

**Fast algorithm for inversion of  
polynomial–Vandermonde matrices related to  
 $(H, m)$ -quasiseparable matrices  
and**

**Classification of recurrence relations for  
polynomials via subclasses of  
 $(H, m)$ -quasiseparable matrices**

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Joint work with **Vadim Olshevsky, Yuli Eidelman,  
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# Introduction

## A problem: Inversion/system solving for the classical Vandermonde matrices

►► **Definition.** A **Vandermonde matrix** is defined by

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

►► **Fast algorithms:**

- **Inversion:** **Traub** (1966) algorithm
- **System solver:** **Björck-Pereyra** (1970) algorithm
- **Cost:**  $O(n^2)$  vs  $O(n^3)$  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*”.
- ▶▶▶ **Björck-Pereyra algorithm. (Higham’s example)**

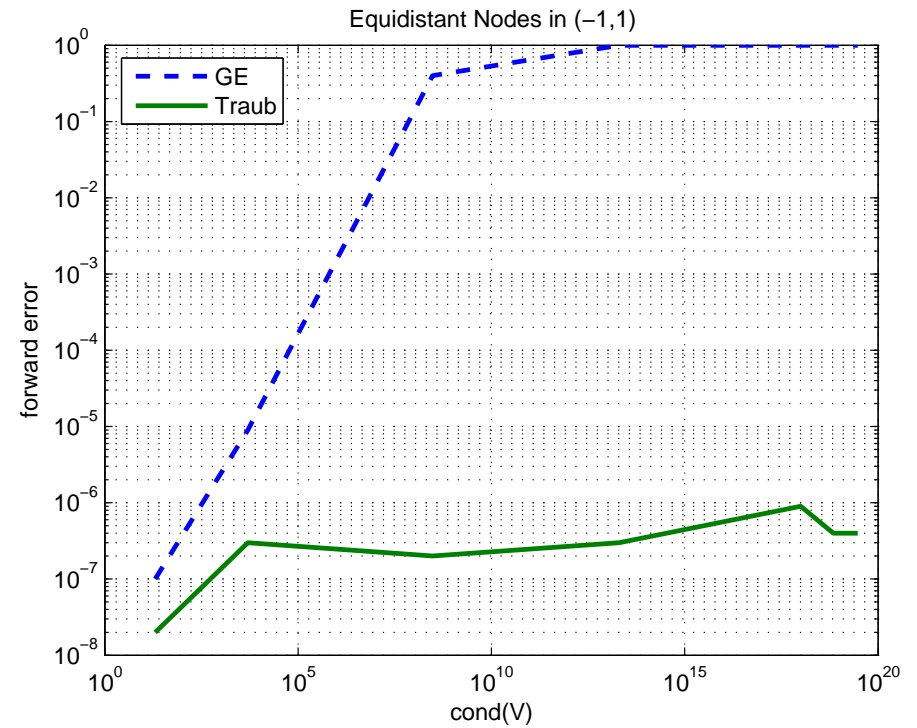
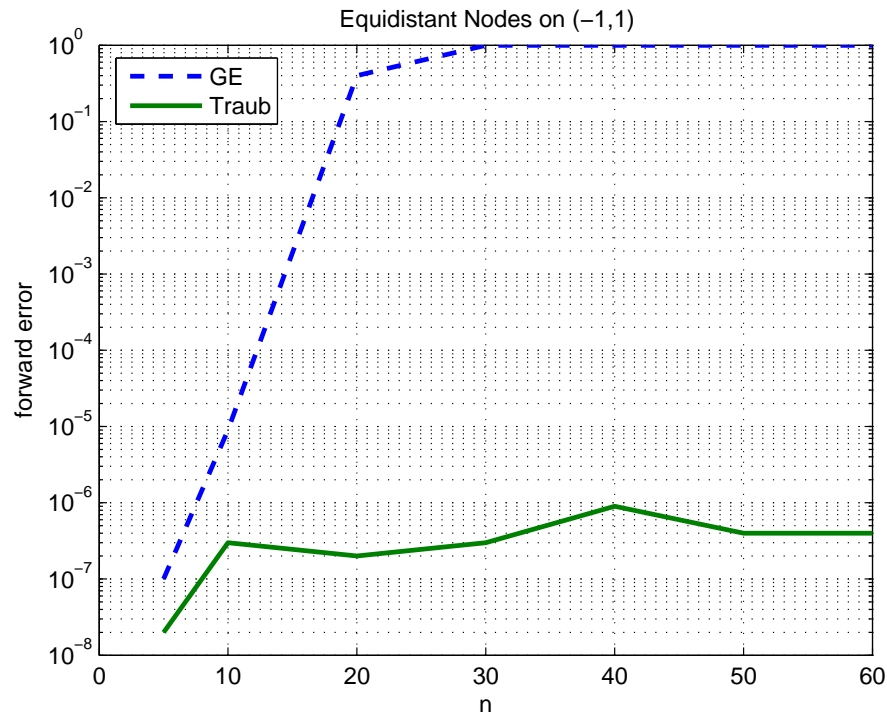
1. Nodes chosen randomly in  $(0, 1)$ ; RHS alternating signs,  $\begin{bmatrix} 1 & -1 & 1 & \cdots \end{bmatrix}^T$
2. Forward error measured by  $e = \frac{\|x - \hat{x}\|_2}{\|x\|_2}$

