

# Fast algorithms for polynomial-Vandermonde matrices related to quasiseparable matrices

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## Outline

### Fast algorithms for Vandermonde matrices

- ▶▶▶ Classical Björck-Pereyra & Traub algorithms
- ▶▶▶ Polynomial-Vandermonde matrices and previous work

### A Björck-Pereyra algorithm for Szegő-Vandermonde matrices

- ▶▶▶ Extension to Szegő-Vandermonde matrices
- ▶▶▶ Computational speedup
- ▶▶▶ Numerical illustrations

### Fast algorithms for quasiseparable-Vandermonde matrices

- ▶▶▶ Quasiseparable matrices & special cases
- ▶▶▶ Björck-Pereyra-like algorithm for quasiseparable matrices
- ▶▶▶ Recurrence relations
- ▶▶▶ Traub-like algorithm for quasiseparable matrices

## Classical Vandermonde matrices

► **Definition.** For a set of nodes, a **Vandermonde matrix** is defined by

$$x = \{x_1, x_2, \dots, x_n\}$$
$$\Downarrow$$
$$V(x) = \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ 1 & x_3 & x_3^2 & \dots & x_3^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{bmatrix}$$

► **Note.** The  $n^2$  entries of  $V(x)$  are defined by only  $n$  parameters. This allows the design of **fast algorithms**.

## Fast Björck-Pereyra algorithm

► The **Björck-Pereyra algorithm** (1970) is based on the formula

$$V(x)^{-1} = U_1^{-1} \dots U_{n-1}^{-1} L_{n-1}^{-1} \dots L_1^{-1},$$

with

$$U_k^{-1} = \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & -x_k & \\ & & 1 & \ddots \\ & & & \ddots & -x_k \\ & & & & 1 \end{array} \right]$$

$$L_k^{-1} = \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & & \frac{1}{x_{k+1}-x_k} & \\ & & & \ddots \\ & & & & \frac{1}{x_n-x_k} \end{array} \right] \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & -1 & 1 & \\ & \vdots & & \ddots \\ & -1 & & 1 \end{array} \right]$$

## Fast Björck-Pereyra algorithm

- ▣ The solution  $a$  of the linear system

$$V(x)a = f$$

is computed by Björck-Pereyra as

$$a = V(x)^{-1}f = U_1^{-1} \dots U_{n-1}^{-1} L_{n-1}^{-1} \dots L_1^{-1} f$$

- ▣ Each matrix in the factorization is **sparse**, and so each matrix-vector product can be computed in  $\mathcal{O}(n)$  operations.
- ▣ **Björck-Pereyra** requires only  $\mathcal{O}(n^2)$  arithmetic operations vs  $\mathcal{O}(n^3)$  of Gaussian elimination.

## Fast Traub algorithm

► The **Traub algorithm** (1966) is based on the formula

$$V(x)^{-1} = \tilde{I} \cdot \begin{bmatrix} r_0(x_1) & r_1(x_1) & r_2(x_1) & \cdots & r_{n-1}(x_1) \\ r_0(x_2) & r_1(x_2) & r_2(x_2) & \cdots & r_{n-1}(x_2) \\ r_0(x_3) & r_1(x_3) & r_2(x_3) & \cdots & r_{n-1}(x_3) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ r_0(x_n) & r_1(x_n) & r_2(x_n) & \cdots & r_{n-1}(x_n) \end{bmatrix}^T \cdot \text{diag}(c_1, c_2, \dots, c_n)$$

where  $\tilde{I}$  is the antidiagonal matrix,  $c_k = \prod_{\substack{j=1 \\ j \neq k}}^n (x_j - x_k)^{-1}$

► The polynomials  $\{r_0(x), \dots, r_{n-1}(x)\}$  are the **Horner (associated) polynomials**, and satisfy **two-term recurrence relations**

$$r_0(x) = \beta_n, \quad r_k(x) = xr_{k-1}(x) + \beta_{n-k}$$

► **Fast:** requires only  $\mathcal{O}(n^2)$  arithmetic operations vs  $\mathcal{O}(n^3)$  of Gaussian elimination.

## Conditioning of Vandermonde matrices

- ▶▶▶ The condition numbers of Vandermonde matrices **grow exponentially with their size** (Tyrtysnikov (1994)).
- ▶▶▶ Björck-Pereyra (1970) : “... *some problems, connected with Vandermonde systems, which traditionally have been considered to be too ill-conditioned to be attacked, actually can be solved with good precision*”.

## Some Numerical Experiments

### ▣▣▣▣ Björck-Pereyra algorithm. (Higham's example)

1. Nodes chosen randomly in  $(0, 1)$ .
2. RHS alternating signs,  $\left[ 1 \quad -1 \quad 1 \quad \cdots \right]^T$
3. Forward error measured by

$$e = \frac{\|x - \hat{x}\|_2}{\|x\|_2}$$

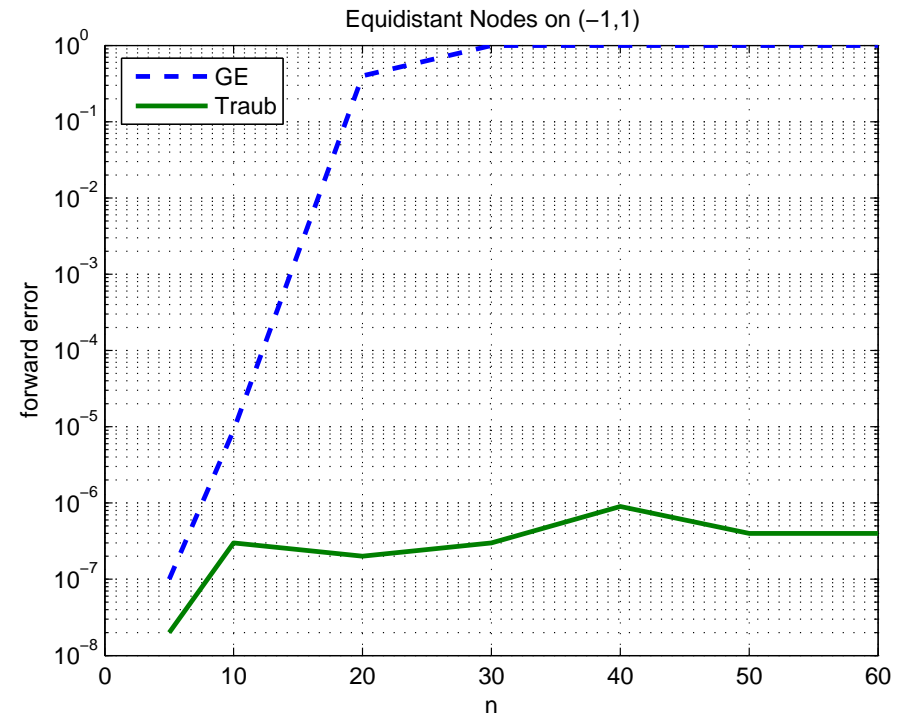
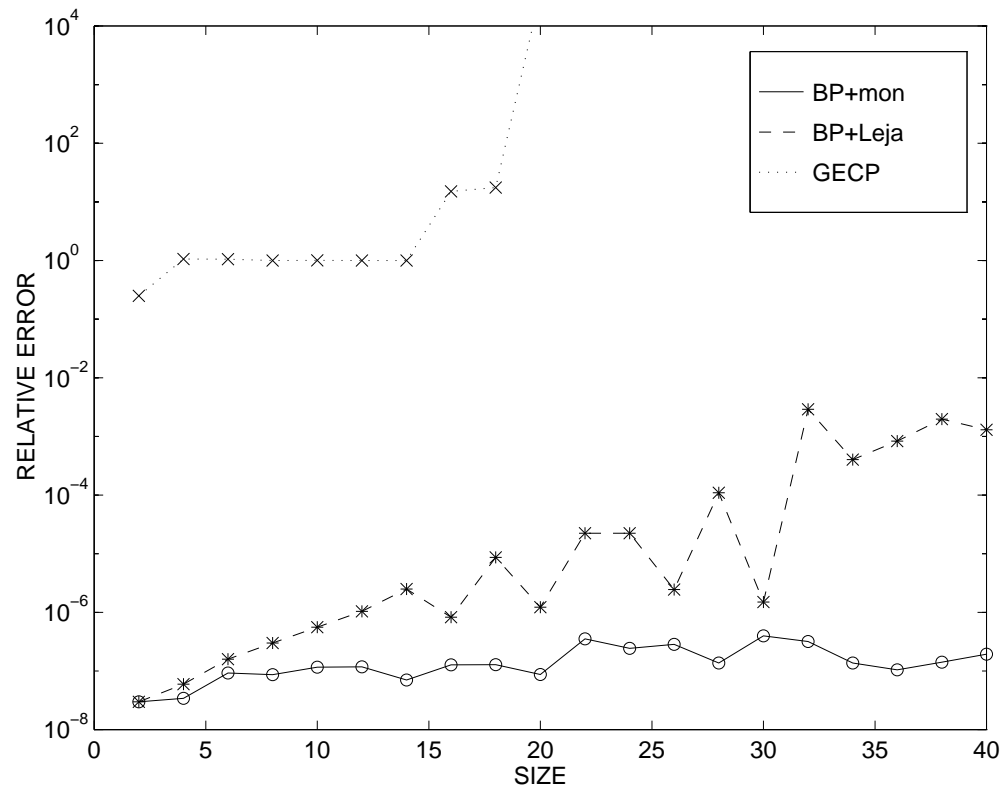
### ▣▣▣▣ Traub algorithm.

1. Nodes chosen randomly in  $(0, 1)$ .
2. Forward error measured by

$$e = \frac{\|A^{-1} - \widehat{A}^{-1}\|_2}{\|A^{-1}\|_2}$$

# Numerical Experiments

## Classical Björck-Pereyra & Traub Algorithms



## Error Analysis

### Björck-Pereyra Algorithm

- ▣▣▣▣ **Higham** (1990) showed that under some conditions (sign-oscillating right-hand-side and monotonic ordering of the nodes, all of which are positive), there is a **provably excellent forward error bound**:

$$\frac{|a - \hat{a}|}{|a|} \leq 5nu + \mathcal{O}(u^2)$$

- ▣▣▣▣ **Conclusion:** Very good forward error is to be expected under some conditions.

