

# An introduction to pseudospectra and application to validated computational dynamics

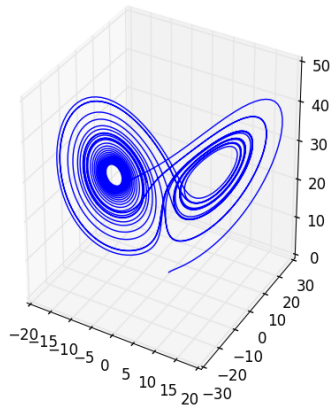
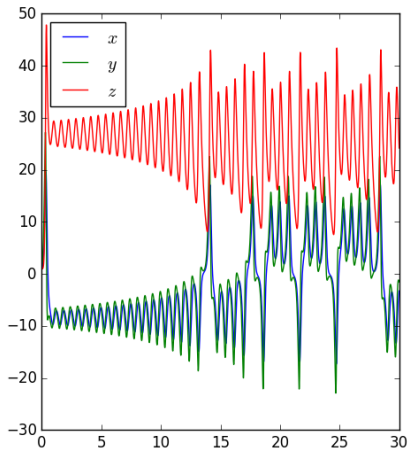
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# Dynamical systems



# Dynamical systems

- **Discrete dynamical system** generated by iteration of a continuous map  $S : X \rightarrow X$
- **Observables**  $\psi : X \rightarrow \mathbb{C}$  can be used to measure statistical behavior
- Evolution of observables is dictated by the operators

Koopman	Perron-Frobenius
$K : L^2 \rightarrow L^2$	$P : \mathcal{M} \rightarrow \mathcal{M}$
$f \mapsto f \circ S$	$\mu \mapsto \mu \circ S^{-1}$

- The spectrum of these operators describe **macroscopic asymptotic** statistics of the system

## Petrov-Galerkin discretization of linear operators

- Given: bounded linear operator on a Banach space  $M : \mathcal{X} \rightarrow \mathcal{X}$
- Approximation space  $U = \{\varphi_i\}_{i=1}^n \subset \mathcal{X}$
- Trial space  $V = \{\psi_j\}_{j=1}^m \subset \mathcal{X}^*$
- Representation matrix  $A_{i,j} = \psi_j(M\varphi_i)$
- Examples:
  - EDMD:  $U =$  Fourier / radial basis functions,  $V =$  point evaluation functionals
  - Ulam's method:  $U = V =$  characteristic functions
  - ...
- We now wish to compute eigenpairs of  $A$

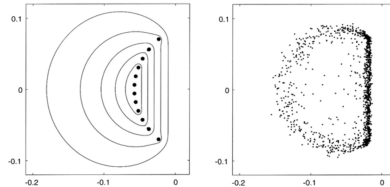
How 'well' does the spectrum of  $A$  reproduce the spectrum of  $M$ ?

# 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

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

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**Fig. 2.2.** Pseudospectra of a  $12 \times 12$  Legendre spectral differentiation matrix. The left side shows the eigenvalues (solid dots) and the boundaries of the 2-norm  $\epsilon$ -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  $\epsilon$ -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. [Dellnitz, Junge]

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

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

PROPOSITION 5.7 Suppose that  $\nu$  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  $\delta$ -almost invariant and  $X - A$  is  $\sigma$ -almost invariant with respect to  $|\nu|$ .

**Theorem 4.** Suppose  $\nu$  is an  $\epsilon$ -pseudoeigenmeasure for the  $\epsilon$ -pseudoeigenvalue  $0 < \lambda < 1$  of a Perron-Frobenius operator  $P$ . Suppose further that  $\nu$  is scaled so  $|\nu| \in \mathcal{M}$  and  $A$  is a set with  $\nu(A) = 1/2$ . Then

$$\delta + \sigma = \lambda + 1 + \text{const} \cdot \epsilon \tag{6}$$

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

## A note on backward-stability

**Theorem 5.** *Let  $(X, d)$  be a metric space with Borel measure. Let  $S, \hat{S} : X \rightarrow X$  be two continuous functions with*

$$d_{\text{ess}}^{\infty}(S, \hat{S}) = \operatorname{ess\,sup}_x d(S(x), \hat{S}(x)) > 0 \quad (7)$$

*Then the induced Perron-Frobenius (pushforward) operators  $P_S, P_{\hat{S}} : \mathcal{M} \rightarrow \mathcal{M}$  satisfy*

$$\|P_S - P_{\hat{S}}\|_{op} \geq 2. \quad (8)$$

*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  $P_S$ .