## Some Numerical Experiments

- ▶▶▶ **Traub algorithm.**
- 1. Nodes are chosen to be equidistant on  $(-1, 1)$ .
- 2. Forward error measured by  $e = \frac{\|A^{-1} - \widehat{A}^{-1}\|_2}{\|A^{-1}\|_2}$

# Numerical Experiments

## Gaussian Elimination & Traub Algorithm



## Polynomial-Vandermonde matrices

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

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

# Fast $\mathcal{O}(n^2)$ algorithms for polynomial-Vandermonde matrices

## Previous work

Polynomials $\{r_k(x)\}$	Traub-like algorithm for <b>inversion</b>	Björck-Pereyra-like <b>system solver</b>
monomials	Traub (1966)	Björck-Pereyra (1970)
Chebyshev	Gohberg-Olshevsky (1994)	Reichel-Opfer (1991)
real-orthogonal	Calvetti-Reichel (1993)	Higham (1988, 90)
Szegö	Olshevsky (2001)	Bella et al (LAA, Jan. 2007)
$(H, 1)$ -quasi-separable	Bella et al (2007)	Bella et al (2007)

All listed algorithms require only  $\mathcal{O}(n^2)$  operations, as opposed to  $\mathcal{O}(n^3)$  required by GE.

# Fast $\mathcal{O}(n^2)$ algorithms for polynomial-Vandermonde matrices

Previous work

Polynomials $\{r_k(x)\}$	Traub-like algorithm for <b>inversion</b>	Björck-Pereyra-like <b>system solver</b>
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<b><math>(H, m)</math>-quasi- separable</b>	<b>2007</b>	<b>Bella et al (2007)</b>

All listed algorithms require only  $\mathcal{O}(n^2)$  operations, as opposed to  $\mathcal{O}(n^3)$  required by GE.

## Part 1. $(H, m)$ -quasiseparable matrices and polynomials

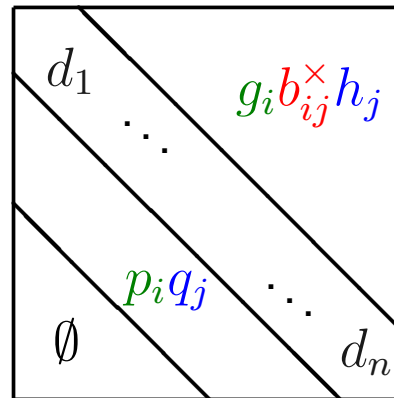
Rank Definition. A matrix  $C$  is  $(H, m)$ -quasiseparable if it is upper Hessenberg and

$$\max \text{Rank} C_{12} = m$$

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

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

Generator Definition The following matrix is Hessenberg and order- $m$  quasiseparable:



where  $\{g_i\}, \{b_i\}, \{h_i\}$  are matrices of sizes  $1 \times r_i, r_{i-1} \times r_i$  and  $r_{i-1} \times 1$  respectively.

$$\max_{1 \leq i \leq N-1} r_i = m$$

# The set of $(H, m)$ -quasiseparable matrices contains:

⇒  $(H, 1)$ -quasiseparable matrices

- Tridiagonal matrices

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

- Unitary Hessenberg matrices

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

⇒ Matrices not covered by  $(H, 1)$ -case!!!

# It is easy to find an example of $(H, m)$ -qs polynomials and matrices!

If polynomials satisfy simplest  $l$ -term recurrence relations

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

then their confederate matrices

$$A = \begin{bmatrix} \frac{a_{0,1}}{\alpha_1} & \dots & \frac{a_{0,l-1}}{\alpha_{l-1}} & 0 & \dots & 0 \\ \frac{1}{\alpha_1} & \frac{a_{1,2}}{\alpha_2} & \dots & \frac{a_{1,l}}{\alpha_l} & \ddots & \vdots \\ 0 & \frac{1}{\alpha_2} & & & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & & \frac{a_{n-(l-1),n}}{\alpha_n} \\ \vdots & & \ddots & \frac{1}{\alpha_{n-2}} & & \vdots \\ 0 & \dots & \dots & 0 & \frac{1}{\alpha_{n-1}} & \frac{a_{n-1,n}}{\alpha_n} \end{bmatrix}$$

are  $(1, l - 2)$ -banded, i.e., they have only one nonzero subdiagonal and  $l - 2$  nonzero superdiagonals. Clearly  $A$  is an  $(H, m)$ -quasiseparable matrix for  $m = l - 2$  by definition.

