

On the HZ Method for PGEP

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OUTLINE

- GEP and PGEP
- Derivation of the algorithm
- Convergence, global and asymptotic
- Stability and relative accuracy

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For such a pair there is a nonsingular matrix F such that

$$F^T A F = \Lambda_A, \quad F^T B F = \Lambda_B,$$

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The **eigenpairs** of (A, B) are: $(\alpha_i/\beta_i, Fe_i)$, $1 \leq i \leq n$; $I_n = [e_1, \dots, e_n]$.

How to solve PGEP?

One can try with the transformation $(A, B) \mapsto (L^{-1}AL^{-T}, I)$, $B = LL^T$ and reduce PGEP to EP for one symmetric matrix.

If L has small singular value(s), then computed $L^{-1}AL^{-T}$ will have **corrupt smallest eigenvalues**.

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$$(A, B) \mapsto (A_\varphi, B_\varphi) = (A \cos \varphi + B \sin \varphi, -A \sin \varphi + B \cos \varphi),$$

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We follow the second path.

Jacobi methods for PGEP

We have two diagonalization methods for PGEP

- **Falk-Langemeyer method** (shorter: **FL method**)
(Elektronische Datenverarbeitung, 1960)
- **Hari-Zimmermann variant of the FL method** (shorter: **HZ method**)
(Hari Ph.D. 1984)

The two methods are connected: the FL method can be viewed as the HZ method with “fast scaled” transformations. So, the FL method seems to be somewhat faster and the HZ method seems to be more robust.

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Numerical tests on large matrices, on parallel machines, have confirmed the advantage of the HZ approach. When implemented as one-sided block algorithm for the GSVD, it is almost perfectly parallelizable, so parallel shared memory versions of the algorithm are highly scalable, and their speedup almost solely depends on the number of cores used.

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This property of $B^{(0)}$ is maintained during the iteration process:

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$$Z_k = \begin{bmatrix} I & & & & \\ & c_k & & -s_k & \\ & & I & & \\ & \tilde{s}_k & & \tilde{c}_k & \\ & & & & I \end{bmatrix} \begin{matrix} i(k) \\ j(k) \end{matrix}, \quad i(k) < j(k) \text{ are pivot indices at step } k,$$

$$c_k^2 + s_k^2 = \tilde{c}_k^2 + \tilde{s}_k^2 = 1 / \sqrt{1 - b_{i(k)j(k)}^2} \quad (\text{Gose 1979}).$$

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The selection of pivot pairs $(i(k), j(k))$ defines pivot strategy.

Derivation of the HZ Method

To describe step k , we assume: $A = A^{(k)}$, $A' = A^{(k+1)}$, $Z_k = Z$,

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\hat{Z} is sought in the form of a product of two Jacobi rotations and one diagonal matrix. We have two possibilities:

\hat{Z} is sought in the form:

$$(a) \begin{bmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{1+b_{ij}}} & 0 \\ 0 & \frac{1}{\sqrt{1-b_{ij}}} \end{bmatrix} \begin{bmatrix} \cos(\theta - \frac{\pi}{4}) & -\sin(\theta - \frac{\pi}{4}) \\ \sin(\theta - \frac{\pi}{4}) & \cos(\theta - \frac{\pi}{4}) \end{bmatrix}$$

$$(b) \begin{bmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \\ -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{1-b_{ij}}} & 0 \\ 0 & \frac{1}{\sqrt{1+b_{ij}}} \end{bmatrix} \begin{bmatrix} \cos(\theta + \frac{\pi}{4}) & -\sin(\theta + \frac{\pi}{4}) \\ \sin(\theta + \frac{\pi}{4}) & \cos(\theta + \frac{\pi}{4}) \end{bmatrix}$$

$$\downarrow \\ \hat{B} \rightarrow \text{diag}$$

$$\downarrow \\ \hat{B} \rightarrow I_2$$

$$\downarrow \\ \hat{A} \rightarrow \text{diag}$$

Essential Part of the Algorithm

$$\xi = \frac{b_{ij}}{\sqrt{1+b_{ij}} + \sqrt{1-b_{ij}}}, \quad \rho = \xi + \sqrt{1-b_{ij}}, \quad \xi^2 + \rho^2 = 1,$$

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$$\tan(2\theta) = \frac{2a_{ij} - (a_{ii} + a_{jj}) b_{ij}}{\sqrt{1 - (b_{ij})^2} (a_{ii} - a_{jj})}, \quad -\frac{\pi}{4} \leq \theta \leq \frac{\pi}{4},$$

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$$\cos \phi = \rho \cos \theta - \xi \sin \theta,$$

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Then

$$\hat{Z} = \frac{1}{\sqrt{1 - b_{ij}^2}} \begin{bmatrix} \cos \phi & \sin \phi \\ \cos \psi & \sin \psi \end{bmatrix}.$$

We have used the following **measure** in the convergence analysis:

$$S(A, B) = [S^2(A) + S^2(B)]^{\frac{1}{2}}, \quad S^2(A) = \|A - \text{diag}(A)\|_F^2.$$

The HZ method **converges globally** if

$$A^{(k)} \rightarrow \Lambda = \text{diag}(\lambda_1, \dots, \lambda_n), \quad B^{(k)} \rightarrow I_n \quad \text{as } k \rightarrow \infty,$$

holds for any initial pair of symmetric matrices (A, B) with $B \succ O$.

Global Convergence

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We are adapting the proof to hold for a new larger class of **generalized serial strategies** which includes the known **weak wavefront strategies**.

Asymptotic Convergence

Let (A, B) have simple eigenvalues:

$$\lambda_1 > \lambda_2 > \cdots > \lambda_n, \quad \mu = \max\{|\lambda_1|, |\lambda_n|\},$$
$$3\delta_i = \min_{\substack{1 \leq i \leq n \\ j \neq i}} |\lambda_i - \lambda_j|, \quad 1 \leq i \leq n; \quad \delta = \min_{1 \leq i \leq n} \delta_i.$$

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Theorem

If $S(B^{(0)}) < \frac{1}{n(n-1)}$ and $S(A^{(0)}, B^{(0)}) < \frac{\delta}{2\sqrt{1+\mu^2}}$,

then for the general cyclic and for the serial strategies it holds, respectively:

$$S(A^{(N)}, B^{(N)}) \leq \sqrt{N(1+\mu^2)} \frac{S^2(A^{(0)}, B^{(0)})}{\delta}, \quad N = n(n-1)/2$$

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In the case of **multiple eigenvalues**, the method is **not quadratically convergent**, but can be modified to be such.