## Polynomial-Vandermonde matrices

►►► **Definition.** For sets of polynomials and nodes, define a **polynomial-Vandermonde matrix**:

$$\begin{aligned}
 x &= \{x_1, x_2, \dots, x_n\} \\
 R &= \{r_0(x), r_1(x), \dots, r_{n-1}(x)\} \\
 &\quad \Downarrow \\
 V_R(x) &= \begin{bmatrix} r_0(x_1) & r_1(x_1) & r_2(x_1) & \cdots & r_{n-1}(x_1) \\ r_0(x_2) & r_1(x_2) & r_2(x_2) & \cdots & r_{n-1}(x_2) \\ r_0(x_3) & r_1(x_3) & r_2(x_3) & \cdots & r_{n-1}(x_3) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ r_0(x_n) & r_1(x_n) & r_2(x_n) & \cdots & r_{n-1}(x_n) \end{bmatrix}
 \end{aligned}$$

►►► **Note.** If the polynomial system  $R$  satisfies nice recurrence relations, then the  $n^2$  entries of  $V_R(x)$  can be defined by only  $\mathcal{O}(n)$  parameters.

## Fast algorithms for polynomial-Vandermonde matrices

### Previous work

polynomial-Vandermonde matrix	Björck-Pereyra-type	Traub-type
Vandermonde matrices	Björck-Pereyra (1970)	Traub (1966)
Vandermonde-like matrices	Kailath-Olshevsky (1996)	Gohberg-Olshevsky (1996)
block Vandermonde matrices	Tang-Golub (1981)	
Chebyshev-Vandermonde matrices	Reichel-Opfer (1991)	Gohberg-Olshevsky (1994)
three-term Vandermonde matrices	Higham (1988,90)	Calvetti-Reichel (1993)
Szegö-Vandermonde matrices	<b>BEGKO (2006)</b>	Olshevsky (2001)

# A Björck-Pereyra algorithm for Szegő-Vandermonde matrices

Joint work with Y.Eidelman, I.Gohberg, I.Koltracht, V.Olshevsky

## Szegő-Vandermonde matrices

- ▣ If the polynomials are Szëgo polynomials, the resulting matrix is a Szëgo-Vandermonde matrix.

$$V_{\Phi} = \begin{bmatrix} \phi_0(x_1) & \phi_1(x_1) & \phi_2(x_1) & \cdots & \phi_{n-1}(x_1) \\ \phi_0(x_2) & \phi_1(x_2) & \phi_2(x_2) & \cdots & \phi_{n-1}(x_2) \\ \phi_0(x_3) & \phi_1(x_3) & \phi_2(x_3) & \cdots & \phi_{n-1}(x_3) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \phi_0(x_n) & \phi_1(x_n) & \phi_2(x_n) & \cdots & \phi_{n-1}(x_n) \end{bmatrix}$$

## Björck-Pereyra-like Formula

➡ based on the formula

$$V_{\Phi}^{-1} = U_1^{-1} \dots U_{n-1}^{-1} L_{n-1}^{-1} \dots L_1^{-1},$$

with

$$U_k^{-1} = \text{diag} \left\{ I_{k-1}, \begin{bmatrix} \frac{1}{\alpha_0} & & & \\ 0 & \boxed{H - x_k I} & & \\ \vdots & & \ddots & \\ 0 & & & \frac{1}{\alpha_{n-k}} \end{bmatrix} \right\}$$

$$L_k^{-1} = \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & & \frac{1}{x_{k+1}-x_k} & \\ & & & \ddots \\ & & & & \frac{1}{x_n-x_k} \end{array} \right] \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & -1 & 1 & \\ & \vdots & & \ddots \\ & -1 & & 1 \end{array} \right]$$

## Unitary Hessenberg Matrices

$$H = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

- ➡ For monomials (classical BP), the algorithm is **fast** because the  $H$ 's are **bidagonal**.
- ➡ For the Szegő case, the algorithm is **fast** using a **factorization** of  $H$  into **Givens rotations**.



## Numerical Illustrations

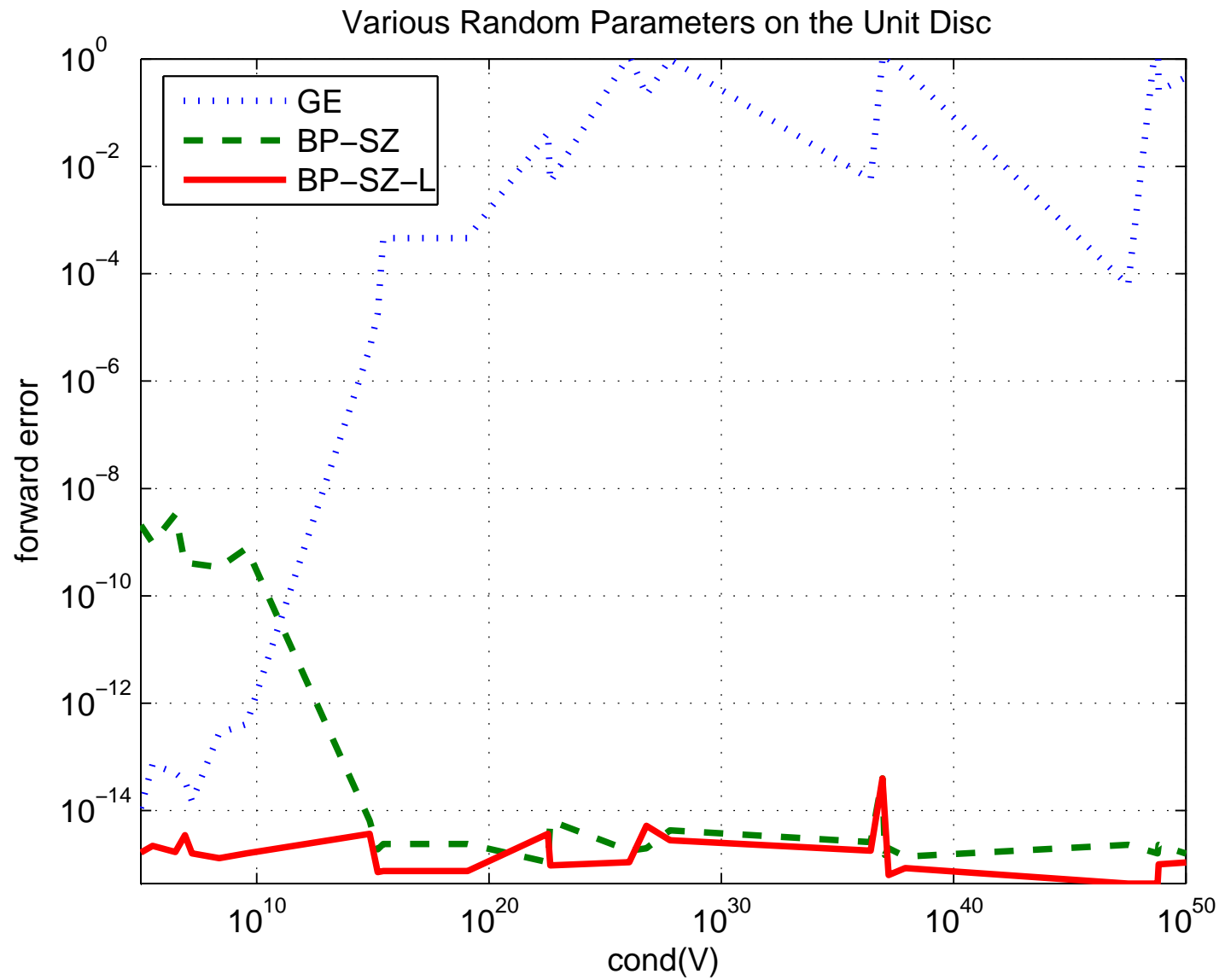
- ▶ We compare the **forward error** of the solutions  $\hat{x}$  from MATLAB in double precision via

$$e = \frac{\|x - \hat{x}\|_2}{\|x\|_2},$$

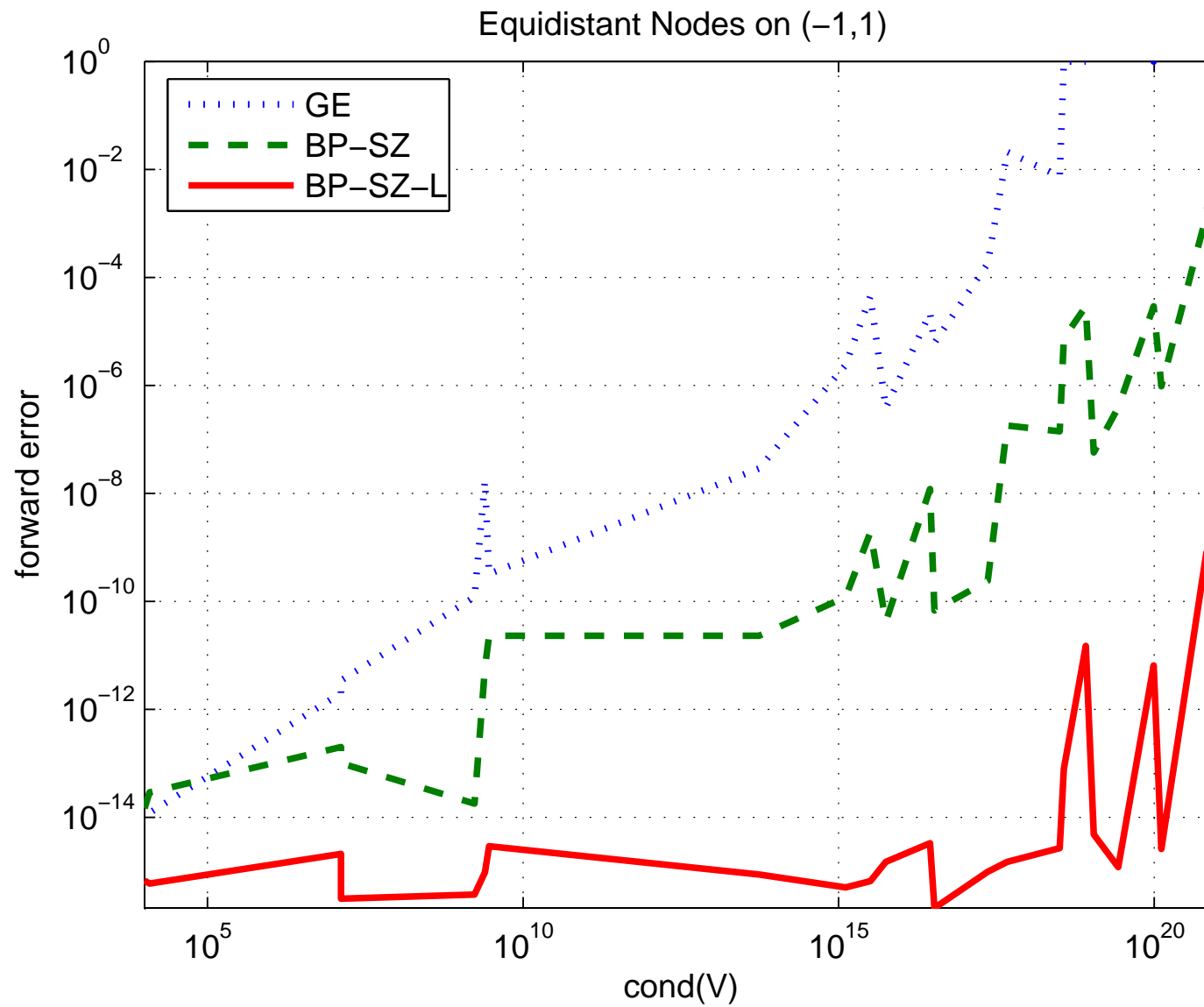
with  $x$ , the “exact” solution using MATLAB’s `vpa ( )` command for software-implemented arbitrary digit arithmetic.

