

# Accurate and Efficient Matrix Computations with Totally Positive Matrices

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# Goals

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

- Linear Equation Solving, Inversion
- Eigenvalues, Eigenvectors
- Singular Values, Singular Vectors
- Schur Complementation, Multiplication

- **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**

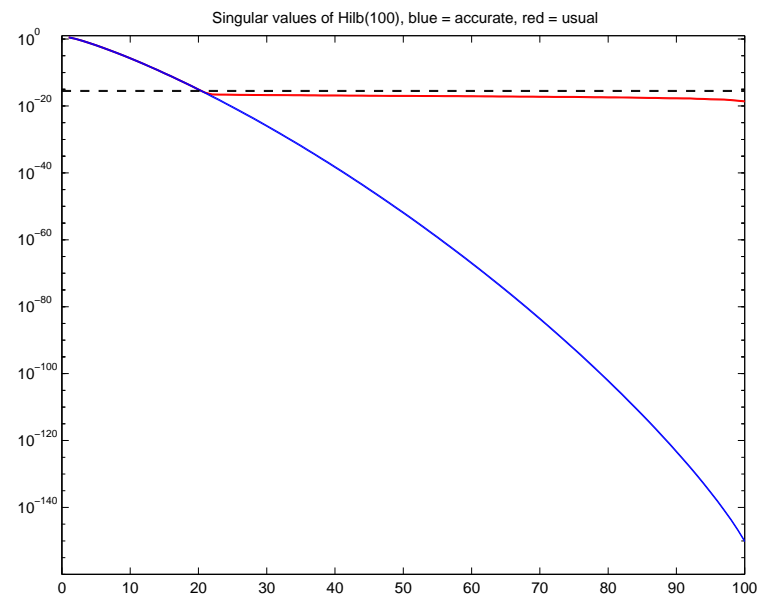
$$|\hat{\lambda}_i - \lambda_i| \leq O(\epsilon)|\lambda_i|, \quad |\hat{\sigma}_i - \sigma_i| \leq O(\epsilon)\sigma_i,$$

- 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: $100 \times 100$ Hilbert Matrix $H = 1/(i + j)$

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## How do we lose relative accuracy in floating point?

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- **MODEL:**  $\text{fl}(a \otimes b) = (a \otimes b)(1 + \delta)$ , no under/overflow
- **ACCURATE:**
  - Products, Quotients, Sums of 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)

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- **POSSIBLE LOSS OF ACCURACY:**
  - **Subtractive Cancellation** when subtracting approximate results:  
$$\begin{array}{r} .12345\mathbf{xxx} \\ - .12345\mathbf{yyy} \\ \hline .00000\mathbf{zzz} \end{array}$$

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 $.12345(1 + \delta_1) - .12345(1 + \delta_2) = .12345(\delta_1 - \delta_2) - \text{BAD}$   
 $.12345(1 + \delta_1) + .12345(1 + \delta_2) = .23490(1 + (\delta_1 + \delta_2)/2) - \text{OK}$

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- **TRICK:** Only compute:
  - $1/(1 + xy)$ ;  $1/x$ ;  $xy$ ;  $1/(x - y)$ ;  $x, y > 0$
  - **never subtract** approximate quantities

# Bidiagonal Matrices

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- 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  $\sigma_i$  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)

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- Def:  $A$  is TP if all minors  $> 0$ :

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

- 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)
- $\text{diag}(\pm 1) \cdot (\text{TP})^{-1} \cdot \text{diag}(\pm 1)$
- 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

# Structure of Totally Positive Matrices

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

- A lot simpler than it seems
- Matrix entries – bad choice of parameters ( $4^n$  conditions)
- Best represented as products of positive bidiagonals

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

$$\underbrace{\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & + & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & + & 1 & \\ & & + & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & + & 1 & \\ & & + & 1 \\ & & & + & 1 \end{bmatrix}}_{L} \underbrace{\begin{bmatrix} + & & & \\ & + & & \\ & & + & \\ & & & + \end{bmatrix}}_{D} \underbrace{\begin{bmatrix} 1 & + & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & + & \\ & & 1 & + \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & + \\ & & & 1 \end{bmatrix}}_{U}$$

- $n^2$  independent positive parameters

# Total Positivity and Neville Elimination – 1

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- 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 – 2

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- 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 – 3

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- Neville Elimination

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

## Total Positivity and Neville Elimination – 4

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- 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 – 5

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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 – 6

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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 – 7

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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 – 8

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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 – 9

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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 – 10

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Neville Elimination – We can choose the pattern

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

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

## Total Positivity and Neville Elimination – 11

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Neville Elimination – We can choose the pattern

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

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

## Total Positivity and Neville Elimination – 12

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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 – 13

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Neville Elimination – We can choose the pattern

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

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

## Total Positivity and Neville Elimination – 14

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Neville Elimination – We can choose the pattern

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

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

Notice: positive bidiagonal decomposition

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



- **Bidiagonal Decomposition** of Any TP Matrix

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

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

- $n^2$  entries of factors **intrinsically** parametrize all TP matrices
- For numerical computation:
  - Discard  $A$   
Instead: Use  $\mathcal{BD}(A)$
  - 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 $\mathcal{BD}(A)$

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- 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 for eigenvalues, SVD, etc. 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

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

## “Commuting” of positive bidiagonals

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- 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

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



## 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

---

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

# 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

---

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

# Applying similarity to bidiagonal form – 6

---

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

DONE!

## Accurate Matrix Computations with TP matrices

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- $\mathcal{BD}(\text{TP} \times \text{TP})$ 
  - 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

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

# Reduction of a nonsymmetric matrix to tridiagonal form – 9

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$$\begin{aligned} & \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & 0 \\ + & + & + & + \\ + & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \\ & \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & 0 & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & 0 \\ 0 & + & + & + \\ 0 & 0 & + & + \end{bmatrix} \end{aligned}$$

# Reduction of a nonsymmetric matrix to tridiagonal form – 10

$$\begin{aligned}
 & \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ + & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & + \\ + & + & + & + \\ + & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & 0 \\ + & + & + & + \\ + & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & + & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \\
 & \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & + & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & + \\ 0 & + & + & + \\ 0 & 0 & + & + \end{bmatrix} \Rightarrow \begin{bmatrix} + & + & 0 & 0 \\ + & + & + & 0 \\ 0 & + & + & + \\ 0 & 0 & + & + \end{bmatrix}
 \end{aligned}$$

... Obtaining  $\mathcal{BD}(\text{TP Tridiagonal}) =$

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



## Accurate Eigenvalues of Tridiagonals

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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 $\mathcal{BD}(A)$

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

– Vandermonde  $V = x_i^{j-1}$ :

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

– Cauchy  $C = 1/(x_i + y_j)$ :

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

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

- 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

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- $O(n^3)$  algorithms for the eigenvalues and the SVD of (*unsymmetric*) TP Matrices to high relative accuracy
- Accurate linear algebra with TP matrices closed under same operations as TP
- Applies to:
  - Oscillatory
  - Totally Nonnegative
  - Sign Regular (Inverses of TP)
- 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  $\epsilon$ .
- This talk and Matlab software: [math.mit.edu/~plamen](http://math.mit.edu/~plamen)