

Accurate and Efficient Matrix Computations with Totally Positive Matrices

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SIAM Annual Meeting, Montreal, June, 2003

Goals

- **Accurate** and **Efficient** Linear Algebra with **Totally Positive** Matrices

- Linear Equation Solving, Inversion
- Eigenvalues, Eigenvectors
- Singular Values, Singular Vectors

- **Efficient** means $O(n^3)$

- Usual arithmetic model $\text{fl}(a \odot b) = (a \odot b)(1 + \delta)$, $|\delta| \leq \epsilon$

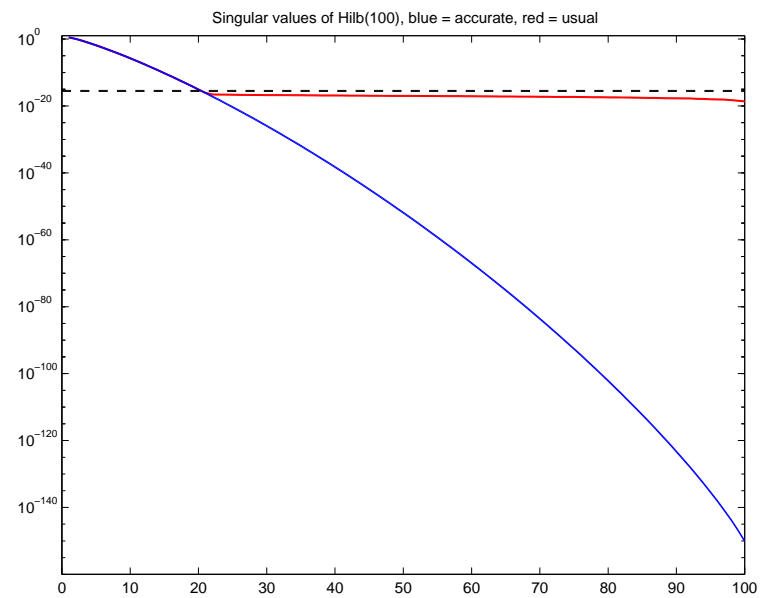
- **Accurate** means correct **sign** and **leading digits**

$$|\hat{\lambda}_i - \lambda_i| \leq O(\epsilon)|\lambda_i|, \quad |\hat{\sigma}_i - \sigma_i| \leq O(\epsilon)\sigma_i,$$

- Contrast: Traditional algorithms

$$|\hat{\lambda}_i - \lambda_i| \leq O(\epsilon) \frac{\|A\|}{|y^*x|}, \quad |\hat{\sigma}_i - \sigma_i| \leq O(\epsilon)\sigma_1$$

Example: 100×100 Hilbert Matrix $H = 1/(i + j)$



How do we lose accuracy in floating point?

- $\text{fl}(a \otimes b) = (a \otimes b)(1 + \delta)$, no under/overflow
- How can we lose accuracy in computing in floating point?
 - OK to multiply, divide, add positive numbers
Proof: $1 + \delta$ factors can be factored out
 - $x_i \pm x_j$, where x_i and x_j are initial data (so exact)
 - Cancellation when subtracting approximate results dangerous:

$$\begin{array}{r} .12345\mathbf{xxx} \\ - .12345\mathbf{yyy} \\ \hline .00000\mathbf{zzz} \end{array}$$

- Trick: We will only compute:
 - $1/(1 + xy)$; $1/x$; xy ; $1/(x_i - y_j)$; $x, y > 0$
 - **never subtract** approximate quantities

Bidiagonal Matrices

- Demmel–Kahan '91

$$B = \begin{bmatrix} a_1 & b_1 & & & & \\ & a_2 & b_2 & & & \\ & & \cdots & \cdots & & \\ & & & a_{n-1} & b_{n-1} & \\ & & & & & a_n \end{bmatrix}$$

- a_i, b_i determine σ_i accurately
- `dqds` preserves the relative accuracy
- Eigenvalues of SPD tridiagonal matrices: $\lambda_i(B^T B) = \sigma_i^2(B)$
 - computable accurately given the Cholesky factors

Total Positivity (TP)

