



Matrix functions and large-scale matrix
equations:
effective numerical solution strategies

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The Problem

Given $v \in \mathbb{R}^n$ and $A \in \mathbb{R}^{n \times n}$, approximate

$$x = f(A) v$$

with f regular function such that $f(A)$ is well defined

Focus:

- A large dimension
- A symmetric pos. (semi)def., or A *positive real*

Context

- A of small dimension:

$$A \text{ symmetric, } A = X\Lambda X^\top \Rightarrow f(A) = Xf(\Lambda)X^\top$$

Similar, but more involved, the definition for A nonsymmetric

- A medium to large dimension:

$$f(A) \quad \text{vs.} \quad f(A)v$$

Applications

Among which:

- Numerical solution of evolution PDEs
(e.g. $\exp(\lambda)$, $\sqrt{\lambda^{-1}}$, $\cos(\lambda)$, $\varphi_k(\lambda)$...)
- Numerical solution of some Inverse Problems ($\exp(\lambda)$, $\cosh(\lambda)$, ...)
- Fluxes on manifolds
- Scientific Computing problems (e.g. QCD, $\text{sign}(\lambda)$)
- (Analysis of) reduced Dynamical System Models
(through Grammian Matrices)

Numerical approximation. I

$$f(A)v \approx \tilde{x} \quad \tilde{x} = ???$$

Two wide paths:

- Substitute f with “simpler” function, $f \approx \mathcal{R}$

$$\|f(A)v - \tilde{x}\| \leq \|f(A)v - \mathcal{R}(A)v\| + \|\mathcal{R}(A)v - \tilde{x}\|$$

and $\Rightarrow \tilde{x} \approx \mathcal{R}(A)v$

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- Approximation by projection: Find V and

$$\tilde{x} \in \text{range}(V), \quad \text{space dim} \ll n$$

Numerical approximation. II

$$f(A)v \approx \tilde{x}$$

Important issues:

- ★ Measure accuracy of approximation?
- ★ Relation between f and quality of approximation
- ★ Relation between A and quality of approximation
- ★ Efficiency ?

Rational Approximation

$$x = f(A)v \approx \mathcal{R}_{\mu,\nu}(A)v$$

$$\mathcal{R}_{\mu,\nu}(\lambda) = \frac{\Phi_{\mu}(\lambda)}{\Psi_{\nu}(\lambda)}, \quad \Phi_{\mu}(\lambda), \Psi_{\nu}(\lambda) \text{ polynomials}$$

- Polynomial Approx., $\nu = 0$
 - Rational Approx.: Padé or Chebyshev, e.g. $\mu = \nu$
 - Rational Approx with multiple pole
 - Quadrature Methods
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We consider the case of partial fraction expansion:

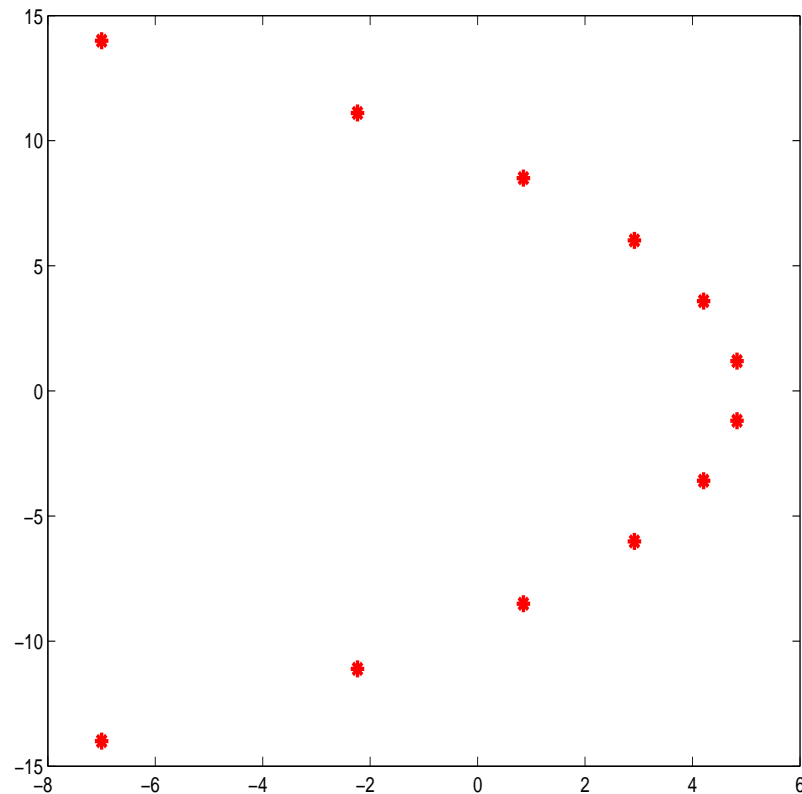
$$\mathcal{R}_{\mu,\nu}(\lambda) = q(\lambda) + \sum_{k=1}^{\nu} \frac{\omega_k}{\lambda - \xi_k} \quad (\mathcal{R}_{\nu} = \mathcal{R}_{\nu,\nu})$$

