover, writing $B_0 = \sigma_0 I - A$, one has the explicit spectral-gap bound

$$\lambda_{m+1}^\uparrow(P_0) \geq \frac{\gamma_H}{\|B_0\|^2}, \quad (4.4)$$

and consequently, for all sufficiently large k ,

$$\|\Pi_k - \Pi_0\| \leq \frac{2\|P_k - P_0\|}{\lambda_{m+1}^\uparrow(P_0)} \leq \frac{2\|B_0\|^2}{\gamma_H} \|P_k - P_0\|. \quad (4.5)$$

Proof. By Lemma 2 and Step 2 of Theorem 1,

$$B_0^* P_0 B_0 = Q_0 := \lambda_{\max}(H(n))I - H(n) \succeq 0.$$

The eigenvalues of Q_0 are 0 with multiplicity m and at least γ_H on $\mathcal{M}(n)^\perp$, so $\lambda_{m+1}^\uparrow(Q_0) = \gamma_H$.

Using the Courant–Fischer characterization with the change of variables $x = B_0 y$, one obtains for every j

$$\lambda_j^\uparrow(P_0) = \min_{\dim S=j} \max_{\substack{x \in S \\ \|x\|=1}} x^* P_0 x = \min_{\dim S=j} \max_{\substack{y \in B_0^{-1}S \\ y \neq 0}} \frac{y^* Q_0 y}{\|B_0 y\|^2} \geq \frac{1}{\|B_0\|^2} \lambda_j^\uparrow(Q_0).$$

Taking $j = m + 1$ gives (4.4).

Next, Theorem 1 gives $\|P_k - P_0\| \rightarrow 0$. Since P_0 has an isolated cluster of m eigenvalues at 0 separated by the gap $\lambda_{m+1}^\uparrow(P_0) > 0$, the Davis–Kahan $\sin \Theta$ theorem for invariant subspaces [10] yields (4.5), and hence $\|\Pi_k - \Pi_0\| \rightarrow 0$. $\square$

4.4. The Spectral-Support Regime for Normal Matrices

Proposition 4 (Normal matrices: explicit eigenvalues near a spectral support point). *Let A be normal with eigenvalues $\lambda_1, \dots, \lambda_d$ (listed with algebraic multiplicity). Fix $\sigma \notin \text{spec}(A)$ and unimodular $n \in \mathbb{C}$. Then*

$$P(\sigma, n) := \text{Re}(n(\sigma I - A)^{-1})$$

is unitarily diagonalizable and its eigenvalues are the scalars

$$p_j(\sigma, n) := \text{Re}\left(\frac{n}{\sigma - \lambda_j}\right) = \frac{\text{Re}(\bar{n}(\sigma - \lambda_j))}{|\sigma - \lambda_j|^2}, \quad j = 1, \dots, d. \quad (4.6)$$

Now fix $\sigma_0 \in \text{spec}(A)$ and let

$$J_0 := \{j \in \{1, \dots, d\} : \lambda_j = \sigma_0\}.$$

Let $\sigma_k \notin \text{spec}(A)$ and unimodular n_k satisfy

$$\sigma_k \rightarrow \sigma_0, \quad n_k \rightarrow n.$$

Write $p_{k,j} := p_j(\sigma_k, n_k)$. Then:

(i) *For every $j \notin J_0$,*

$$p_{k,j} \rightarrow p_j(\sigma_0, n) = \text{Re}\left(\frac{n}{\sigma_0 - \lambda_j}\right).$$

(ii) For every $j \in J_0$ one has the exact identity

$$p_{k,j} = \operatorname{Re}\left(\frac{n_k}{\sigma_k - \sigma_0}\right) = \frac{\operatorname{Re}(\overline{n_k}(\sigma_k - \sigma_0))}{|\sigma_k - \sigma_0|^2}.$$

In particular, if there exists $c > 0$ such that

$$\operatorname{Re}(\overline{n_k}(\sigma_k - \sigma_0)) \geq c |\sigma_k - \sigma_0| \quad \text{for all sufficiently large } k, \quad (4.7)$$

then $p_{k,j} \rightarrow +\infty$ for every $j \in J_0$.

