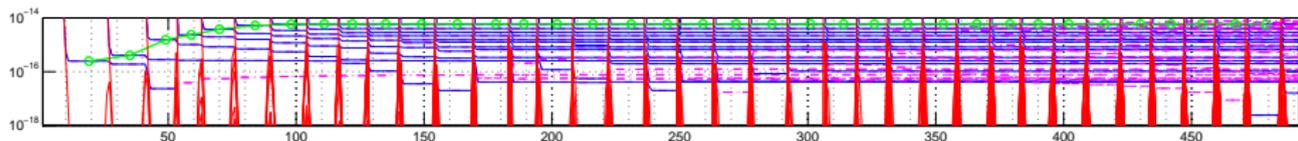


# Towards a deeper understanding of Chris Paige's error analysis of the finite precision Lanczos process

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# Outline

## Some history

### Hessenberg matrices

- Hessenberg decompositions

- Hessenberg eigenvectors

### Chris Paige's approach

- On the length of the Ritz vectors

- Eigenvector sensitivity

- Closer to the original

### Our approach

- The shifted decomposition

- About higher derivatives

- The polynomial point of view

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Thus far, this is maybe the **only** successful error analysis ever carried out for a perturbed short-term Krylov subspace method.

# An example: Lanczos' method

We used the diagonal matrix

$$\mathbf{A} = \text{diag}([\text{linspace}(0, 1, 50), 3])$$

and the starting vector

$$\mathbf{e} = \text{ones}(51, 1)$$

in an implementation of Lanczos' method in MATLAB on a PC conforming to ANSI/IEEE 754 with machine precision  $\text{eps}(1) = 2^{-52} \approx 2.2204 \cdot 10^{-16}$ .

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Eigenvalues and eigenvectors are computed using **MRRR**, i.e., LAPACK's routine `DSTEGR`, since MATLAB's `eig` (using LAPACK's `DSYEV`, i.e., the QR algorithm implemented as `DSTEQR`) fails in delivering accurate **eigenvectors**.

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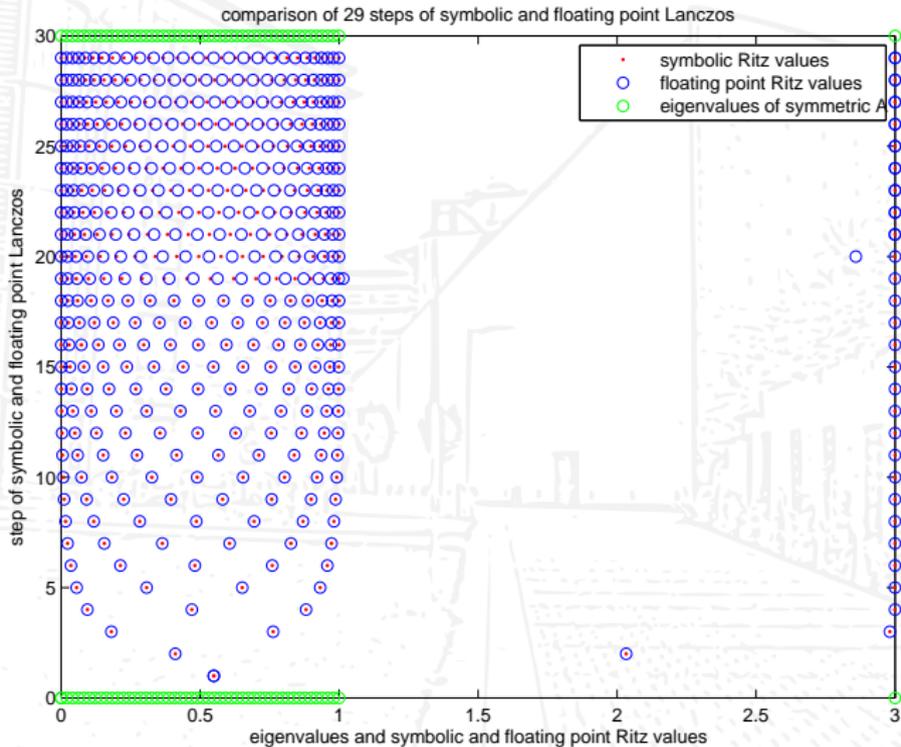
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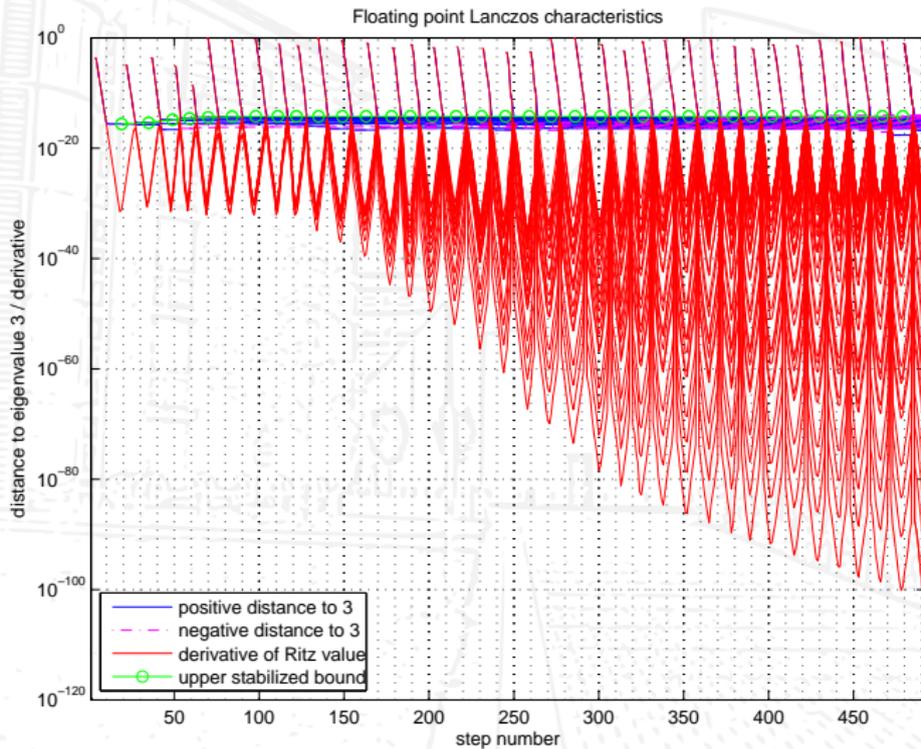
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# Hessenberg decompositions

Essential features of Krylov subspace methods can be described by a **Hessenberg decomposition**

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The matrix  $\mathbf{H}_k$  of the perturbed variant will, in general, still be unreduced.

# Hessenberg decompositions

In (Z, 2007) we did consider in an abstract manner the matrix equation

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and came up with polynomial expressions in  $\mathbf{A}$  for

- ▶ the basis vectors  $\mathbf{q}_j$ ,
- ▶ the Ritz vectors  $\mathbf{y}_j := \mathbf{Q}_k\mathbf{s}_j$ , where  $\mathbf{s}_j$  is an eigenvector of  $\mathbf{H}_k$  to the eigenvalue  $\theta_j$ ,
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This talk: Application to the (symmetric) Lanczos process (in finite precision);  
 Aim: generalize (Paige, 1971; Paige, 1972; Paige, 1976; Paige, 1980) to the  
 general (non-symmetric) Lanczos process (with general perturbations).

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In case of the symmetric Lanczos process we have

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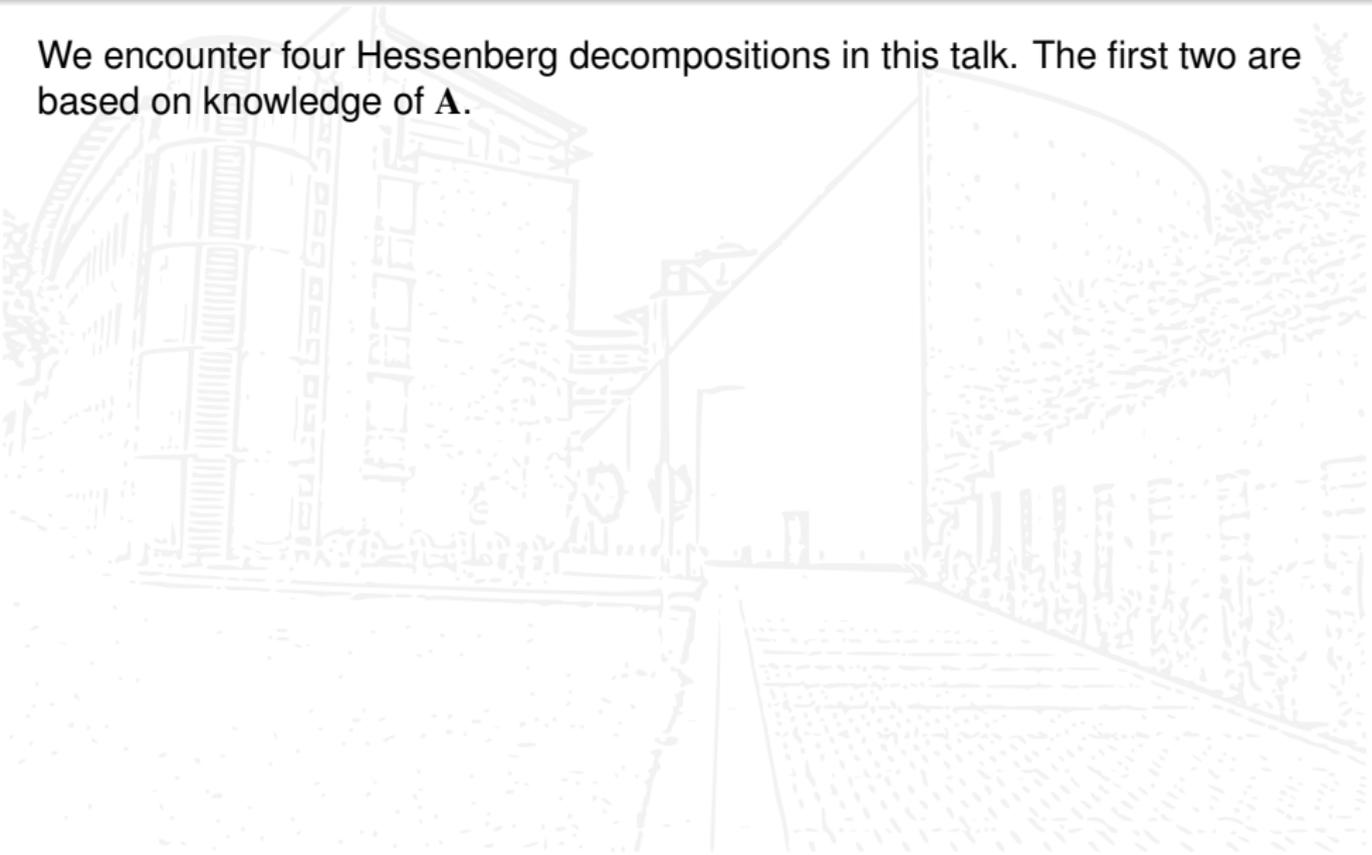
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(If off-diagonal elements were negative, impose diagonal scaling.)

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This Hessenberg decomposition is interesting especially in case that  $\lambda_i - \theta_j$  is “small”, i.e., “comparable” to  $\mathbf{F}_k$ .

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Let  $\mathbf{W}_{k+1} := \mathbf{Q}_k^H \mathbf{Q}_{k+1}$ , define  $\mathbf{G}_k := \mathbf{e}_k \beta_k \mathbf{q}_{k+1}^H \mathbf{Q}_k + \mathbf{Q}_k^H \mathbf{F}_k - \mathbf{F}_k^H \mathbf{Q}_k$ . Then

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with  $\mathbf{E}_k$  upper triangular and small if  $\mathbf{F}_k$  is small and local orthogonality is preserved.

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We have that  $(\tilde{\nu}(z))^T = \hat{\nu}(z)^H$

$$(z\mathbf{I}_k - \mathbf{T}_k)\boldsymbol{\nu}(z) = \mathbf{e}_1 \frac{\chi(z)}{\beta_{1:k-1}}, \quad \tilde{\nu}(z)^T(z\mathbf{I}_k - \mathbf{T}_k) = \frac{\chi(z)}{\beta_{1:k-1}} \mathbf{e}_n^T \quad (7)$$

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In (Z, 2006) we used differentiation and the above relations to construct eigenvectors and corresponding principal vectors.