Rational Approximation: poles

$$f(\lambda) = \exp(-\lambda)$$

\mathcal{R}_ν : ℓ_∞ best approx
in $[0, \infty)$, Chebyshev

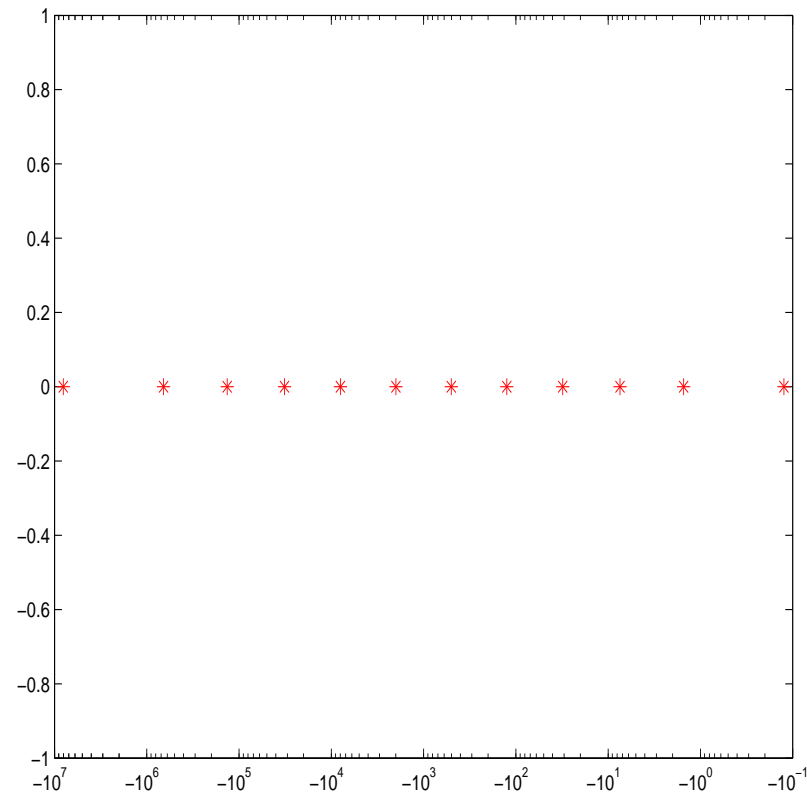
$$\|f - \mathcal{R}_\nu\|_\infty \approx 10^{-\nu}$$



$$f(\lambda) = \lambda^{-\frac{1}{2}}$$

\mathcal{R}_ν : Zolotarev approx
in $[a, b] \subseteq (0, \infty)$

$$\|f - \mathcal{R}_\nu\| \approx e^{-\pi\sqrt{2\nu}}$$



Matrix Rational approximation

$$\begin{aligned} f(A)v &\approx \mathcal{R}_\nu(A)v = \sum_{k=1}^{\nu} \omega_k (A - \xi_k I)^{-1} v \\ &\approx \sum_{k=1}^{\nu} \omega_k \tilde{x}_k \end{aligned}$$

- $\forall k, (A - \xi_k I)$ “Shifted” matrix, $\xi_k \in \mathbb{C}$
- $\xi_{2j-1} = \bar{\xi}_{2j}, j = 1, \dots, \lfloor \nu/2 \rfloor$
- $\forall k, \tilde{x}_k$ approx solution

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\Rightarrow Iterative Methods for **shifted** linear systems

Error estimates

\tilde{x}_k : Krylov subspace methods: $\sum_{k=1}^{\nu} \omega_k (A - \xi_k I)^{-1} v \approx \sum_{k=1}^{\nu} \omega_k \tilde{x}_k$

$$\|\mathcal{R}_{\nu}(A)v - \sum_{k=1}^{\nu} \omega_k \tilde{x}_k\| = ??$$

Error estimate during iteration : (Frommer & S., '08)

- Estimate for real symmetric A and complex poles
- Lower estimate for A spd and real negative poles