(iii) Assume in addition that n is a supporting direction for $W(A) = \operatorname{conv}\{\lambda_1, \dots, \lambda_d\}$ at σ_0 , i.e.

$$\operatorname{Re}(\overline{n} \lambda_j) \leq \operatorname{Re}(\overline{n} \sigma_0), \quad j = 1, \dots, d. \quad (4.8)$$

Then for every $j \notin J_0$,

$$p_j(\sigma_0, n) = \frac{\operatorname{Re}(\overline{n}(\sigma_0 - \lambda_j))}{|\sigma_0 - \lambda_j|^2} = \frac{\operatorname{Re}(\overline{n} \sigma_0) - \operatorname{Re}(\overline{n} \lambda_j)}{|\sigma_0 - \lambda_j|^2} \geq 0,$$

and $p_j(\sigma_0, n) = 0$ if and only if λ_j lies on the same supporting line $\{z : \operatorname{Re}(\overline{n}z) = \operatorname{Re}(\overline{n} \sigma_0)\}$. If moreover (4.7) holds (so that all $p_{k,j} \rightarrow +\infty$ for $j \in J_0$), then

$$\lambda_{\min}(P(\sigma_k, n_k)) \longrightarrow \min_{j \notin J_0} p_j(\sigma_0, n),$$

which is strictly positive if and only if no eigenvalue $\lambda_j \neq \sigma_0$ lies on the supporting line $\{z : \operatorname{Re}(\overline{n}z) = \operatorname{Re}(\overline{n} \sigma_0)\}$.

Proof. Since A is normal, $A = U \operatorname{diag}(\lambda_1, \dots, \lambda_d) U^*$ for some unitary U , hence

$$(\sigma I - A)^{-1} = U \operatorname{diag}\left(\frac{1}{\sigma - \lambda_1}, \dots, \frac{1}{\sigma - \lambda_d}\right) U^*.$$

Therefore,

$$\begin{aligned} P(\sigma, n) &= \operatorname{Re}(n(\sigma I - A)^{-1}) \\ &= U \operatorname{Re}\left(\operatorname{diag}\left(\frac{n}{\sigma - \lambda_1}, \dots, \frac{n}{\sigma - \lambda_d}\right)\right) U^* \\ &= U \operatorname{diag}\left(\operatorname{Re}\left(\frac{n}{\sigma - \lambda_1}\right), \dots, \operatorname{Re}\left(\frac{n}{\sigma - \lambda_d}\right)\right) U^*, \end{aligned}$$

which proves (4.6). The limit in (i) follows by continuity of the map $(\sigma, n) \mapsto \operatorname{Re}(n/(\sigma - \lambda_j))$ when $\sigma_0 \neq \lambda_j$. For (ii), if $\lambda_j = \sigma_0$ then

$$\operatorname{Re}\left(\frac{n_k}{\sigma_k - \sigma_0}\right) = \operatorname{Re}\left(\frac{n_k \overline{\sigma_k - \sigma_0}}{|\sigma_k - \sigma_0|^2}\right) = \frac{\operatorname{Re}(\overline{n_k}(\sigma_k - \sigma_0))}{|\sigma_k - \sigma_0|^2},$$

and (4.7) implies $p_{k,j} \geq c/|\sigma_k - \sigma_0| \rightarrow +\infty$.

Finally, (4.8) implies $\operatorname{Re}(\overline{n}(\sigma_0 - \lambda_j)) \geq 0$ for all j , giving the nonnegativity (and the characterization of equality) in (iii). If additionally (4.7) holds, then $p_{k,j} \rightarrow +\infty$ for all $j \in J_0$ while $p_{k,j} \rightarrow p_j(\sigma_0, n) \in [0, \infty)$ for $j \notin J_0$, so for large k the minimum eigenvalue is attained among indices $j \notin J_0$, yielding the stated limit and positivity criterion. $\square$

Example 1 (Nondegeneracy at a spectral support point). Let $A = \text{diag}(0, 1)$, so $W(A) = [0, 1]$. Take $\sigma = 1 + \varepsilon$ with $\varepsilon > 0$ and $n = 1$. Then

$$P(\sigma, n) = \text{Re}((\sigma I - A)^{-1}) = \text{diag}\left(\frac{1}{1+\varepsilon}, \frac{1}{\varepsilon}\right),$$

so $\lambda_{\min}(P(\sigma, n)) = \frac{1}{1+\varepsilon} \rightarrow 1$ as $\varepsilon \downarrow 0$. Thus the smallest eigenvalue does not degenerate when the limiting support point is spectral and unique on the support face.