# How to compute the pseudospectrum

## General inner approximation

**Lemma 1.** *Let  $M : \mathcal{X} \rightarrow \mathcal{X}$  be a closed linear operator,  $(\Pi_d)_d$  be a collection of projections which converge pointwise to the identity<sup>1</sup>. Let*

$$(\lambda, x) \text{ be an } \epsilon\text{-pseudoeigenpair for } \Pi_d M \Pi_d. \quad (9)$$

*Then for every  $\delta$  there exists a  $D = D(\delta, x)$  such that  $\lambda \in \sigma_{\epsilon+\delta}(M)$  for all  $d > D$ .*

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

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<sup>1</sup>  $V_d x \xrightarrow{d \rightarrow \infty} x \quad \forall x$

# How to compute the pseudospectrum

## EDMD [Williams, Kevrekidis, Rowley]

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

$$\Psi_X = \Psi.(\mathbf{x}) = \begin{pmatrix} \psi_1(x^1) & \cdots & \psi_N(x^1) \\ \vdots & & \vdots \\ \psi_1(x^m) & \cdots & \psi_N(x^m) \end{pmatrix} \quad (10)$$

(11)

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

## 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

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

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

# How to compute the pseudospectrum

## Residual Ulam

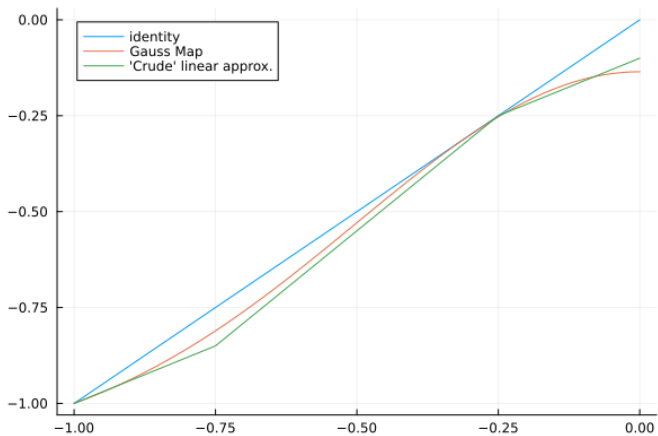
**Theorem 6.** *Let  $S : X \rightarrow X$  be an a.e. diffeomorphism onto its image,  $P = P_S$  the induced transfer operator. Consider a box partition  $\mathcal{P} = \{A_1, \dots, A_N\}$  of the phase space  $X$ . Then*

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

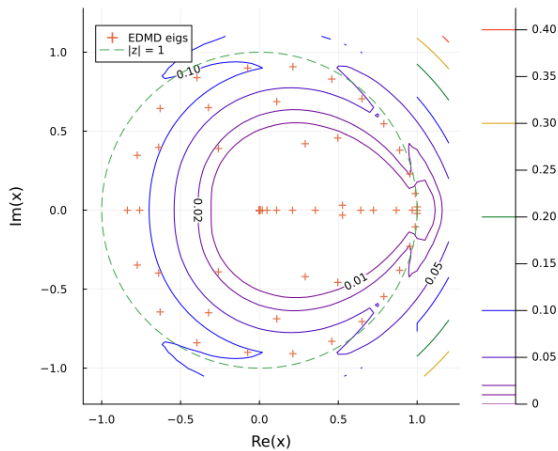
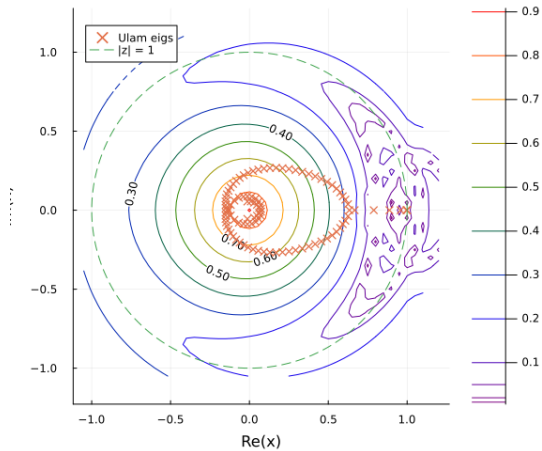
where

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

# Numerical results



# Numerical results





# Numerical results