# Hessenberg eigenvectors and eigenvector derivatives

In case of **Hermitean/symmetric** matrices  $\mathbf{A}$  and  $\mathbf{T}_k$  we know that the left and right eigenvector are parallel and can be scaled to unit length.

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where

$$\chi_{i:j}(z) := \det(z\mathbf{I}_{j-i+1} - \mathbf{T}_{i:j}) \quad \text{and} \quad \beta_{i:j} := \prod_{\ell=i}^j \beta_\ell, \quad 0 \leq i \leq j < k. \quad (10)$$

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To be more precise:  $\boldsymbol{\nu}_k(z) \equiv \mathbf{1} \equiv \check{\boldsymbol{\nu}}_1(z)$ .

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Unit length eigenvectors  $\mathbf{s}_j$  of  $\mathbf{T}_k$  to the eigenvalue  $\theta_j$  are **defined** by

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# Outline

## Some history

## Hessenberg matrices

- Hessenberg decompositions

- Hessenberg eigenvectors

## Chris Paige's approach

- On the length of the Ritz vectors

- Eigenvector sensitivity

- Closer to the original

## Our approach

- The shifted decomposition

- About higher derivatives

- The polynomial point of view

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We might guess that it is indeed a perturbation of the eigenvector that causes the deviation. But where to look for this perturbation? **Where do we find the underlying sensitivity analysis?**

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$$y_j^{(k)T} R_k y_j^{(k)} = - \sum_{t=1}^{k-1} \eta_{t+1, j}^{(k)} \sum_{r=1}^t \frac{\varepsilon_{rr}^{(t)}}{\beta_{t+1} \eta_{tr}^{(t)}} y_j^{(k)T} \begin{bmatrix} y_r^{(t)} \\ 0 \end{bmatrix} \quad (3.19)$$

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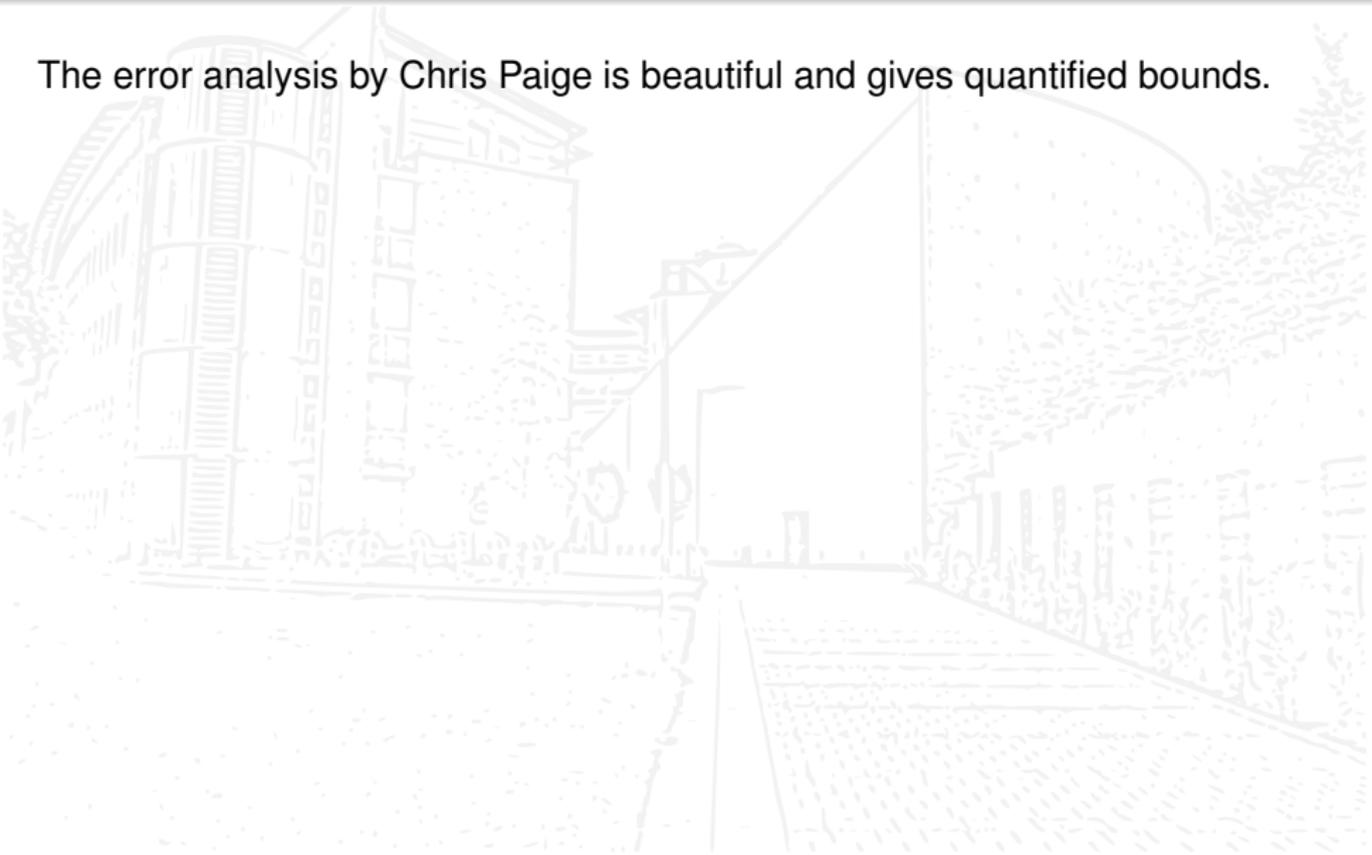
Caution: notational changes!

C. Paige: this talk:

$$\begin{aligned} z_j^{(k)} &\Leftrightarrow y_j \\ y_j^{(k)} &\Leftrightarrow s_j \\ \beta_{k+1} \eta_{kj}^{(k)} &\Leftrightarrow \beta_k s_{kj} \\ \mu_j^{(k)} &\Leftrightarrow \theta_j^{(k)} = \theta_j \end{aligned}$$

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$$\mathbf{T}_k \mathbf{W}_k + \mathbf{G}_k = \mathbf{W}_{k+1} \mathbf{T}_k = \mathbf{W}_k \mathbf{T}_k + \mathbf{w}_{k+1} \beta_k \mathbf{e}_k^T. \quad (\text{HessT1})$$

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Here, the basis vectors  $\mathbf{w}_j$  describe the loss of orthogonality and the perturbation term has a **large** rank-one part (i.e., large last row),

$$\begin{aligned} \mathbf{W}_{k+1} &:= \mathbf{Q}_k^H \mathbf{Q}_{k+1}, \\ \mathbf{G}_k &:= \mathbf{e}_k \beta_k \mathbf{q}_{k+1}^H \mathbf{Q}_k + \mathbf{Q}_k^H \mathbf{F}_k - \mathbf{F}_k^H \mathbf{Q}_k. \end{aligned} \quad (16)$$

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The derivation of (HessT1) is really simple: Multiplication of (HessA1),

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and (17)–(17)<sup>H</sup> gives

$$\mathbf{T}_k\mathbf{W}_k + \mathbf{G}_k = \mathbf{W}_{k+1}\underline{\mathbf{T}}_k = \mathbf{W}_k\mathbf{T}_k + \mathbf{w}_{k+1}\beta_k\mathbf{e}_k^T \quad (\text{HessT1})$$

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$$\mathbf{G}_k = \mathbf{e}_k\beta_k\mathbf{q}_{k+1}^H\mathbf{Q}_k + \mathbf{Q}_k^H\mathbf{F}_k - \mathbf{F}_k^H\mathbf{Q}_k, \quad (18)$$

since  $\mathbf{A} = \mathbf{A}^H$  and  $\mathbf{T}_k = \mathbf{T}_k^T$  are self-adjoint.

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We can use the results of (Z, 2007) on the angles between eigenvectors and Ritz vectors to obtain the following formula:

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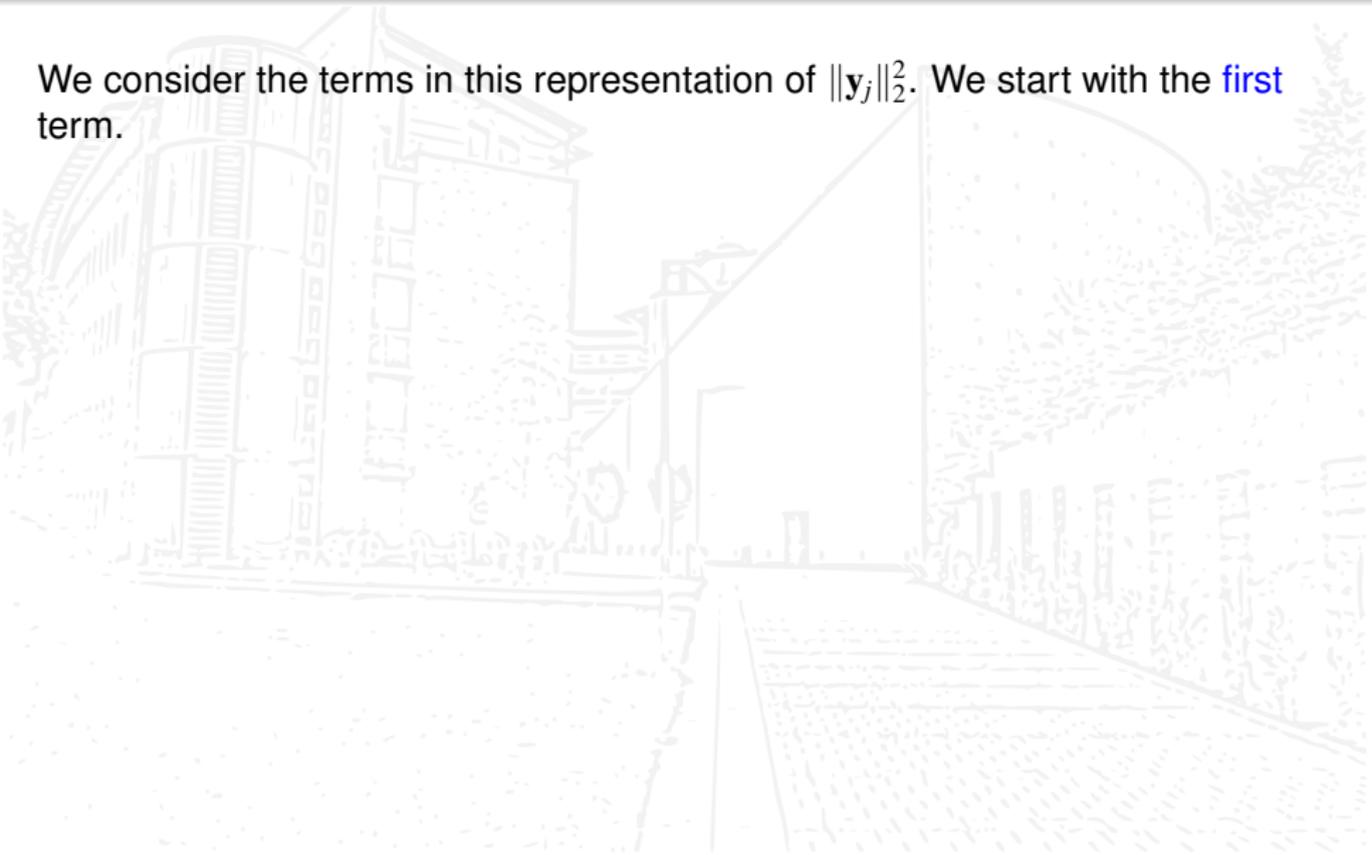
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 &= \hat{\mathbf{v}}(\theta_j)^H \mathbf{Q}_k^H \mathbf{q}_1 + \sum_{\ell=1}^k \frac{\beta_{1:k-1}}{\omega(\theta_j)} \nu'_\ell(\theta_j) \hat{\mathbf{v}}(\theta_j)^H \mathbf{g}_\ell \\
 &= \hat{\mathbf{v}}(\theta_j)^H \mathbf{Q}_k^H \mathbf{q}_1 + \frac{\mathbf{v}(\theta_j)^H \mathbf{G}_k \mathbf{v}'(\theta_j)}{\mathbf{v}(\theta_j)^H \mathbf{v}(\theta_j)} \\
 &= \hat{\mathbf{v}}(\theta_j)^H \mathbf{Q}_k^H \mathbf{q}_1 + \frac{\beta_k \mathbf{q}_{k+1}^H \mathbf{Q}_k \mathbf{v}'(\theta_j)}{\mathbf{v}(\theta_j)^H \mathbf{v}(\theta_j)} + \frac{\mathbf{v}(\theta_j)^H (\mathbf{Q}_k^H \mathbf{F}_k - \mathbf{F}_k^H \mathbf{Q}_k) \mathbf{v}'(\theta_j)}{\mathbf{v}(\theta_j)^H \mathbf{v}(\theta_j)}.
 \end{aligned} \tag{19}$$

Here,  $\omega(\theta_j) := \chi'(\theta_j)$  and  $\mathcal{A}_{\ell+1:k}(z, w) := \chi_{\ell+1:k}[z, w] = \beta_{\ell:k-1} \nu_\ell[z, w]$ .

