# On efficient numerical approximation of the bilinear form $c^*\mathbf{A}^{-1}b$

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joint work with

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## Formulation of the problem

Given a nonsingular matrix  ${\bf A}$  and vectors b and c.

We want to approximate

$$c^* \mathbf{A}^{-1} b$$
.

Equivalently, we look for an approximation to

$$c^*x$$
 such that  $\mathbf{A}x = b$ .

#### Motivation

- Approximation of the jth component of the solution
  - $\bullet$  i.e., we want to approximate  $e_j^T \mathbf{A}^{-1} b.$
- Signal processing (the scattering amplitude)
  - b and c represent incoming and outgoing waves, respectively, and the operator A relates the incoming and scattered fields on the surface of an object,
  - $\mathbf{A}x = b$  determines the field x from the signal b. The signal is received on an antenna c. The signal received by the antenna is then  $c^*x$ . The value  $c^*x$  is called the scattering amplitude.
- Optimization
- Nuclear physics, quantum mechanics, other disciplines

# Krylov subspace methods approach

Projection of the original problem onto Krylov subspaces

$$\mathcal{K}_n(\mathbf{A}, b) = \operatorname{span}\{b, \mathbf{A}b, \dots \mathbf{A}^{n-1}b\}.$$

A possible approach: compute  $x_n$  using a Krylov subspace method,

$$c^* \mathbf{A}^{-1} b = c^* x \approx c^* x_n .$$

- Due to finite precision arithmetic, the explicit numerical computation of  $c^*x_n$  can be *highly inefficient*. [HPD case: Strakoš & T. '02, '05]
- If  ${\bf A}$  is HPD and c=b, there are several efficient methods. [Golub & Meurant '94, '97, Axelsson & Kaporin '01, Strakoš & T. '02, '05]
- How to generalize ideas from the HPD case to a general case?

### Outline

Vorobyev moment problem

2 Approximation of the bilinear form  $c^*\mathbf{A}^{-1}b$ 

3 Numerical experiments

# Vorobyev moment problem, Vorobyev '58, '65

Popularized by Brezinski '97, Strakoš '08

Find a linear operator  $\mathbf{A}_n$  on  $\mathcal{K}_n(\mathbf{A},v)$  such that

$$\mathbf{A}_{n} v = \mathbf{A} v,$$

$$\mathbf{A}_{n}^{2} v = \mathbf{A}^{2} v,$$

$$\vdots$$

$$\mathbf{A}_{n}^{n-1} v = \mathbf{A}^{n-1} v,$$

$$\mathbf{A}_{n}^{n} v = \mathbf{Q}_{n} \mathbf{A}^{n} v,$$

where  $\mathbf{Q}_n$  is a given linear projection operator.

- Some Krylov subspace methods can be identified with the Vorobyev moment problem.
- Useful formulation for understanding approximation properties of Krylov subspace methods.

# Non-Hermitian Lanczos and Arnoldi algorithms

Given a nonsingular A, v and w.

Non-Hermitian Lanczos algorithm is represented by

$$\mathbf{A}\mathbf{V}_{n} = \mathbf{V}_{n}\mathbf{T}_{n} + \delta_{n+1}v_{n+1}e_{n}^{T},$$
  
$$\mathbf{A}^{*}\mathbf{W}_{n} = \mathbf{W}_{n}\mathbf{T}_{n}^{*} + \eta_{n+1}^{*}w_{n+1}e_{n}^{T},$$

where  $\mathbf{W}_n^* \mathbf{V}_n = \mathbf{I}$  and  $\mathbf{T}_n = \mathbf{W}_n^* \mathbf{A} \mathbf{V}_n$  is tridiagonal.

Arnoldi algorithm is represented by

$$\mathbf{AV}_n = \mathbf{V}_n \mathbf{H}_n + h_{n+1,n} v_{n+1} e_n^T,$$

where  $\mathbf{V}_n^*\mathbf{V}_n=\mathbf{I}$ , and  $\mathbf{H}_n=\mathbf{V}_n^*\mathbf{A}\mathbf{V}_n$  is upper Hessenberg.