Error estimates

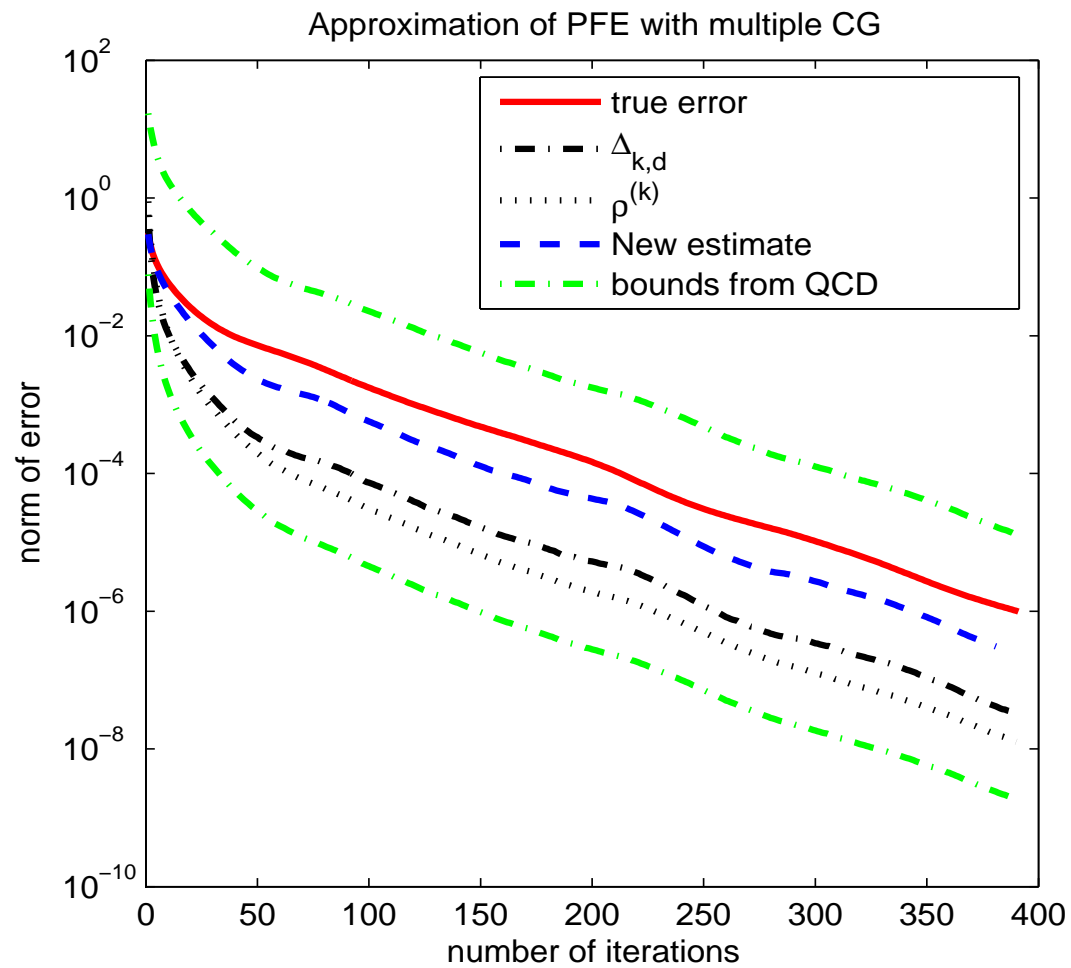
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Error estimate during **iteration** : (Frommer & S., '08)

- Estimate for real symmetric A and complex poles
- Lower estimate for A spd and real negative poles
- ★ Does not require spectral info
- ★ Computational cost only 3-5 additional iterations

CG for A spd and $f(\lambda) = \text{sign}(\lambda) = (\lambda^2)^{-\frac{1}{2}} \lambda$: $\text{sign}(A)v$



Projection-type methods

\mathcal{K} approximation space, $m = \dim(\mathcal{K})$

$V \in \mathbb{R}^{n \times m}$ s.t. $\mathcal{K} = \text{range}(V)$

$$x = f(A)v \approx x_m = Vf(V^\top AV)(V^\top v)$$

Question: Which \mathcal{K} ?

Some explored alternatives for \mathcal{K}

- Krylov subspace, $\mathcal{K} = K_m(A, v)$
- Shift-Invert Krylov subspace, $\mathcal{K} = K_m((I + \gamma A)^{-1}, v)$ for some γ
- Rational Krylov subspace, for some $\omega_1, \omega_2, \dots$
$$\mathcal{K} = \text{span}\{v, (A - \omega_1 I)^{-1}v, (A - \omega_2 I)^{-1}v, \dots\}$$
- Extended Krylov subspace, $\mathcal{K} = K_m(A, v) + K_m(A^{-1}, A^{-1}v)$
- Restarted Krylov subspace

Note: In all cases, A nonsymmetric.