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$$\hat{\nu}(\theta_j)^H \mathbf{Q}_k^H \mathbf{q}_1 = 1, \quad \text{since} \quad \hat{\nu}_1(z) \equiv 1. \quad (20)$$

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In the **perturbed case** the elements in the scalar product are given by

$$\hat{\mathbf{v}}(\theta_j)^H \mathbf{Q}_k^H \mathbf{q}_1 = \sum_{l=1}^k \frac{\chi_{1:l-1}(\theta_j)}{\beta_{1:l-1}} \mathbf{q}_l^H \mathbf{q}_1. \quad (21)$$

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The term should be of order one plus “small” times “sensitivity”, the ratio measures the “closeness” of older Ritz values to  $\theta_j$ . At “sensitive” steps we can have a large loss of orthogonality. It is **not known** how we should prove this assertion.

# Outline

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## Hessenberg matrices

- Hessenberg decompositions

- Hessenberg eigenvectors

## Chris Paige's approach

- On the length of the Ritz vectors

## Eigenvector sensitivity

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Both other terms in our expression for  $\|\mathbf{y}_j\|_2^2$  are of the form

$$\frac{\boldsymbol{\nu}(\theta_j)^H \mathbf{X}_k \boldsymbol{\nu}'(\theta_j)}{\boldsymbol{\nu}(\theta_j)^H \boldsymbol{\nu}(\theta_j)} = \frac{\hat{\boldsymbol{\nu}}(\theta_j)^H \mathbf{X}_k \boldsymbol{\nu}'(\theta_j)}{\hat{\boldsymbol{\nu}}(\theta_j)^H \boldsymbol{\nu}(\theta_j)}. \quad (22)$$

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For those not familiar with eigenvector perturbations:

$$|\sin \angle(\boldsymbol{\nu}(\theta_j + \Delta\theta_j), \boldsymbol{\nu}(\theta_j))| = \frac{\|\mathbf{P}_{\boldsymbol{\nu}(\theta_j)^\perp} \boldsymbol{\nu}(\theta_j + \Delta\theta_j)\|_2}{\|\boldsymbol{\nu}(\theta_j + \Delta\theta_j)\|_2} \quad (23)$$

measures the **sensitivity of the eigenvector** to structured perturbations affecting “only” the Ritz value.

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measures the **sensitivity of the eigenvector** to structured perturbations affecting “only” the Ritz value. The right eigenvector polynomial is not affected if we alter the elements in the **first row** of  $\mathbf{T}_k$ .

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Using Taylor expansion we obtain

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It turns out to be easy to obtain **analytic expressions** for

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Since  $\hat{\nu}(\theta_j)$  and  $\nu(\theta_j)$  are parallel, by the **Cauchy-Schwarz (in)equality**

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$$\begin{aligned} \frac{|\nu(\theta_i)^H \nu'(\theta_j)|}{\|\nu(\theta_i)\|_2 \|\nu(\theta_j)\|_2} &= \frac{\|\hat{\nu}(\theta_j)\|_2 |\hat{\nu}(\theta_i)^H \nu'(\theta_j)|}{\|\hat{\nu}(\theta_i)\|_2 |\hat{\nu}(\theta_j)^H \nu(\theta_j)|} \\ &= \begin{cases} \frac{\|\hat{\nu}(\theta_j)\|_2}{\|\hat{\nu}(\theta_i)\|_2} \frac{1}{|\theta_j - \theta_i|}, & j \neq i, \\ \left| \sum_{\ell \neq j} \frac{1}{\theta_j - \theta_\ell} \right|, & j = i. \end{cases} \end{aligned} \quad (28)$$

# Chris Paige's approach

Observe that the norms of the eigenvectors

$$\|\hat{\mathbf{v}}(\theta_j)\|_2^2 = \frac{1}{s_{1j}^2} \quad (29)$$

are related to the **squares of the first components** of the normalized eigenvectors, which are the **weights** in Gaussian quadrature.

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In general, we can make **use** of the **relations**

$$\begin{aligned} s_{kj}^2 &= \frac{\chi_{1:k-1}(\theta_j)}{\omega(\theta_j)} = \frac{1}{\|\boldsymbol{v}(\theta_j)\|_2^2}, \\ s_{1j}^2 &= \frac{\chi_{2:k}(\theta_j)}{\omega(\theta_j)} = \frac{1}{\|\hat{\boldsymbol{v}}(\theta_j)\|_2^2}, \end{aligned} \quad (30)$$

where the reduced polynomial  $\omega = \omega_j$  is defined as before by

$$\omega(z) = \prod_{\ell \neq j} (z - \theta_\ell). \quad (31)$$

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By classical perturbation theory

$$|\sin \angle(\hat{\boldsymbol{v}}(\theta_j), \boldsymbol{v}(\theta_j) + \boldsymbol{v}'(\theta_j)\Delta\theta_j)| \lesssim \frac{|\Delta\theta_j|}{\min_{\ell \neq j} |\theta_j - \theta_\ell|}. \quad (32)$$

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This is **not easy** to deduce here, we only have seen thus far that

$$\begin{aligned} \sin^2 \angle(\hat{\boldsymbol{\nu}}(\theta_j), \boldsymbol{\nu}(\theta_j) + \boldsymbol{\nu}'(\theta_j)\Delta\theta_j) &= \frac{\|\mathbf{P}_{\hat{\boldsymbol{\nu}}(\theta_j)} + \boldsymbol{\nu}'(\theta_j)\|_2^2}{\|\boldsymbol{\nu}(\theta_j)\|_2^2} |\Delta\theta_j|^2 + O(|\Delta\theta_j|^3) \\ &= \frac{|\Delta\theta_j|^2}{s_{1j}^2} \sum_{\ell \neq j} \frac{s_{1\ell}^2}{(\theta_j - \theta_\ell)^2} + O(|\Delta\theta_j|^3). \end{aligned} \quad (33)$$

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Maybe the **relations** collected on the following slides will provide helpful.

# Chris Paige's approach

A first **tool of trade** that works in the symmetric case is the identity

$$\beta_{1:k-1}^2 = \chi_{1:k-1}(\theta_j) \cdot \chi_{2:k}(\theta_j), \quad (34)$$

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A **relation without squares** follows easily using (Z, 2006), (Z, 2007) and Cauchy-Schwarz, we have

$$s_{1j}s_{kj} = \frac{\beta_{1:k-1}}{\omega(\theta_j)} = \frac{1}{\hat{\nu}(\theta_j)^H \nu(\theta_j)}. \quad (36)$$

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There can **not** be **two consecutive zeros** in an eigenvector of a tridiagonal matrix, as then the three-term recurrence would construct **only zeros**,

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Thus,  $|\omega(\theta_j)| = |\chi'(\theta_j)| > 2\beta_{1:k-1}$ .

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Inserting the identity matrix gives

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Again, we have to treat the **norms of the eigenvector polynomials** in some (not specified) manner to make this a successful approach.

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We only used the first Hessenberg decomposition with  $\mathbf{T}_k$ . We can stick closer to what Chris Paige did, and use the second one:

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Here,  $\mathbf{E}_k$  is upper triangular, and  $\mathbf{W}_{k+1} = \mathbf{R}_k^H + \mathbf{D}_k + \mathbf{R}_{k+1}$  with  $\mathbf{R}_{k+1} = \text{sut}(\mathbf{W}_{k+1})$  and  $\mathbf{D}_k$  diagonal.

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Based on the identity

$$\|\mathbf{y}_j\|_2^2 - 1 = \mathbf{s}_j^H (\mathbf{D}_k - \mathbf{I}_k) \mathbf{s}_j + 2 \text{Re} (\mathbf{s}_j^H \mathbf{R}_k \mathbf{s}_j) \quad (42)$$

Chris Paige bounded the deviation of  $\|\mathbf{y}_j\|$  from one.

# Chris Paige's approach

We can again use the characterization of the angles to compute his results in terms of the derivative,

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Thus,

$$\|\mathbf{y}_j\|_2^2 - 1 = \mathbf{s}_j^H (\mathbf{D}_k - \mathbf{I}_k) \mathbf{s}_j + 2\operatorname{Re} \left( \frac{\hat{\mathbf{v}}(\theta_j)^H \mathbf{E}_k \boldsymbol{\nu}'(\theta_j)}{\hat{\mathbf{v}}(\theta_j)^H \boldsymbol{\nu}(\theta_j)} \right). \quad (44)$$

# Chris Paige's approach

We can reformulate this by our “perturbation analysis”:

$$\frac{\hat{\boldsymbol{\nu}}(\theta_j)^H \mathbf{E}_k \boldsymbol{\nu}'(\theta_j)}{\hat{\boldsymbol{\nu}}(\theta_j)^H \boldsymbol{\nu}(\theta_j)} = \sum_{\ell \neq j} \frac{1}{\theta_j - \theta_i} \left( \frac{\hat{\boldsymbol{\nu}}(\theta_j)^H \mathbf{E}_k \boldsymbol{\nu}(\theta_i)}{\hat{\boldsymbol{\nu}}(\theta_i)^H \boldsymbol{\nu}(\theta_i)} + \frac{\hat{\boldsymbol{\nu}}(\theta_j)^H \mathbf{E}_k \boldsymbol{\nu}(\theta_j)}{\hat{\boldsymbol{\nu}}(\theta_j)^H \boldsymbol{\nu}(\theta_j)} \right). \quad (45)$$

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proves that **loss of orthogonality** and “**convergence**” go hand in hand,

$$\epsilon_{jj}^{(k)} = \mathbf{s}_j^H \mathbf{Q}_k^H \mathbf{q}_{k+1} \beta_k \mathbf{e}_k^T \mathbf{s}_j = \mathbf{y}_j^H \mathbf{q}_{k+1} \beta_k s_{kj}. \quad (47)$$

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Perhaps we need to better understand the **derivative of the eigenvector polynomial**. In (Z, 2006) it was proven that this vector is the first principal vector if the eigenvalue is multiple, which is never true in our setting.

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Perhaps we need to better understand the **derivative of the eigenvector polynomial**. In (Z, 2006) it was proven that this vector is the first principal vector if the eigenvalue is multiple, which is never true in our setting.

It turns out that the derivative of the eigenvector polynomial is in some sense obtained by **inverse iteration** with shifted  $\mathbf{A}$ . This can be seen with the aid of the shifted Hessenberg decomposition.

# Outline

## Some history

### Hessenberg matrices

- Hessenberg decompositions

- Hessenberg eigenvectors

### Chris Paige's approach

- On the length of the Ritz vectors

- Eigenvector sensitivity

- Closer to the original

### Our approach

- The shifted decomposition**

- About higher derivatives

- The polynomial point of view

# A new approach

Consider the **shifted Lanczos Hessenberg decomposition**

$$\tilde{\mathbf{A}}\mathbf{Q}_k + \tilde{\mathbf{F}}_k = \mathbf{Q}_{k+1}\mathbf{T}_k = \mathbf{Q}_k\mathbf{T}_k + \mathbf{q}_{k+1}\beta_k\mathbf{e}_k^T \quad (\text{HessA2})$$

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where for a given eigenpair  $\mathbf{A}\mathbf{v}_i = \mathbf{v}_i\lambda_i$  and a given Ritz value  $\theta_j$  we defined

$$\tilde{\mathbf{A}} := \mathbf{A} - (\lambda_i - \theta_j)\mathbf{v}_i\mathbf{v}_i^H \quad \text{and} \quad \tilde{\mathbf{F}}_k := (\lambda_i - \theta_j)\mathbf{v}_i\mathbf{v}_i^H\mathbf{Q}_k + \mathbf{F}_k. \quad (48)$$

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This definitions ensure that the Hessenberg decomposition **is still balanced** and that now

$$\mathbf{v}_i^H\tilde{\mathbf{A}} = \mathbf{v}_i^H(\mathbf{A} - (\lambda_i - \theta_j)\mathbf{v}_i\mathbf{v}_i^H) = \lambda_i\mathbf{v}_i^H - (\lambda_i - \theta_j)\mathbf{v}_i^H\mathbf{v}_i\mathbf{v}_i^H = \theta_j\mathbf{v}_i^H, \quad (49)$$

i.e.,  $\mathbf{v}_i$  is a **left eigenvector to the eigenvalue  $\theta_j$** .