## Two-term recurrence relations for polynomials

Let  $C$  be an  $(H, m)$ -quasiseparable matrix. Then the system of polynomials

$$\mathbf{r}_k(x) = \frac{1}{c_{2,1}c_{3,2} \cdots c_{n,n-1}} \det(xI - C_{k \times k})$$

associated with  $C$  satisfy the recurrence relation

$$\begin{bmatrix} F_k(x) \\ \mathbf{r}_k(x) \end{bmatrix} = \frac{1}{p_{k+1}q_k} \begin{bmatrix} p_k q_k & b_k^T & -q_k g_k^T & F_{k-1}(x) \\ p_k & h_k^T & x - d_k & \mathbf{r}_{k-1}(x) \end{bmatrix}$$

where  $F_k(x)$  - the vector of auxiliary polynomials.

## Part 2. Fast Traub-like algorithm for $(H, m)$ -quasiseparable Vandermonde matrices

$(H, m)$ -quasiseparable-Vandermonde matrices

► **Definition.** An  $(H, m)$ -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  $r_k(x)$  defined by

$$r_k(x) = \frac{1}{c_{2,1}c_{3,2} \cdots c_{n,n-1}} \det(xI - C_{k \times k})$$

correspond to an  $(H, m)$ -quasiseparable matrix  $C$ .

► The class of  $(H, m)$ -quasiseparable polynomials contains as subclasses the classes of **real-orthogonal polynomials** and **Szegő polynomials**.

## Traub-like algorithm for $(H, m)$ -quasiseparable-Vandermonde matrices

### Traub-like algorithm.

Based on the formula

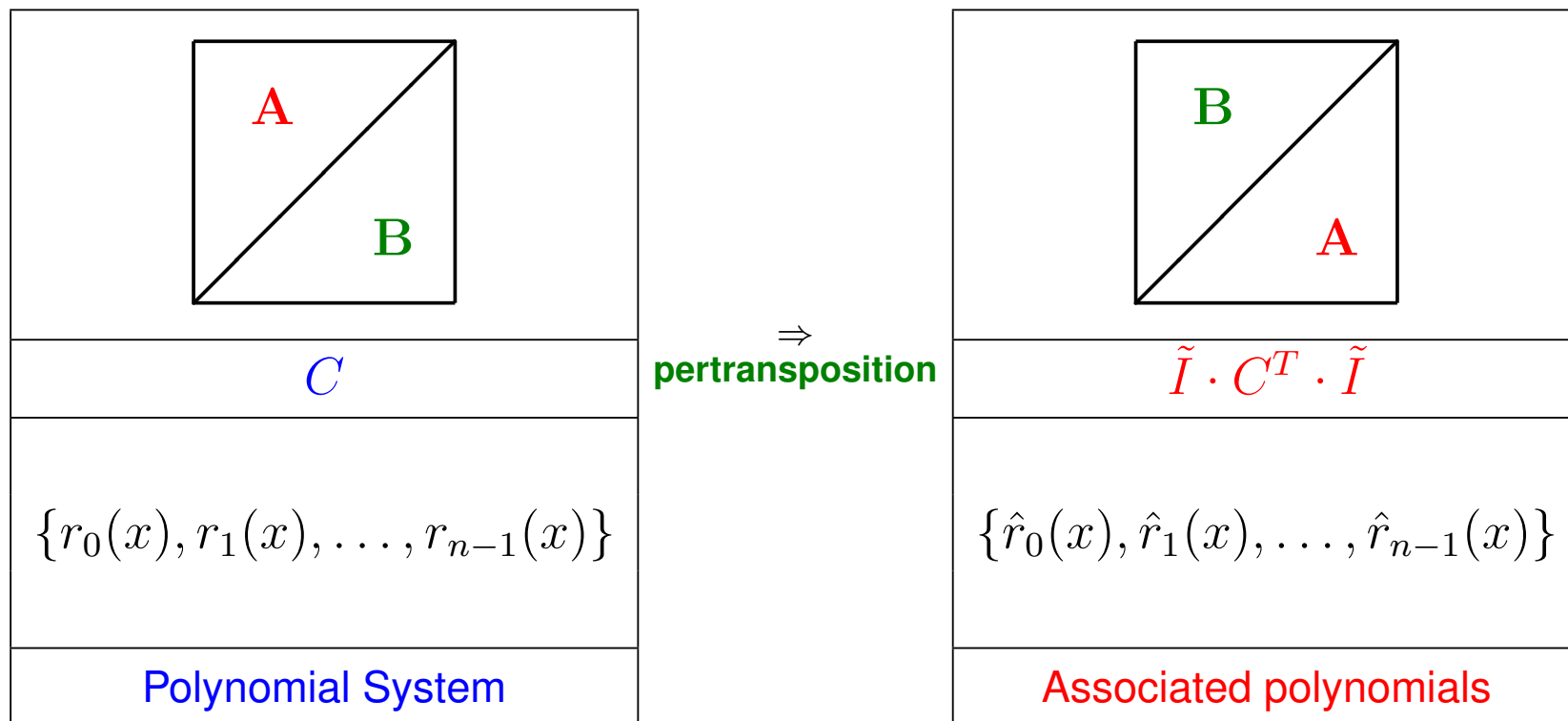
$$V_R^{-1} = \begin{bmatrix} r_0(x_1) & r_1(x_1) & \cdots & r_{n-1}(x_1) \\ r_0(x_2) & r_1(x_2) & \cdots & r_{n-1}(x_2) \\ \vdots & \vdots & \ddots & \vdots \\ r_0(x_n) & r_1(x_n) & \cdots & r_{n-1}(x_n) \end{bmatrix}^{-1} = \tilde{I} \cdot V_{\hat{R}}^T \cdot \text{diag}(c_1, c_2, \dots, c_n)$$