- ▶ **GE** - Gaussian elimination via MATLAB’s backslash command.
- ▶ **BP-SZ** - Björck-Pereyra-like algorithm.
- ▶ **BP-SZ-L** - Björck-Pereyra-like algorithm with nodes ordered via the **Leja ordering**. (Reichel, Higham)
- ▶ Use of the **Leja ordering** exploits the Vandermonde structure to produce a partial pivoting ordering of the rows of  $V$ . This is at a cost of  $\mathcal{O}(n^2)$ .

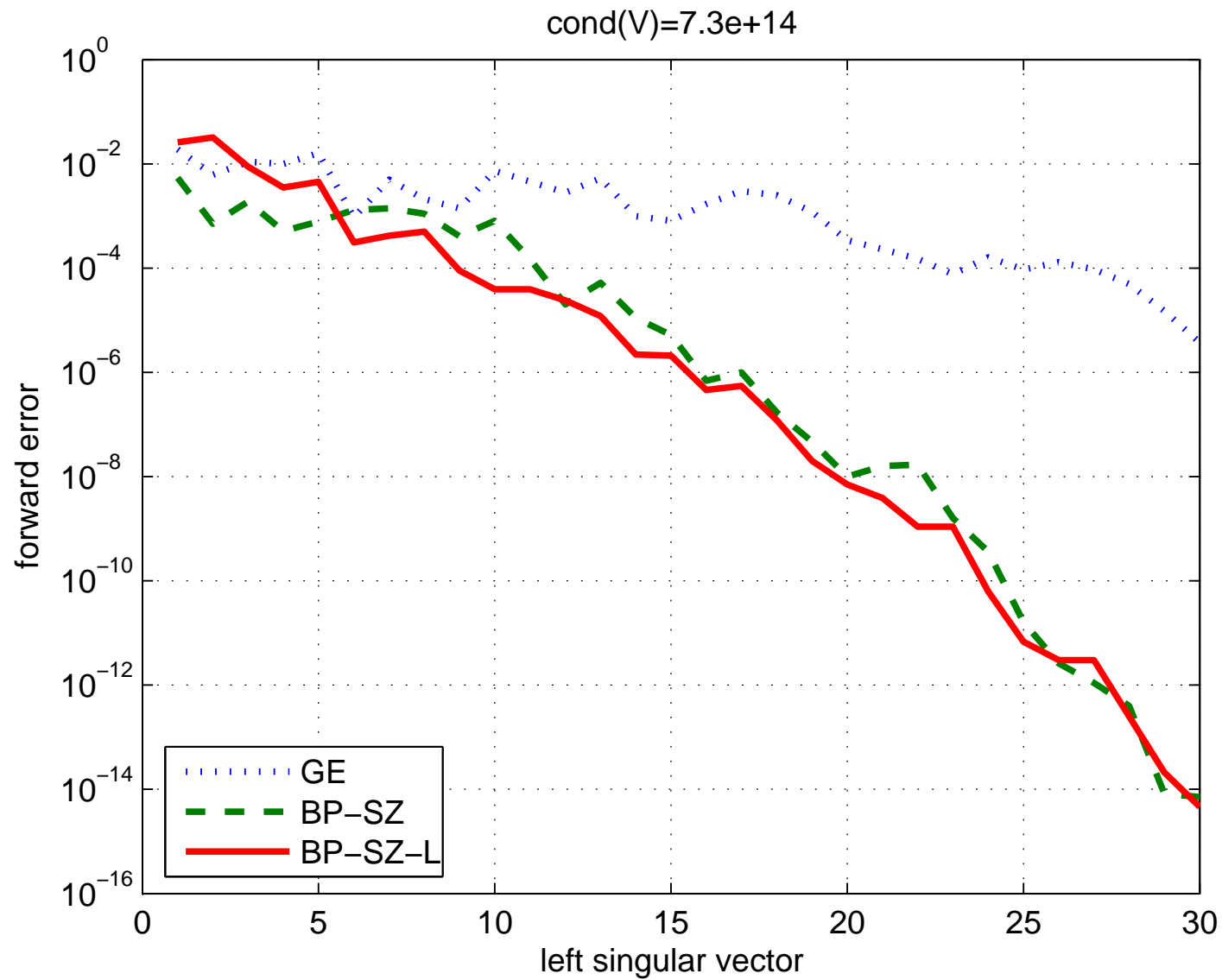
## Numerical Illustrations - Experiment 1



## Numerical Illustrations - Experiment 2



## Numerical Illustrations - Experiment 3



# Fast algorithms for quasiseparable-Vandermonde matrices

Joint work with Y.Eidelman, I.Gohberg, I.Koltracht, V.Olshevsky

## Fast algorithms for polynomial-Vandermonde matrices

### Previous work

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Quasiseparable-Vandermonde	??????????????????	??????????????????

## Quasiseparable Matrices

▣▣▣ **Definition.** A matrix  $C$  is **quasiseparable of order one** if

$$\max \text{Rank} C_{12} = \max \text{Rank} C_{21} = 1$$

where the maxima are taken over all symmetric partitions of the form

$$C = \left[ \begin{array}{c|c} * & C_{12} \\ \hline C_{21} & * \end{array} \right]$$

▣▣▣ **Previous Work.** Gohberg-Kaashoek-Lerer, Dewilde, Gohberg-Eidelman, Van Barel et al, Tyrtyshnikov et al, Bini et al, Gu-Chandrasekaran et al.

▣▣▣ Henceforth we consider only **Hessenberg** matrices  $C$ .

## Important Special Cases

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

▣▶ The system of polynomials  $r_k(x) = \det(xI - C_{k \times k})$  associated with  $C$  is the **monomials** with recurrence relations

$$r_k(x) = xr_{k-1}(x)$$

▣▶ The matrix  $C$  is **quasiseparable**.

## Important Special Cases

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

## Important Special Cases

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

## Important Special Cases

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

## Important Special Cases

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

## Important Special Cases

### Tridiagonal

$$C = \begin{bmatrix} d_1 & g_1 & 0 & 0 & 0 \\ q_1 & d_2 & g_2 & 0 & 0 \\ 0 & q_2 & d_3 & g_3 & 0 \\ 0 & 0 & q_3 & d_4 & g_4 \\ 0 & 0 & 0 & q_4 & d_5 \end{bmatrix}$$

- The system of polynomials  $r_k(x) = \det(xI - C_{k \times k})$  associated with  $C$  are **real orthogonal polynomials** with recurrence relations

$$r_k(x) = \frac{1}{q_k} (x - d_k) r_{k-1}(x) - \frac{g_{k-1}}{q_k} r_{k-2}(x)$$

- The matrix  $C$  is **quasiseparable**.

## Important Special Cases

### Tridiagonal

$$C = \begin{bmatrix} d_1 & g_1 & 0 & 0 & 0 \\ q_1 & d_2 & g_2 & 0 & 0 \\ 0 & q_2 & d_3 & g_3 & 0 \\ 0 & 0 & q_3 & d_4 & g_4 \\ 0 & 0 & 0 & q_4 & d_5 \end{bmatrix}$$

## Important Special Cases

### Tridiagonal

$$C = \begin{bmatrix} d_1 & g_1 & 0 & 0 & 0 \\ q_1 & d_2 & g_2 & 0 & 0 \\ 0 & q_2 & d_3 & g_3 & 0 \\ 0 & 0 & q_3 & d_4 & g_4 \\ 0 & 0 & 0 & q_4 & d_5 \end{bmatrix}$$

## Important Special Cases

Tridiagonal

$$C = \begin{bmatrix} d_1 & g_1 & 0 & 0 & 0 \\ q_1 & d_2 & g_2 & 0 & 0 \\ 0 & q_2 & d_3 & g_3 & 0 \\ 0 & 0 & q_3 & d_4 & g_4 \\ 0 & 0 & 0 & q_4 & d_5 \end{bmatrix}$$

## Important Special Cases

Tridiagonal

$$C = \begin{bmatrix} d_1 & g_1 & 0 & 0 & 0 \\ q_1 & d_2 & g_2 & 0 & 0 \\ 0 & q_2 & d_3 & g_3 & 0 \\ 0 & 0 & q_3 & d_4 & g_4 \\ 0 & 0 & 0 & q_4 & d_5 \end{bmatrix}$$

## Important Special Cases

### Unitary Hessenberg

$$C = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

► The system of polynomials  $r_k(x) = \det(xI - C_{k \times k})$  associated with  $C$  are the **Szegő polynomials** with recurrence relations

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{\mu_k} \begin{bmatrix} 1 & -\rho_k^* \\ -\rho_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ xr_{k-1}(x) \end{bmatrix}$$

► The matrix  $C$  is **quasiseparable**.