# Vorobyev moment problem,

Non-Hermitian Lanczos and Arnoldi

**Lanczos**:  $\mathbf{Q}_n$  projects onto  $\mathcal{K}_n(\mathbf{A},v)$  orthog. to  $\mathcal{K}_n(\mathbf{A}^*,w)$ ,

$$\mathbf{Q}_n = \mathbf{V}_n \mathbf{W}_n^*, \qquad \mathbf{A}_n = \mathbf{V}_n \mathbf{T}_n \mathbf{W}_n^*.$$

**Arnoldi**:  $\mathbf{Q}_n$  projects onto  $\mathcal{K}_n(\mathbf{A},v)$  orthog. to  $\mathcal{K}_n(\mathbf{A},v)$ ,

$$\mathbf{Q}_n = \mathbf{V}_n \mathbf{V}_n^*, \qquad \mathbf{A}_n = \mathbf{V}_n \mathbf{H}_n \mathbf{V}_n^*.$$

#### Matching moments property:

$$w^* \mathbf{A}^k v = w^* \mathbf{A}_n^k v$$
,

 $k = 0, \dots, \frac{2n-1}{n}$  for Lanczos,  $k = 0, \dots, \frac{n-1}{n}$  for Arnoldi.

[Gragg & Lindquist '83, Villemagne & Skelton '87]

[Gallivan & Grimme & Van Dooren '94, Antoulas '05]

[a simple proof using the Vorobyev moment problem - Strakoš '08]

# Approximation of $c^*\mathbf{A}^{-1}b$

General framework, Strakoš & T. '10

Vorobyev moment problem:  $\mathbf{A} \rightarrow \mathbf{A}_n$ 

Define approximation:  $c^*\mathbf{A}^{-1}b \approx c^*\mathbf{A}_n^{-1}b$ 

 $\mathbf{A}_n^{-1}$  is the matrix representation of the inverse of the reduced order operator  $\mathbf{A}_n$  which is restricted onto  $\mathcal{K}_n(\mathbf{A},b)$ .

#### **Examples**:

- ullet  ${f A}_n^{-1} = {f V}_n {f T}_n^{-1} {f W}_n^*$  (Non-Hermitian Lanczos)
- $\bullet$   $\mathbf{A}_n^{-1} = \mathbf{V}_n \mathbf{H}_n^{-1} \mathbf{V}_n^*$  (Arnoldi)

#### Questions:

- How to compute  $c^* \mathbf{A}_n^{-1} b$  efficiently?
- Relationship to the existing approximations?

We concentrate only to non-Hermitian Lanczos approach.

# Non-Hermitian Lanczos approach

Define

$$v_1 = \frac{b}{\|b\|}$$
,  $w_1 = \frac{c}{c^*v_1}$ , i.e.  $w_1^*v_1 = 1$ .

Then

$$c^* \mathbf{A}_n^{-1} b = c^* \mathbf{V}_n \mathbf{T}_n^{-1} \mathbf{W}_n^* b = (c^* v_1) \|b\| (\mathbf{T}_n^{-1})_{1,1}.$$

Let  $x_0=0$ . We also know that  $x_n=\|b\|\mathbf{V}_n\mathbf{T}_n^{-1}e_1$  is the approximate solution computed via BiCG. Therefore,

$$c^* \mathbf{A}_n^{-1} b = c^* ||b|| \mathbf{V}_n \mathbf{T}_n^{-1} \mathbf{W}_n^* \mathbf{V}_n e_1 = c^* x_n.$$

- BiCG can be used for computing  $c^* \mathbf{A}_n^{-1} b!$
- We used the global biorthogonality!
   Do the identities hold in finite precision computations?

#### The BiCG method

Simultaneous solving of

$$\mathbf{A}x = b, \qquad \mathbf{A}^*y = c.$$

input 
$${\bf A}$$
,  $b$ ,  $c$ 

$$x_0 = y_0 = 0$$
  
 $r_0 = p_0 = b$ ,  $s_0 = q_0 = c$ 

for 
$$n = 0, 1, ...$$

$$\begin{split} & \alpha_n = \frac{s_n^* r_n}{q_n^* \mathbf{A} p_n} \,, \\ & x_{n+1} = x_n + \alpha_n \, p_n \,, \qquad y_{n+1} = y_n + \alpha_n^* \, q_n \,, \\ & r_{n+1} = r_n - \alpha_n \, \mathbf{A} p_n \,, \qquad s_{n+1} = s_n - \alpha_n^* \, \mathbf{A}^* q_n \,, \\ & \beta_{n+1} = \frac{s_{n+1}^* r_{n+1}}{s_n^* r_n} \,, \\ & p_{n+1} = r_{n+1} + \beta_{n+1} \, p_n \,, \qquad q_{n+1} = s_{n+1} + \beta_{n+1}^* \, q_n \end{split}$$

end

# An efficient approximation based on the BiCG method

How to compute  $c^*\mathbf{A}_n^{-1}b$  in BiCG without using the global biorthogonality?