- Def: A is TP if all minors > 0
 - Hilbert $H = 1/(i + j)$
 - Cauchy $C = 1/(x_i + y_j)$, if $0 < x_i, y_j$ -increasing
 - Generalized Vandermonde $V = x_i^{a_j}$, $0 < x_i, a_i$ - increasing
 - TP·TP
 - Schur-complement(TP)
 - $\text{diag}(\pm 1) \cdot \text{TP} \cdot \text{diag}(\pm 1)$
 - See books by Karlin, Gantmacher, Krein
- Theory extends to Totally Nonnegative, Oscillatory, Sign-regular
- Motivation:
 - Eigenvalues and SVD well determined
 - Right answer costs $O(n^3)$, same as the wrong
- Trick: choose the **right parametrization: Bidiagonal**

Total Positivity and Neville Elimination – 1

- Gaussian Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix}$$

Total Positivity and Neville Elimination – 2

- Gaussian Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 0 & 3 & 15 & 63 \end{bmatrix}$$

Total Positivity and Neville Elimination – 3

- Gaussian Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 0 & 2 & 8 & 26 \\ 0 & 3 & 15 & 63 \end{bmatrix}$$

Total Positivity and Neville Elimination – 4

- Gaussian Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 2 & 8 & 26 \\ 0 & 3 & 15 & 63 \end{bmatrix}$$

- In Matrix form:

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} = \begin{bmatrix} 1 & & & \\ 1 & 1 & & \\ 1 & 0 & 1 & \\ 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 2 & 8 & 26 \\ 0 & 3 & 15 & 63 \end{bmatrix}$$

Total Positivity and Neville Elimination – 5

- Neville Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix}$$

Total Positivity and Neville Elimination – 6

- Neville Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 0 & 1 & 7 & 37 \end{bmatrix}$$

Total Positivity and Neville Elimination – 7

- Neville Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 0 & 1 & 5 & 19 \\ 0 & 1 & 7 & 37 \end{bmatrix}$$

- Neville Elimination

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 1 & 5 & 19 \\ 0 & 1 & 7 & 37 \end{bmatrix}$$

Total Positivity and Neville Elimination – 9

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix}$$

Total Positivity and Neville Elimination – 10

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 0 & 1 & 7 & 37 \end{bmatrix}$$

Total Positivity and Neville Elimination – 11

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & 1 & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 0 & 1 & 5 & 19 \\ 0 & 1 & 7 & 37 \end{bmatrix}$$

Total Positivity and Neville Elimination – 12

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & 1 & 1 & \\ & & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 0 & 1 & 5 & 19 \\ 0 & 0 & 2 & 18 \end{bmatrix}$$

Total Positivity and Neville Elimination – 13

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 1 & 5 & 19 \\ 0 & 0 & 2 & 18 \end{bmatrix}$$

Total Positivity and Neville Elimination – 14

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & 1 & 1 & \\ & & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & \mathbf{1} & 1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 0 & \mathbf{2} & \mathbf{12} \\ 0 & 0 & \mathbf{2} & \mathbf{18} \end{bmatrix}$$

Total Positivity and Neville Elimination – 15

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & 1 & 1 & \\ & & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & 1 & 1 \\ & & & \mathbf{1} & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 0 & 2 & 12 \\ 0 & 0 & 0 & \mathbf{6} \end{bmatrix}$$

Total Positivity and Neville Elimination – 16

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & 1 & 1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 2 & \\ & & & 6 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 7 \\ 0 & 0 & 1 & 6 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Total Positivity and Neville Elimination – 17

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & 1 & 1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 2 & \\ & & & 6 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & 1 & 3 & 4 \\ 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & 1 \\ & & & 1 \end{bmatrix}$$

Neville Elimination – We can choose the pattern

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & 1 & 1 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 2 & \\ & & & 6 \end{bmatrix} \begin{bmatrix} 1 & 1 & & \\ & 1 & 2 & \\ & & 1 & 3 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & 1 & \\ & & 1 & 2 \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & 1 \\ & & & 1 \end{bmatrix}$$

Notice: positive bidiagonal decomposition

$$U_{i,i+1}^{(k)} = \frac{\det(A(1:k, i-k+2:i+1))}{\det(A(1:k-1, i-k+2:i))} \cdot \frac{\det(A(1:k-1, i-k+1:i-1))}{\det(A(1:k, i-k+1:i))}.$$