Example 2 (Degeneracy at a spectral point with a flat support face). Let $A = \text{diag}(1, 1 + i)$ and take $\sigma = 1 + \varepsilon$, $n = 1$. Then

$$P(\sigma, n) = \text{diag}\left(\frac{1}{\varepsilon}, \text{Re}\left(\frac{1}{\varepsilon - i}\right)\right) = \text{diag}\left(\frac{1}{\varepsilon}, \frac{\varepsilon}{\varepsilon^2 + 1}\right),$$

so $\lambda_{\min}(P(\sigma, n)) = \frac{\varepsilon}{\varepsilon^2 + 1} \sim \varepsilon \downarrow 0$. Here the supporting functional $\text{Re}(z)$ is maximized by more than one eigenvalue, and degeneracy persists at $\sigma_0 = 1 \in \text{spec}(A)$.

4.5. A fully explicit 2×2 example: a nilpotent Jordan block

Example 3 (Exact Poisson kernel and exact degeneracy rate for a disk exhaustion). Let

$$A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}.$$

Then $W(A) = \{z \in \mathbb{C} : |z| \leq \frac{1}{2}\}$. For $r > \frac{1}{2}$, let $\Omega_r := \{z \in \mathbb{C} : |z| < r\}$ and choose $\sigma = re^{it} \in \partial\Omega_r$. The outward normal at σ is $n_{\Omega_r}(\sigma) = e^{it}$ and

$$(\sigma I - A)^{-1} = \begin{pmatrix} \frac{1}{\sigma} & \frac{1}{\sigma^2} \\ 0 & \frac{1}{\sigma} \end{pmatrix}.$$

Hence

$$P_{\Omega_r}(\sigma, A) = \text{Re}(e^{it}(\sigma I - A)^{-1}) = \begin{pmatrix} \frac{1}{r} & \frac{e^{-it}}{2r^2} \\ \frac{e^{it}}{2r^2} & \frac{1}{r} \end{pmatrix},$$

whose eigenvalues are $\lambda_{\pm}(r) = \frac{2r \pm 1}{2r^2}$. In particular,

$$\lambda_{\min}(P_{\Omega_r}(\sigma, A)) = \frac{2r - 1}{2r^2} = \frac{r - \frac{1}{2}}{r^2},$$

so the degeneracy is linear as $r \downarrow \frac{1}{2}$.

Moreover, a min-eigenvector is $u(r, t) \propto (-e^{-it}, 1)^{\top}$ (independent of r). For the support direction $n = e^{it}$,

$$H(n) = \text{Re}(\bar{n}A) = \frac{1}{2} \begin{pmatrix} 0 & e^{-it} \\ e^{it} & 0 \end{pmatrix}, \quad \mathcal{M}(n) = \text{span}\{(e^{-it}, 1)^{\top}\}.$$

At $\sigma_0 = \frac{1}{2}e^{it} \in \partial W(A)$,

$$(\sigma_0 I - A)(e^{-it}, 1)^{\top} = \frac{1}{2}(-1, e^{it})^{\top} \propto (-e^{-it}, 1)^{\top},$$

in agreement with Theorem 1.

4.6. Numerical Experiments

This section provides numerical illustrations of: (i) the linear degeneracy predicted by Corollary 3 (and, in offset form, Proposition 2), (ii) the global coercivity collapse of Corollary 2, and (iii) the contrasting behavior at spectral support points for normal matrices (Proposition 4 and Examples 1–2).

Sampling model for an “outer offset” exhaustion. Fix a unimodular direction $n \in \mathbb{C}$. Let $H(n) = \operatorname{Re}(\bar{n}A)$ and let $v \in \mathcal{M}(n)$ be a unit vector in the maximal eigenspace of $H(n)$ (Lemma 1). The corresponding numerical-range support point is

$$z_0(n) := v^* A v \in \partial W(A), \quad \operatorname{Re}(\bar{n} z_0(n)) = \lambda_{\max}(H(n)).$$

For $\varepsilon > 0$ we define the offset boundary point

$$\sigma_\varepsilon(n) := z_0(n) + \varepsilon n.$$

Then $\operatorname{Re}(\bar{n} \sigma_\varepsilon(n)) = \lambda_{\max}(H(n)) + \varepsilon$, so the support gap equals $\delta(\sigma_\varepsilon(n), n) = \varepsilon$ (cf. Proposition 2). Moreover, $\sigma_\varepsilon(n) \notin W(A)$, hence $\sigma_\varepsilon(n) \notin \operatorname{spec}(A)$ (because $\operatorname{spec}(A) \subset W(A)$; Remark 1), so the resolvent is well-defined.