Using local biorthogonality we can show that

$$s_j^* \mathbf{A}^{-1} r_j - s_{j+1}^* \mathbf{A}^{-1} r_{j+1} = \alpha_j s_j^* r_j.$$

Consequently,

$$c^* \mathbf{A}^{-1} b = \sum_{j=0}^{n-1} \alpha_j s_j^* r_j + s_n^* \mathbf{A}^{-1} r_n.$$

Moreover, it can be shown that (using global biorthogonality) that

$$c^* \mathbf{A}^{-1} b = c^* x_n + s_n^* \mathbf{A}^{-1} r_n.$$

Finally,

$$c^* \mathbf{A}_n^{-1} b = c^* x_n = \sum_{j=0}^{n-1} \alpha_j s_j^* r_j.$$

# Approximations based on the BiCG method and finite precision arithmetic

It holds that

$$c^* \mathbf{A}^{-1} b \; = \; \sum_{j=0}^{n-1} \alpha_j s_j^* r_j \; \; + \; \underbrace{s_n^* \mathbf{A}^{-1} r_n}_{\text{error}} \; \cdot \; \underbrace{s_n^* \mathbf{A}^{-1} r_n}_{\sim \; \|y - y_n\| \| \|r_n\|} \; .$$

It can be shown that

$$c^* \mathbf{A}^{-1} b = c^* x_n + \underbrace{y_n^* r_n + s_n^* \mathbf{A}^{-1} r_n}_{\text{error} \sim \|y_n\| \|r_n\|}.$$

In exact arithmetic  $y_n^* r_n = 0$ .

If the global biorthogonality is lost, one can expect that

$$|y_n^*r_n| \sim ||y_n|| ||r_n||.$$

# Yet another approach

#### Hybrid BiCG methods

We know that

$$c^* \mathbf{A}_n^{-1} b \; = \; \sum_{j=0}^{n-1} \frac{\alpha_j}{s_j^* r_j} \quad \text{ and } \quad s_j^* r_j \; = \; (c^* b) \prod_{k=0}^{j-1} \frac{\beta_k}{\beta_k} \, .$$

In hybrid BiCG methods like CGS, BiCGStab, BiCGStab( $\ell$ ), the BiCG coefficients are available, i.e. we can compute the approximation  $e^*\mathbf{A}_n^{-1}b$  during the run of these method.

**Question**: Hybrid BiCG methods produce approximations  $\mathbf{x}_n$ , better than  $x_n$  produced by BiCG.

Is  $c^*\mathbf{x}_n$  a better approximation of  $c^*\mathbf{A}^{-1}b$  than  $c^*x_n$ ?

No. We showed that mathematically [Strakoš & T. '10],

$$c^*\mathbf{x}_n = c^*x_n.$$

# Summary (non-Hermitian Lanczos approach)

How to compute  $c^* \mathbf{A}_n^{-1} b$ ?

#### Algorithm of choice:

- non-Hermitian Lanczos
- BiCG
- hybrid BiCG methods

#### Way of computing the approximation:

- $\bullet$   $c^*x_n$
- $(c^*v_1) \|b\| (\mathbf{T}_n^{-1})_{1,1}$
- complex Gauss quadrature (Saylor-Smolarski approach)
- from the BiCG coefficients, or, in BiCG using

$$\xi_n^B \equiv \sum_{j=0}^{n-1} \alpha_j \, s_j^* r_j \,.$$

# Numerical experiments

Diffraction of light on periodic structures, RCWA method

[Hench & Strakoš '08]

$$\mathbf{A} \, x \equiv \left[ \begin{array}{cccc} -\mathbf{I} & \mathbf{I} & e^{\mathbf{i}\sqrt{\mathbf{C}}\varrho} & 0 \\ \mathbf{Y}_I & \sqrt{\mathbf{C}} & -\sqrt{\mathbf{C}}e^{\mathbf{i}\sqrt{\mathbf{C}}\varrho} & 0 \\ 0 & e^{\mathbf{i}\sqrt{\mathbf{C}}\varrho} & I & -\mathbf{I} \\ 0 & \sqrt{\mathbf{C}}e^{\mathbf{i}\sqrt{\mathbf{C}}\varrho} & -\sqrt{\mathbf{C}} & -\mathbf{Y}_{\mathrm{II}} \end{array} \right] \, x \, = \, b \, ,$$

 $\mathbf{Y}_{\mathrm{I}},\ \mathbf{Y}_{\mathrm{II}},\mathbf{C}\in\mathbb{C}^{(2M+1)\times(2M+1)}$ ,  $\varrho>0$ , M is the discretization parameter representing the number of Fourier modes used for approximation of the electric and magnetic fields as well as the material properties.