Total Non-Negativity is preserved in Schur complementation

- Neville Elimination yield **Bidiagonal Decomposition** of Any Totally Positive Matrix

$$\mathcal{BD} \left(\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} \right) =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & + & 1 \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & & & \\ & + & 1 & \\ & & + & 1 \\ & & & + \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}$$

- n^2 entries of factors **intrinsically** parametrize all TP matrices
- For numerical computation:
 - Discard A
Instead: Use $\mathcal{BD}(A)$ above
 - Let $A \rightarrow A'$ be a linear transformation
 - * Subtract a row from another, apply Givens rot, etc.
 - We will compute $\mathcal{BD}(A')$ from $\mathcal{BD}(A)$
without losing relative accuracy

Accurate Linear Algebra with $\mathcal{BD}(A)$

- With $\mathcal{BD}(A)$ we can do accurately
 - Positive diagonal scaling – trivial
 - Subtract a row/column from **next** to create a zero
 - Multiply by a positive bidiagonal matrix
- All linear algebra can be “assembled” from above

Subtract a row/column from next to create a zero – 1

- Equivalent to setting an entry in $\mathcal{BD}(A)$ to 0 and doing **no arithmetic**

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} =$$

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 1 & & & \\ & 1 & & \\ & + & 1 & \\ & & + & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ + & 1 & & \\ & + & 1 & \\ & & + & 1 \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}}_{\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}}$$

Subtract a row/column from next to create a zero – 2

- Equivalent to setting an entry in $\mathcal{BD}(A)$ to 0 and doing **no arithmetic**

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & + & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}}_{\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}}$$

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & + & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}$$

Accurate TP×BIDIAGONAL

$$\mathcal{BD} \left(\mathcal{BD}(A) \times \begin{bmatrix} 1 & & & \\ & 1 & & \\ & + & 1 & \\ & & & 1 \end{bmatrix} \right) = ?$$

“Commuting” of positive bidiagonals

- Assume $x, y, z > 0, b_i \geq 0$. Then

$$\begin{bmatrix} 1 & b_1 & & & & \\ & 1 & b_2 & & & \\ & & \cdots & \cdots & & \\ & & & 1 & b_{n-1} & \\ & & & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & & & \\ & 1 & & & & \\ & & y & & & \\ & & x & z & & \\ & & & & 1 & \\ & & & & & 1 \end{bmatrix} = \begin{bmatrix} 1 & & & & & \\ & 1 & & & & \\ & & y' & & & \\ & & x' & z' & & \\ & & & & 1 & \\ & & & & & 1 \end{bmatrix} \begin{bmatrix} 1 & b'_1 & & & & \\ & 1 & b'_2 & & & \\ & & \cdots & \cdots & & \\ & & & 1 & b'_{n-1} & \\ & & & & & 1 \end{bmatrix}$$

where

$$\begin{aligned} x' &= x \\ y' &= y + b_1 x \\ z' &= 1/y' \\ b'_{i-1} &= b_{i-1} y \\ b'_i &= b_i z / y_1 \end{aligned}$$

- No loss of relative accuracy
- “Commute accurately”.

“Commuting” of positive bidiagonals - 2

- Assume $x > 0$, $b_i \geq 0$. Then

$$\begin{bmatrix} 1 & & & & & & \\ b_1 & 1 & & & & & \\ & \dots & \dots & & & & \\ & & b_{n-2} & 1 & & & \\ & & & b_{n-1} & 1 & & \end{bmatrix} \begin{bmatrix} 1 & & & & & & \\ & 1 & & & & & \\ & & 1 & & & & \\ & & x & 1 & & & \\ & & & & 1 & & \\ & & & & & 1 & \\ & & & & & & 1 \end{bmatrix} = \begin{bmatrix} 1 & & & & & & \\ & 1 & & & & & \\ & & 1 & & & & \\ & & & 1 & & & \\ & & & & x' & 1 & \\ & & & & & 1 & \\ & & & & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & & & & \\ b'_1 & 1 & & & & & \\ & \dots & \dots & & & & \\ & & b'_{n-2} & 1 & & & \\ & & & b'_{n-1} & 1 & & \end{bmatrix}$$

where

$$\begin{aligned} x' &= xb_{i+1}/(b_i + x) \\ b'_i &= b_i + x \\ b'_{i+1} &= xb_i/b'_i \end{aligned}$$