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Theory mostly for field of values of A in \mathbb{C}^+

Field of values: $W(A) = \{x^* Ax, x \in \mathbb{C}^n, \|x\| = 1\}$

Krylov subspace approximation

“Classical” approach:

$$\mathcal{K} = K_m(A, v) = \text{span}\{v, Av, \dots, A^{m-1}v\}$$

For $H_m = V_m^\top AV_m$, $v = V_m e_1$ and $V_m^\top V_m = I_m$:

$$x_m = V_m f(H_m) e_1$$

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Polynomial approximation: $x_m = p_{m-1}(A)v$
(p_{m-1} interpolates f at eigenvalues of H_m)

★ Numerical and theoretical results since mid '80s

(van der Vorst'87, Saad'92, Hochbruck & Lubich'97, ..., Beckermann & Reichel'10)

Application. Evolution Problem

$$\left\{ \begin{array}{l} \frac{\partial u(x,y,t)}{\partial t} = \Delta u, \quad (x,y) \in (0,1)^2 \quad t \in [0,0.1] \\ u(x,y,t) = 0, \quad (x,y) \in \partial([0,1]^2) \\ u(x,y,0) = 1, \quad (x,y) \in [0,1]^2 \end{array} \right.$$

Implicit Euler: $u_{i+1} = (I + \delta t A)^{-1} u_i, \quad i = 0, 1, \dots$

Exponential Integrator: $u(t) = \exp(-tA)u_0 \quad t = 0.1$

Application. Evolution Problem

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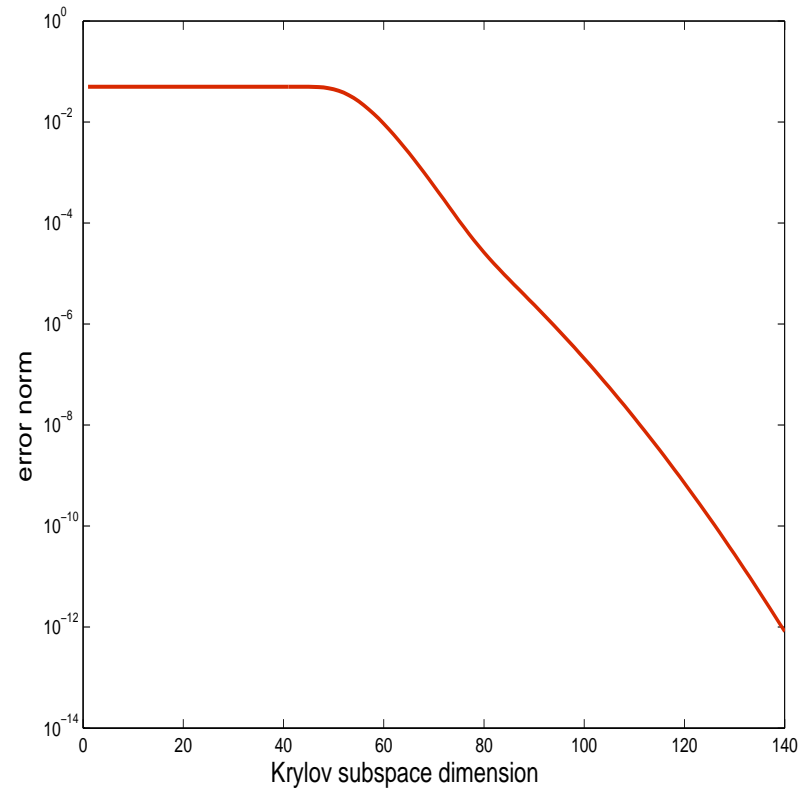
	Euler		Exp	
step δt	CPU	error	CPU	error (#its*)
0.001	1.9	$2 \cdot 10^{-3}$	0.09	$9 \cdot 10^{-4}$ (37)
0.005	0.4	$1 \cdot 10^{-2}$	0.07	$4 \cdot 10^{-3}$ (28)
0.01	0.2	$2 \cdot 10^{-2}$	0.05	$1 \cdot 10^{-2}$ (25)

* : Stopping criterion tolerance related to Euler timestep

⇒ More general exponential integrators (see Hochbruck's talk)