We evaluate the pointwise kernel

$$P_\varepsilon(n) := \operatorname{Re}\left(n(\sigma_\varepsilon(n)I - A)^{-1}\right),$$

and track $\lambda_{\min}(P_\varepsilon(n))$ as $\varepsilon \downarrow 0$. In the generic (bounded-resolvent) regime $z_0(n) \notin \operatorname{spec}(A)$, Corollary 3 predicts the linear scaling $\lambda_{\min}(P_\varepsilon(n)) = \Theta(\varepsilon)$ and convergence of min-eigenvectors to $(z_0(n)I - A)\mathcal{M}(n)$ (Theorem 1).

Experiment 1: exact linear rate for the nilpotent Jordan block. We revisit Example 3 with $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ and the disk exhaustion $\Omega_r = \{z : |z| < r\}$, $r > \frac{1}{2}$. Writing $\varepsilon = r - \frac{1}{2}$, one has the exact formula $\lambda_{\min}(P_{\Omega_r}(\sigma, A)) = \frac{\varepsilon}{r^2}$, hence linear degeneracy as $\varepsilon \downarrow 0$. Figure 1 compares the computed smallest eigenvalue to the exact expression.

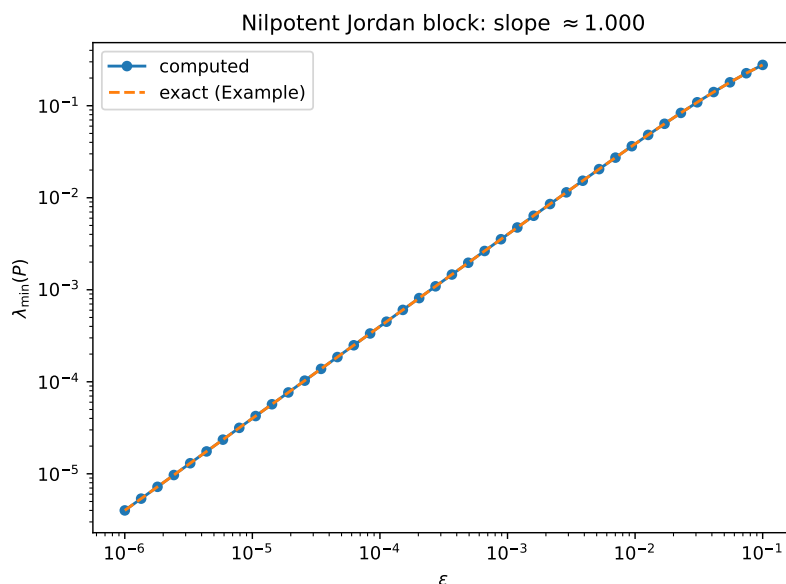


Figure 1. Nilpotent Jordan block (Example 3): log–log plot of $\lambda_{\min}$ versus $\varepsilon = r - \frac{1}{2}$, showing the predicted linear scaling.

Experiment 2: generic nonnormal matrix—linear degeneracy and eigenvector convergence. We generate a fixed random complex matrix $A \in \mathbb{C}^{5 \times 5}$ (seeded for reproducibility), fix one direction

$n = e^{i\theta}$, and form $\sigma_\varepsilon(n) = z_0(n) + \varepsilon n$ as above. Figure 2 shows $\lambda_{\min}(P_\varepsilon(n))$ against ε on a log-log scale, together with a reference $\propto \varepsilon$ line; the observed slope is ≈ 1 on the plotted range. Figure 3 tracks the distance of a min-eigenvector u_ε of $P_\varepsilon(n)$ to the predicted limiting subspace $(z_0(n)I - A)\mathcal{M}(n)$, quantified by $\|(I - \Pi)u_\varepsilon\|$ where Π is the orthogonal projector onto $(z_0(n)I - A)\mathcal{M}(n)$ (consistent with Theorem 1).