## Important Special Cases

### Unitary Hessenberg

$$C = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

## Important Special Cases

### Unitary Hessenberg

$$C = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

## Important Special Cases

### Unitary Hessenberg

$$C = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

## Important Special Cases

### Unitary Hessenberg

$$C = \begin{bmatrix} -\rho_0^* \rho_1 & -\rho_0^* \mu_1 \rho_2 & -\rho_0^* \mu_1 \mu_2 \rho_3 & -\rho_0^* \mu_1 \mu_2 \mu_3 \rho_4 & -\rho_0^* \mu_1 \mu_2 \mu_3 \mu_4 \rho_5 \\ \mu_1 & -\rho_1^* \rho_2 & -\rho_1^* \mu_2 \rho_3 & -\rho_1^* \mu_2 \mu_3 \rho_4 & -\rho_1^* \mu_2 \mu_3 \mu_4 \rho_5 \\ 0 & \mu_2 & -\rho_2^* \rho_3 & -\rho_2^* \mu_3 \rho_4 & -\rho_2^* \mu_3 \mu_4 \rho_5 \\ 0 & 0 & \mu_3 & -\rho_3^* \rho_4 & -\rho_3^* \mu_4 \rho_5 \\ 0 & 0 & 0 & \mu_4 & -\rho_4^* \rho_5 \end{bmatrix}$$

## Quasiseparable-Vandermonde matrices

► Definition. A **Quasiseparable-Vandermonde matrix** is of the form

$$V_R = \begin{bmatrix} r_0(x_1) & r_1(x_1) & r_2(x_1) & \cdots & r_{n-1}(x_1) \\ r_0(x_2) & r_1(x_2) & r_2(x_2) & \cdots & r_{n-1}(x_2) \\ r_0(x_3) & r_1(x_3) & r_2(x_3) & \cdots & r_{n-1}(x_3) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ r_0(x_n) & r_1(x_n) & r_2(x_n) & \cdots & r_{n-1}(x_n) \end{bmatrix}$$

where the polynomials  $r_k(x) = \det(xI - C_{k \times k})$  correspond to an **order-one quasiseparable Hessenberg** matrix  $C$ .

## Björck-Pereyra-like algorithm for quasiseparable-Vandermonde matrices

▣ based on the formula

$$V_R^{-1} = U_1^{-1} \cdots U_{n-1}^{-1} L_{n-1}^{-1} \cdots L_1^{-1},$$

with

$$U_k^{-1} = \text{diag} \left\{ I_{k-1}, \begin{bmatrix} \frac{1}{\alpha_0} & & & \\ 0 & \boxed{C - x_k I} & & \\ \vdots & & \ddots & \\ 0 & & & \frac{1}{\alpha_{n-k}} \end{bmatrix} \right\}$$

$$L_k^{-1} = \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & & \frac{1}{x_{k+1}-x_k} & \\ & & & \ddots \\ & & & & \frac{1}{x_n-x_k} \end{array} \right] \left[ \begin{array}{c|ccc} I_{k-1} & & & \\ \hline & 1 & & \\ & -1 & 1 & \\ & \vdots & & \ddots \\ & -1 & & & 1 \end{array} \right]$$

## Complexity of the Björck-Pereyra-like algorithm

- ▣ To design a fast algorithm, we need **fast multiplication of  $C$  by a vector**.  
In the **monomial** case,  $C$  is **bidiagonal**.  
In the **Szegö** case,  $C$  is **unitary Hessenberg**, and hence admits a convenient **factorization**.
- ▣ In the **quasiseparable** case,  $C$  can be multiplied by a vector in  $\mathcal{O}(n)$  operations using an algorithm of **Eidelman and Gohberg** (1999).
- ▣ Thus the cost of the algorithm is  $\mathcal{O}(n^2)$  arithmetic operations.

## Special cases of the Björck-Pereyra-like algorithm

- ▶▶▶ **Tridiagonal** case: This algorithm reduces to the **Higham algorithm** (1990).
  - **Stage I** corresponds to computing  $L_{n-1}^{-1} \dots L_1^{-1} f = x$ .
  - **Stage II** corresponds to computing  $U_1^{-1} \dots U_{n-1}^{-1} x$ .
- ▶▶▶ **Unitary Hessenberg** case: This reduces to the **BEGKO algorithm** (2006).

## Numerical Illustrations - Björck-Pereyra

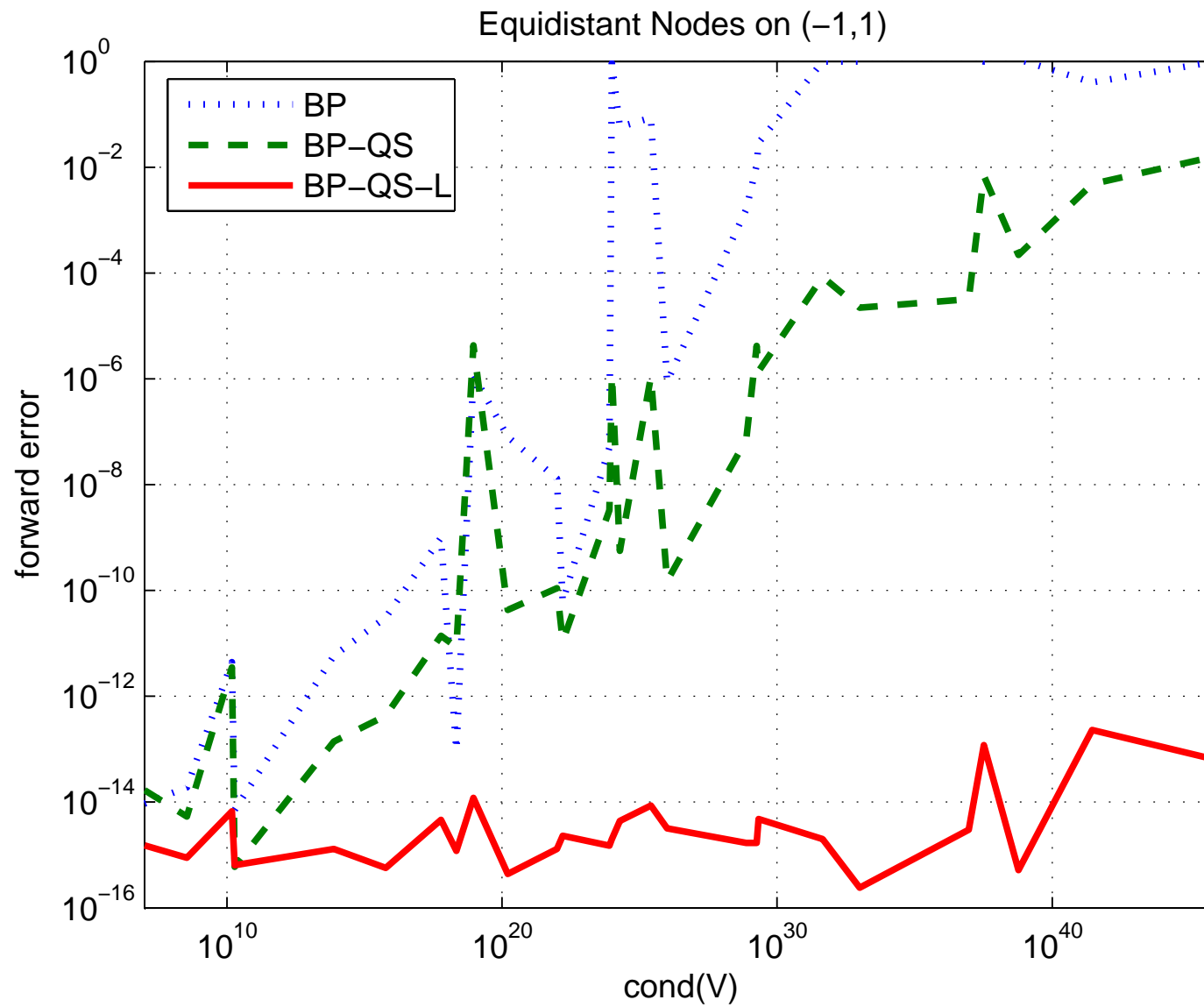
- ▶▶▶ We compare the **forward error** of the solutions  $\hat{x}$  from MATLAB in double precision via

$$e = \frac{\|x - \hat{x}\|_2}{\|x\|_2},$$

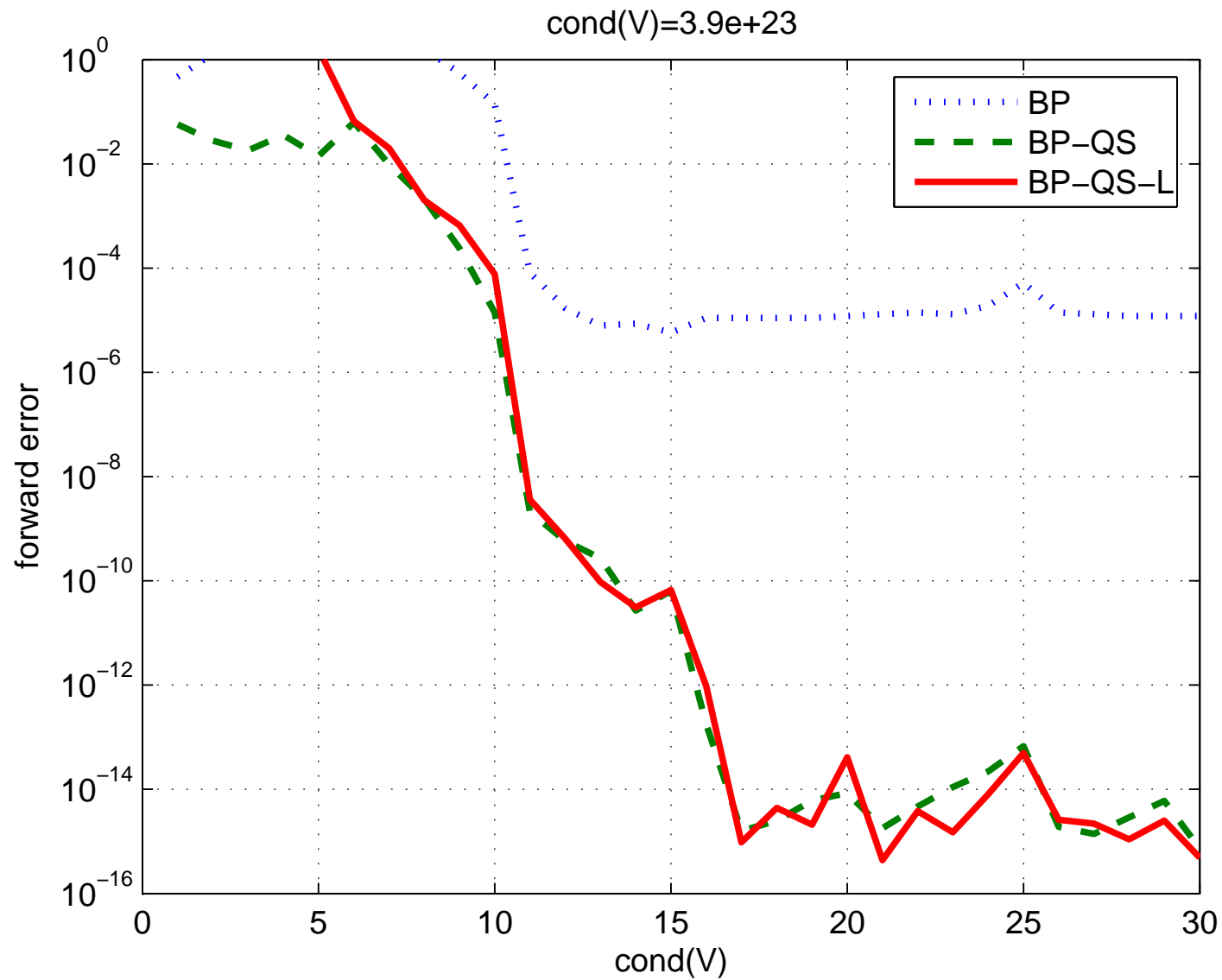
with  $x$ , the “exact” solution using MATLAB’s `vpa ( )` command for software-implemented arbitrary digit arithmetic.