...When things are not so easy

$$\| \exp(-A)v - V_m \exp(-H_m)e_1 \| \quad A \in \mathbb{R}^{400 \times 400}, \|A\| = 10^5$$



$$\| \exp(-A)v - V_m \exp(-H_m)e_1 \| \leq 10e^{-m^2/(5\rho)}, \quad \sqrt{4\rho} \leq m \leq 2\rho$$

where $\sigma(A) \subseteq [0, 4\rho]$

Acceleration Procedures: Shift-Invert Krylov

Choose γ s.t. $(I + \gamma A)$ is invertible, and construct

$$\mathcal{K} = K_m((I + \gamma A)^{-1}, v), \quad \text{Moret-Novati '04, van den Eshof-Hochbruck '06}$$

with $T_m = V_m^\top (I + \gamma A)^{-1} V_m$, $v = V_m e_1$ and $V_m^\top V_m = I_m$

$$x_m = V_m f\left(\frac{1}{\gamma}(T_m^{-1} - I_m)\right)e_1$$

Rational approximation: $x_m = p_{m-1}((I + \gamma A)^{-1})v$

Choice of γ : A spd, $\gamma = \frac{1}{\sqrt{\lambda_{\min} \lambda_{\max}}}$ (Moret, '09)

A nonsym, (Beckermann & Reichel'10)

Acceleration Procedures: Extended Krylov

For A nonsingular,

$$\mathcal{K} = K_{m_1}(A, v) + K_{m_2}(A^{-1}, A^{-1}v), \quad \text{Druskin \& Knizhnerman'98, } A \text{ sym.}$$

Note: $\mathcal{K} = A^{-m_2} K_{m_1+m_2}(A, v)$

Algorithm (augmentation-style)

- Fix $m_2 \ll m_1$
- Run m_2 steps of Inverted Lanczos
- Run m_1 steps of Standard Lanczos + orth.

Extended Krylov: an effective implementation

$m_1 = m_2 = m$ **not** fixed a priori

$$\begin{aligned}\mathcal{K} &= K_m(A, v) + K_m(A^{-1}, A^{-1}v) \\ &= \text{span}\{v, A^{-1}v, Av, A^{-2}v, A^2v, \dots\}\end{aligned}$$

★ *Block* Arnoldi-type recurrence:

- $U_1 \leftarrow \text{orth}([v, A^{-1}v])$

- $U_{j+1} \leftarrow [AU_j(:, 1), A^{-1}U_j(:, 2)] + \text{orth} \quad j = 1, 2, \dots$

★ Recurrence to cheaply compute $\mathcal{T}_m = \mathcal{U}_m^\top A \mathcal{U}_m$, $\mathcal{U}_m = [U_1, \dots, U_m]$

★ Compute $x_m = \mathcal{U}_m f(\mathcal{T}_m) e_1$

Simoncini '07

Extended Krylov: Convergence theory I

f satisfying $f(z) = \int_{-\infty}^0 \frac{1}{z - \zeta} d\mu(\zeta), \quad z \in \mathbb{C} \setminus]-\infty, 0]$

(with convenient measure $d\mu(\zeta)$)

Extended Krylov: Convergence theory II

Nonsingular A , with $0 \notin W(A)$.

