

# On the convergence of the residual inverse iteration

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# Nonlinear eigenvalue problems

## Mathematical formulation.

For analytic  $T : \Omega \rightarrow \mathbb{C}^{n \times n}$ ,  $\Omega \subseteq \mathbb{C}$ , find eigenvalues  $\lambda \in \mathbb{C}$ :

$$T(\lambda)x = 0, \quad x \neq 0 \quad (\text{NLEVP})$$

Several exciting recent algorithmic developments:

- ▶ nonlinear Arnoldi [Voss'2004]
- ▶ rational linearization [Su/Bai'2011]
- ▶ block Newton [K.'2009]
- ▶ infinite Arnoldi [Jarlebring/Meerbergen/Michiels'2010/2012]
- ▶ contour integral method  
[Asakura/Sakurai/Tadano/Ikegami/Kimura'2009, Beyn'2012, ...]
- ▶ interpolation based approaches [Effenberger/K.'2012, Van Beeumen/Meerbergen/Michiels'2013, K./Roman'2013, Güttel/Van Beeumen/Meerbergen/Michiels'2013]
- ▶ Jacobi-Davidson with eigenvalue locking [Effenberger'2013]
- ▶ ...

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- ▶ Jacobi-Davidson with eigenvalue locking [Effenberger'2013]
- ▶ ...

Not the topic of this talk!

# Residual inverse iteration

Plain Newton applied to nonlinear system of equations

$$T(\lambda)x = 0, \quad w^*x = 1$$

yields:

$$y_{j+1} = T(\lambda_j)^{-1} T'(\lambda_j) v_j$$

$$\lambda_{j+1} = \lambda_j - 1/w^* y_{j+1}$$

$$v_{j+1} = y_{j+1}/w^* y_{j+1}$$

↪ Need derivative and need to refactorize  $T(\lambda_j)$  in every step.

# Residual inverse iteration

Quasi-Newton applied to nonlinear system of equations

$$T(\lambda)x = 0, \quad w^*x = 1$$

yields:

$$y_{j+1} = T(\lambda_j)^{-1} \frac{1}{\lambda_{j+1} - \lambda_j} (T(\lambda_{j+1}) - T(\lambda_j)) v_j$$

$$\lambda_{j+1} = \lambda_j - 1/w^* y_{j+1}$$

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- ▶ In the linear Hermitian case: Rayleigh quotient solves  $v_j^* A v_j - \lambda v_j^* v_j = 0$ .
- ▶ In the nonlinear Hermitian case:  $\lambda_{j+1}$  solves  $v_j^* T(\lambda) v_j = 0$ .



# Residual inverse iteration

for  $j = 0, 1, 2, \dots$

Compute solution  $\lambda_{j+1}$  of  $v_j^* T(\lambda) v_j = 0$ .

$$v_{j+1} = v_j - T(\lambda_j)^{-1} T(\lambda_{j+1}) v_j$$

Normalize  $v_{j+1}$ .

end for

↪ Still need to refactorize  $T(\lambda_j)$  in every step!

# Residual inverse iteration

Choose fixed  $\sigma$  sufficiently close to eigenvalue of interest.

for  $j = 0, 1, 2, \dots$

    Compute solution  $\lambda_{j+1}$  of  $v_j^* T(\lambda) v_j = 0$ .

$$v_{j+1} = v_j - T(\sigma)^{-1} T(\lambda_{j+1}) v_j$$

    Normalize  $v_{j+1}$ .

end for

Finally...

... this is RESINVIT (residual inverse iteration) by Neumaier [1985].

# Residual inverse iteration

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    Normalize  $v_{j+1}$ .

end for

- For linear  $T(\lambda) = \lambda I - A$ : Odd way of writing inverse iteration.

When using preconditioner  $P \approx -T(\sigma) \rightsquigarrow$  PINVIT.

Preconditioner for nonlinear eigenvalue problems?

# Model problem

Vibrating string with elastically attached mass [Solov'ev'2006]:

$$-u''(x) = \lambda u(x), \quad u(0) = 0, \quad u'(1) + \phi(\lambda)u(1) = 0, \quad \phi(\lambda) = \frac{\lambda \eta m}{\lambda - \eta}.$$

Discretization with pw linear FEs  $\rightsquigarrow$

$$(K(\lambda) - \lambda M)v = 0$$

with  $K(\lambda) = K_0 + \phi(\lambda)\mathbf{e}_n\mathbf{e}_n^\top$  and

$$K_0 = \frac{1}{h} \begin{bmatrix} 2 & -1 & & \\ -1 & \ddots & \ddots & \\ & \ddots & 2 & -1 \\ & & -1 & 1 \end{bmatrix}, \quad M = \frac{h}{6} \begin{bmatrix} 4 & 1 & & \\ 1 & \ddots & \ddots & \\ & \ddots & 4 & 1 \\ & & 1 & 2 \end{bmatrix}.$$

# Model problem

- ▶ Aim to compute smallest eigenvalue  $\lambda_1 = 4.482024295 \dots$
- ▶ Apply RESINVIT with preconditioner for  $T(\sigma)$  = one W-cycle of multigrid with Jacobi smoother.
- ▶ Stop when residual  $\leq 10^{-6}$ .