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The angle between the eigenvector  $\mathbf{v}_i$  and a scaled Ritz vector is given by

$$\frac{\beta_{1:k-1}}{\omega(\theta_j)} \mathbf{v}_i^H \mathbf{Q}_k \boldsymbol{\nu}(\theta_j) = \mathbf{v}_i^H \mathbf{q}_1 + \frac{\beta_{1:k-1}}{\omega(\theta_j)} \mathbf{v}_i^H \tilde{\mathbf{F}}_k \boldsymbol{\nu}'(\theta_j), \quad (50)$$

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Remark: This relation is correct, **no matter how close** or far away  $\lambda_i$  and  $\theta_j$  are. The relation can be obtained using **any** eigenvalue and **any** Ritz value.

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Sorting gives the following **anti-Taylor-like** approximation,

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There is a good chance that  $\mathbf{Q}_k \boldsymbol{\nu}'(\theta_j)$  is a **better candidate** for a “Ritz vector” if  $\mathbf{Q}_k \boldsymbol{\nu}(\theta_j)$  is “small” and  $\theta_j$  is close to an eigenvalue of  $\mathbf{A}$ .

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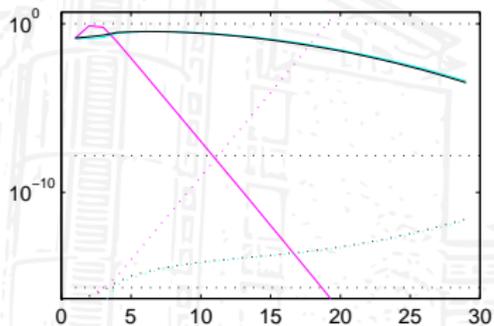
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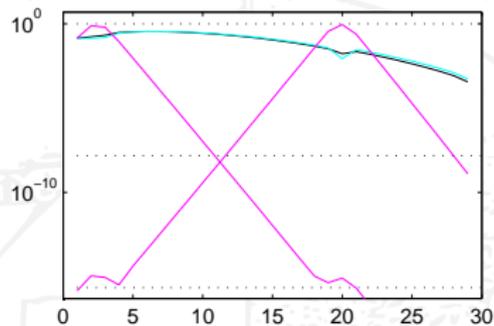
A mixed **numerical-symbolic** computation I presented at the GAMM annual meeting 2006 does support this idea in case of a second Ritz copy.

# An example from my 2006 GAMM talk

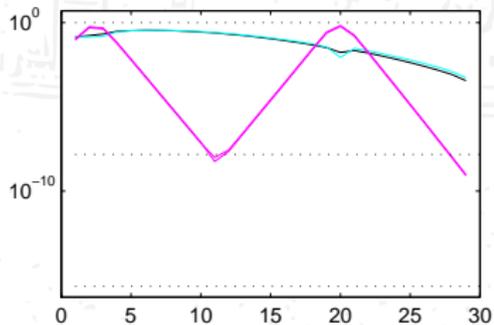
symbolic Lanczos for 29 steps



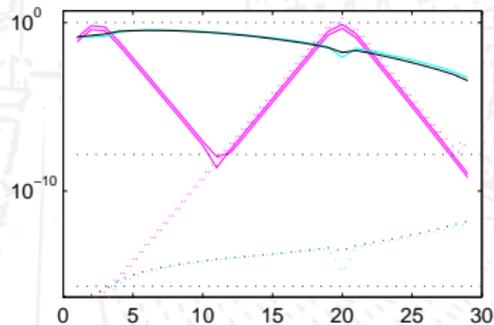
finite precision Lanczos for 29 steps; Matlab 7.2.0.294 (R2006a)



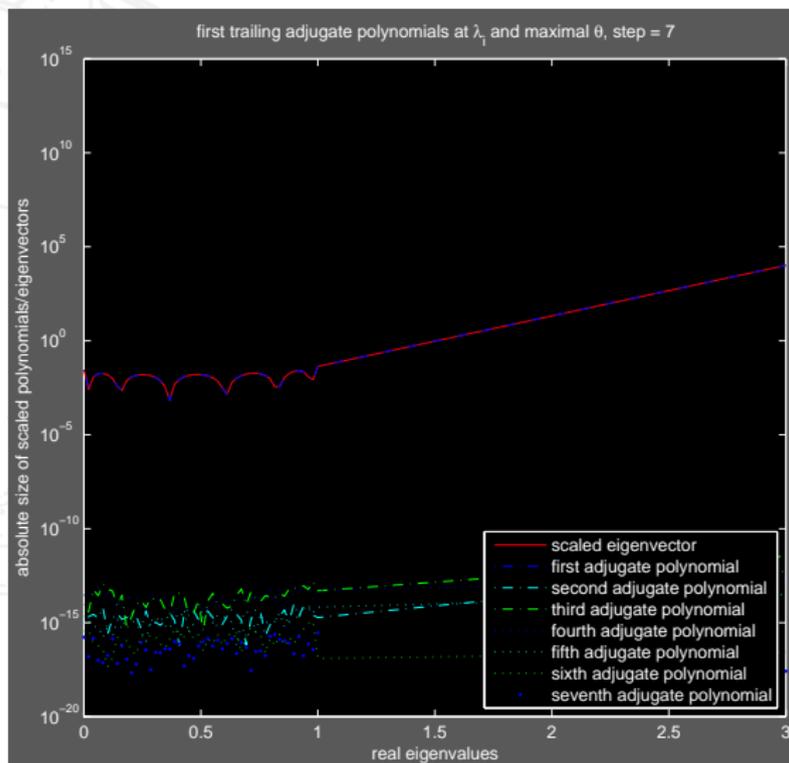
finite precision Lanczos for 29 steps; older version of MRRR



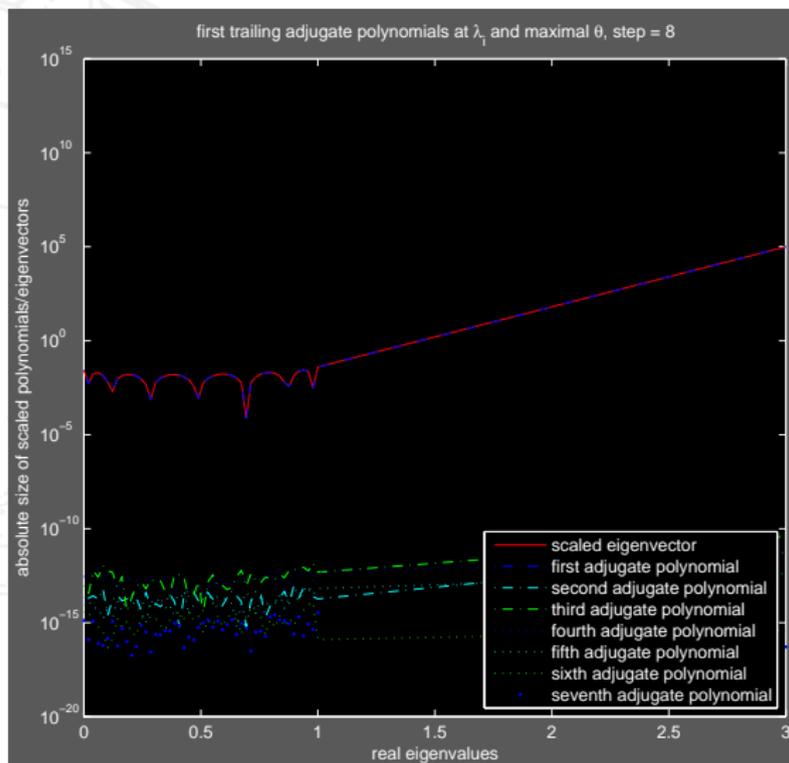
finite precision Lanczos for 29 steps; exact eigenvectors