Let Φ_1, Φ_2 be the conformal mappings for $W(A)$ and $W(A)^{-1}$

There exists $a > 0$ s.t. $|\Phi_1(-a)| = |\Phi_2(-\frac{1}{a})|$ so that

$$\|f(A)v - \mathcal{U}_m f(\mathcal{T}_m)e_1\| \leq \frac{c}{|\Phi_1(-a)|^m}$$

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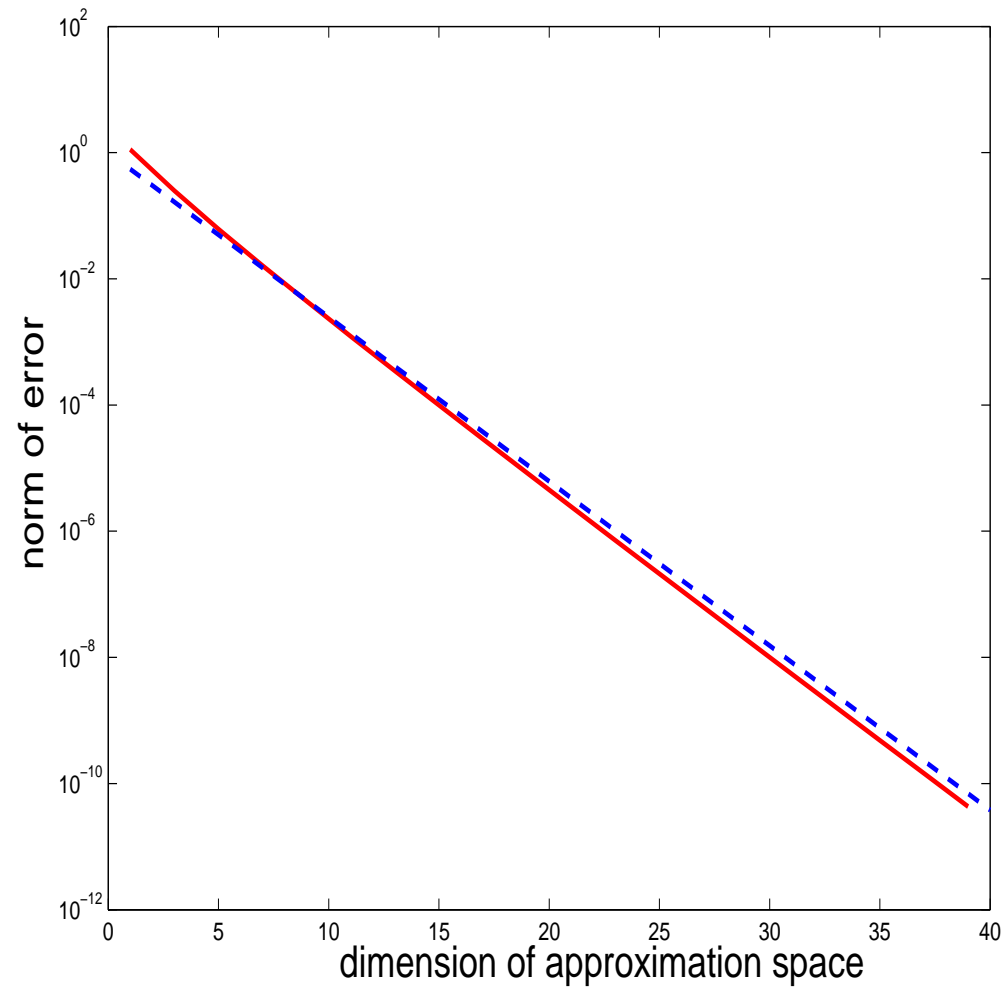
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e.g. for A symmetric (Φ_1, Φ_2 known, $a = \sqrt{\lambda_{\min} \lambda_{\max}}$) :

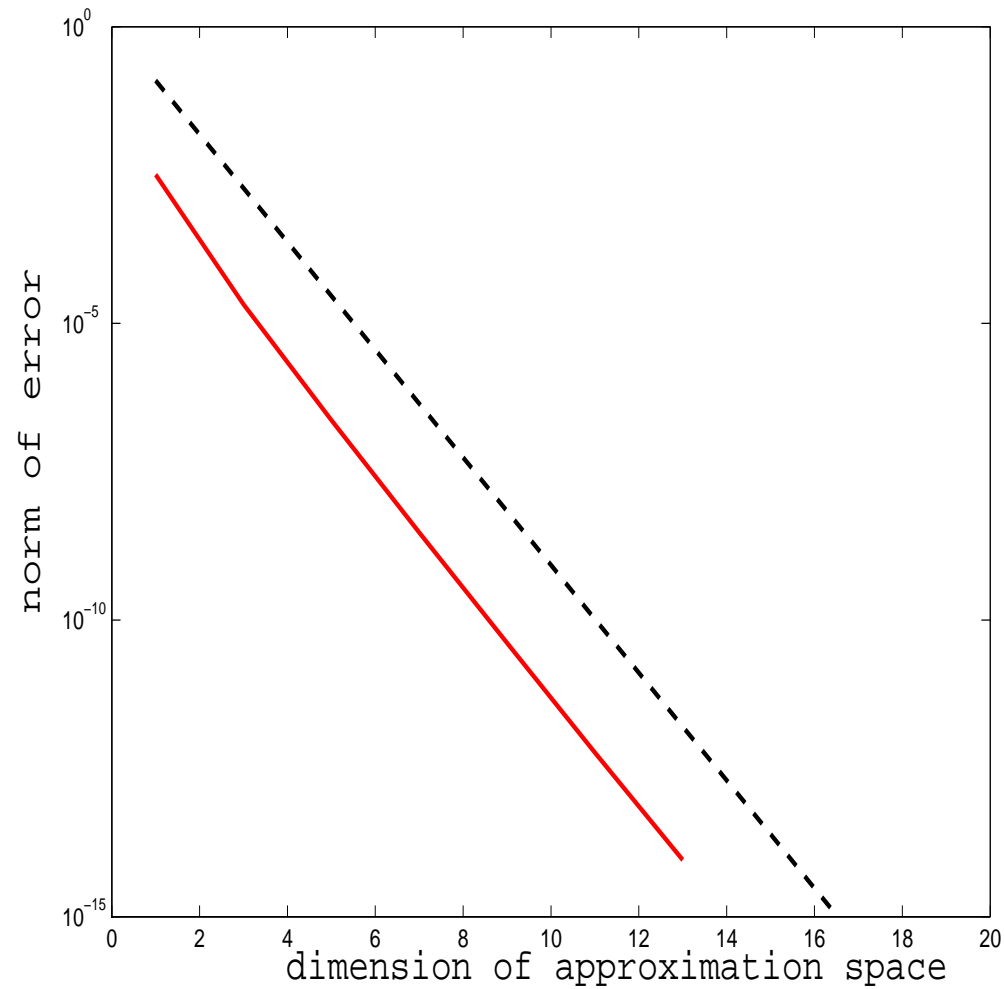
$$\|x - x_m\| = O(\exp(-2m \sqrt[4]{\frac{\lambda_{\min}}{\lambda_{\max}}}))$$