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

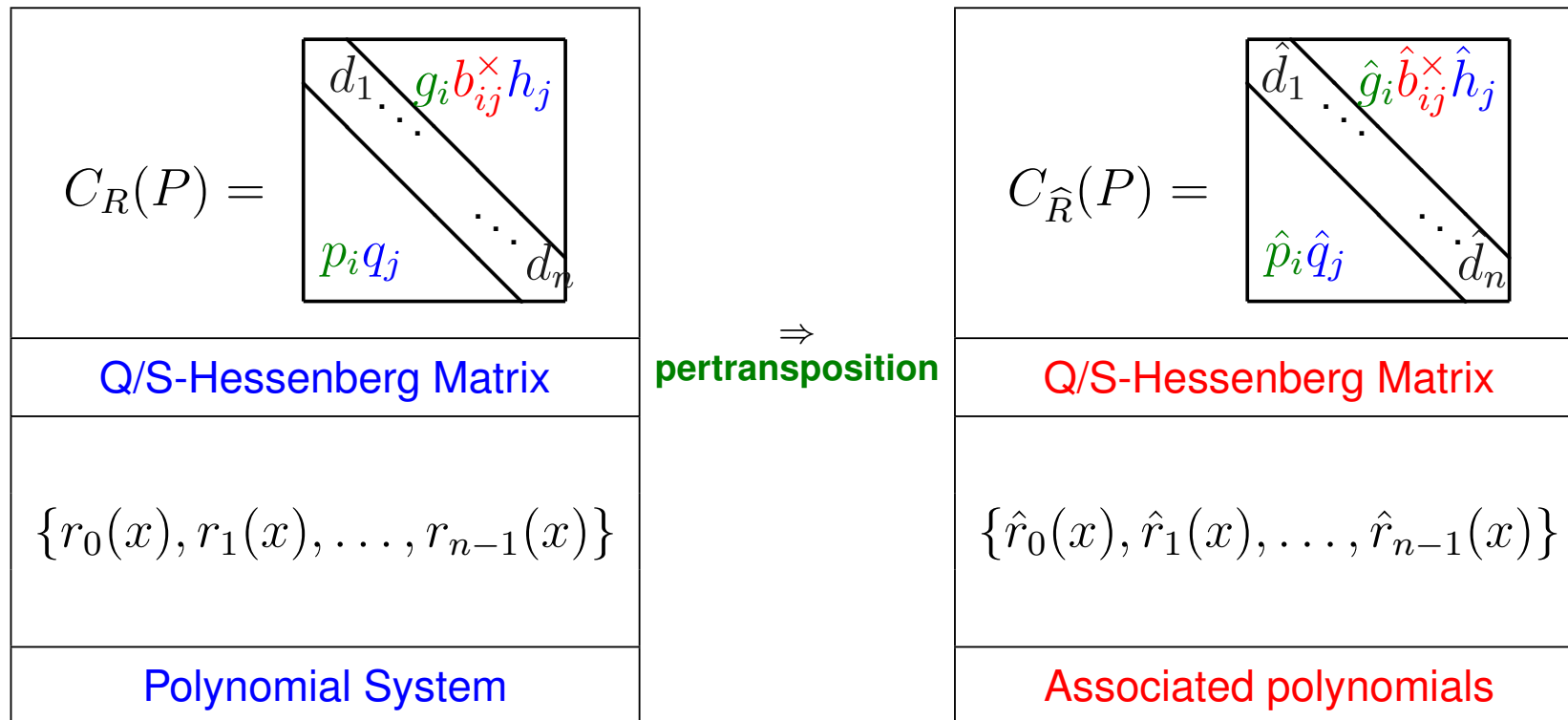
- ▶▶▶  $\hat{R}$  is the system of **Horner-like** polynomials corresponding to the polynomial system  $R$ .  
 (When  $R$  is the monomial basis, this is the classical Traub (1966))
- ▶▶▶ **How do we evaluate the polynomials  $\hat{r}_k$  at the nodes?**

## Via Pertransposition!

- ➔ **Pertransposition** of the related matrix provides a relation between the matrix corresponding to a system of polynomials  $R$  and the matrix corresponding to the **associated** polynomials  $\hat{R}$ .