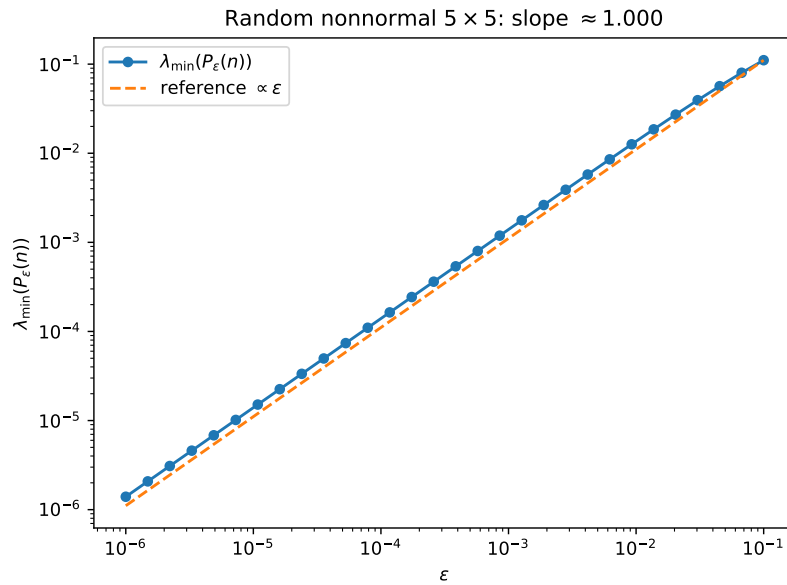


Figure 2. Random nonnormal $A \in \mathbb{C}^{5 \times 5}$ (fixed seed), fixed direction n : $\lambda_{\min}(P_\varepsilon(n))$ scales linearly in ε (Corollary 3).

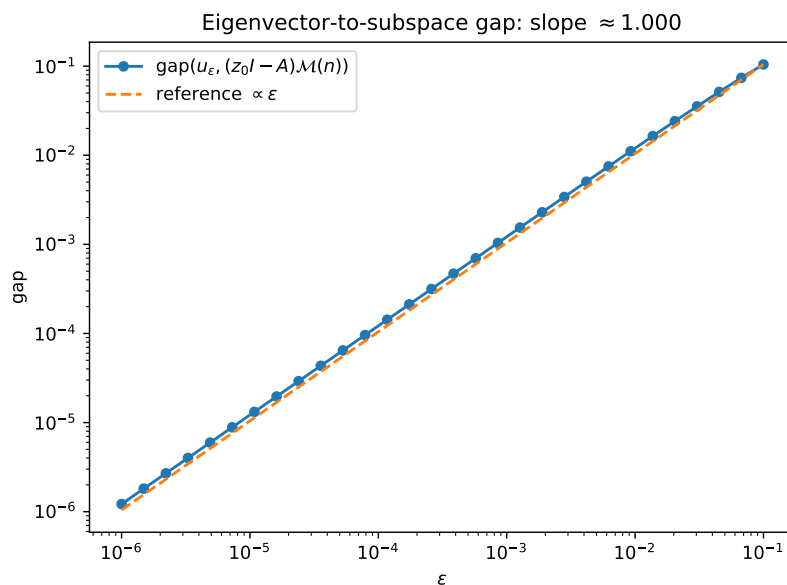


Figure 3. Same setup as Figure 2: the min-eigenvector direction converges to $(z_0(n)I - A)\mathcal{M}(n)$ as $\varepsilon \downarrow 0$ (Theorem 1). The plotted “gap” is $\|(I - \Pi)u_\varepsilon\|$.

Experiment 3: approximate global coercivity collapse. For the same A we approximate the global coercivity constant

$$c(\varepsilon) = \inf_{\sigma \in \partial\Omega_\varepsilon} \lambda_{\min}(P_{\Omega_\varepsilon}(\sigma, A))$$

by sampling a fine grid of directions $\{n_j\}$ and using the offset model $\sigma_\varepsilon(n_j) = z_0(n_j) + \varepsilon n_j$. Figure 4 plots the sampled minimum $\min_j \lambda_{\min}(P_\varepsilon(n_j))$ versus ε , illustrating the collapse asserted by Corollary 2. (Here the offset model has $\Delta(\Omega_\varepsilon) = \varepsilon$, so Corollary 1 also predicts that uniform coercivity cannot persist as $\varepsilon \downarrow 0$.)