- No loss of relative accuracy

Applying similarity to bidiagonal form – 1

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & + & & \\ & & 1 & \\ & & & + 1 \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + 1 \end{bmatrix}$$

Applying similarity to bidiagonal form – 2

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & +1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & +1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix}
 \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & y & \\ & & +z & \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}$$

Applying similarity to bidiagonal form – 3

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & + & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix}
 \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & y' & \\ & & + & z' \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

Applying similarity to bidiagonal form – 4

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & y'' & \\ & & & z'' \end{bmatrix}
 \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

Applying similarity to bidiagonal form – 5

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & + & 1 \\ & & & + & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & + & 1 & \\ & & + & 1 \\ & & & + & 1 \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + & 1 \end{bmatrix}
 \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + & + \end{bmatrix}
 \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 & + \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 & + \end{bmatrix}
 \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 & + \end{bmatrix}$$

Applying similarity to bidiagonal form – 6

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & + 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & + 1 & & \\ & & + 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ + & 1 & & \\ & + 1 & & \\ & & 1 & \\ & & & + 1 \end{bmatrix} \begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix} \begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}$$

DONE!

Accurate Matrix Computations with TP matrices

- $\mathcal{BD}(\text{TP} \times \text{TP})$
 - suffices $\mathcal{BD}(\text{TP} \times \text{BIDIAGONAL})$ – done
- $\mathcal{BD}(\text{diag}(\pm 1) \cdot (\text{TP})^{-1} \cdot \text{diag}(\pm 1))$
 - OK, still a product of positive bidiagonals
- $\mathcal{BD}(\text{SchurComplement})$ – also OK, but barely more complicated
- $\mathcal{BD}(\text{Givens} \cdot TP)$

$$\begin{bmatrix} c & s \\ -s & c \end{bmatrix} = \begin{bmatrix} + & + \\ 0 & + \end{bmatrix} \begin{bmatrix} 1 & 0 \\ - & 1 \end{bmatrix}$$

- SVD is then OK: $\text{TP} \rightarrow \text{BIDIAGONAL} \rightarrow \text{SVD}$
 - no loss of accuracy
- Eigenvalues: First reduce to tridiagonal

Reduction of a nonsymmetric matrix to tridiagonal form – 1

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 2

$$\begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & -\mathbf{1} & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix}$$

$$= \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 3

$$\begin{aligned}
 & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & -\mathbf{1} & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & + & \mathbf{1} \end{bmatrix} \\
 & = \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & + & \mathbf{1} \end{bmatrix}
 \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 4

$$\begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & -\mathbf{1} & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & + & \mathbf{1} \end{bmatrix}$$

$$= \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & \\ & & + & \mathbf{1} \end{bmatrix}$$

$$= \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 5

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 6

$$\begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix}$$
$$= \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 7

$$\begin{aligned} & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\ = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 8

$$\begin{aligned}
 & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\
 = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \\
 & = \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}
 \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 9

$$\begin{aligned}
 & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\
 = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \\
 = & \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ \mathbf{0} & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix}
 \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 10

$$\begin{aligned} & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\ = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \\ = & \begin{bmatrix} + & + & \mathbf{0} & \mathbf{0} \\ + & + & + & + \\ \mathbf{0} & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 11

$$\begin{aligned} & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\ = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \\ = & \begin{bmatrix} + & + & \mathbf{0} & \mathbf{0} \\ + & + & + & + \\ \mathbf{0} & + & + & + \\ \mathbf{0} & \mathbf{0} & + & + \end{bmatrix} \end{aligned}$$