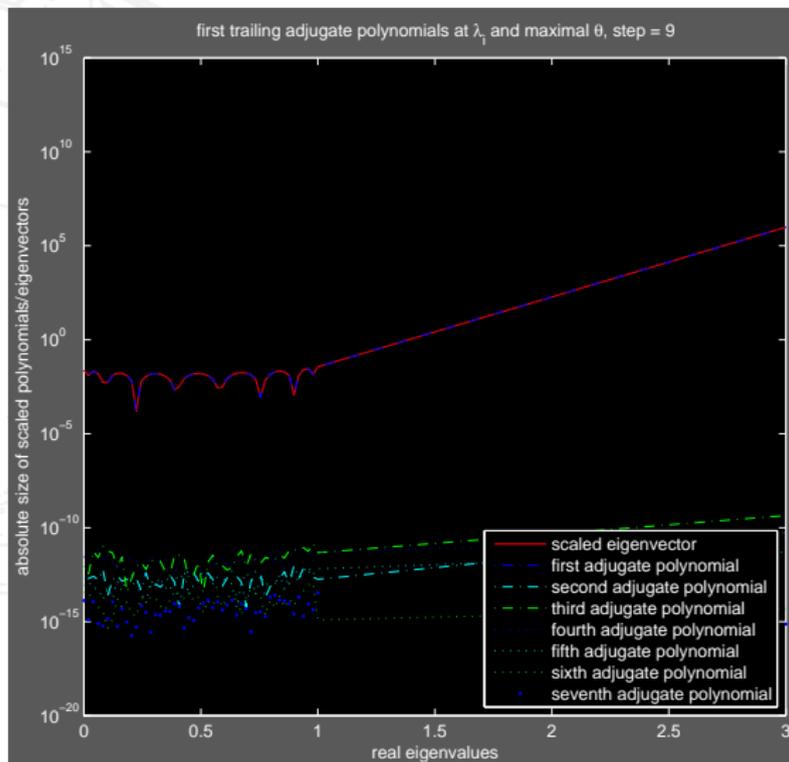
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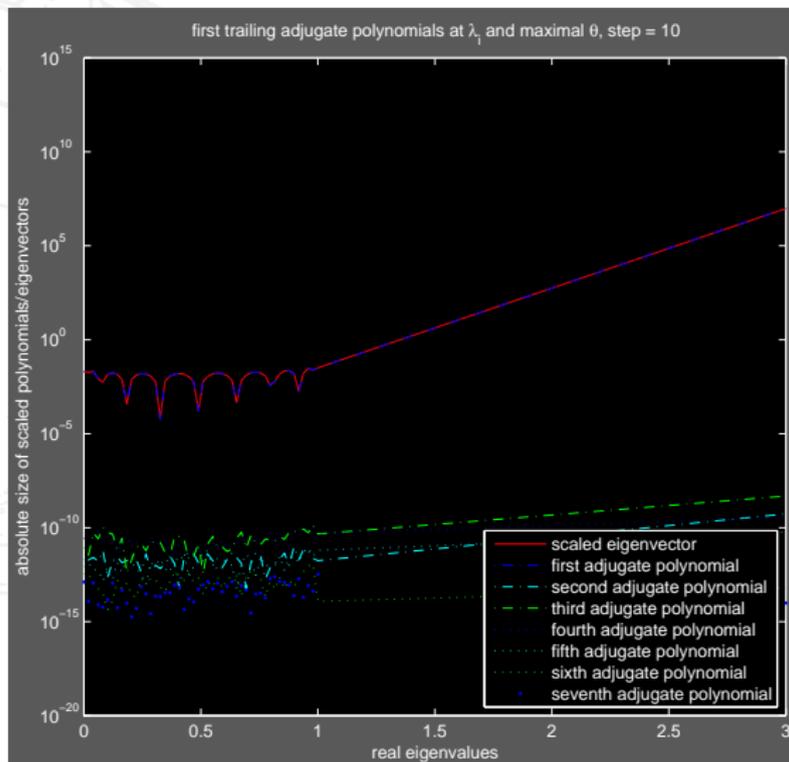
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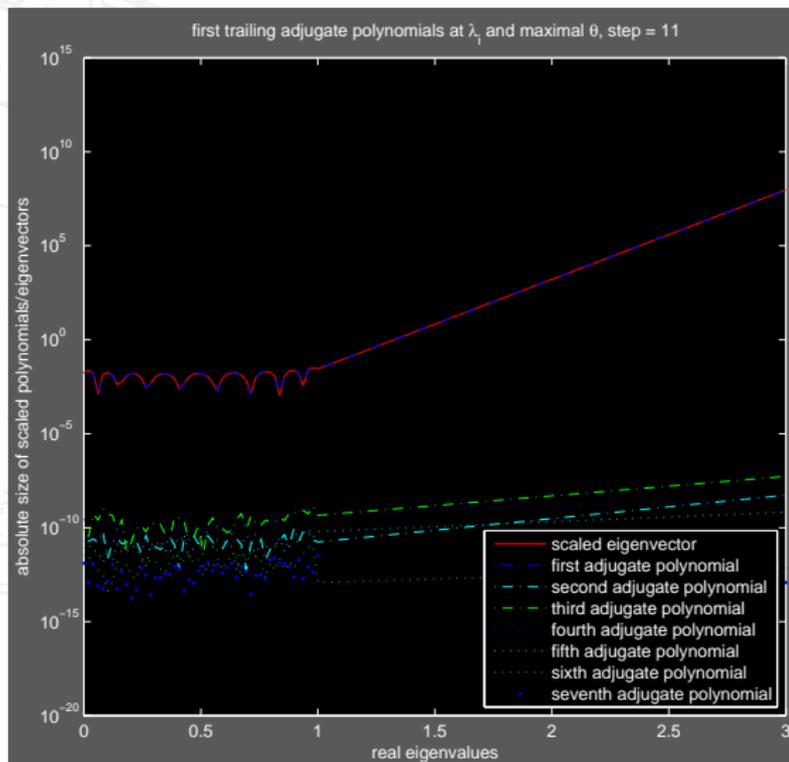
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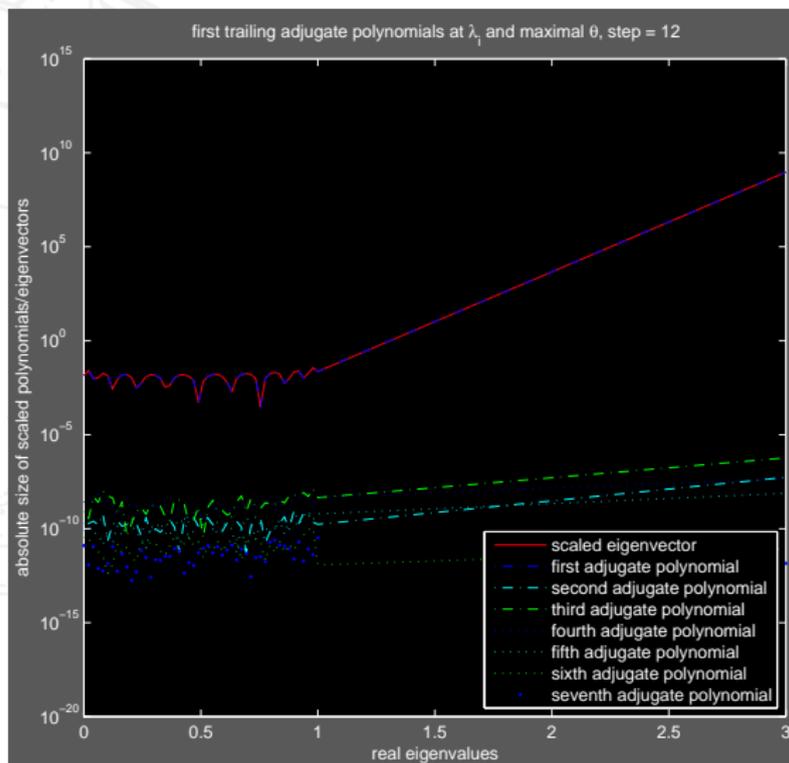
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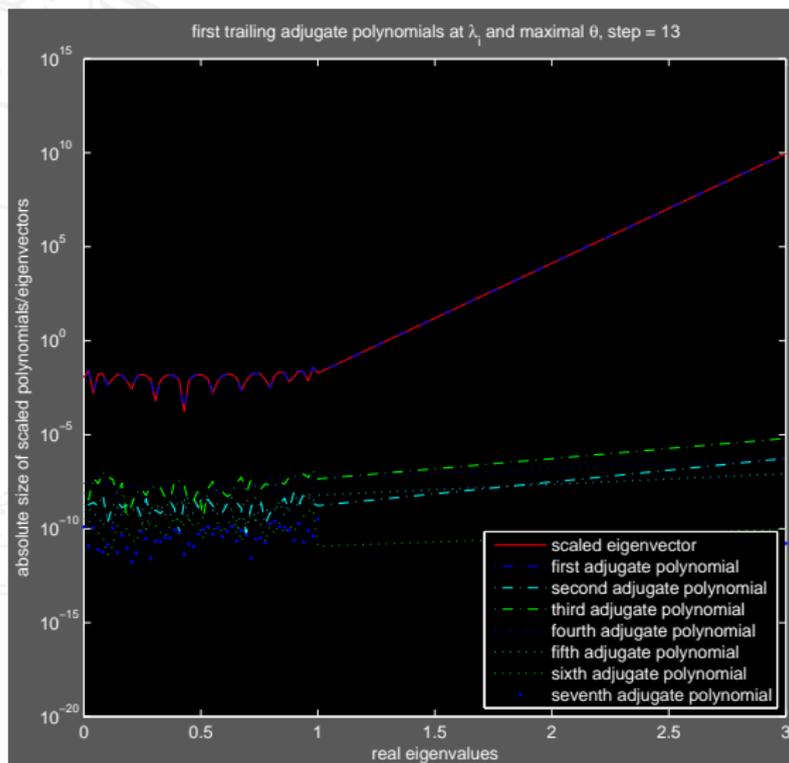
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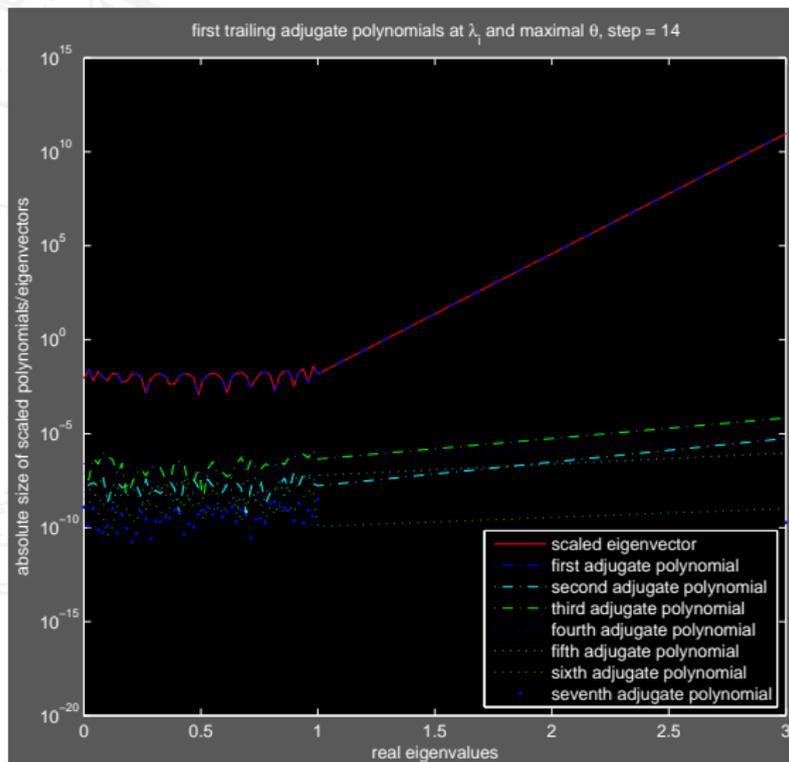
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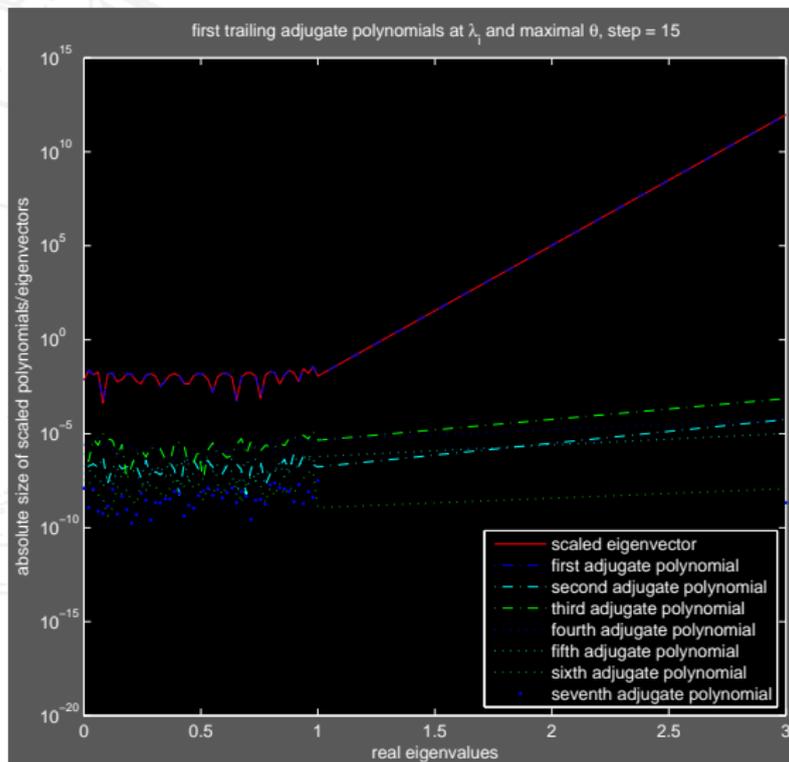
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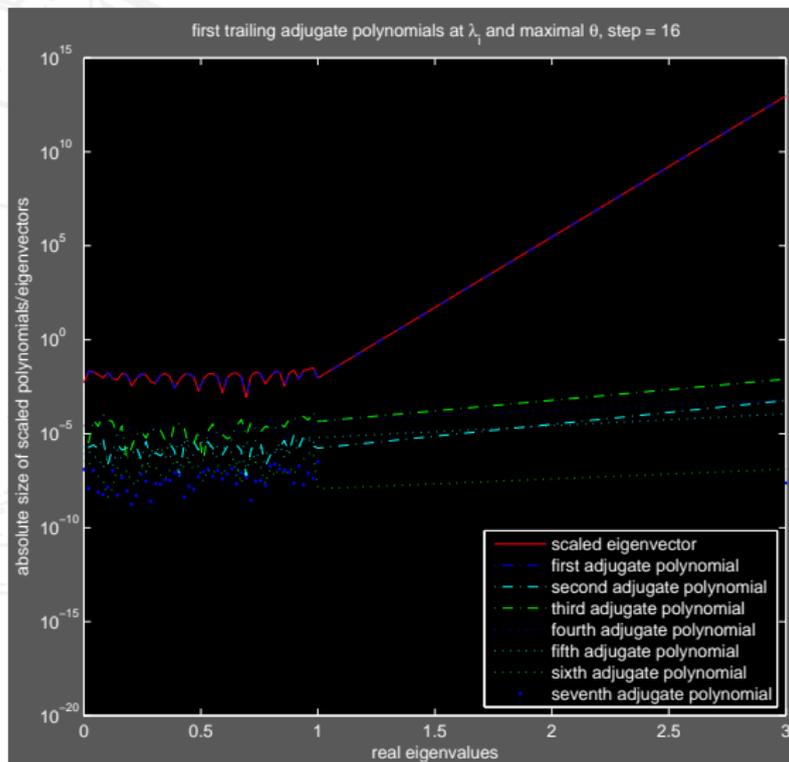
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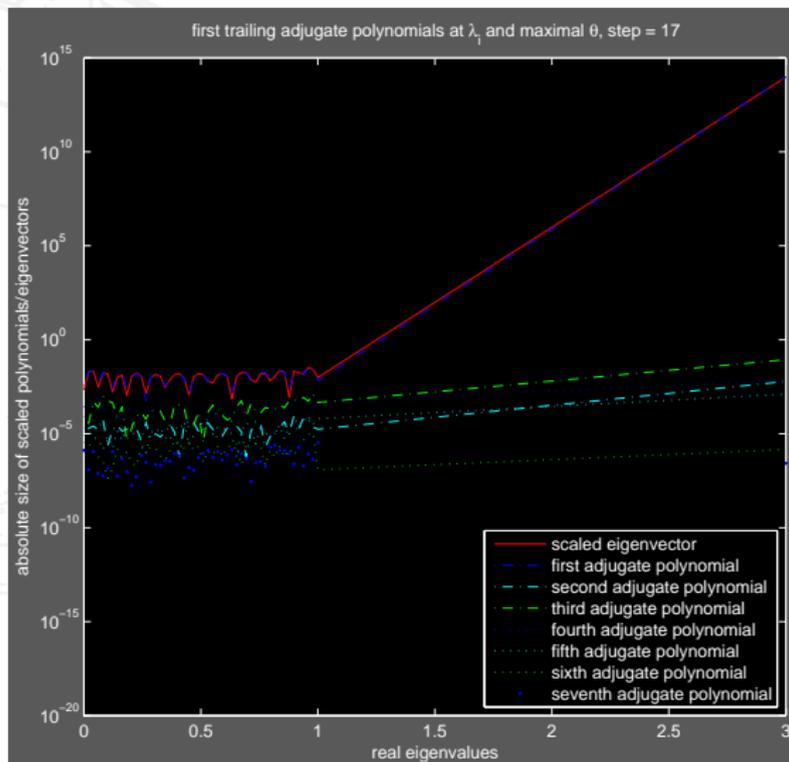