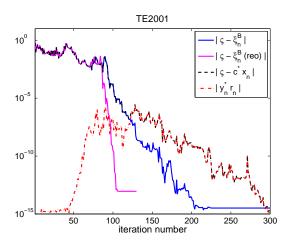
Typically, one needs only the dominant (M+1)st component

$$e_{M+1}^* \mathbf{A}^{-1} b.$$

In our experiments M=20, i.e.  $\mathbf{A}\in\mathbb{C}^{164\times164}$ . [Strakoš & T. '10]

## Approximations based on the BiCG method

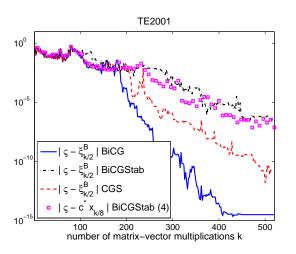
Mathematically equivalent approximations  $\xi_n^B$  and  $c^*x_n$ ,  $\varsigma \equiv c^*\mathbf{A}^{-1}b$ 



$$|c^* \mathbf{A}^{-1} b - c^* x_n| = |y_n^* r_n + s_n^* \mathbf{A}^{-1} r_n|,$$
  
 $|c^* \mathbf{A}^{-1} b - \xi_n^B| = |s_n^* \mathbf{A}^{-1} r_n|.$ 

# Non-Hermitian Lanczos approach

Mathematically equivalent approximations based on hybrid BiCG methods



The BiCGStab and CGS approximations are significantly more affected by rounding errors than the BiCG approximations.

#### Conclusions

- Generalization of the HPD case:
  - ullet Via Vorobyev moment problem o very natural and general.
    - no assumptions on  $\mathbf{A}$ , based on approximation properties
  - Complex Gauss Quadrature approach
    - A has to be diagonalizable, just a formalism
- We proved mathematical equivalence of the existing approximations based on Non-Hermitian Lanczos.
- Preferable approximation

$$\xi_n^B \equiv \sum_{j=0}^{n-1} \alpha_j \, s_j^* r_j \,.$$

It is simple and numerically better justified.

In finite precision arithmetic, the identities need not hold.
 A justification is necessary (e.g. local biorthogonality).

## Related papers

- Z. Strakoš and P. Tichý, [On efficient numerical approximation of the bilinear form  $c^*\mathbf{A}^{-1}b$ , submitted to SISC, 2009].
- G. H. Golub, M. Stoll, and A. Wathen, [Approximation of the scattering amplitude and linear systems, Electron. Trans. Numer. Anal., 31 (2008), pp. 178–203].
- Z. Strakoš and P. Tichý, [On error estimation in the conjugate gradient method and why it works in finite precision computations, Electron. Trans. Numer. Anal., 13 (2002), pp. 56–80].
- P. E. Saylor and D. C. Smolarski, [Why Gaussian quadrature in the complex plane?, Numer. Algorithms, 26 (2001), pp. 251–280].
- G. H. Golub and G. Meurant, [Matrices, moments and quadrature, in Numerical analysis 1993 (Dundee, 1993), vol. 303 of Pitman Res. Notes Math. Ser., Longman Sci. Tech., Harlow, 1994, pp. 105–156].

#### More details

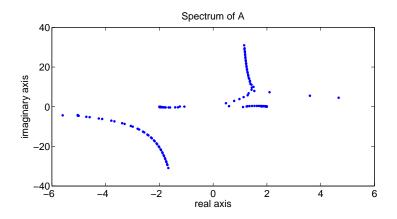
#### More details can be found at

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http://www.cs.cas.cz/strakos
http://www.cs.cas.cz/tichy
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Thank you for your attention!

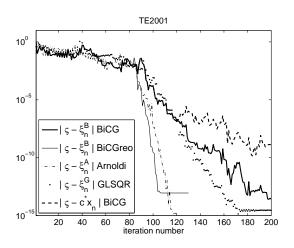
#### The matrix A

#### Spectrum of ${\bf A}$ computed via the Matlab command ${\tt eig}$



Some eigenvalues have large imaginary parts in comparison to the real parts,  $\kappa(\mathbf{A}) \approx 104$ .

## Non-Hermitian Lanczos, Arnoldi, GLSQR



GLSQR: [Golub & Stoll & Wathen '08], [Saunders & Simon & Yip '88]

# Different approaches with preconditioning

Non-Hermitian Lanczos, Arnoldi, GLSQR