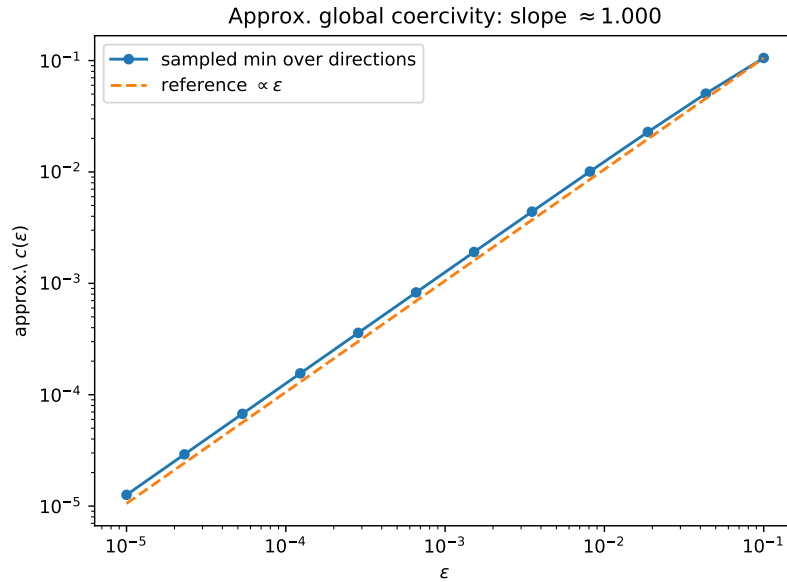


Figure 4. Approximate global coercivity constant $c(\varepsilon)$ computed by sampling directions and using the offset model $\sigma_\varepsilon(n) = z_0(n) + \varepsilon n$. The minimum over directions tends to 0 with $\varepsilon \downarrow 0$ (Corollary 2).

Experiment 4: normal matrices at spectral support points. We reproduce the contrasting behavior in Examples 1–2 by evaluating $P(1 + \varepsilon, 1) = \operatorname{Re}(((1 + \varepsilon)I - A)^{-1})$ for two diagonal (hence normal) matrices: $A = \operatorname{diag}(0, 1)$ and $A = \operatorname{diag}(1, 1 + i)$. Figure 5 shows that the former remains bounded away from 0 as $\varepsilon \downarrow 0$, while the latter degenerates linearly, consistent with Proposition 4.

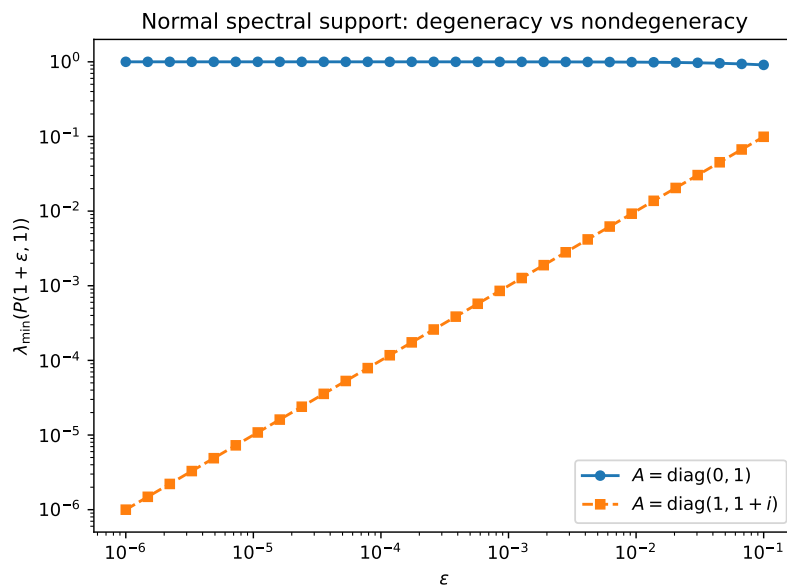


Figure 5. Normal matrices at a spectral support point: nondegeneracy for $A = \operatorname{diag}(0, 1)$ versus linear degeneracy for $A = \operatorname{diag}(1, 1 + i)$, consistent with Proposition 4.