- ▶▶▶ **GE** - Gaussian elimination via MATLAB’s backslash command.
- ▶▶▶ **BP-QS** - Björck-Pereyra-like algorithm.
- ▶▶▶ **BP-QS-L** - Björck-Pereyra-like algorithm with nodes ordered via the **Leja ordering**. (Reichel, Higham)

## Numerical Illustrations - BP Experiment 1



## Numerical Illustrations - BP Experiment 2



## Fast algorithms for polynomial-Vandermonde matrices

### Previous work

polynomial-Vandermonde matrix	Björck-Pereyra-type	Traub-type
Vandermonde matrices	Björck-Pereyra(1970)	Traub (1966)
Vandermonde-like matrices	Kailath-Olshevsky (1996)	Gohberg-Olshevsky (1996)
block Vandermonde matrices	Tang-Golub (1981)	
Chebyshev-Vandermonde matrices	Reichel-Opfer (1991)	Gohberg-Olshevsky (1994)
three-term Vandermonde matrices	Higham (1988,90)	Calvetti-Reichel (1993)
Szegö-Vandermonde matrices	BEGKO (2006)	Olshevsky (2001)
Quasiseparable-Vandermonde	BEGKO (2006)	??????????????????

## A generator representation

► The following matrix is **Hessenberg** and **order-one quasiseparable**:

$$\begin{bmatrix} d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\ p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\ 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\ 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\ 0 & 0 & 0 & p_5 q_4 & d_5 \end{bmatrix}$$

## A generator representation

➡ The following matrix is **Hessenberg** and **order-one quasiseparable**:

$$\begin{bmatrix}
 d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\
 p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\
 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\
 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\
 0 & 0 & 0 & p_5 q_4 & d_5
 \end{bmatrix}$$

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 p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\
 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\
 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\
 0 & 0 & 0 & p_5 q_4 & d_5
 \end{bmatrix}$$

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$$\begin{bmatrix}
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 p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\
 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\
 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\
 0 & 0 & 0 & p_5 q_4 & d_5
 \end{bmatrix}$$

## A generator representation

➡ The following matrix is **Hessenberg** and **order-one quasiseparable**:

$$\begin{bmatrix}
 d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\
 p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\
 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\
 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\
 0 & 0 & 0 & p_5 q_4 & d_5
 \end{bmatrix}$$

## A generator representation

► The following matrix is **Hessenberg** and **order-one quasiseparable**:

$$\begin{bmatrix} d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\ p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\ 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\ 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\ 0 & 0 & 0 & p_5 q_4 & d_5 \end{bmatrix}$$

► This **generator representation** exists for any **order-one quasiseparable** matrix.

Order-one quasiseparable-  
Hessenberg

⇔

Generator representation

## Recurrence relations satisfied by polynomials related to quasiseparable matrices

► **Theorem.** A matrix  $C$  is **order-one quasiseparable-Hessenberg** if and only if the system of polynomials  $r_k(x) = \det(xI - C_{k \times k})$  satisfy the recurrence relations

### Two-term recurrence relations

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{p_{k+1}q_k} \begin{bmatrix} \alpha_k & -g_{k+1} \\ h_k/b_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ u_k(x)r_{k-1}(x) \end{bmatrix}$$

### Three-term recurrence relations

$$r_k(x) = -\frac{1}{p_{k+1}q_k} \psi_k(x)r_{k-1}(x) - \frac{1}{p_{k+1}p_kq_{k-1}q_k} \varphi_k(x)r_{k-2}(x)$$

## Special cases of these recurrence relations

Matrix	Polynomial system
Shift matrix	monomials
Tridiagonal matrix	real orthogonal polynomials
Unitary Hessenberg matrix	Szegő polynomials

## Special cases of these recurrence relations

Matrix	Polynomial system
Shift matrix	monomials
Tridiagonal matrix	real orthogonal polynomials
Unitary Hessenberg matrix	Szegő polynomials

## Special cases of these recurrence relations

### Monomial case

$$r_k(x) = -\frac{1}{p_{k+1}q_k}\psi_k(x)r_{k-1}(x) - \frac{1}{p_{k+1}p_kq_{k-1}q_k}\varphi_k(x)r_{k-2}(x)$$

⇓

$$p_k = 1, \quad q_k = 1, \quad d_k = 0, \quad g_k = 0, \quad b_k = 0, \quad h_k = 1$$

⇓

$$r_k(x) = x \cdot r_{k-1}(x)$$

## Special cases of these recurrence relations

Matrix	Polynomial system
Shift matrix	monomials
Tridiagonal matrix	real orthogonal polynomials
Unitary Hessenberg matrix	Szegő polynomials

## Special cases of these recurrence relations

### Real orthogonal polynomial case

$$r_k(x) = -\frac{1}{p_{k+1}q_k}\psi_k(x)r_{k-1}(x) - \frac{1}{p_{k+1}p_kq_{k-1}q_k}\varphi_k(x)r_{k-2}(x)$$

$$\Downarrow$$

$$p_k = 1, \quad b_k = 0, \quad h_k = 1$$

$$\Downarrow$$

$$r_k(x) = \frac{1}{q_k}(x-d_k)r_{k-1}(x) - \frac{g_{k-1}}{q_k}r_{k-2}(x)$$

## Special cases of these recurrence relations

Matrix	Polynomial system
Shift matrix	monomials
Tridiagonal matrix	real orthogonal polynomials
Unitary Hessenberg matrix	Szegő polynomials

## Special cases of these recurrence relations

### Szegő polynomial case

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{p_{k+1}q_k} \begin{bmatrix} \alpha_k & -g_{k+1} \\ h_k/b_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ u_k(x)r_{k-1}(x) \end{bmatrix}$$

 $\Downarrow$ 

$$p_k = 1, \quad q_k = \mu_k, \quad d_k = -\rho_{k-1}^* \rho_k, \quad g_k = \rho_{k-1}^*, \quad b_k = \mu_{k-1}, \quad h_k = -\mu_{k-1} \rho_k$$

 $\Downarrow$ 

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{\mu_k} \begin{bmatrix} 1 & -\rho_k^* \\ -\rho_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ x r_{k-1}(x) \end{bmatrix}$$

## A Class of Polynomials

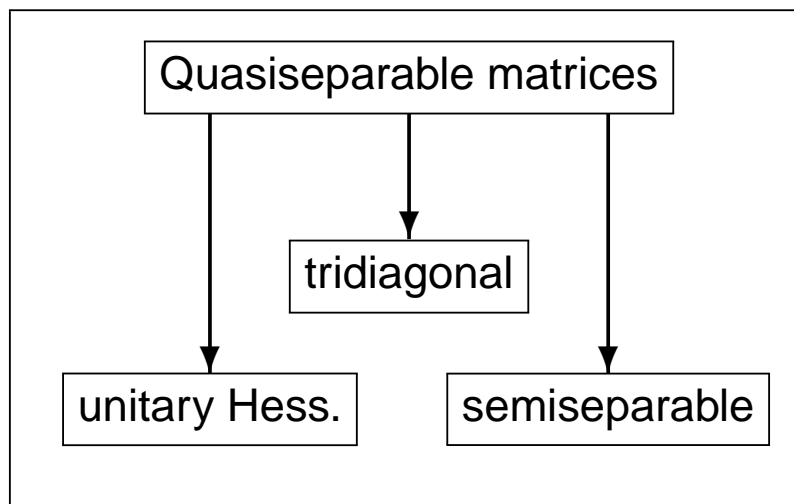
- Systems of polynomials associated with a **quasiseparable matrix**.

(**Quasiseparable polynomials**)?

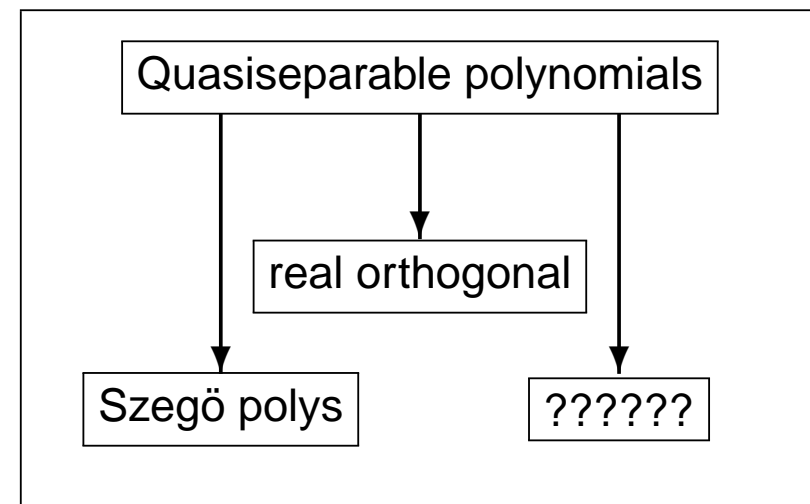
Tridiagonal matrices, unitary Hessenberg matrices, semiseparable matrices are special cases of quasiseparable matrices

⇒

Polynomials orthogonal on (i) a real interval, and (ii) on the unit circle are special cases of quasiseparable polynomials



⇒



## Traub-like algorithm for quasiseparable-Vandermonde matrices

▣▣▣ **Traub-like algorithm.** Based on the formula

$$V_R^{-1} = \begin{bmatrix} \hat{r}_0(x_1) & \hat{r}_1(x_1) & \cdots & \hat{r}_{n-1}(x_1) \\ \hat{r}_0(x_2) & \hat{r}_1(x_2) & \cdots & \hat{r}_{n-1}(x_2) \\ \vdots & \vdots & \ddots & \vdots \\ \hat{r}_0(x_n) & \hat{r}_1(x_n) & \cdots & \hat{r}_{n-1}(x_n) \end{bmatrix}^{-1} = \tilde{I} \cdot V_{\hat{R}}^T \cdot \text{diag}(c_1, c_2, \dots, c_n)$$

where  $\tilde{I}$  is the antidiagonal matrix,  $c_k = \prod_{\substack{j=1 \\ j \neq k}}^n (x_j - x_k)^{-1}$

