Highlights of Paul Willems' thesis

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Notation for error analysis

Need to bound

$$\sqrt{(1+\alpha_1)(1+\alpha_2)(1+\alpha_3)(1+\alpha_4)}, |\alpha_i| \le \varepsilon.$$
Get
$$1+2\varepsilon + O(\varepsilon^2)$$

Question

How to avoid $O(\cdot)$ but keep it simple?

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Higham. $\gamma_n = n\varepsilon/(1-n\varepsilon)$.

Good for sums and products, but

$$\sqrt{1+\gamma_4} \neq 1+\gamma_2 !!$$

Willems. $\varepsilon^{[k]}(n) := n\varepsilon/(1 - kn\varepsilon)$. Top expression $\leq 1 + \varepsilon^{[2]}(2) = 1 + 2\varepsilon/(1 - 4\varepsilon)$. In general,

 $(1 + \varepsilon^{[k]}(n))^{-1} = 1 + \varepsilon^{[k+1]}(n).$

 $(1 + \varepsilon^{[k]}(n))^s = 1 + \varepsilon^{[k/s]}(sn), \ 0 < s < 1,$

Representations of T (without twists)

- 1. (T) matrix entries $\{c_i, e_i\}$
- 2. (N) $\{d_i, l_i\}$, $T = LDL^*$. Pro: defines tiny eigenvalues to high relative accuracy.

Con: does not always exist. Element growth.

- 3. (e) $\{d_i, e_i\}, e_i = l_i d_i$.
- 4. (Z) $\{d_i, I | d_i\}$, $I | d_i = d_i I_i^2 = \text{Schur complements}$. $c_{i+1} = d_{i+1} + I | d_i$.

Can convert between representations with ${f no}$ adds or subtracts. Square Roots for (Z).

Comparison of Representations

The computation of eigenvectors is organized in the Representation Tree.

Each internal node requires a new representation which defines a specific subset of shifted eigenvalues to high relative accuracy.

Accuracy: (Z) is best. max is 3 ulps for (Z) versus 4 for (e).

Speed: (e), when properly optimized. (N) almost as good.

Conclusion: Always use (Z) but switch to (N) or (e) when a node contains a singleton.

Shift of Origin

$$L_{+}D_{+}(L_{+})^{T} = LDL^{T} - \tau I$$

$$(I^{+}_{i})^{2}d^{+}_{i} + d^{+}_{i+1} = I_{i}^{2}d_{i} + d_{i+1} - \tau$$

$$d^{+}_{i+1} = e_{i}^{2}/d_{i} + d_{i+1} - e_{i}^{2}/d^{+}_{i} - \tau$$

Define s_i by $s_i - \tau = d^+{}_i - d_i$ to find

$$s_{i+1} = e_i^2(s_i - \tau)/d_i d_i^+,$$

an important recurrence.

a sample error analysis

Algorithm dstqds, unblocked, e-rep

1.
$$d_i^+ = d_i + (s_i - \tau)$$

2.
$$e_i^+ = e_i$$

3.
$$s_{i+1} = e_i^2(s_i - \tau)/d_i d_i^+$$

Keep the computed s_i sacred.

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Keep the computed
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$$egin{aligned} 1. \ {d_i}^+/(1+lpha_+) &= d_i + (s_i - au)(1+\sigma), \ \left(rac{{d_i}^+}{(1+lpha_+)(1+\sigma)}
ight) &= \left(rac{d_i}{1+\sigma}
ight) + (s_i - au). \end{aligned}$$

3.
$$s_{i+1} = e_i^2 (s_i - \tau)(1 + \sigma)(1 + \alpha_s)/d_i d_i^+$$

1 + $\alpha_s = (1 + \varepsilon)^4$ 3 mults 1 divide

$$\begin{split} 1+\alpha_{\mathfrak{s}} &= (1+\varepsilon)^4, \quad \text{3 mults, 1 divide.} \\ \tilde{e}_i &= e_i \sqrt{\frac{(1+\alpha_{\mathfrak{s}})(1+\sigma)}{(1+\sigma)^2(1+\alpha_+)}}, \quad \text{cancel } 1+\sigma, \\ &= e_i [1+\varepsilon^{[4]}(3)], \quad \text{Willems notation .} \end{split}$$

Figure: Mixed relative error analysis for dtwqds.

Perturbing the shift

Outer Perturbations $NGN^* \rightarrow DNGN^*D, D \approx I$

Inner Perturbations $NGN^* \rightarrow NDGDN^*, D \approx I$

Ostrowski's Theorem. $A^* = A$ $|\lambda(FAF^*) - \lambda(A)| \le |\lambda(A)| \cdot ||FF^* - I||$. Outer perturbations make tiny relative changes.