Reduction of a nonsymmetric matrix to tridiagonal form – 12

$$\begin{aligned} & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & - \\ & & & \mathbf{1} \end{bmatrix} \\ \\ = & \begin{bmatrix} \mathbf{1} & & & \\ & \mathbf{1} & & \\ & & \mathbf{1} & + \\ & & & \mathbf{1} \end{bmatrix} \begin{bmatrix} + & + & + & \mathbf{0} \\ + & + & + & + \\ + & + & + & + \\ \mathbf{0} & + & + & + \end{bmatrix} \\ \\ & = \begin{bmatrix} + & + & \mathbf{0} & \mathbf{0} \\ + & + & + & \mathbf{0} \\ \mathbf{0} & + & + & + \\ \mathbf{0} & \mathbf{0} & + & + \end{bmatrix} \end{aligned}$$

Accurate Eigenvalues of Tridiagonals

The eigenvalues of

$$\begin{bmatrix} 1 & & & & & \\ b_1 & 1 & & & & \\ & \dots & \dots & & & \\ & & b_{n-2} & 1 & & \\ & & & b_{n-1} & 1 & \end{bmatrix} \begin{bmatrix} d_1 & & & & & \\ & d_2 & & & & \\ & & \dots & & & \\ & & & d_{n-1} & & \\ & & & & d_n & \end{bmatrix} \begin{bmatrix} 1 & c_1 & & & & \\ & 1 & c_2 & & & \\ & & \dots & \dots & & \\ & & & 1 & c_{n-1} & \\ & & & & 1 & \end{bmatrix}$$

are the squares of the singular values of

$$\begin{bmatrix} \sqrt{d_1} & \sqrt{d_1 b_1 c_1} & & & & \\ & \sqrt{d_2} & \sqrt{d_2 b_2 c_2} & & & \\ & & \dots & \dots & & \\ & & & \sqrt{d_{n-1}} & \sqrt{d_{n-1} b_{n-1} c_{n-1}} & \\ & & & & \sqrt{d_n} & \end{bmatrix}$$

- Thus the eigenvalue problem of a TP matrix is reduced to the SVD problem for a bidiagonal matrix
- Forward stable
- Total cost = $8n^3$.

Computing Accurate $\mathcal{BD}(A)$

- Bidiagonal decompositions intrinsic for all TP matrices
- Often accurate formulas:

– Vandermonde:

$$\prod_{j=1}^{i-1} (x_i - x_j), \quad \prod_{j=i-k+1}^{i-1} \frac{x_{i+1} - x_{j+1}}{x_i - x_j}, \quad x_{i+n-k}$$

– Cauchy:

$$\prod_{k=1}^{n-1} \frac{(x_n - x_k)(y_n - y_k)}{(x_n + y_k)(y_n + x_k)}, \quad \frac{x_{n-k} + y_{i-n+k+1}}{x_i + y_{i-n+k+1}} \prod_{l=n-k}^{i-1} \frac{x_{i+1} - x_{l+1}}{x_i - x_l} \cdot \prod_{l=1}^{i-n+k-1} \frac{x_i + y_l}{x_{i+1} + y_l}$$

$$\frac{y_{n-k} + x_{i-n+k+1}}{y_i + x_{i-n+k+1}} \prod_{l=n-k}^{i-1} \frac{y_{i+1} - y_{l+1}}{y_i - y_l} \cdot \prod_{l=1}^{i-n+k-1} \frac{y_i + x_l}{y_{i+1} + x_l},$$

- Equivalent to Björck-Pereyra Methods
- Can be computed iff **initial minors** can be computed accurately

$$U_{i,i+1}^{(k)} = \frac{\det(A(1 : k, i - k + 2 : i + 1))}{\det(A(1 : k - 1, i - k + 2 : i))} \cdot \frac{\det(A(1 : k - 1, i - k + 1 : i - 1))}{\det(A(1 : k, i - k + 1 : i))}.$$

initial minors = contiguous, include the first row or column

Conclusions

- $O(n^3)$ algorithms for the eigenvalues and the SVD of Totally Positive Matrices to high relative accuracy
- Accurate linear algebra with TP matrices closed under same operations as TP
- Applies to:
 - Oscillatory
 - Totally Nonnegative Matrices
 - Sign Regular (Inverses of TP matrices)
- Open problem: If we perturb an entry in
 $(\text{PositiveBidiagonal}) \cdot (\text{PositiveBidiagonal})$
How much do the singular vectors change?
Singular values provably change by at most ϵ .
- This talk and Matlab software: math.mit.edu/~plamen