- ▣▣▣  $\hat{R}$  is the system of **Horner-like** polynomials corresponding to the polynomial system  $R$ .  
(When  $P$  is the monomial basis, this is the classical Traub (1966))
- ▣▣▣ **How do we evaluate the polynomials  $\hat{r}_k$  at the nodes?**

## Three-Term Recurrence Relations

Original polynomials forming  $V_R$

$$r_k(x) = -\frac{1}{p_{k+1}q_k}\psi_k(x)r_{k-1}(x) - \frac{1}{p_{k+1}p_kq_{k-1}q_k}\varphi_k(x)r_{k-2}(x)$$

Horner-like polynomials forming  $V_R^{-1}$

$$\hat{r}_k(x) = -\frac{1}{\hat{p}_{k+1}\hat{q}_k}\hat{\psi}_k(x)\hat{r}_{k-1}(x) - \frac{1}{\hat{p}_{k+1}\hat{p}_k\hat{q}_{k-1}\hat{q}_k}\hat{\varphi}_k(x)\hat{r}_{k-2}(x) - \hat{z}_k$$

$$\psi_k(x) = (d_k - x) - \frac{p_k h_k q_{k-1} b_{k-1}}{h_{k-1}} \quad \varphi_k(x) = \frac{p_k h_k q_{k-1}}{h_{k-1}} ((d_k - x)b_k - h_k g_k)$$

$$z_k = \frac{h_k b_{k-1}}{p_{k+1} q_k h_{k-1}} \beta_{n-k+1} - \frac{1}{p_{k+1} q_k} \beta_{n-k}$$

## Two-Term Recurrence Relations

Original polynomials forming  $V_R$

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{p_{k+1}q_k} \begin{bmatrix} \alpha_k & -g_{k+1} \\ h_k/b_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ u_k(x)r_{k-1}(x) \end{bmatrix}$$

Horner-like polynomials forming  $V_R^{-1}$

$$\begin{bmatrix} \hat{G}_k(x) \\ \hat{r}_k(x) \end{bmatrix} = \frac{1}{\hat{p}_{k+1}\hat{q}_k} \begin{bmatrix} \hat{\alpha}_k & -\hat{g}_{k+1} \\ \hat{h}_k/\hat{b}_k & 1 \end{bmatrix} \begin{bmatrix} \hat{G}_{k-1}(x) \\ \hat{u}_k(x)\hat{r}_{k-1}(x) + \beta_{n-k} \end{bmatrix}$$

$$\alpha_k = p_{k+1}b_{k+1}q_k - \frac{g_{k+1}h_k}{b_k} \quad u_k(x) = x - d_k + \frac{g_k h_k}{b_k}$$

## What are the coefficients $\beta_k$ ?

- ▣ The difference between the recurrence relations for the original polynomials and those for the Horner-like polynomials is the presence of the coefficients  $\beta_k$ .
- ▣  $\beta_k$  is the coefficient of  $r_k(x)$  in the decomposition of the **master polynomial**  $\beta(x) = \prod_{i=1}^n (x - x_i)$  into the  $\{r_k\}$  basis:

$$\prod_{i=1}^n (x - x_i) = \beta_0 r_0(x) + \beta_1 r_1(x) + \cdots + \beta_n r_n(x)$$

- ▣ These coefficients can be computed in  $\mathcal{O}(n^2)$  arithmetic operations.

## Complexity of the Traub-like algorithm

- ▶ Each Horner-like polynomial can be evaluated at all of the nodes in  $\mathcal{O}(n)$  operations.
- ▶ The coefficients  $\beta_k$  can be computed in  $\mathcal{O}(n^2)$  operations.
- ▶ The total cost of the algorithm is  $\mathcal{O}(n^2)$  operations. Comparing this to the complexity of Gaussian elimination,  $\mathcal{O}(n^3)$ , we have the algorithm is **FAST!**

## Special cases of the Traub-like algorithm

- ▶ **Tridiagonal** case: This reduces to the algorithm of **Calvetti-Reichel** (1993).
- ▶ **Unitary Hessenberg** case: This reduces to the algorithm of **Olshevsky** (2001).

# Fast algorithms for polynomial-Vandermonde matrices related to quasiseparable matrices

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Joint work with Yuli Eidelman, Israel Gohberg, Israel Koltracht, & Vadim Olshevsky.

## Supplemental Slides

## Confederate Matrices

► **Definition** For polynomials  $R = \{r_0(x), r_1(x), \dots, r_n(x)\}$  satisfying  $n$ -term recurrence relations

$$r_k(x) = \alpha_k \cdot x r_{k-1}(x) - a_{k-1,k} \cdot r_{k-1}(x) - a_{k-2,k} \cdot r_{k-2}(x) - \dots - a_{0,k} \cdot r_0(x).$$

and the polynomial

$$b(x) = b_0 \cdot r_0(x) + b_1 \cdot r_1(x) + \dots + b_{n-1} \cdot r_{n-1}(x) + b_n \cdot r_n(x)$$

define the **confederate matrix** of  $b$  with respect to  $R$  by

$$C_R(b) = \begin{bmatrix} \frac{a_{01}}{\alpha_1} & \frac{a_{02}}{\alpha_2} & \frac{a_{03}}{\alpha_3} & \dots & \dots & \frac{a_{0,n}}{\alpha_n} & -\frac{1}{\alpha_n} & \frac{b_0}{b_n} \\ \frac{1}{\alpha_1} & \frac{a_{12}}{\alpha_2} & \frac{a_{13}}{\alpha_3} & \dots & \dots & \frac{a_{1,n}}{\alpha_n} & -\frac{1}{\alpha_n} & \frac{b_1}{b_n} \\ 0 & \frac{1}{\alpha_2} & \frac{a_{23}}{\alpha_3} & \dots & \dots & \frac{a_{2,n}}{\alpha_n} & -\frac{1}{\alpha_n} & \frac{b_2}{b_n} \\ 0 & 0 & \frac{1}{\alpha_3} & \ddots & \dots & \dots & \vdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \dots & \vdots & \vdots \\ 0 & 0 & \dots & 0 & \frac{1}{\alpha_{n-1}} & \frac{a_{n-1,n}}{\alpha_n} & -\frac{1}{\alpha_n} & \frac{b_{n-1}}{b_n} \end{bmatrix}$$

## Confederate Matrices

$$r_1(x) = \alpha_1 \cdot xr_0(x) - a_{0,1} \cdot r_0(x)$$

$$r_2(x) = \alpha_2 \cdot xr_1(x) - a_{1,2} \cdot r_1(x) - a_{0,2} \cdot r_0(x)$$

$$r_3(x) = \alpha_3 \cdot xr_2(x) - a_{2,3} \cdot r_2(x) - a_{1,3} \cdot r_1(x) - a_{0,3} \cdot r_0(x)$$

$$\vdots$$

$$\left[ \begin{array}{cccccc} \frac{a_{01}}{\alpha_1} & \frac{a_{02}}{\alpha_2} & \frac{a_{03}}{\alpha_3} & \cdots & \cdots & \frac{a_{0,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_0}{b_n} \\ \frac{1}{\alpha_1} & \frac{a_{12}}{\alpha_2} & \frac{a_{13}}{\alpha_3} & \cdots & \cdots & \frac{a_{1,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_1}{b_n} \\ 0 & \frac{1}{\alpha_2} & \frac{a_{23}}{\alpha_3} & \cdots & \cdots & \frac{a_{2,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_2}{b_n} \\ 0 & 0 & \frac{1}{\alpha_3} & \ddots & & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 & \frac{1}{\alpha_{n-1}} & \frac{a_{n-1,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_{n-1}}{b_n} \end{array} \right]$$

## Confederate Matrices

$$r_1(x) = \alpha_1 \cdot xr_0(x) - a_{0,1} \cdot r_0(x)$$

$$r_2(x) = \alpha_2 \cdot xr_1(x) - a_{1,2} \cdot r_1(x) - a_{0,2} \cdot r_0(x)$$

$$r_3(x) = \alpha_3 \cdot xr_2(x) - a_{2,3} \cdot r_2(x) - a_{1,3} \cdot r_1(x) - a_{0,3} \cdot r_0(x)$$

$$\vdots$$

$$\left[ \begin{array}{cccccc} \frac{a_{01}}{\alpha_1} & \frac{a_{02}}{\alpha_2} & \frac{a_{03}}{\alpha_3} & \cdots & \cdots & \frac{a_{0,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_0}{b_n} \\ \frac{1}{\alpha_1} & \frac{a_{12}}{\alpha_2} & \frac{a_{13}}{\alpha_3} & \cdots & \cdots & \frac{a_{1,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_1}{b_n} \\ 0 & \frac{1}{\alpha_2} & \frac{a_{23}}{\alpha_3} & \cdots & \cdots & \frac{a_{2,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_2}{b_n} \\ \vdots & \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & 0 & \frac{1}{\alpha_3} & \ddots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 & \frac{1}{\alpha_{n-1}} & \frac{a_{n-1,n}}{\alpha_n} - \frac{1}{\alpha_n} \cdot \frac{b_{n-1}}{b_n} \end{array} \right]$$

## Confederate Matrices

$$r_1(x) = \alpha_1 \cdot xr_0(x) - a_{0,1} \cdot r_0(x)$$

$$r_2(x) = \alpha_2 \cdot xr_1(x) - a_{1,2} \cdot r_1(x) - a_{0,2} \cdot r_0(x)$$

$$r_3(x) = \alpha_3 \cdot xr_2(x) - a_{2,3} \cdot r_2(x) - a_{1,3} \cdot r_1(x) - a_{0,3} \cdot r_0(x)$$

$$\vdots$$

$$\begin{bmatrix} \frac{a_{01}}{\alpha_1} & \frac{a_{02}}{\alpha_2} & \frac{a_{03}}{\alpha_3} & \cdots & \cdots & \frac{a_{0,n}}{\alpha_n} & - & \frac{1}{\alpha_n} & \cdot & \frac{b_0}{b_n} \\ \frac{1}{\alpha_1} & \frac{a_{12}}{\alpha_2} & \frac{a_{13}}{\alpha_3} & \cdots & \cdots & \frac{a_{1,n}}{\alpha_n} & - & \frac{1}{\alpha_n} & \cdot & \frac{b_1}{b_n} \\ 0 & \frac{1}{\alpha_2} & \frac{a_{23}}{\alpha_3} & \cdots & \cdots & \frac{a_{2,n}}{\alpha_n} & - & \frac{1}{\alpha_n} & \cdot & \frac{b_2}{b_n} \\ 0 & 0 & \frac{1}{\alpha_3} & \ddots & & & & & \vdots & \\ \vdots & \vdots & \ddots & \ddots & \ddots & & & & \vdots & \\ 0 & 0 & \cdots & 0 & \frac{1}{\alpha_{n-1}} & \frac{a_{n-1,n}}{\alpha_n} & - & \frac{1}{\alpha_n} & \cdot & \frac{b_{n-1}}{b_n} \end{bmatrix}$$

