An Algorithm for the Generalized Symmetric Tridiagonal Eigenvalue Problem

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Generalized Eigenvalue Problem

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Def
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We consider *only*

symmetric and tridiagonal

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Def (Problem)

Find λ such that $Tx = \lambda Sx$.

T, S symmetric implies $\lambda \in \mathbb{R}$.

Algorithm philosophy

We find zeros of the polynomial equation

$$\mathcal{F}_{(T,S)}(\lambda) = \det(T - \lambda S) = 0$$

using an iterative method, living on real line.

Brainstorming

We want:

Fast and secure iterative method.

Starting points for our method.

Scalability.

We have: Lague

Laguerre's method.

Cuppen's divide and conquer method.

Symmetric tridiagonal matrices.

We add:

Unreducible condition.

Dynamic programming (Bottom-up).

Efficient matrix storing.

Unreducible pencil

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Def ( as in [1] ) (T, S) is an unreducible pencil if t_{i,i+1}^2 + s_{i,i+1}^2 \neq 0 for i = 1, 2, \ldots, n-1.
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Matrix storin

$$T = trid(sub, diag, super)$$
 But T is symmetric, so $sub = super$. We define and use integer , parameter :: dp = kind(1.d0) real(dp), dimension(1:n,0:1) :: T, S with $T(:,0) = diag$ and $T(:,1) = super$. Oss We don't use $T(n,1)$ and $S(n,1)$.

 $\mathcal{F}_{T,S}(\lambda)$ is a polynomial with only real zeros; we call them

$$\lambda_1 \leq \lambda_2 \leq \cdots \leq \lambda_n$$

where we count with multiplicity.

If λ_m and λ_{m+1} are simple zeros (mlt=1), then we consider the quadric

$$g_u(X) = (x - X)^2 \sum_{i=1}^n \frac{(u - \lambda_i)^2}{(x - \lambda_i)^2} - (u - X)^2$$

if $u \neq x$ then $g_u(x) < 0$ and $g_u(\lambda_m) > 0$.

So we have two sign change.

Bolzano's Theorem tell us that there are two zeros of g_u between λ_m and λ_{m+1} . We call them X_-, X_+ .

$$\lambda_m < X_- < x < X_+ < \lambda_{m+1}$$

We have one freedom: the u parameter. Calling $\hat{X}_- = min_u X_-$ and $\hat{X}_+ = max_u X_+$ we can obtain

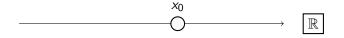
$$\lambda_m \approx \hat{X}_- < x < \hat{X}_+ \approx \lambda_{m+1}$$

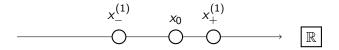
With algebraic manipulations we obtain

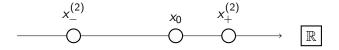
$$\hat{X}_{-}, \hat{X}_{+} = L_{\pm}(x) = x + \frac{n}{-\frac{f'}{f} \pm \sqrt{(n-1)[(n-1)(-\frac{f'}{f}) - n\frac{f''}{f}]}}$$

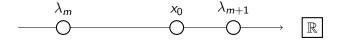
with
$$\frac{f'}{f} = \frac{(\mathcal{F}_{\mathcal{T},S}(\lambda))'}{\mathcal{F}_{\mathcal{T},S}(\lambda)}$$
.











Def (Laguerre's iteration)

If
$$mlt(\lambda_m) = mlt(\lambda_{m+1}) = 1$$
 then

$$x_{+}^{(k)} = L_{+}^{k}(x) = L_{+}(L_{+}(\dots(x_{0})))$$

 $x_{-}^{(k)} = L_{-}^{k}(x) = L_{-}(L_{-}(\dots(x_{0})))$

else we have a similar expression.

We can prove that

$$\lambda_m \leftarrow \dots x_-^{(2)} < x_-^{(1)} < x_0 < x_+^{(1)} < x_+^{(2)} \cdots \rightarrow \lambda_{m+1}$$

It's clear that we need a powerfull method to obtain x_0 and an algorithm to estimate $mlt(\lambda_m)$.

Overstimate $mlt(\lambda_m)$ (as we can read in [2]) causes no trouble, so the most importan aspects of our calculation are: good x_0 and good evaluation of $L_+(x)$.

Split

Consider (\hat{T}, \hat{S}) with

$$\hat{T} = \begin{bmatrix} T_0 & 0 \\ 0 & T_1 \end{bmatrix}$$

$$\hat{S} = \begin{bmatrix} S_0 & 0 \\ 0 & S_1 \end{bmatrix}$$

and let be

$$\hat{\lambda}_1 \leq \hat{\lambda}_2 \leq \cdots \leq \hat{\lambda}_n$$

eigenvalues of (\hat{T}, \hat{S}) , then

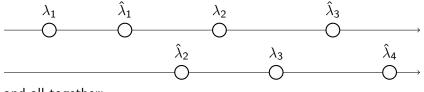
Split

$$-\infty < \lambda_1 \le \hat{\lambda}_1$$
$$\hat{\lambda}_{i-1} \le \lambda_i \le \hat{\lambda}_{i+1}$$
$$\hat{\lambda}_n \le \lambda_n < \infty$$

with i = 2, 3, ..., n - 1.

Oss

It's possible that



and all together:

Grazie per l'attenzione.