Reproducibility. All figures are generated by the accompanying scripts `poisson_utils.py` and `run_numerical_experiments.py`, which require only NumPy and Matplotlib and save PDF figures into a `figs/` folder.

4.7. Discussion and Open Problems: the Nonnormal Spectral-Support Regime

Remark 6 (Beyond the normal case at spectral support points). *Theorem 1* treats the bounded-resolvent regime $\sigma_0 \notin \text{spec}(A)$, while *Proposition 4* gives a complete description of what can happen at a spectral support point $\sigma_0 \in \text{spec}(A) \cap \partial W(A)$ for normal matrices.

For nonnormal matrices, the regime $\sigma_0 \in \text{spec}(A) \cap \partial W(A)$ appears substantially more delicate: the resolvent $(\sigma I - A)^{-1}$ typically diverges as $\sigma \rightarrow \sigma_0$, and the interplay between (i) the approach geometry $\sigma \rightarrow \sigma_0$ along $\partial\Omega_\epsilon$, (ii) the supporting direction n , and (iii) the Jordan/pseudospectral behavior of A near σ_0 can produce several qualitatively different limits for $\lambda_{\min}(P_\Omega(\sigma, A))$.

Natural questions suggested by the present results include:

- Can one classify (or even bound) the possible asymptotic behavior of $\lambda_{\min}(P_{\Omega_\epsilon}(\sigma_\epsilon, A))$ as $\sigma_\epsilon \rightarrow \sigma_0 \in \text{spec}(A) \cap \partial W(A)$, in terms of the local spectral data of A (e.g. Jordan structure) and the support direction n ?
- In analogy with *Proposition 4*, is there a purely geometric/spectral criterion characterizing when degeneracy must occur at a spectral support point for a general (possibly defective) A ?
- How do these boundary effects interact with quantitative constants in boundary-integral functional calculi and with conditioning of numerical schemes based on C^1 domain approximations $\Omega_\epsilon \downarrow W(A)$?

We leave these questions for future work.

References

1. O. Toeplitz, *Das algebraische Analogon zu einem Satze von Fejér*, Math. Z. **2** (1918), no. 1–2, 187–197. doi:10.1007/BF01212904
2. F. Hausdorff, *Der Wertevorrat einer Bilinearform*, Math. Z. **3** (1919), no. 1, 314–316. doi:10.1007/BF01292610
3. M. Crouzeix, *Bounds for analytical functions of matrices*, Integral Equations Operator Theory **48** (2004), 461–477. doi:10.1007/s00020-002-1188-6
4. M. Crouzeix, *Numerical range and functional calculus in Hilbert space*, J. Funct. Anal. **244** (2007), no. 2, 668–690. doi:10.1016/j.jfa.2006.10.013
5. M. Crouzeix and C. Palencia, *The numerical range is a $(1 + \sqrt{2})$ -spectral set*, SIAM J. Matrix Anal. Appl. **38** (2017), no. 2, 649–655. doi:10.1137/17M1116672
6. B. Delyon and F. Delyon, *Generalization of von Neumann’s spectral sets and integral representation of operators*, Bull. Soc. Math. France **127** (1999), no. 1, 25–41. doi:10.24033/bsmf.2340
7. C. Badea, M. Crouzeix, and B. Delyon, *Convex domains and K -spectral sets*, Math. Z. **252** (2006), no. 2, 345–365. doi:10.1007/s00209-005-0857-y
8. F. L. Schwenninger and J. de Vries, *The double-layer potential for spectral constants revisited*, Integr. Equ. Oper. Theory **97** (2025), 13. doi:10.1007/s00020-025-02800-2
9. M. Crouzeix and A. Greenbaum, *Spectral Sets: Numerical Range and Beyond*, SIAM J. Matrix Anal. Appl. **40** (2019), no. 3, 1087–1101. doi:10.1137/18M1198417
10. C. Davis and W. M. Kahan, *The rotation of eigenvectors by a perturbation. III*, SIAM J. Numer. Anal. **7** (1970), 1–46. doi:10.1137/0707001

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