## Associated (Horner-like) polynomials



## Relation of generators

➡ **Before pertransposition: polynomials**  $R = \{r_0(x), \dots, r_{n-1}(x)\}$

$$C_R(P) = \begin{bmatrix} d_1 & g_1 h_2 & g_1 b_2 h_3 & g_1 b_2 b_3 h_4 & g_1 b_2 b_3 b_4 h_5 \\ p_2 q_1 & d_2 & g_2 h_3 & g_2 b_3 h_4 & g_2 b_3 b_4 h_5 \\ 0 & p_3 q_2 & d_3 & g_3 h_4 & g_3 b_4 h_5 \\ 0 & 0 & p_4 q_3 & d_4 & g_4 h_5 \\ 0 & 0 & 0 & p_5 q_4 & d_5 \end{bmatrix} - \begin{bmatrix} P_0 \\ P_1 \\ P_2 \\ P_3 \\ P_4 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & \frac{1}{P_5} \end{bmatrix}$$

➡ **After pertransposition: associated polynomials**  $\hat{R} = \{\hat{r}_0(x), \dots, \hat{r}_{n-1}(x)\}$

$$C_{\hat{R}}(P) = \begin{bmatrix} d_5 & g_4 h_5 & g_3 b_4 h_5 & g_2 b_3 b_4 h_5 & g_1 b_2 b_3 b_4 h_5 \\ p_5 q_4 & d_4 & g_3 h_4 & g_2 b_3 h_4 & g_1 b_2 b_3 h_4 \\ 0 & p_4 q_3 & d_3 & g_2 h_3 & g_1 b_2 h_3 \\ 0 & 0 & p_3 q_2 & d_2 & g_1 h_2 \\ 0 & 0 & 0 & p_2 q_1 & d_1 \end{bmatrix} - \begin{bmatrix} \frac{1}{P_5} \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} P_4 & P_3 & P_2 & P_1 & P_0 \end{bmatrix}$$

**Theorem: recurrence relations for the associated polynomials**

$$\left[ \begin{array}{c} \widehat{F}_k(x) \\ \hline \widehat{r}_k(x) \end{array} \right] = \left[ \begin{array}{c} 0 \\ \hline P_n \end{array} \right],$$

$$\left[ \begin{array}{c} \widehat{F}_k(x) \\ \hline \widehat{r}_k(x) \end{array} \right] = \frac{1}{\widehat{p}_{k+1}\widehat{q}_k} \left[ \begin{array}{c|c} \widehat{p}_k\widehat{q}_k \widehat{b}_k^T & -\widehat{q}_k \widehat{g}_k^T \\ \hline \widehat{p}_k \widehat{h}_k^T & x - \widehat{d}_k \end{array} \right] \left[ \begin{array}{c} \widehat{F}_{k-1}(x) \\ \hline \widehat{r}_k(x) \end{array} \right] + \frac{1}{\widehat{p}_{k+1}\widehat{q}_k} \left[ \begin{array}{c} 0 \\ \hline P_{n-k} \end{array} \right]$$

## What are the coefficients $P_k$ ?

- ▶ The difference between the recurrence relations for the original polynomials and those for the Horner-like polynomials is the presence of the coefficients  $P_k$ .
- ▶  $P_k$  is the coefficient of  $r_k(x)$  in the decomposition of the **master polynomial**

$$P(x) = \prod_{i=1}^n (x - x_i)$$

into the  $\{r_k\}$  basis:

$$\prod_{i=1}^n (x - x_i) = P_0 r_0(x) + P_1 r_1(x) + \cdots + P_n r_n(x)$$

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

## Complexity of the $(H, m)$ -Traub-like algorithm

- ▣ Each Horner-like polynomial can be evaluated at all of the nodes in  $\mathcal{O}(n)$  operations.
- ▣ The coefficients  $P_k$  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!**

## Part 3. Numerical experiments

## Description

- ▶▶▶ We compare the **forward error** of the inverse  $A_s^{-1}$  from C++ code in single precision via

$$e = \frac{\|A_d^{-1} - A_s^{-1}\|_2}{\|A_d^{-1}\|_2},$$

with  $A_d^{-1}$ , the “exact” solution computed in double precision.