## Motivation for Horner-like Polynomials

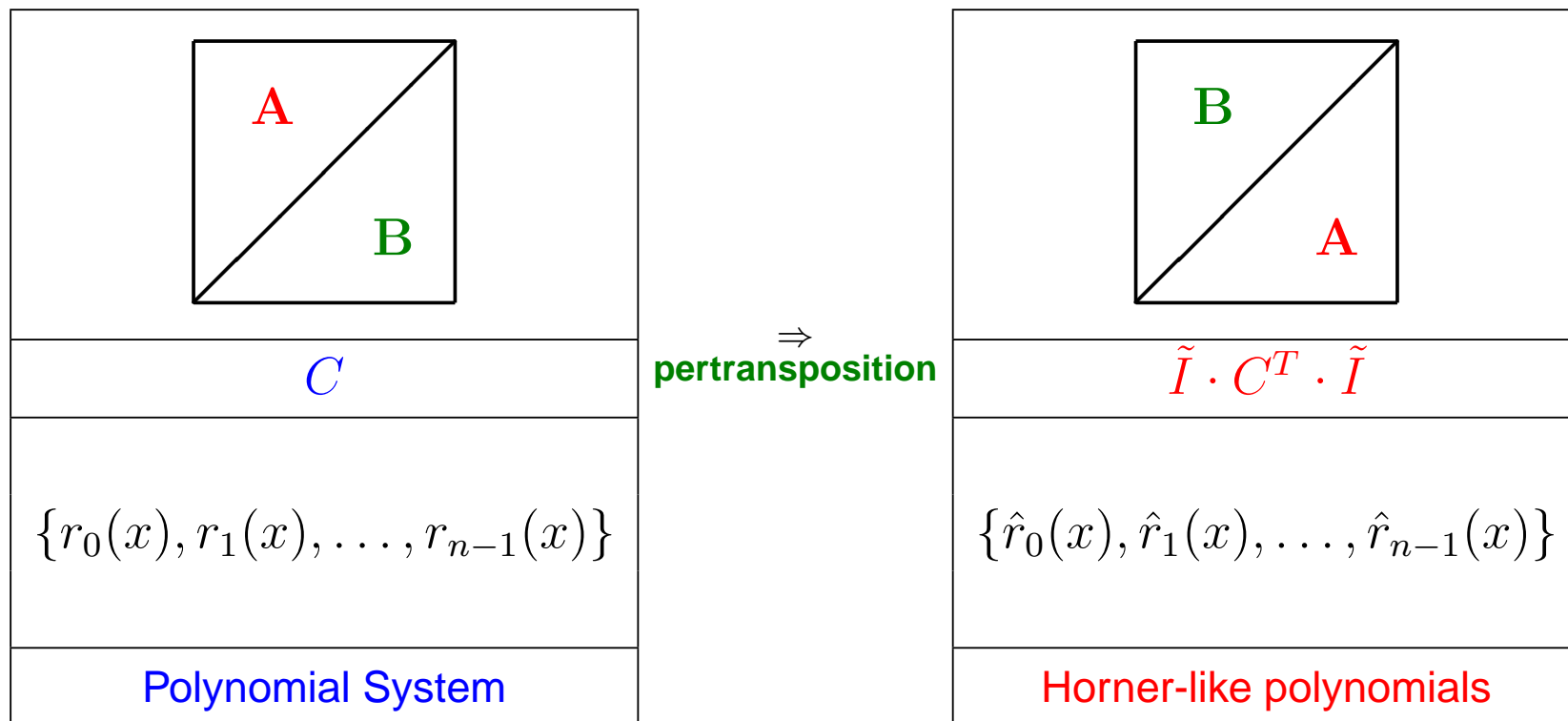
- The confederate matrix  $C(\mathbf{b})$  for a polynomial  $b(x) = b_0 + b_1x + \cdots + b_nx_n$  in the **monomial basis** reduces to the companion matrix, and the confederate matrix  $C_R(\hat{\mathbf{p}}_n)$  for the **Horner polynomials** is:

$$\mathbf{C}(\mathbf{b}) = \begin{bmatrix} 0 & 0 & \cdots & 0 & -b_0 \\ 1 & 0 & \cdots & 0 & -b_1 \\ 0 & 1 & \ddots & \vdots & \vdots \\ \vdots & & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 1 & -b_{n-1} \end{bmatrix} \quad \mathbf{C}_R(\hat{\mathbf{p}}_n) = \begin{bmatrix} -b_{n-1} & -b_{n-2} & \cdots & -b_1 & -b_0 \\ 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & \ddots & \vdots & 0 \\ \vdots & & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 1 & 0 \end{bmatrix}$$

- **Observation:**  $\mathbf{C}_R(\hat{\mathbf{p}}_n) = \tilde{I} \cdot \mathbf{C}(\mathbf{b})^T \cdot \tilde{I}$

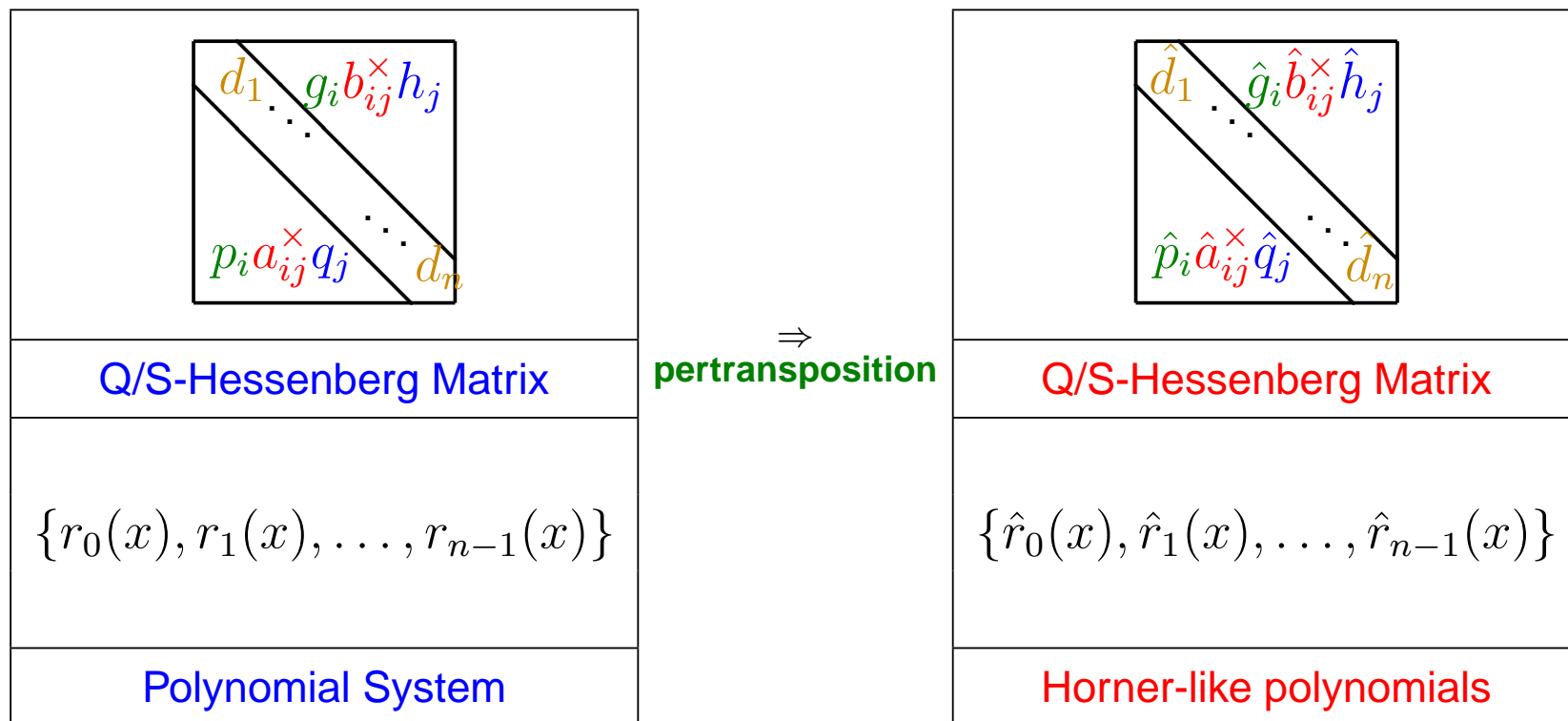
## Horner-like Polynomials

- ➡ A relation between the matrix corresponding to a system of polynomials  $R$  and the matrix corresponding to the **Horner-like** polynomials  $\hat{R}$ .



## Horner-like Polynomials

► **Fact.** A pertransposed quasiseparable matrix is again a quasiseparable matrix.



