## The Envelope Method

for computing orthogonal eigenvectors belonging to isolated clusters of very close eigenvalues of a symmetric tridiagonal matrix

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## Eigenvectors - The pure mathematician's solution

## Givens' method

$$\begin{pmatrix} (T - \lambda I)\mathbf{x} = \mathbf{0} \\ \beta_1 & \alpha_2 - \lambda & \beta_2 \\ \beta_2 & \alpha_3 - \lambda & \beta_3 \\ \vdots \\ \beta_{n-2} & \alpha_{n-1} - \lambda & \beta_{n-1} \\ \beta_{n-1} & \alpha_n - \lambda \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \mathbf{0}$$

Assumption:  $\beta_i \neq 0$ , all *i* 

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Theorem 1.  $\beta_i \neq 0$ , all  $i \implies$  eigenvalues distinct Theorem 2.  $\beta_i \neq 0$ , all  $i \implies x_1 x_n \neq 0$ Givens method: set  $x_1 = 1$ Eqn. 1 determines  $x_2$ Egn. 2  $X_3$ . . . Eqn. *n* – 1 " Xn [Eqn. n is redundant] Normalize, if required (could start with  $x_n = 1$  and proceed in reverse)

Properties. Perfect in exact arithmetic with exact  $\lambda$ .

Defect 1. What if 2 eigs agree to working precision? What if 4 eigs agree too working precision?

Defect 2. Can fail even for an isolated eigenvalue  $(W_{21}^{-})$ 

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Is it difficult to compute numerically orthogonal eigenvectors belonging to an isolated cluster of very close eigenvalues?

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Ans: yes, if you compute them one by one.

Ans: no , if you compute them all together.

## Envelopes

The envelope of a vector v is

$$|\mathbf{v}| = [|\mathbf{v}_1|, |\mathbf{v}_2|, \dots, |\mathbf{v}_n|]^t.$$

Let *Q* be any orthonormal basis for a subspace  $S \subseteq \mathbb{R}^n$  of dimension *p*. Then its envelope  $\mathcal{E}_s$  is given by

$$\mathcal{E}_{s}(i) = \|Q(i, 1:p)\|_{2}, i = 1, 2, ..., n$$
  
 $\|\mathcal{E}_{s}\|_{2} = \|Q\|_{F} = \sqrt{p}.$ 

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The envelopes of <u>invariant</u> subspaces belonging to isolated clusters of very close eigenvalues of tridiagonal T have hills and valleys.

The smaller is

cluster width cluster gap

the deeper are the valleys.

The number of hills  $\leq$  the number of eigenvalues in the cluster.

Distinguished sparse basis

To each hill associate a vector that is zero except for the hill (suitably extended).

Some hills may have two vectors assigned to them.



Figure: An eigenvector of a tridiagonal: most of its entries are negligible

Matrix comes from nuclear chemistry George Fann matrix, n = 966



Figure: Snapshot of Envelope (108 eigenvalues)

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## n = 2053, cluster size 108 Cluster determined by submatrix 1:515

# How to construct a distinguished basis from the envelope

Take each hill down to the valley on each side and extend smoothly to zero to obtain the indices *first* and *last* 



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This gives one submatrix T(first : last) per hill.

In pratice, we will not have the envelope.

We want to create the index pair (first:last) from the tridiagonal itself.

There are several methods.

We will describe a new inexpensive one later.

To create the basis vector(s)

- Use the submatrix T(first.last)
- Compute the eigenpair(s) with eigenvalue in the cluster interval.

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#### Table: Selected eigenvalues of W<sup>+</sup><sub>21</sub>





Figure: Envelope for  $\lambda_{12}$ ,  $\lambda_{13}$  from  $W_{21}^+$ 



Figure: Vectors  $z_+$  and  $z_-$  for the pair near 6 on a log scale

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 $0 < x < 1, x \longrightarrow -1/\log_{10} x$ , correct sign normalized



Figure: Bisectors of  $z_+$  and  $z_-$  on a log scale



Figure: Basis eigenvectors of  $W_{100}$  for  $\{\lambda_{93}, \lambda_{94}, \lambda_{95}\}$ 

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 Glued Wilkinson matrix

 4 copies of  $W_{25}^+$ , glue= 0.3

 Cluster structure 3 - 2 - 3 

 12.5778...

 12.74619...

 2

 12.93911...

 $G_2(W_{25}^+, 4, 0.3)$ 



Figure: Basis eigenvectors of  $W_{100}$  for  $\{\lambda_{96}, \lambda_{97}\}$ 



Figure: Basis eigenvectors of  $W_{100}$  for { $\lambda_{98}, \lambda_{99}, \lambda_{100}$ }

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Figure:  $G_2(W_{201}^+, 2, \sqrt{\varepsilon})$ ,  $\lambda_3, \lambda_4 \sim 0.25$ 



Figure: Eigenvectors  $x_{79}$ ,  $x_{80}$ ,  $x_{81}$ ,  $x_{82}$  for  $T_{87}$  on a log scale

Lanczos

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## Wanted:

# a cheap way to approximate the submatrices

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Results complementary to Gersgorin's Circle Theorem  $B \in \mathbb{C}^{n \times n}$ 

$$G_{j}^{row} := \left\{ \zeta : |\zeta - b_{jj}| \le \sum_{k \ne j} |b_{jk}| \right\}$$

G's Theorem. Each  $\lambda \in eig(B)$  is located in at least one G-disk.