$h$	#iterations		
	$\sigma = 0$	$\sigma = 2$	$\sigma = 4$
$2^{-5}$	12	8	5
$2^{-6}$	12	8	5
$2^{-7}$	12	8	5
$2^{-8}$	12	7	4
$2^{-9}$	12	7	4
$2^{-10}$	12	7	4
$2^{-11}$	11	7	4
$2^{-12}$	11	7	4
$2^{-13}$	11	6	4
$2^{-14}$	10	6	4
$2^{-15}$	10	6	4
$2^{-16}$	10	6	4
$2^{-17}$	9	6	3
$2^{-18}$	9	5	3

# Goal of this talk

Derive convergence result that allows to conclude

mesh-independent asymptotic convergence rates

when using a spectrally equivalent preconditioner for  $T(\sigma)$ .

[Solov'ëv'2006] has established such a result for “monotonic” nonlinear eigenvalue problems

$$T(\lambda) = A(\lambda) - \lambda B(\lambda)$$

with spd  $A(\lambda), B(\lambda)$ .

# Convergence result by Neumaier

- ▶ Consider simple eigenvalue  $\lambda_1$  with eigenvector  $x_1$  normalized s.t.  $w^*x_1 = 1$ .
- ▶ Assume that  $\sigma$  is sufficiently close to  $\lambda_1$ .
- ▶ Exact preconditioner  $T(\sigma)^{-1}$ .

Then RESINVIT converges with

$$\frac{\|v_{j+1} - x_1\|_2}{\|v_j - x_1\|_2} = O(|\sigma - \lambda_1|), \quad |\lambda_{j+1} - \lambda_1| = O(\|v_{j+1} - x_1\|_2).$$

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**But:** Not helpful if constant in  $O(|\sigma - \lambda_1|)$  deteriorates as  $h \rightarrow 0$ .

# Convergence rate

Detailed analysis of convergence rate by [Jarlebring/Michiels'2011]:

$$\text{asympt convergence rate} = |\sigma - \lambda_1| \rho(B_1) + o(|\sigma - \lambda_1|)$$

with

$$B_0 = I - \frac{T'(\lambda_1)x_1x_1^*}{x_1^*T'(\lambda_1)x_1}$$

$$B_1 = B_0 \left( -T'(\lambda_1)T(\lambda_1)^+B_0 - \frac{T''(\lambda_1)x_1x_1^*}{x_1^*T'(\lambda_1)x_1} \right) \left( I + \frac{T'(\lambda_1)x_1x_1^*}{x_1^*T'(\lambda_1)x_1} \right)^{-1}$$

Inexact variant of RESINVIT analyzed by [Szyld/Xue'2013].

# PINVIT (Preconditioned inverse iteration)

for  $j = 0, 1, 2, \dots$

$$v_{j+1} = v_j + P^{-1}((v_j^* A v_j I - A)v_j)$$

Normalize  $v_{j+1}$ .

end for

Elegant convergence analyses by [D'yakonov/Orekhov'1980], [Knyazev'1998], [Neymeyr'2001], [Knyazev/Neymeyr'2003], ...:

- ▶ Setting: **symmetric positive definite  $A$**  and aim at smallest eigenvalue  $\lambda_1$ .
- ▶ Neymeyr's analysis relies on mini-dimensional analysis for Rayleigh quotient.
- ▶ Allow to conclude mesh-independent convergence rates.

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- ▶ Neymeyr's analysis relies on mini-dimensional analysis for Rayleigh quotient.
- ▶ Allow to conclude mesh-independent convergence rates.

**Difficult to extend this analysis to  
general nonlinear eigenvalue problems.**

Instead: Extend poor men's approach by interpreting PINVIT as perturbed inverse iteration.

# Rayleigh functionals

Back to nonlinear Hermitian  $T(\lambda)$ .