## Generators of a Quasiseparable matrix

► Can be represented in terms of their **generators**:

**Diagonal entries**

$$d_k \quad k = 1, \dots, n$$

**Lower Generators**

$$p_k \quad k = 2, \dots, n$$

$$a_k \quad k = 2, \dots, n - 1$$

$$q_k \quad k = 1, \dots, n - 1$$

**Upper Generators**

$$g_k \quad k = 1, \dots, n - 1$$

$$b_k \quad k = 2, \dots, n - 1$$

$$h_k \quad k = 2, \dots, n$$

► **Example.** In terms of generators, with  $n = 5$ ,

$$\begin{bmatrix} d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\ p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\ p_3 a_2 q_1 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\ p_4 a_3 a_2 q_1 & p_4 a_3 q_2 & p_4 q_3 & d_4 & g_4 h_5 \\ p_5 a_4 a_3 a_2 q_1 & p_5 a_4 a_3 q_2 & p_5 a_4 q_3 & p_5 q_4 & d_5 \end{bmatrix}$$

# Recurrence relations that generalize Szegő recurrence relations

## Three-Term Recurrence Relations

$$r_0(x) = 1 \quad r_1(x) = \frac{1}{p_2 q_1} (x - d_1)$$

$$r_k(x) = -\frac{1}{p_{k+1} q_k} \psi_k(x) r_{k-1}(x) - \frac{1}{p_{k+1} p_k q_{k-1} q_k} \varphi_k(x) r_{k-2}(x)$$

$$\Downarrow$$

$$p_k = 1, \quad q_k = \mu_k, \quad d_k = -\rho_{k-1}^* \rho_k, \quad g_k = \rho_{k-1}^*, \quad b_k = \mu_{k-1}, \quad h_k = -\mu_{k-1} \rho_k$$

$$\Downarrow$$

$$r_0(x) = 1 \quad r_1(x) = \frac{1}{\mu_1} (x + \rho_0^* \rho_1)$$

$$r_k(x) = -\frac{1}{\mu_k} \left( x + \frac{\rho_k}{\rho_{k-1}} \right) r_{k-1}(x) - \left( \frac{\rho_k \mu_{k-1}}{\rho_{k-1} \mu_k} \right) x \cdot r_{k-2}(x)$$

# Recurrence relations that generalize Szegő recurrence relations

## Perturbed Two-Term Recurrence Relations

$$\begin{bmatrix} G_0(x) \\ r_0(x) \end{bmatrix} = \frac{1}{p_2 q_1} \begin{bmatrix} -g_1 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{p_{k+1} q_k} \begin{bmatrix} \alpha_k & -g_{k+1} \\ h_k/b_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ u_k(x) r_{k-1}(x) \end{bmatrix}$$

⇓

$$p_k = 1, \quad q_k = \mu_k, \quad d_k = -\rho_{k-1}^* \rho_k, \quad g_k = \rho_{k-1}^*, \quad b_k = \mu_{k-1}, \quad h_k = -\mu_{k-1} \rho_k$$

⇓

$$\begin{bmatrix} G_0(x) \\ r_0(x) \end{bmatrix} = \frac{1}{\mu_1} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \frac{1}{\mu_k} \begin{bmatrix} 1 & -\rho_k^* \\ -\rho_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ x r_{k-1}(x) \end{bmatrix}$$

## Numerical Illustrations - Traub

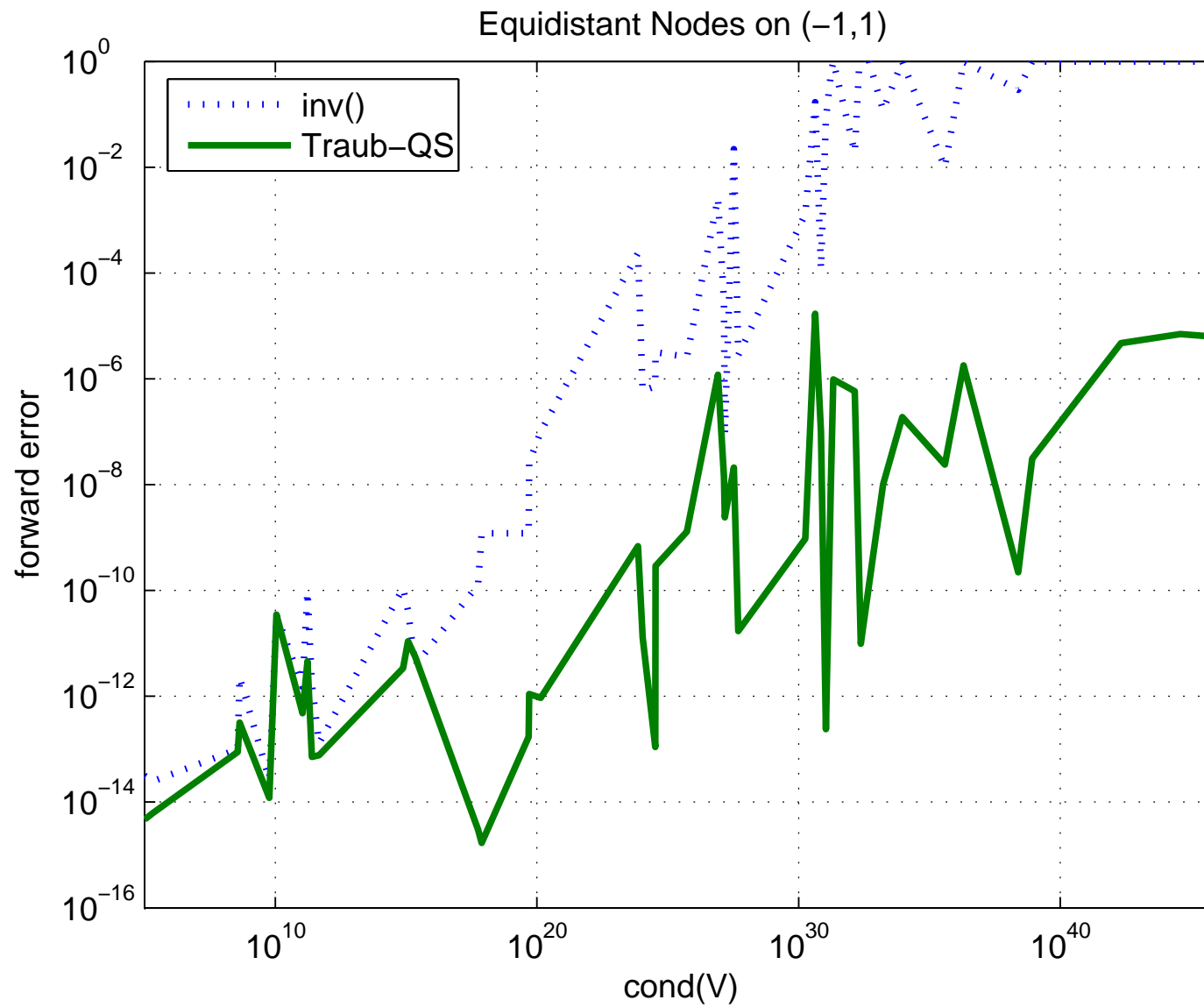
- ▶▶▶ We compare the **forward error** of the inverse  $\widehat{A}^{-1}$  from MATLAB in double precision via

$$e = \frac{\|A^{-1} - \widehat{A}^{-1}\|_2}{\|A^{-1}\|_2},$$

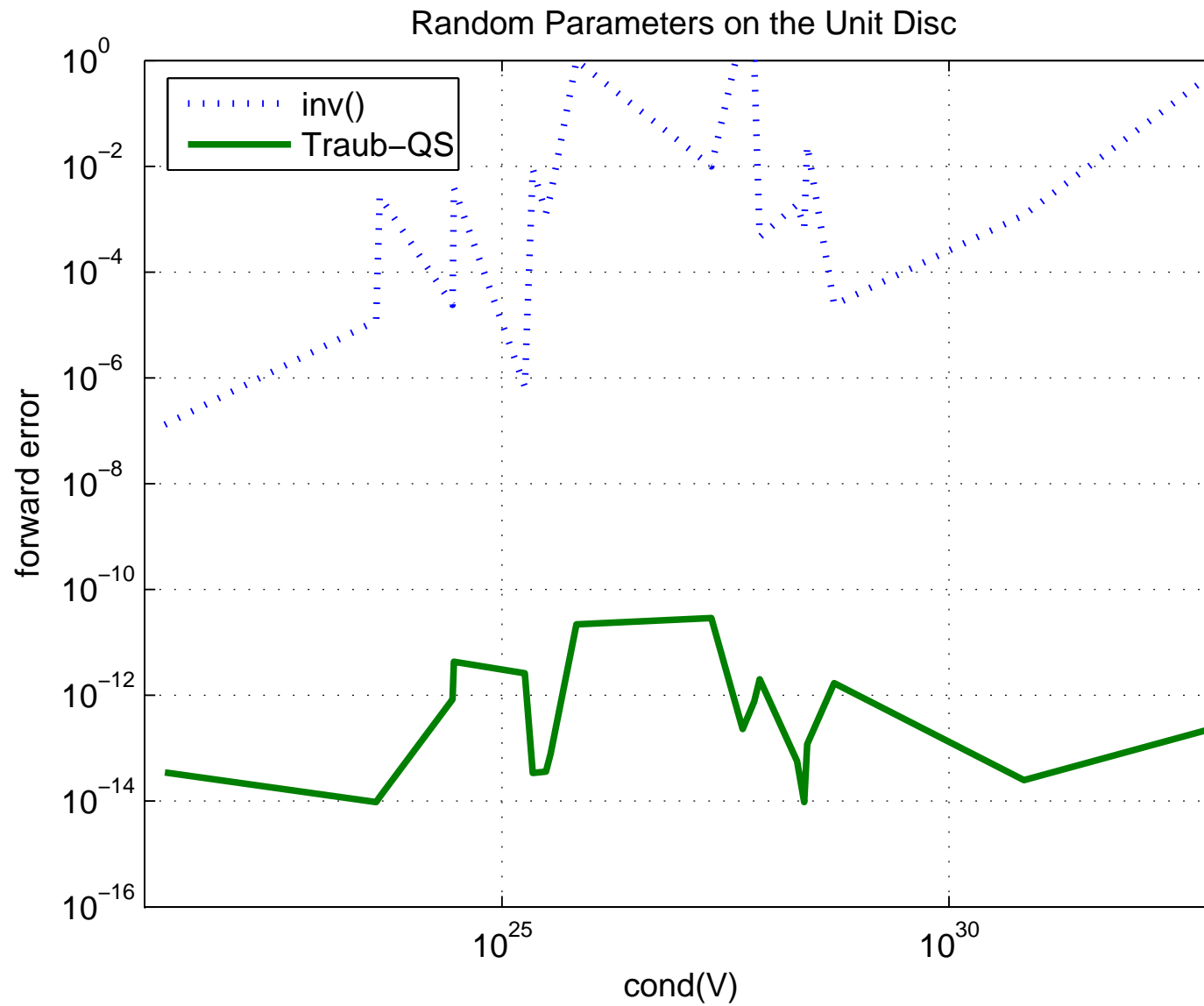
with  $A^{-1}$ , the “exact” solution using MATLAB’s `vpa ( )` command for software-implemented arbitrary digit arithmetic.

- ▶▶▶ **GE** - Gaussian elimination via MATLAB’s `inv ( )` command.
- ▶▶▶ **Traub-QS** - Traub-like algorithm with nodes ordered via the **Leja ordering**. (Reichel, Higham)

# Numerical Illustrations - Traub Experiment 1



## Numerical Illustrations - Traub Experiment 2



# Fast algorithms for polynomial-Vandermonde matrices related to quasiseparable matrices

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Based on joint work with Yuli Eidelman, Israel Gohberg, & Vadim Olshevsky.

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