A complementary result

lf

$$\lambda \notin G_j^{row}$$
 for  $j = p : q, p < q$ 

then

 $\lambda$ 's column eigenvector decays in entries p : q.

(the direction of decay varies, sometimes from p to q, sometimes q to p) Corollary. The (local) maximal elements of an eigenvector occur only for indices *k* for which  $\lambda \in G_k$ .

For large *n* this corollary makes searching for a maximal entry more efficient.

E.g. If  $\lambda \notin G_i$ , i = 1, ..., 500, then start computing at index 501.

Are there any uses for indices *i* for which  $\lambda \notin G_i$ ?

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## Symmetric Tridiagonal Case



 $(T - \lambda I)z = \mathbf{0} \Longrightarrow L^t z = \mathbf{e}_n z_n \Longrightarrow z_j = -I_j z_{j+1}, \ j = n-1, \dots, 1$ 

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Lemma. If  $\lambda \notin G_j$  and  $|I_{j-1}| < 1$  then  $|I_j| < 1$ .

Proof.

$$\begin{split} l_{j} &= \frac{\beta_{j}}{d_{j}} \\ d_{j} &= \alpha_{j} - \lambda - l_{j-1}\beta_{j-1} \\ |d_{j}| &\geq |\alpha_{j} - \lambda| - |l_{j-1}| |\beta_{j-1}| \\ &> |\alpha_{j} - \lambda| - |\beta_{j-1}|, \quad \text{since} |l_{j-1}| < 1 \\ |l_{j}| &< \frac{|\beta_{j}|}{|\alpha_{j} - \lambda| - |\beta_{j-1}|} \\ &< 1, \quad \text{since} \lambda \notin G_{j} \quad \Box \end{split}$$

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Theorem. If  $\lambda \notin G_j$  for j = p : q, p < q, and  $|I_{p-1}| < 1$  then

$$|l_j| < 1$$
 for  $j = p - 1 : q$ 

and

$$|z_j| < \left(\prod_{i=j}^q |I_i|\right) |z_{q+1}|$$
 for  $j = p-1:q$ .

Similar results for  $T - \lambda I = U \mathring{D} U^t$ .

<u>Heuristic:</u> replace  $d_i^+$  and  $d_i^-$  by  $\alpha_i - \lambda$ Hence

$$\prod_{i=j}^{q} |I_i| \approx \frac{\prod_{i=j}^{q} \beta_i}{\prod_{i=j}^{q} (\alpha_i - \lambda)} =: \frac{\operatorname{num}}{\operatorname{den}}$$

Test: num  $\leq$  tol.den tol = 10<sup>-12</sup> or 10<sup>-15</sup>

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**General Case** 

$$B - \lambda I = L^+ D^+ U^+$$

$$u_{kj}^{+} = \frac{b_{kj}}{d_{k}^{+}}, \qquad k = 1, 2, \dots, j - 1$$
  
$$d_{j}^{+} = (b_{jj} - \lambda) - \sum_{k < j} b_{jk} u_{kj}^{+}$$
  
$$u_{jl}^{+} = \frac{b_{jl}}{d_{j}^{+}}, \qquad l = j + 1 : n$$

$$\sum_{l>j} |u_{jl}^{+}| = \frac{1}{|d_{j}^{+}|} \sum_{l>j} |b_{jl}| \leq \frac{\sum_{l>j} |b_{jl}|}{|b_{jj} - \lambda| - (\sum_{k< j} |b_{jk}|) \max_{k < j} |u_{kj}^{+}|} \\ < \frac{\sum_{l>j} |b_{jl}|}{|b_{jj} - \lambda| - \sum_{k < j} |b_{jk}|}, \text{ if } \max_{k < j} |u_{kj}^{+}| < 1$$

< 1, if  $\lambda \notin G_{j}^{row}$ 

Theorem. If 
$$\max_{k < p} |u_{kp}^+| < 1$$
 and  $\lambda \notin G_j^{row}$  for  $j = p : q$  then  
 $\sum_{l > m} |u_{ml}^+| < 1$  for  $m = p : q$ .

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Figure:  $W_{51}^+$ ,  $\lambda_{21}$ 



Figure:  $W_{51}^+$ ,  $\lambda_{31}$ 



Figure:  $W_{51}^+$ ,  $\lambda_{41}$ 



Figure:  $W_{51}^+$ ,  $\lambda_{51}$ 

## Conclusion

For symmetric tridiagonal matrices the computation of numerically orthogonal eigenvectors for

isolated cluster of close eigenvalues

is

easy and rapid

if the cluster is treated as a whole and

almost impossible

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if they are computed one by one.