Assume that scalar nonlinear equation

$$x^* T(\lambda) x = 0$$

admits unique solution  $\lambda \in \Omega$  for every  $x$  in open set  $D \in \mathbb{C}^n$ .

Defines **Rayleigh functional**

$$\rho : D \rightarrow \Omega, \quad x \mapsto \lambda.$$

Additional assumption:

$$x^* T'(\rho(x)) x > 0 \quad \forall x \in D.$$

- ▶ Allows to carry over variational principles [Haderer'1968], [Voss/Werner'1982], ....
- ▶ For example:

$$\lambda_1 = \inf_{x \in D} \rho(x).$$

$\rightsquigarrow -T(\sigma)$  is spd (on  $D$ ) for any  $\sigma < \lambda_1$ .

# Rayleigh functional for model problem

$$T(\lambda) = \lambda M - K_0 - \phi(\lambda) e_n e_n^T.$$

with spd  $M$ ,  $K_0$  and  $\phi(\lambda) = \frac{\lambda \eta m}{\lambda - \eta}$ .

- ▶  $x^* T(\lambda) x = 0$  has exactly one solution in  $] \eta, \infty ]$  for every  $x \in \mathbb{C}^n \setminus \{0\}$ .

↪  $\exists$  Rayleigh functional for  $\Omega = ] \eta, \infty ]$  and  $D = \mathbb{C}^n \setminus \{0\}$ .

# One step of residual inverse iteration

RESINVIT with Rayleigh functional  $\rho$ :

$$v^+ = v + P^{-1} T(\rho(v)) v$$

for some spd preconditioner  $P$ .

Ignore scaling: will work with angles.

What is the right geometry?

# Geometry induced by $P$

RESINVIT with Rayleigh functional  $\rho$ :

$$v^+ = v + P^{-1} T(\rho(v)) v$$

for some spd preconditioner  $P$ .

- ▶ Geometry induced by  $P$ :

$$\langle v, w \rangle_P := v^* P w,$$

and induced norm  $\|v\|_P := \sqrt{v^* P v}$ .

- ▶ Angle  $\phi_P(v, x_1)$  between  $v$  and  $x_1$ :

$$\cos \phi_P(v, x_1) := \frac{\operatorname{Re} \langle v, x_1 \rangle_P}{\|v\|_P \|x_1\|_P}$$

- ▶ Given  $P$ -orthogonal decomposition

$$v = v_1 + v_\perp, \quad v_1 \in \operatorname{span}\{x_1\}, \quad \langle v_1, v_\perp \rangle_P = 0,$$

$\rightsquigarrow$

$$\tan \phi_P(v, x_1) = \|v_\perp\|_P / \|v_1\|_P.$$



# Convergence result

Theorem [Effenberger/K.'2013].

$$\tan \phi_P(v^+, x_1) \leq \gamma \cdot \tan \phi_P(v, x_1) + O(\varepsilon^2)$$

with  $\varepsilon = \tan \phi_P(v, x_1)$  and convergence factor

$$\gamma = \max_{\substack{z \neq 0 \\ \langle z, x_1 \rangle_P = 0}} \left| 1 + \frac{z^* T(\lambda_1) z}{z^* P z} \right|.$$

Proof based on expansion  $T(\rho(v)) = T(\lambda_1) + O(\varepsilon^2)$ .

- ▶ Linear case  $T(\lambda) = \lambda I - A$  with exact preconditioner  $P = A \rightsquigarrow$

$$\gamma = 1 + \frac{\lambda_1 - \lambda_2}{\lambda_2} = \frac{\lambda_1}{\lambda_2}$$

- ▶ Sufficient condition for convergence:  $\exists 0 \leq \tilde{\gamma} < 1$  such that

$$(1 - \tilde{\gamma}) z^* P z \leq -z^* T(\lambda_1) z \leq (1 + \tilde{\gamma}) z^* P z \quad \forall z \perp_P x_1.$$

Satisfied, e.g., by multigrid for  $-T(\sigma)$  with  
 $\sigma$  sufficiently close to  $\lambda_1$ .

# Summary

RESINVIT for Hermitian  $T(\lambda)$  with preconditioner  $P \approx -T(\sigma)$  has linear convergence rate

$$\gamma = \max_{\substack{z \neq 0 \\ \langle z, x_1 \rangle_P = 0}} \left| 1 + \frac{z^* T(\lambda_1) z}{z^* P z} \right|.$$

- ▶ For  $P = -T(\sigma)$ :  $\gamma < 1$  if  $\lambda_1$  simple eigenvalue and  $\sigma \approx \lambda_1$
- ▶ **Aim:**  
Show that  $\gamma$  stays away from 1 as  $h \rightarrow 0$  for discretized PDE.

