

An introduction to pseudospectra and application to validated computational dynamics

April Herwig

Department of Mathematics
Technical University of Munich

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Pseudospectra

Definition 1. [Trefethen, Landau, Varah, Godunov, Hinrichsen & Pritchard] The pseudospectrum of a closed linear operator $M : \mathcal{X} \rightarrow \mathcal{X}$ over a Banach space \mathcal{X} is the set

$$\sigma_\epsilon(M) = \bigcup_{\|E\|_{op} \leq \epsilon} \sigma(M + E) \quad (1)$$

228 L.N. Trefethen

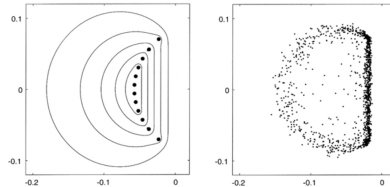


Fig. 2.2. Pseudospectra of a 12×12 Legendre spectral differentiation matrix. The left side shows the eigenvalues (solid dots) and the boundaries of the 2-norm ϵ -pseudospectra for $\epsilon = 10^{-3}, 10^{-4}, \dots, 10^{-7}$ (from outer to inner). The right side shows 1200 of the 10^{-3} -pseudo-eigenvalues of A —specifically, a superposition of the eigenvalues of 100 randomly perturbed matrices $A + E$, where each E is a matrix with independent normally distributed complex entries of mean 0 scaled so that $\|E\| = 10^{-3}$. If all possible perturbations with $\|E\| = 10^{-3}$ were considered instead of just 100 of them, the dots on the right would exactly fill the outermost curve on the left.

Pseudospectra

Theorem 1. [Trefethen] *The following formulations are equivalent:*

(i)

$$\sigma_{\epsilon}(M) = \bigcup_{\|E\|_{op} \leq \epsilon} \sigma(M + E), \quad (2)$$

(ii)

$$\sigma_{\epsilon}(M) = \left\{ z \in \mathbb{C} \mid \|(M - z)^{-1}\|_{op} \geq 1/\epsilon \right\}, \quad (3)$$

(iii)

$$\sigma_{\epsilon}(M) = \left\{ z \in \mathbb{C} \mid \inf_{\|v\|=1} \|(M - z)v\| \leq \epsilon \right\}, \quad (4)$$

(iv) *If $\dim(\mathcal{X}) < \infty$ is an inner product space,*

$$\sigma_{\epsilon}(M) = \{z \in \mathbb{C} \mid s_{min} \leq \epsilon\}. \quad (5)$$

Theoretical results

Geometry of pseudospectra

Theorem 2. [Trefethen]

(i) If M is normal, $\sigma_\epsilon(M) = \sigma(M) + B_\epsilon$ are ϵ -balls around the spectrum.

(ii) $\bigcap_{\epsilon>0} \sigma_\epsilon(M) = \sigma(M)$ and conversely $\sigma_{\epsilon+\delta}(M) \supset \sigma_\epsilon + B_\delta$

Theoretical results

Almost invariant sets

Theorem 3. [Junge]

From now on we assume that $\lambda \neq 1$ is an eigenvalue of P_ϵ with corresponding real valued eigenmeasure $\nu \in \mathcal{M}_\mathbb{C}$, that is,

$$P_\epsilon \nu = \lambda \nu.$$

PROPOSITION 5.7 *Suppose that ν is scaled so that $|\nu| \in \mathcal{M}$, and let $A \subset X$ be a set with $\nu(A) = \frac{1}{2}$. Then*

$$\delta + \sigma = \lambda + 1, \tag{5.3}$$

if A is δ -almost invariant and $X - A$ is σ -almost invariant with respect to $|\nu|$.

Theorem 4. *Suppose ν is an ϵ -pseudoeigenmeasure for the pseudoeigenvalue $\lambda \neq 1$ of a Perron-Frobenius operator P , where $\epsilon = o(1 - \lambda)$. Suppose further that ν is scaled so $|\nu| \in \mathcal{M}$ and A is a set with $\nu(A) = 1/2$. Then*

$$\delta + \sigma = 1 + \lambda + o(1 - \lambda) \tag{6}$$

if A is δ -almost invariant and $X - A$ is σ -almost invariant with respect to $|\nu|$.

A note on backward-stability

Theorem 5. *Let X be a metric space with Borel measure dx . Let $S, \hat{S} : X \rightarrow X$ be two continuous functions with $d_{ess}^\infty(S, \hat{S}) > 0$ ¹. Then the induced Perron-Frobenius (pushforward) operators $P_S, P_{\hat{S}} : \mathcal{M} \rightarrow \mathcal{M}$ satisfy*

$$\|P_S - P_{\hat{S}}\|_{op} \geq 2. \quad (7)$$

This remains true (under adjustment of the const 2) if $P_S, P_{\hat{S}}$ are induced by (sufficiently) small random perturbations in the sense of [Kifer].