- ▶▶▶ **GE** - Gaussian elimination via Lapack subroutine SGESV.
- ▶▶▶ For all additional matrix operations and algorithms such as matrix multiplication and SVD we used Lapack subroutines.
- ▶▶▶ **Traub**( $H, m$ ) - Traub-like algorithm with nodes ordered via the **Leja ordering**. (Reichel, Higham)

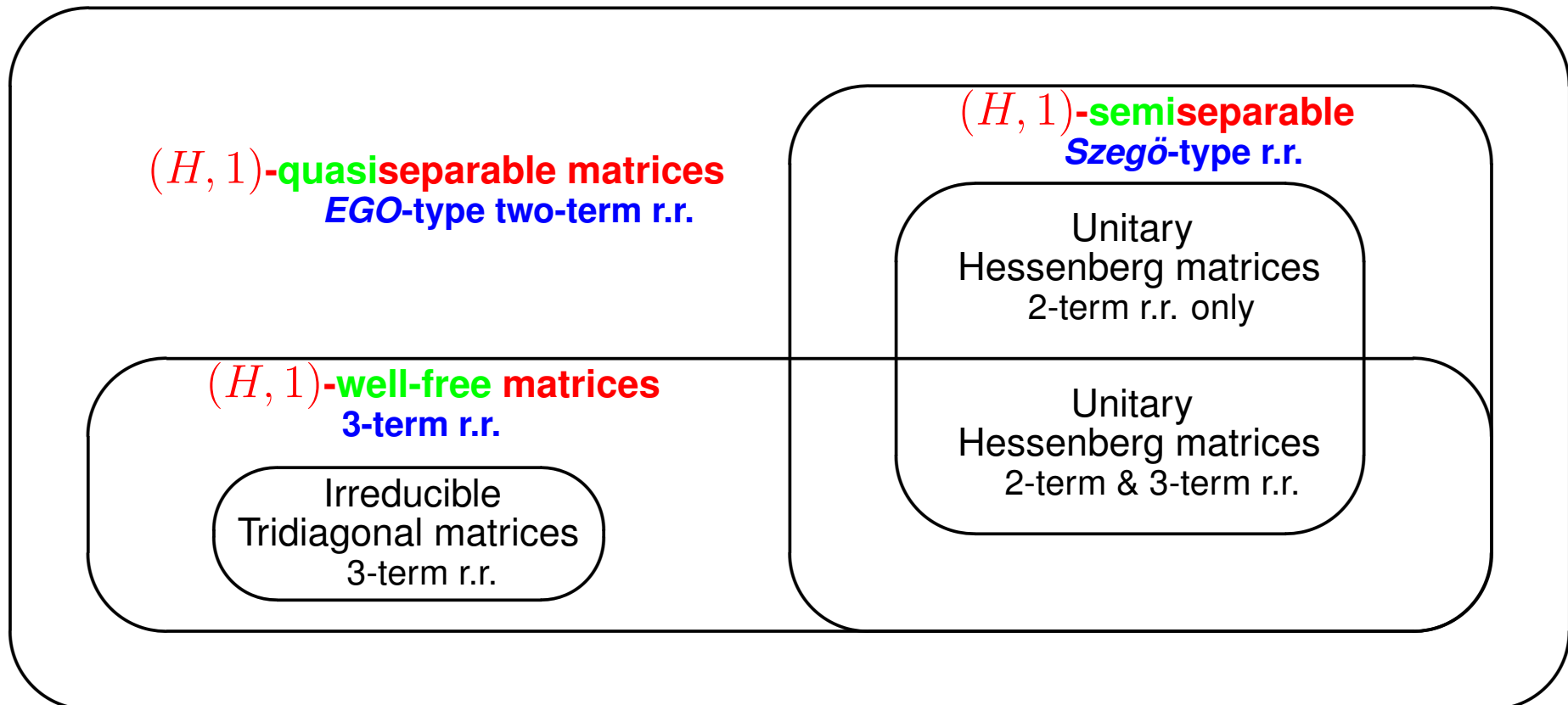
## Random generators on (0,1) & equidistant nodes on (-1,1)

<b>m</b>	n	cond(V)	GE	<b>Traub(<math>H, m</math>)</b>
<b>1</b>	10	1.8e+007	1.5e-006	1.4e-006
	20	2.2e+020	5.9e-001	5.0e-007
	30	7.2e+029	1.2e+000	2.6e-006
	40	2.1e+034	1.0e+000	1.6e-005
	50	1.5e+039	1.0e+000	3.9e-007
<b>2</b>	10	5.6e+007	4.6e-005	5.4e-007
	20	1.6e+019	2.6e+000	1.9e-007
	30	3.4e+025	9.2e-001	2.7e-006
	40	3.3e+033	1.0e+000	1.9e-006
	50	3.7e+038	1.0e+000	7.6e-001
<b>3</b>	20	1.0e+021	5.6e-002	2.3e-006
	30	2.9e+029	1.0e+000	2.0e-006
	40	4.1e+029	1.0e+000	1.1e-003
	50	2.6e+041	1.0e+000	4.5e-004

<b>m</b>	n	cond(V)	GE	<b>Traub(<math>H, m</math>)</b>
<b>4</b>	20	6.1e+020	2.8e+000	5.5e-006
	30	7.5e+026	1.0e+000	1.5e-006
	40	4.5e+028	1.0e+000	3.5e-007
	50	2.0e+037	1.0e+000	3.8e-001
<b>5</b>	30	5.0e+024	1.0e+000	1.4e-006
	40	1.2e+031	1.0e+000	7.2e-007
	50	1.7e+037	1.0e+000	5.7e-007
<b>6</b>	30	2.5e+026	1.0e+000	5.1e-007
	40	3.5e+032	1.0e+000	3.3e-006
	50	8.0e+038	1.0e+000	9.2e-005
<b>7</b>	40	6.0e+027	1.0e+000	1.7e-004
	50	1.7e+038	1.0e+000	9.3e-007
<b>8</b>	40	7.8e+031	1.0e+000	5.0e-007
	50	7.5e+036	1.0e+000	4.7e-007