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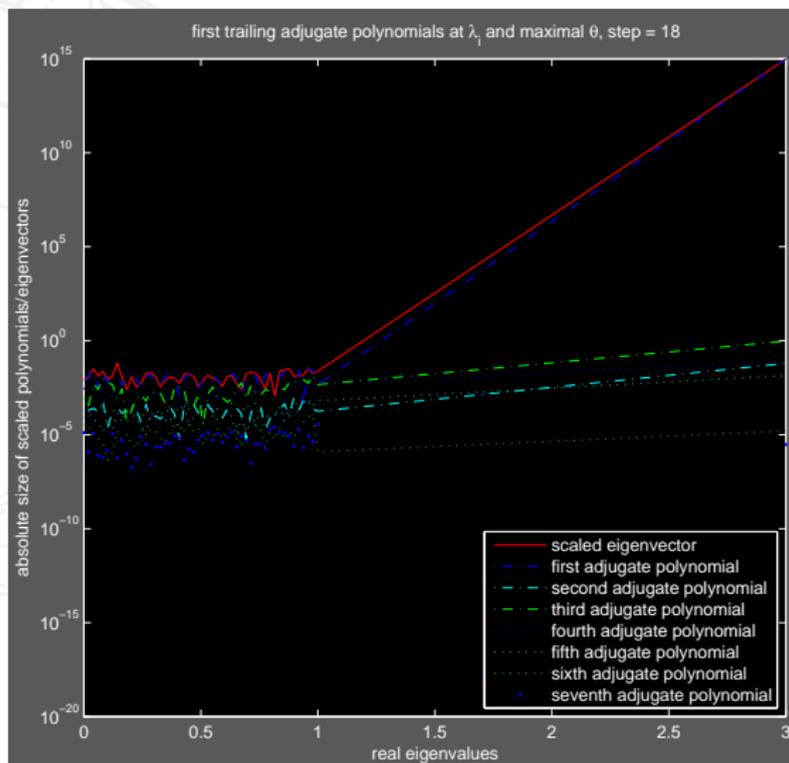
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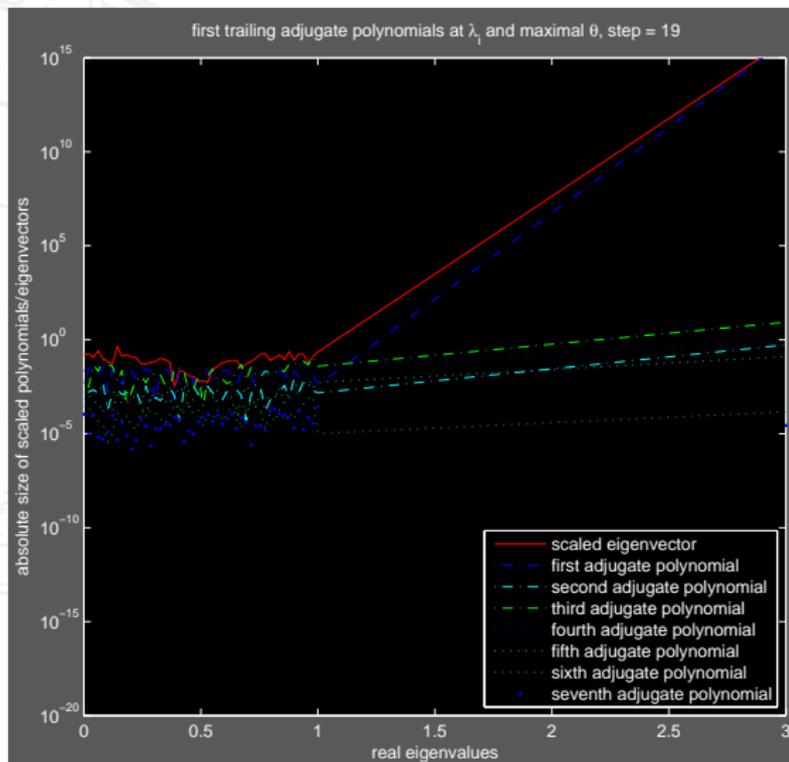
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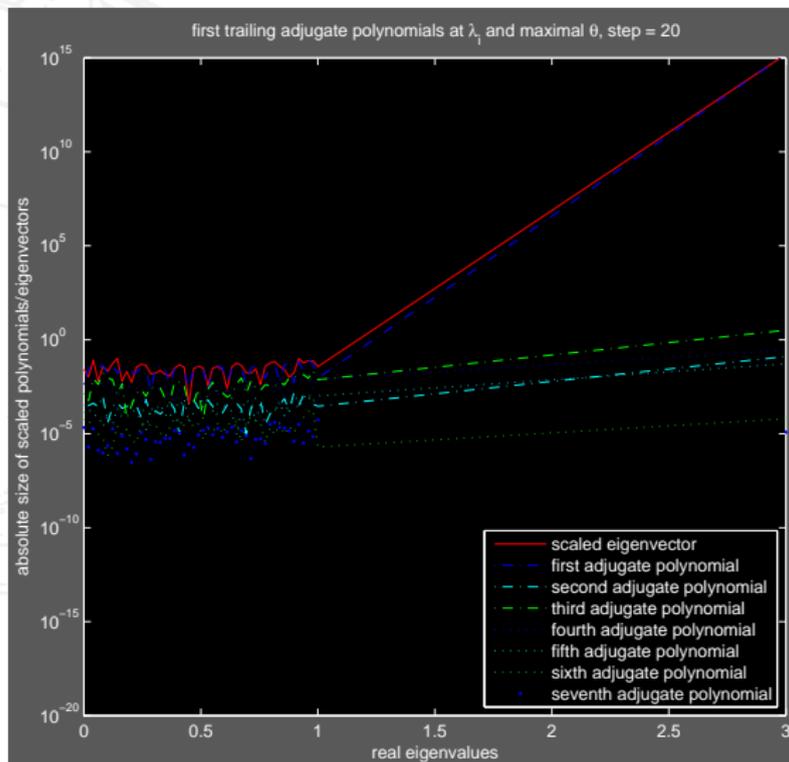
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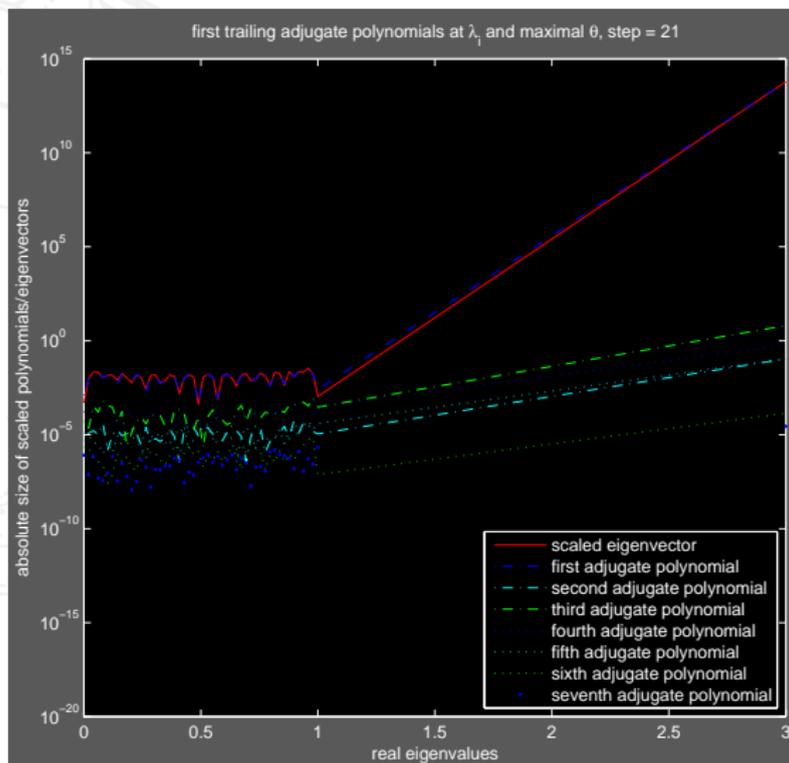
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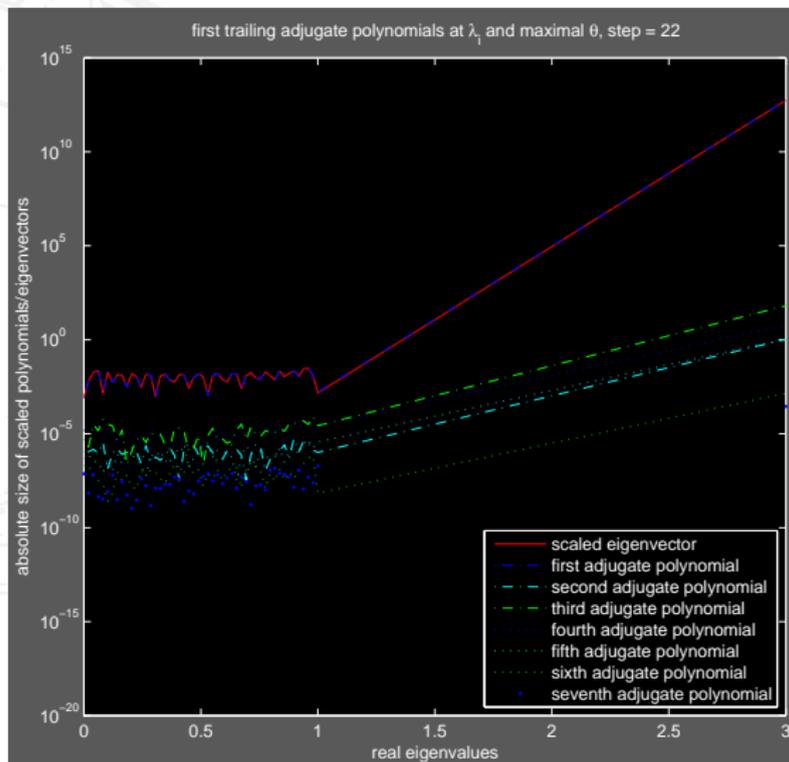
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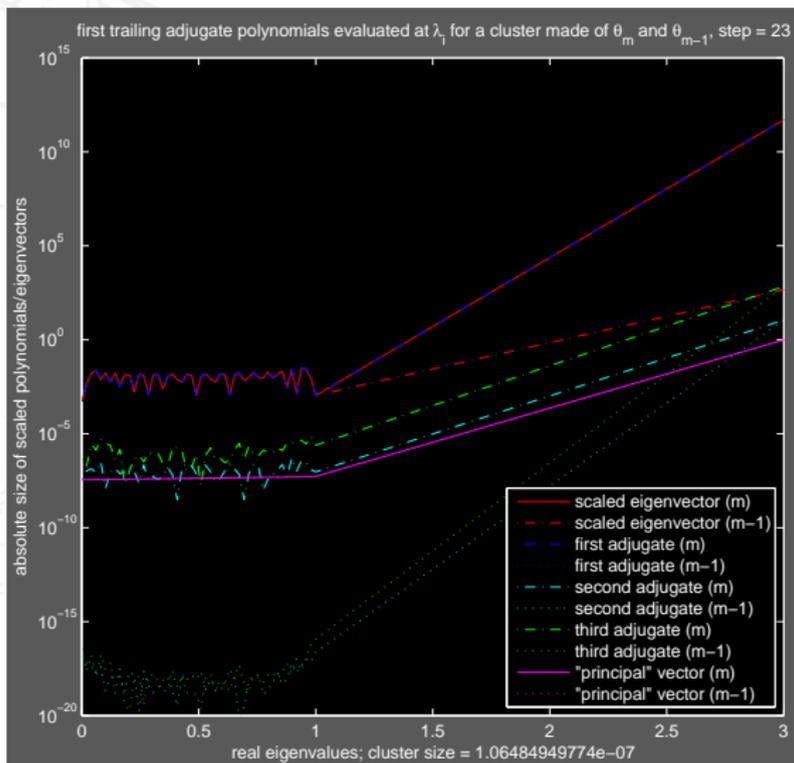
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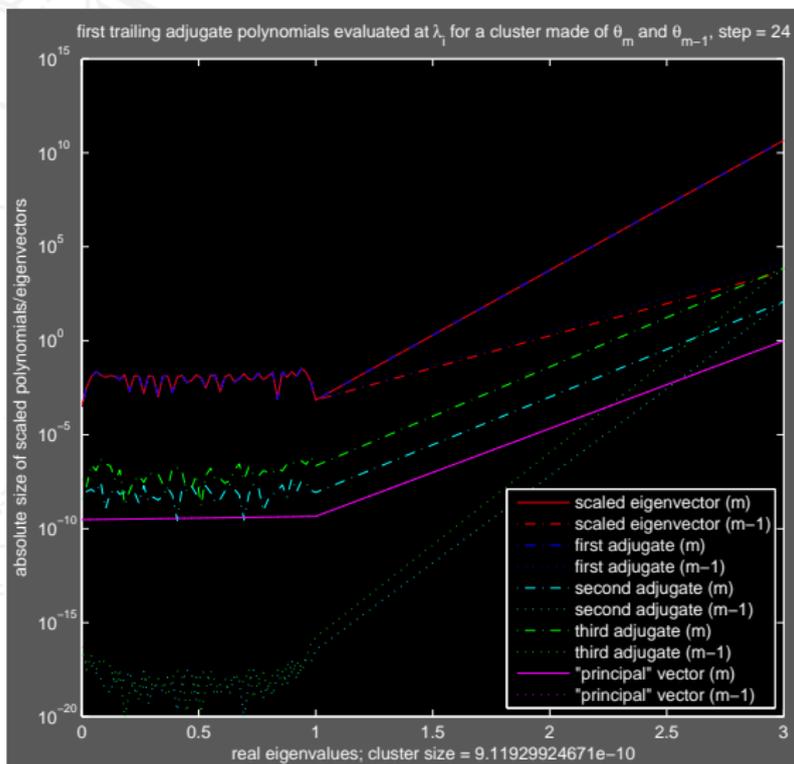
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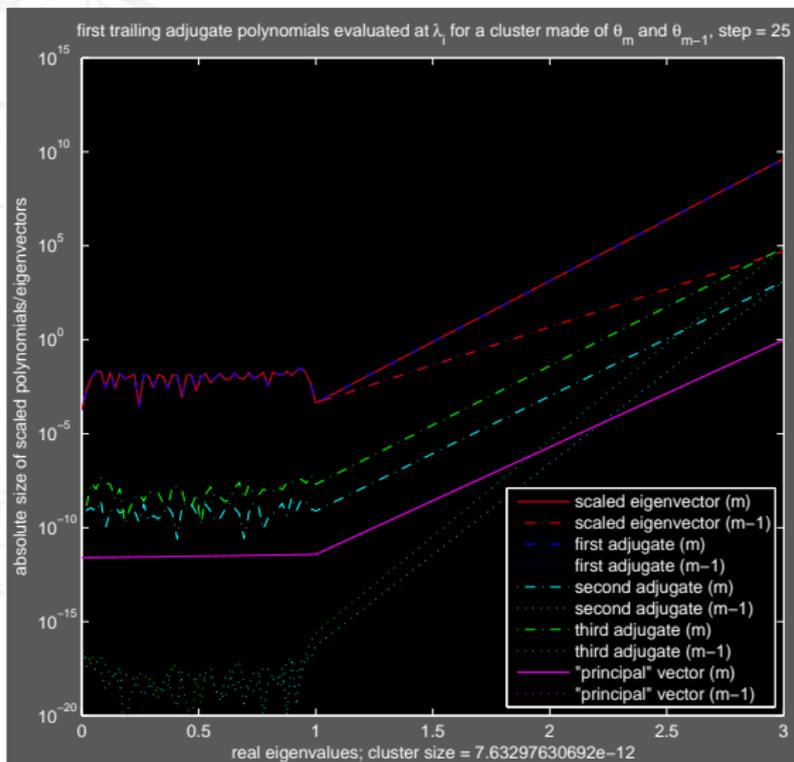