Note that this does **not** contradict the continuous dependence of eigenvalues of P_S .

¹ $d_{ess}^\infty(S, \hat{S}) = \text{ess sup}_x d(S(x), \hat{S}(x))$ refers to the essential supremum distance

A note on backward-stability

Proposition 1. ² Let X be a metric space with Borel measure dx . Let $S, \hat{S} : X \rightarrow X$ be two continuous functions. Then

$$d_{ess}^\infty(S, \hat{S}) < \epsilon \quad \Leftrightarrow \quad W_1(P_S \varphi, P_{\hat{S}} \varphi) < \epsilon \quad \forall \varphi \geq 0, \|\varphi\|_{L^1} = 1. \quad (8)$$

Conjecture 1. This remains true (under potential adjustments of constants) if $P_S, P_{\hat{S}}$ are induced by (sufficiently) small random perturbations in the sense of [Kifer].

This seems to suggest considering a new dynamically motivated error metric

$$\min_{\|\varphi\|=1} W_1(P_S \varphi, \lambda \varphi) \quad (9)$$

² W_1 is the Wasserstein-1 metric.

How to compute the pseudospectrum

General inner approximation

Lemma 1. *Let $(V_d)_d$ be a collection of projections which converge pointwise to the identity*
³. *Let*

$$(\lambda, x) \text{ be an } \epsilon\text{-pseudoeigenpair for } V_d M V_d. \quad (10)$$

Then $\lambda \in \sigma_{\epsilon+\delta}(M)$ for all δ and sufficiently large $d = d(\delta, x)$.

Note that this does **not** necessarily imply that $\sigma_\epsilon(V_d M V_d) \nearrow \sigma_\epsilon(M)$ as $d \rightarrow \infty$.

³ $V_d x \xrightarrow{d \rightarrow \infty} x \quad \forall x$

How to compute the pseudospectrum

ResDMD [Colbrook, Townsend]

- Given:
 - quadrature scheme $(w^i, x^i)_{i=1}^m$
 - dictionary $(\psi_j)_{j=1}^N$ of L^2 observables
- Data matrices:

$$\Psi_X = \Psi \cdot (\mathbf{x}) = (\psi_1 \dots \psi_N) \cdot \begin{pmatrix} x^1 \\ \vdots \\ x^m \end{pmatrix} = \begin{pmatrix} \psi_1(x^1) & \dots & \psi_N(x^1) \\ \vdots & & \vdots \\ \psi_1(x^m) & \dots & \psi_N(x^m) \end{pmatrix} \quad (11)$$

$$\Psi_Y = (\Psi \cdot \circ S) \cdot (\mathbf{x}) \quad (12)$$

How to compute the pseudospectrum

ResDMD [Colbrook, Townsend]

$$\begin{array}{ccc}
 \text{Graham matrix} & \text{EDMD matrix} & \text{ResDMD matrix} \\
 G = \Psi_X' W \Psi_X & A = \Psi_X' W \Psi_Y & L = \Psi_Y' W \Psi_Y \\
 \downarrow & \downarrow & \downarrow \\
 \langle \psi_i, \psi_j \rangle_{i,j} & \langle \psi_i, K\psi_j \rangle_{i,j} & \langle K\psi_i, K\psi_j \rangle_{i,j}
 \end{array}
 \quad \begin{array}{cc}
 m \rightarrow \infty & m \rightarrow \infty
 \end{array}
 \quad (13)$$

Now

$$\inf_{\|v\|_{L^2}=1} \|(K - \lambda)v\|^2 = \inf_{\|v\|_{L^2}=1} \langle v, Kv \rangle - \bar{\lambda} \langle v, Kv \rangle - \lambda \langle Kv, v \rangle + |\lambda|^2 \langle v, v \rangle \quad (14)$$

$$\approx \inf_{\mathbf{v}' G \mathbf{v} = 1} \mathbf{v}' (L - \bar{\lambda} A - \lambda A' + |\lambda|^2 G) \mathbf{v} \quad (15)$$

How to compute the pseudospectrum

Residual Ulam

Theorem 6. *Let S be an a.e. diffeomorphism onto its image, $P = P_S$ the induced transfer operator. Consider a box partition $\mathcal{P} = \{A_1, \dots, A_N\}$ of the phase space X . Then*

$$\inf_{\|v\|_{L^2}=1} \|(P - \lambda)v\|^2 \approx \inf_{\mathbf{v}' G \mathbf{v}=1} v'(L - \bar{\lambda} A - \lambda A' + |\lambda|^2 G)v \quad (16)$$

where

$$L_{i,j} = \int_{A_i \cap A_j} \frac{dx}{|\det DS(x)|}, \quad A_{i,j} = \underbrace{m(A_i \cap S^{-1}(A_j))}_{(scaled) \text{ Ulam matrix}}, \quad G_{i,j} = m(A_i \cap A_j). \quad (17)$$