**Part 4. Classification of recurrence relations for polynomials via subclasses of  $(H, m)$ -quasiseparable matrices**

## Subclasses of $(H, 1)$ -qs matrices and polynomials



## Quasiseparable matrices and polynomials

$(H, 1)$ -quasiseparable matrices



EGO-type recurrence relations

$$\begin{bmatrix} F_0(x) \\ r_0(x) \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad \begin{bmatrix} F_k(x) \\ r_k(x) \end{bmatrix} = \begin{bmatrix} \alpha_k & \beta_k \\ \gamma_k & \delta_k x + \theta_k \end{bmatrix} \begin{bmatrix} F_{k-1}(x) \\ r_{k-1}(x) \end{bmatrix}$$

## Quasiseparable matrices and polynomials

$(H, m)$ -quasiseparable matrices



Block *EGO*-type recurrence relations

$$\begin{array}{c}
 \left[ \begin{array}{c} F_0(x) \\ \hline r_0(x) \end{array} \right] = \begin{array}{c} 0 \\ \vdots \\ 0 \\ \hline 1 \end{array}, \quad \left[ \begin{array}{c} F_k(x) \\ \hline r_k(x) \end{array} \right] = \left[ \begin{array}{c|c} \alpha_k & \beta_k \\ \hline \gamma_k & \delta_k x + \theta_k \end{array} \right] \left[ \begin{array}{c} F_{k-1}(x) \\ \hline r_{k-1}(x) \end{array} \right]
 \end{array}$$

## Subclasses of $(H, m)$ -qs matrices and polynomials

$(H, m)$ -**quasiseparable** matrices  
Block *EGO*-type two-term r.r.

## Semiseparable matrices and polynomials

$(H, 1)$ -semiseparable matrices



Szegő-type recurrence relations

$$\begin{bmatrix} G_k(x) \\ r_k(x) \end{bmatrix} = \begin{bmatrix} \alpha_k & \beta_k \\ \gamma_k & 1 \end{bmatrix} \begin{bmatrix} G_{k-1}(x) \\ (\delta_k x + \theta_k) r_{k-1}(x) \end{bmatrix}$$

## Semiseparable matrices and polynomials

$(H, m)$ -semiseparable matrices



Block Szegő-type recurrence relations

$$\begin{array}{c}
 \left[ \begin{array}{c} G_0(x) \\ \hline r_0(x) \end{array} \right] = \begin{array}{c} 1 \\ \vdots \\ 1 \\ \hline 1 \end{array}, \quad \left[ \begin{array}{c} G_k(x) \\ \hline r_k(x) \end{array} \right] = \left[ \begin{array}{c|c} \alpha_k & \beta_k \\ \hline \gamma_k & 1 \end{array} \right] \left[ \begin{array}{c} G_{k-1}(x) \\ \hline (\delta_k x + \theta_k) \times \\ r_{k-1}(x) \end{array} \right]
 \end{array}$$

## Subclasses of $(H, m)$ -qs matrices and polynomials

$(H, m)$ -**quasiseparable** matrices  
Block *EGO*-type two-term r.r.

$(H, m)$ -**semiseparable**  
Block Szegő-type r.r.

## Well-free matrices and polynomials

Polynomials satisfying **three-term** recurrence relations

$$r_k(x) = (\alpha_k x - \delta_k) r_{k-1}(x) - (\boxed{\beta_k} x + \boxed{\gamma_k}) r_{k-2}(x)$$

contain both:

▣▣▣▣ Real-orthogonal polynomials:  $\boxed{\beta_k} = 0$

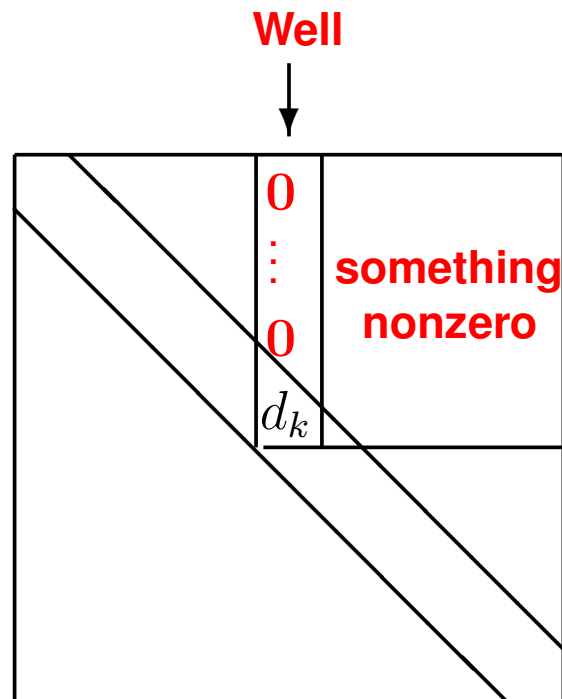
$$r_k(x) = (\alpha_k x - \delta_k) r_{k-1}(x) - \boxed{\gamma_k} r_{k-2}(x)$$