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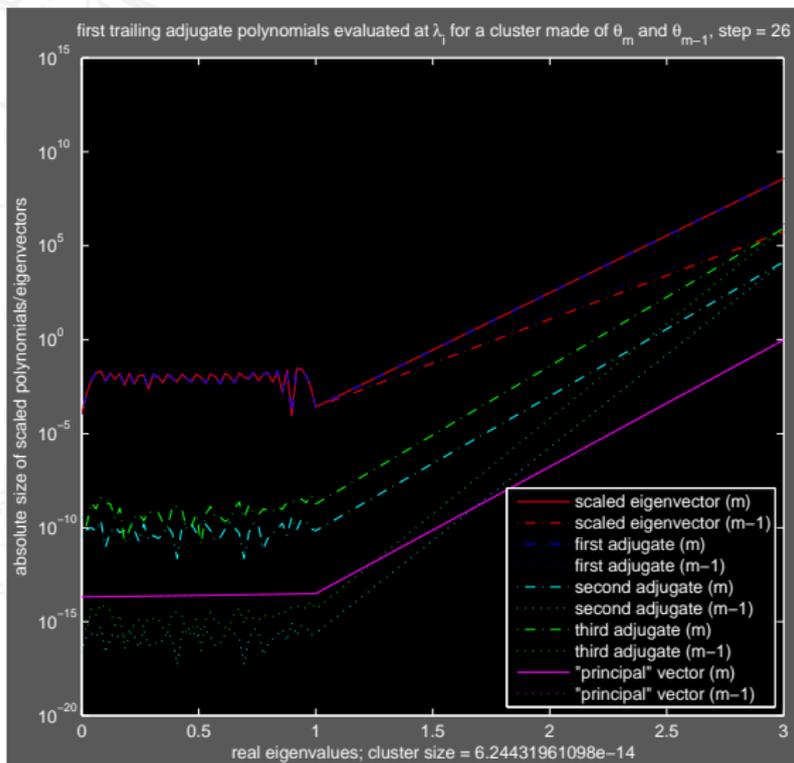
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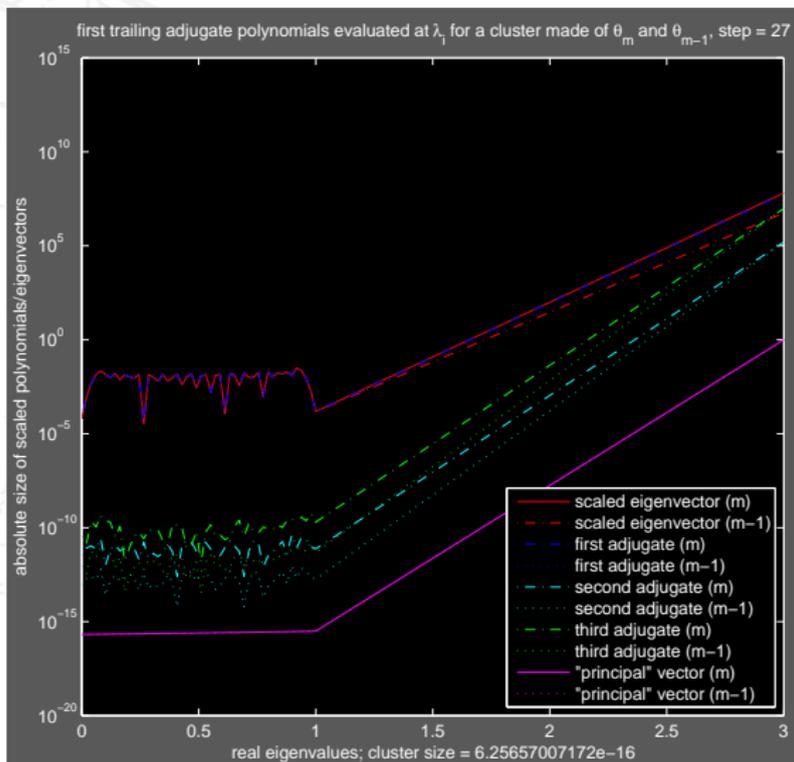
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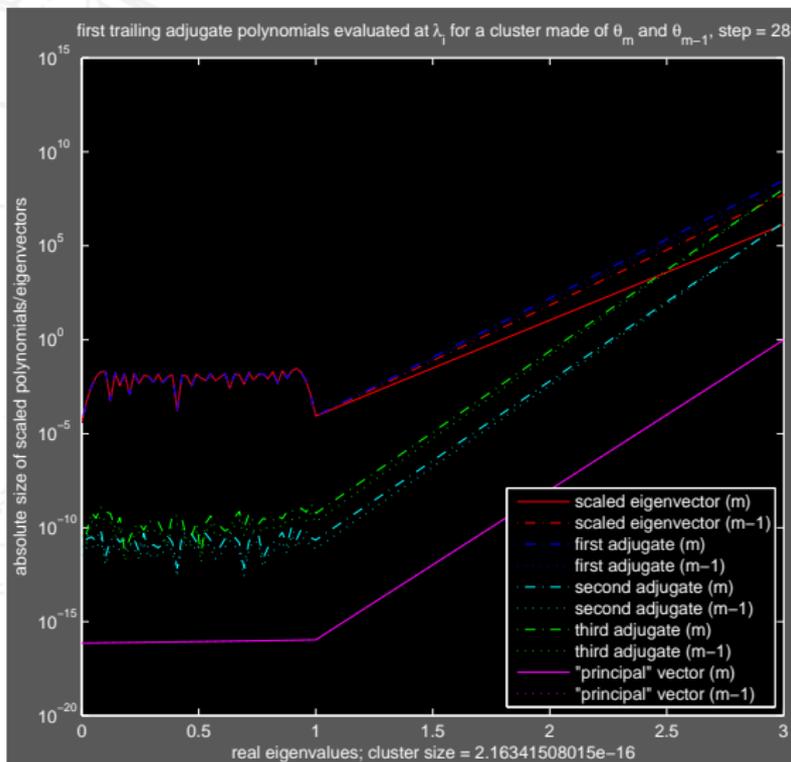
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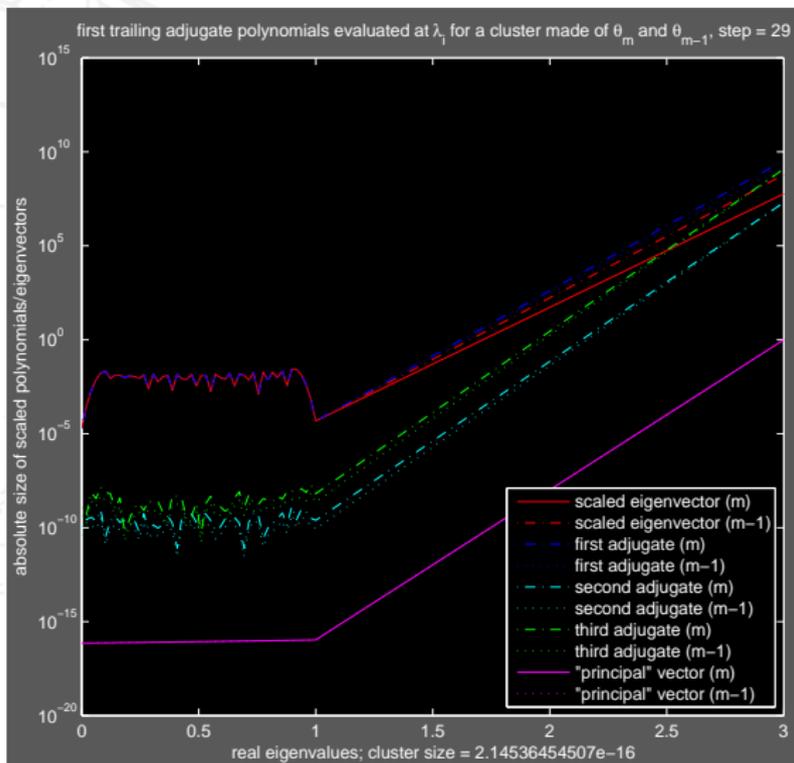
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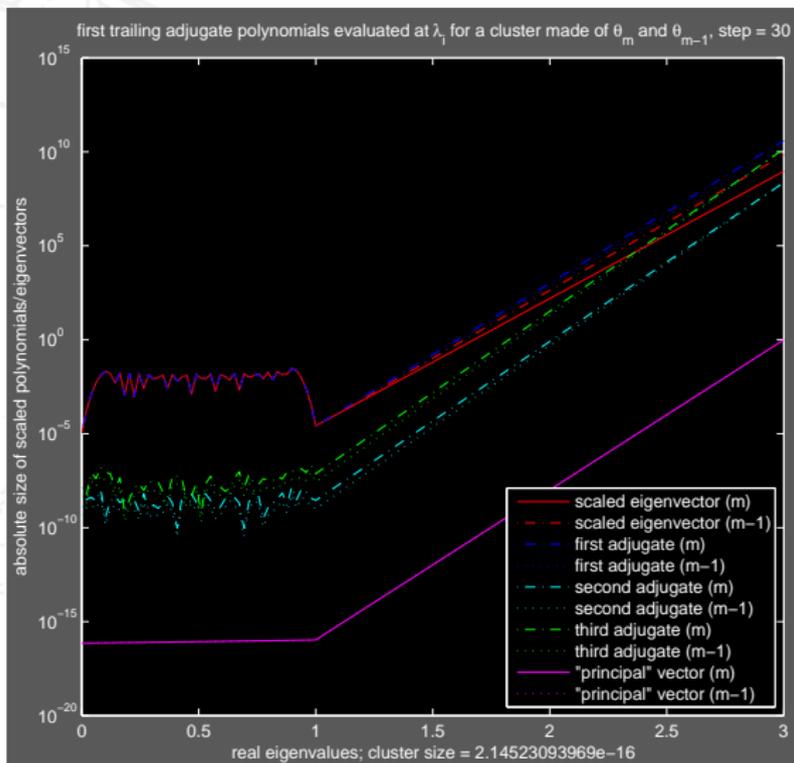
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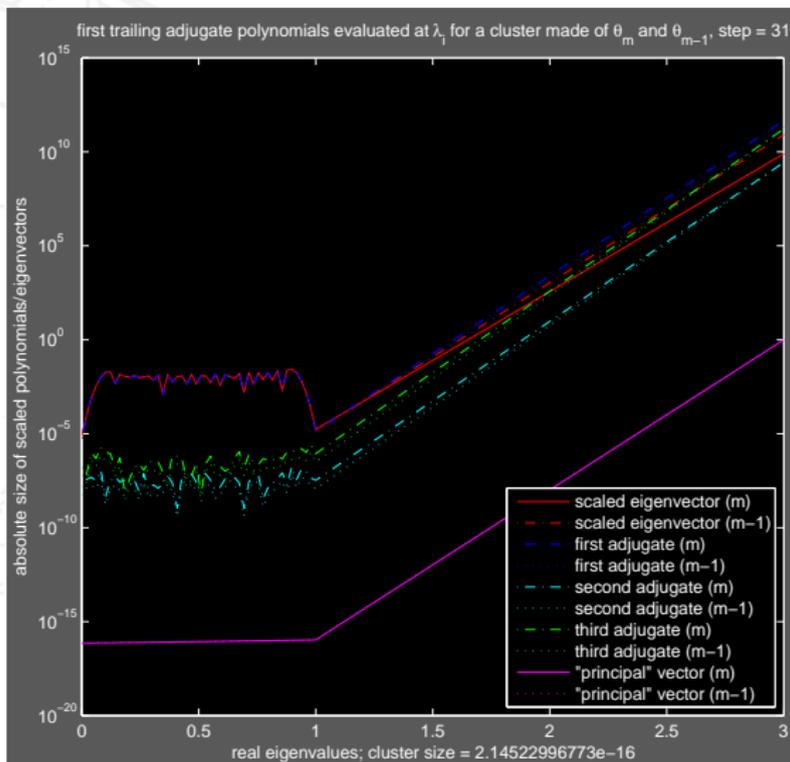
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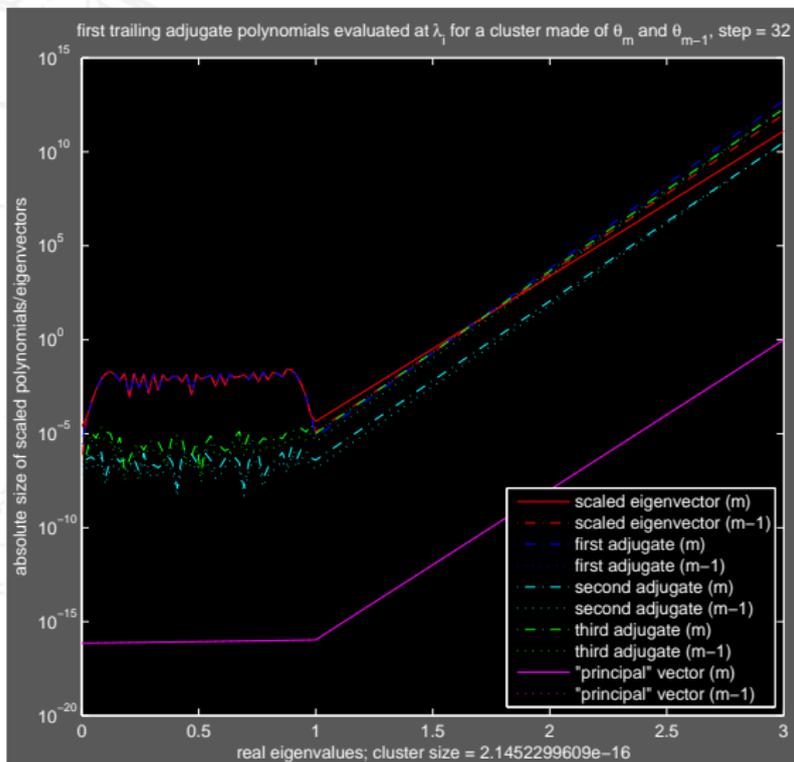
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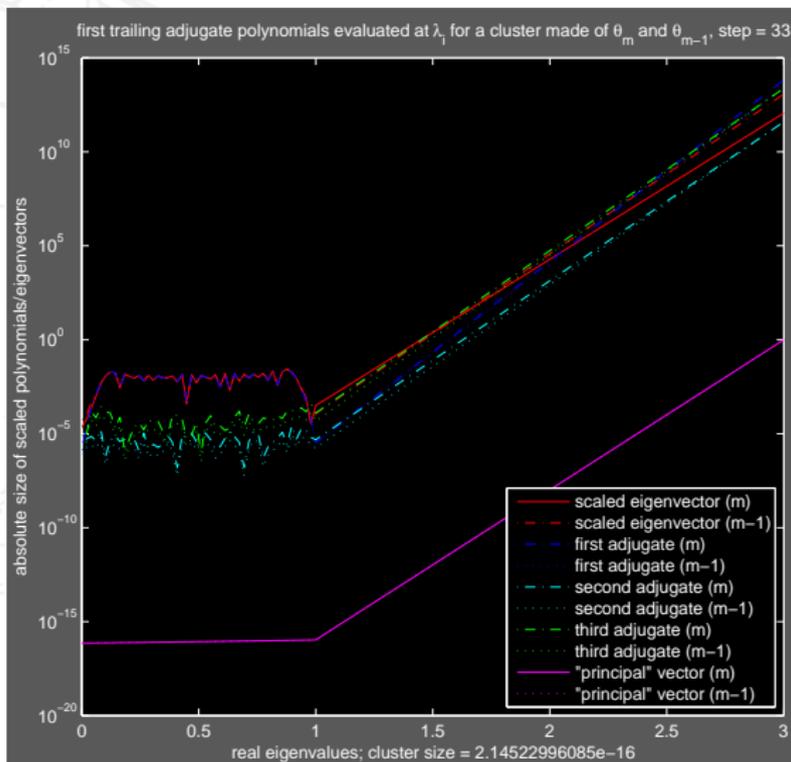
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# Outline