Convergence rate. $A \in \mathbb{R}^{400 \times 400}$ normal. $f(\lambda) = \lambda^{-\frac{1}{2}}$



$\sigma(A)$ on an elliptic curve in \mathbb{C}^+ with center on real axis

Rate. $A \in \mathbb{R}^{200 \times 200}$ Jordan block, $\sigma(A) = \{4\}$. $f(\lambda) = \lambda^{\frac{1}{2}}$



$W(A)$ disk centered at 4 and unit radius

Large-scale numerical experiments

A from FD discretization of

$$\mathcal{L}_1(u) = -100u_{x_1x_1} - u_{x_2x_2} + 10x_1u_{x_1},$$

$$\mathcal{L}_2(u) = -100u_{x_1x_1} - u_{x_2x_2} - u_{x_3x_3} + 10x_1u_{x_1},$$

$$\mathcal{L}_3(u) = -e^{-x_1x_2}u_{x_1x_1} - e^{x_1x_2}u_{x_2x_2} + \frac{1}{x_1 + x_2}u_{x_1},$$

$$\mathcal{L}_4(u) = -\operatorname{div}(e^{3x_1x_2}\operatorname{grad}u) + \frac{1}{x_1 + x_2}u_{x_1}$$

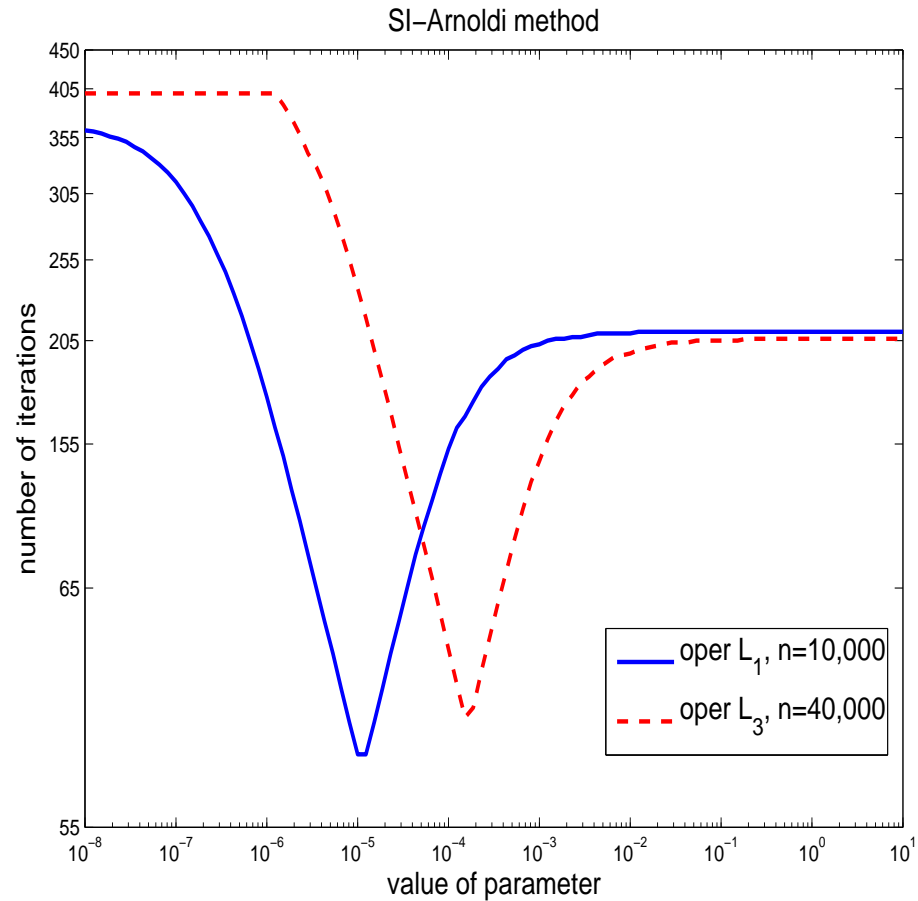
on unit square/cube, Dirichlet hom. bc.