▣▣▣▣ Szegő polynomials (orthogonal on the unit circle):  $\boxed{\gamma_k} = 0$

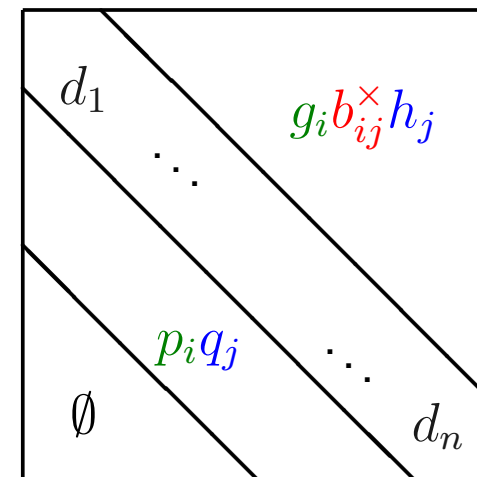
$$r_k(x) = \left( \frac{1}{\mu_k} x + \frac{\rho_k}{\rho_{k-1}} \frac{1}{\mu_k} \right) r_{k-1}(x) - \left( \frac{\rho_k}{\rho_{k-1}} \frac{\mu_{k-1}}{\mu_k} \cdot x \right) r_{k-2}(x)$$

## Well-free matrices and polynomials

Well free matrix in the  $(H, 1)$ -case is defined by



$$\text{Well-free} \Leftrightarrow h_j \neq 0$$



- ▣ Irreducible tridiagonal are well-free.
- ▣ Unitary Hessenberg ( $\rho_k \neq 0$ ) are well-free.

## Well-free matrices and polynomials

In order to generalize **three-term** recurrence relations

$$r_k(x) = (\alpha_k x - \delta_k)r_{k-1}(x) - (\beta_k x + \gamma_k)r_{k-2}(x)$$

have a look at the simplest  **$l$ -term** ones

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

and ...

**Combine!**

## Well-free matrices and polynomials

$(H, m)$ -well-free matrices ( $m = l - 2$ )



General  $l$ -term recurrence relations

$$r_0(x) = 1, \quad r_k(x) = \sum_{i=1}^k (\delta_{ik}x + \varepsilon_{ik})r_{i-1}(x), \quad k = 1, 2, \dots, l - 2$$

$$r_k(x) = \sum_{i=k-l+2}^k (\delta_{ik}x + \varepsilon_{ik})r_{i-1}(x), \quad k = l - 1, l, \dots, n$$

But what are  $(H, m)$ -well-free matrices?

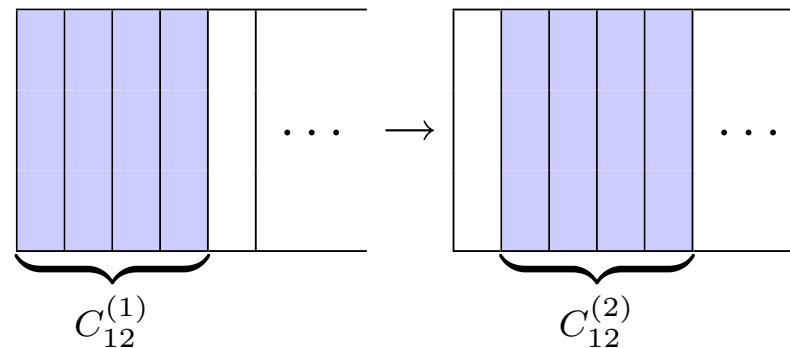
## Well-free matrices and polynomials

### $(H, m)$ -well-free matrices characterization

Let  $C$  be an  $(H, m)$ -**quasiseparable matrix**. Have a look at all its symmetric partitions  $C_{12}$  of the form

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

and a sliding window of  $m$  columns in  $C_{12}$ :



then  $C$  is  $(H, m)$ -**well-free** if

$$\text{Rank } C_{12}^{(i+1)} \leq \text{Rank } C_{12}^{(i)}$$

## Subclasses of $(H, m)$ -qs matrices and polynomials

$(H, m)$ -**quasiseparable** matrices  
Block *EGO*-type two-term r.r.

$(H, m)$ -**semiseparable**  
Block Szegő-type r.r.

$(H, m)$ -**well-free** matrices  
 $(m + 2)$ -term r.r.

## Final Venn diagram of subclasses of $(H, m)$ -quasiseparable matrices and polynomials