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$$(z\mathbf{I}_k - \mathbf{T}_k) \frac{\boldsymbol{\nu}^{(\ell)}(z)}{\ell!} + \frac{\boldsymbol{\nu}^{(\ell-1)}(z)}{(\ell-1)!} = \mathbf{e}_1 \frac{\chi^{(\ell)}(z)}{\beta_{1:k-1}}. \quad (54)$$

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This implies that

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We have used the fact that the **last  $\ell$  components** of  $\boldsymbol{\nu}^{(\ell)}(z)$  are zero.

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We could use any **linear combination** of the derivatives for a fixed  $z$ , as everything is linear,

$$\begin{aligned}
 ((z\mathbf{I}_n - \mathbf{A})\mathbf{Q}_k - \mathbf{F}_k) \left( \sum_{\ell=0}^p a_\ell \frac{\nu^{(\ell)}(z)}{\ell!} \right) + \mathbf{Q}_k \left( \sum_{\ell=1}^p a_\ell \frac{\nu^{(\ell-1)}(z)}{(\ell-1)!} \right) \\
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We could try to find a linear combination

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Another example: Choosing  $p = k$  and  $a_{\ell} = a_{\ell}(z)$  appropriately gives the **Taylor approximation** to, say, the characteristic polynomial of  $\mathbf{A}$  at  $\lambda$ .

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# Polynomial view on Chris Paige's result

We can consider the **parameter-dependent relation**

$$(\mathbf{T}_k - z\mathbf{I}_k)\mathbf{R}_k + \mathbf{E}_k = \mathbf{R}_k(\mathbf{T}_k - z\mathbf{I}_k) + \mathbf{r}_{k+1}\beta_k\mathbf{e}_k^T. \quad (59)$$

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This is an exact polynomial relation with polynomials of degree  $k - 1$ , i.e., these are  $k$  linear equations:

$$\hat{\nu}(z)^H \mathbf{E}_k \nu(z) = \begin{pmatrix} 1 & \cdots & z^{k-1} \end{pmatrix} \begin{pmatrix} \star & \cdots & \star \\ & \ddots & \vdots \\ & & \star \end{pmatrix} \begin{pmatrix} z^{k-1} \\ \vdots \\ 1 \end{pmatrix} \quad (61)$$

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This gives the **complete characterization** of the loss of orthogonality

$$\mathbf{r}_{k+1}\beta_k = \mathbf{Q}_k^H \mathbf{q}_{k+1}\beta_k \quad (62)$$

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Well known is this result when  $z = \theta_j$  is any **Ritz value**, but we could compare, say, the **coefficients of the highest term**  $z^{k-1}$ :

$$\text{trace}(\mathbf{E}_k)z^{k-1} + \dots = \hat{\mathbf{v}}(z)^H \mathbf{E}_k \mathbf{v}(z) = \hat{\mathbf{v}}(z)^H \mathbf{r}_{k+1}\beta_k = \mathbf{q}_k^H \mathbf{q}_{k+1}\beta_k z^{k-1} + \dots \quad (63)$$

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This is correct. It does not give further insights, but proves that the relation is sound. The diagonal of  $\mathbf{E}_k$  is closely related to the local loss of orthogonality.

# Polynomial view on Chris Paige's result

Maybe of interest in **CG or other OR methods** is the relation involving the constant terms, namely

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By definition of  $\boldsymbol{\nu}(z)$ ,  $\mathbf{z}_k$  defined by

$$\mathbf{z}_k \frac{\chi(0)}{\|\mathbf{r}_0\| \beta_{1:k-1}} := -\boldsymbol{\nu}(0) = -(-\mathbf{T}_k)^{-1} \frac{\chi(0)}{\beta_{1:k-1}} \mathbf{e}_1, \quad (65)$$

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At this point the talk comes to its end. The true research can **start** here.

# Conclusion and Outlook

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- ▶ We have shown that the analytic representation of eigenvectors as polynomial vectors evaluated at the eigenvalues results in simpler expressions. These are based on **differentiation**.
- ▶ We failed to give a complete error analysis based **solely** on our polynomial description.
- ▶ The presented relations mostly carry over to the **unsymmetric Lanczos process**, portions of it should help in distinguishing different implementations of the unsymmetric Lanczos process.

# The final slide . . .



Děkuji.

# The final slide . . .



Děkuji. Once Again.

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