Inner system solves:

- Extended Krylov: systems with A solved with GMRES/AMG
- SI-Arnoldi: $\operatorname{span}\{v, (I + \gamma A)^{-1}v, (I + \gamma A)^{-2}v, \dots\}$
systems with $I + \gamma A$ solved with IDR(s)/ILU

An intermezzo

SI-Arnoldi requires getting the parameter γ :



Number of SI-Arnoldi iterations as a function of the parameter for $f(\lambda) = \lambda^{\frac{1}{2}}$

Comparisons: CPU Time in Matlab (space dim.)

f	Oper.	n	SI-Arnoldi	EKSM	Std Krylov
$\lambda^{\frac{1}{2}}$	\mathcal{L}_1	2500	0.9 (59)	0.6 (48)	7 (193)
		10000	4.0 (66)	3.6 (68)	*46 (300)
		160000	642.9(246)	219.7(122)	*458(300)
	\mathcal{L}_2	27000	10.8 (55)	7.4 (40)	6.7(119)
		125000	86.7 (60)	65.3 (52)	138.7(196)
	\mathcal{L}_3	40000	26.3 (75)	21.1 (72)	*87 (300)
		160000	318.5(144)	173.3 (96)	*442(300)
	\mathcal{L}_4	40000	41.1(117)	25.4(106)	*89 (300)
		160000	580.2(442)	231.2(144)	*461 (300)

Comparisons: CPU Time in Matlab (space dim.)

f	Oper.	n	SI-Arnoldi	EKSM	Std Krylov
$\lambda^{-\frac{1}{3}}$	\mathcal{L}_1	2500	0.6 (43)	0.4 (30)	2.2(131)
		10000	2.6 (46)	1.8 (38)	26.2(252)
		160000	79.3 (48)	99.7 (64)	*460(300)
	\mathcal{L}_2	27000	7.8 (41)	4.8 (26)	3.1 (82)
		125000	64.8 (45)	38.9 (32)	67.5(138)
	\mathcal{L}_3	40000	20.7 (61)	13.7 (48)	*88 (300)
		160000	116.5 (62)	105.2 (62)	*460 (300)
	\mathcal{L}_4	40000	35.8(104)	14.2 (66)	*88 (300)
		160000	208.1(104)	112.2 (84)	*461 (300)

Stopping criterion

Unlike linear systems: **no equation \Rightarrow no residual**

Estimate of the error:

(first suggested for $f(\lambda) = e^{-\lambda}$ by van den Eshof-Hochbruck '06)

$$\frac{\|x - x_m\|}{\|x_m\|} \approx \frac{\delta_{m+j}}{1 - \delta_{m+j}}, \quad \delta_{m+j} = \frac{\|x_{m+j} - x_m\|}{\|x_m\|}$$

Stopping criterion:

$$\mathbf{if} \frac{\delta_{m+j}}{1 - \delta_{m+j}} \leq \mathbf{tol} \mathbf{ then stop}$$

Computational costs awareness: inexact solves in EKSM

systems with A : GMRES with **relaxed** inner tolerance

$$\epsilon_m^{(\text{inner})} = \frac{\text{tolin}}{\|x - x_{m-1}\|}.$$

Final outer error (# outer its / # inner its)

tolin	fixed inner tol	relaxed inner tol
1e-10	6.97e-11 (24/901)	6.58e-11 (24/559)
1e-12	6.48e-11 (24/1052)	6.48e-11 (24/716)

$$\mathcal{L}(u) = -u_{xx} - u_{yy} - u_{zz} + 50(x + y)u_x$$

$$f(\lambda) = \lambda^{-\frac{1}{3}} \quad \epsilon^{(\text{outer})} = 10^{-10}$$

References

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4. L. Knizhnerman and V. Simoncini, *A new investigation of the extended Krylov subspace method for matrix function evaluations*, July 2008, pp. 1-17. To appear in Numerical Linear Algebra w/App. l.