Application. $X \approx I, Y = X^{-1/2}$

$$N_{+}G_{+}(N_{+})^{*} = NGN^{*} - \tau X$$

 $Y(N_{+}G_{+}(N_{+})^{*})Y = YNGN^{*}Y - \tau I$

 $G \rightarrow YGY, N \rightarrow YNY^{-1}$.

Preserving Tridiagonal Form

Allow 2 × 2 blocks in
$$D$$

$$LDL^{T}$$

$$L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ k_1 & k_2 & 1 & 0 \\ 0 & 0 & k_3 & 1 \end{bmatrix}, D = \begin{bmatrix} d_1 & 1 & 0 & 0 \\ 1 & c_2 & 0 & 0 \\ 0 & 0 & d_3 & 0 \\ 0 & 0 & 0 & d_4 \end{bmatrix}.$$

(3,1) entry:
$$k_1d_1 + l_2 \cdot 1 + 1 \cdot 0$$

Cannot perturb k_1 and l_2 independently and preserve tridiagonal form

Blocks in LDL*

Example.
$$T = GK - \alpha I$$
, $\alpha = O(\varepsilon)$.

$$T = \begin{bmatrix} -\alpha & 1 & 0 & 0 \\ 1 & -\alpha & 1 & 0 \\ 0 & 1 & -\alpha & \alpha \\ 0 & 0 & \alpha & -\alpha \end{bmatrix}, L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ k_1 & k_2 & 1 & 0 \\ 0 & 0 & k_3 & 1 \end{bmatrix}$$

$$D = \begin{pmatrix} d_1 & 1 \\ 1 & c_2 \end{pmatrix} \oplus \begin{pmatrix} d_3 & 0 \\ 0 & d_4 \end{pmatrix}.$$

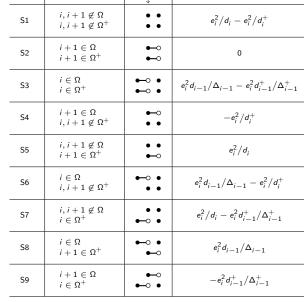
$$k_1 = 1/(1-\alpha^2), \ k_2 = \alpha/(1-\alpha^2) \ k_3 = -(2-\alpha^2)/(1-\alpha^2), \ \Omega = \{2\}.$$

Willems Representation: $D, e, and \Omega = \{ i \text{ where a } 2 \times 2 \text{ ends } \}.$ Secondary Data: $k_i, l_i, \Delta_i = d_i c_{i+1} - e_i^2, inv_D(i).$

Auxiliary Quantities: $s_i = D_+(i, i) - D(i, i) + \tau$ (straightforward) BUT

 s_{i+1} is much more commplicated (9 cases).

What did I miss?



 s_{i+1}

Case

Description

Table: Standard formulae for the next adjustment s_{i+1} .

LDL*			$L^+D^+(L^+)^*$	$i \notin \Omega$	$i \in \Omega$
$d_i \rightsquigarrow \widetilde{d}_i$	1	_	$d_i^+ \leadsto \widetilde{d}_i^+$	2	5
$c_i \leadsto \widetilde{c}_i$	4				

$$i \notin \Omega$$
 $i \in \Omega$
 $e_i \leadsto \widetilde{e}_i$ 3 10
 $\tau \leadsto \widetilde{\tau}_i$ 0 12

Table: Error bounds to achieve mixed relative stability for Algorithm 4.5, for the concrete parameters R=3, $K_{\square}=1/8$, cf. Theorem 4.11 and Figure 4.2. Only first-order bounds are shown, i.e., an entry p stands for a bound $p\epsilon_{\diamond}+\mathcal{O}(\epsilon_{\diamond}^2)$.

Recap on MRRR

- Users want orthogonality among eigenvectors.
 Constraint. No Gram-Schmidt (distributive computing)
 George Fann example.
- Extravagant accuracy delivers orthogonality.
- How to achieve $||Tz z\lambda|| = O(n\varepsilon)|\lambda|$? Replace T with an RRR (for λ). Not enough. Need $relgap(\lambda) > 10^{-3}$.
- Not enough. Need $relgap(\lambda) \ge 10^{-3}$.
- ► Make relgaps large by shifting origin (to clusters). Hence Multiple Representations.
- Organize computation in a Representation Tree.
- ► Twisted factors permit residual norms proportional to $|\lambda|$. Solve $N_k G_k N_k^T z = e_k \gamma_k$. $N_k^T z = e_k$. It is e_k that yields z by products only. Can check that $|\gamma_k| = O(n\varepsilon)|\lambda|$.
- Differential qd algorithms allow for roundoff between representations.
 Eigenvectors invariant under exact shifts.