# Function space setting

- ▶ Hilbert spaces  $V$  and  $H$  with dense and compact embedding  $V \hookrightarrow H$  (think of  $H^1$  and  $L^2$ )
- ▶ Variational formulation of NLEVP:

$$\text{Find } (u, \lambda) \in V \times \Omega \text{ such that} \\ a(u, v, \lambda) = 0 \quad \forall v \in V.$$

with bounded bilinear form  $a(\cdot, \cdot, \lambda)$  depending holomorphically on  $\lambda$ .

- ▶ Fix  $\lambda$  and consider linear eigenvalue problem associated with  $\bar{a}(\cdot, \cdot, \lambda) := -a(\cdot, \cdot, \lambda)$ :

$$\text{Find } (u, \mu) \in V \times \mathcal{D} \text{ such that} \\ \bar{a}(u, v, \lambda) = \mu \langle u, v \rangle_H \quad \forall v \in V.$$

Assume

$$\exists c, s \quad \text{such that} \quad \bar{a}(v, v, \lambda) \geq c \|v\|_V^2 - s \|v\|_H^2 \quad \forall v \in V.$$

- ▶ Modification of setting considered by [Voss'2013]:  
Existence of Rayleigh functional  $\rightsquigarrow$  variational principles for eigenvalues.

# Function space setting for model problem

For model problem  $V = H^1(]0, 1[)$ ,  $H = L^2(]0, 1[)$ , and

$$a(u, v, \lambda) = \int_{\Omega} \nabla u \cdot \nabla v \, dx + \varphi(\lambda) u(1) v(1) - \lambda \int_{\Omega} uv \, dx.$$

$\rightsquigarrow$  all assumptions are satisfied.

## Further assumptions

- ▶  $0 = \mu_1(\lambda_1) < \mu_2(\lambda_1) \leq \mu_3(\lambda_1) \leq \dots$
- ▶  $a$  bounded on  $V \times V$
- ▶  $\frac{\partial}{\partial \lambda} a$  bounded on  $H \times H$
- ▶  $\frac{\partial}{\partial \lambda} a$  Lipschitz continuous wrt  $\lambda$  in neighbourhood of  $\lambda_1$

# Attaining mesh-independent convergence rate

Outline of proof:

1. For eigenpair  $(\lambda_1, u_1)$  choose  $\sigma < \lambda_1$  sufficiently close to  $\lambda_1$   
 $\rightsquigarrow$  (shifted)  $\bar{a}(\cdot, \cdot, \sigma)$  is elliptic on  $V$ .
2. Assumptions imply existence of  $0 \leq \delta < 1$  such that

$$(1 - \delta)\bar{a}(v, v, \sigma) \leq \bar{a}(v, v, \lambda_1) \leq (1 + \delta)\bar{a}(v, v, \sigma), \quad (1)$$

holds for all  $v \neq 0$  with  $a(u_1, v, \sigma) = 0$ .

3. Show that (1) holds asymptotically ( $h \rightarrow 0$ ) for

$$(1 - \delta)\bar{a}_h(v, v, \sigma) \leq \bar{a}_h(v, v, \lambda_{1,h}) \leq (1 + \delta)\bar{a}_h(v, v, \sigma), \quad (2)$$

with  $a_h(\sigma)$ ,  $a_h(\lambda_{1,h})$  obtained by Galerkin projection on  $V_h \subset V$ .

**Main problem:**  $\lambda_{1,h}$  is the eigenvalue of the discrete problem!

4. Show that (1) remains true when  $a_h(\sigma)$  is replaced by spectrally equivalent preconditioner  $P$ .

Establishes  $h$ -independent convergence rate  $\delta$ .

# Conclusions

- ▶ Proposed and analyzed preconditioned variant of RESINVIT for nonlinear eigenvalue problems admitting a Rayleigh functional.
- ▶ Surrogate for understanding convergence of more advanced algorithms, e.g., nonlinear Arnoldi.
- ▶ Analysis requires preconditioner of  $T(\sigma)$  with  $\sigma$  sufficiently close to  $\lambda_1$ . In practice: Not a severe limitation.
- ▶ Ongoing work (with A. Miedlar): Adaptive FEM.

## References:

C. Effenberger, D. K. On the convergence of the residual inverse iteration for nonlinear eigenvalue problems admitting a Rayleigh functional. Preprint available from <http://anchp.epfl.ch>.

C. Effenberger. Robust solution methods for nonlinear eigenvalue problems. PhD thesis, EPFL, 2013.