

# Fast matrix arithmetic for rank structured matrices using non hierarchical representations

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# Quasiseparable matrices

## Definition

$M \in \mathbb{K}^{n \times n}$  is  $(r_L, r_U)$ -quasiseparable if  $\begin{cases} \text{rank}(M_{k+1..n,1..k}) \leq r_L \\ \text{rank}(M_{1..k,k+1..n}) \leq r_U \end{cases}$

## Properties

- ▶  $(s, t)\text{-QS} \times (u, v)\text{-QS} = (s + u, t + v)\text{-QS}$
- ▶  $((s, t)\text{-QS})^{-1} = (s, t)\text{-QS}$

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

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(Note: The matrix on the right is partitioned into four quadrants: top-left is light green, top-right is light purple, bottom-left is light blue, and bottom-right is light orange.)













## Applications

Numerical linear algebra: naturally occur in solving

- ▶ generalized eigenvalue problems,
- ▶ Fast Multipole Method ( $n$ -body simulation)

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## Efficient representation: take advantage of the structure

Space: ideally, storage linear in  $n$  and  $s$

Structure: recursive (hierarchical) or flat

Arithmetic cost for most common ops:

- ▶  $QS \times \text{Vect.}$
- ▶  $QS \times \text{Tall-Skinny matrix } (s \text{ Vect.})$ .
- ▶  $QS \times QS$ .
- ▶ LU, QR factorization, LinSys

# Structured representation of a quasiseparable matrix

Sequentially SemiSeparable (SSS) [Eidelman Gohberg 99] [5]

$$M_{i,j} = \begin{cases} p(i)^T a(i-1) \dots a(j+1) q(j), & 1 \leq j < i \leq n \\ d(i), & 1 \leq i = j \leq n \\ g(i)^T b(i+1) \dots b(i-1) h(j), & 1 \leq i < j \leq n \end{cases}$$

where  $p(i), q(i), g(i), h(i) \in \mathbb{K}^s$ ,  $a(i), b(i) \in \mathbb{K}^{s \times s}$  for  $1 \leq i \leq n-1$

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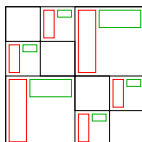
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LU/QR/LinSys	$O(s^3 n)$

# Structured representation of a quasiseparable matrix

## Hierarchical ( $\mathcal{H}$ ) [6]



	SSS [5]	$\mathcal{H}$ [6]
Size	$O(s^2n)$	$O(sn \lg \frac{n}{s})$
Construction	$O(s^2n^2)$	$O(sn^2)$
QS $\times$ Vec	$O(s^2n)$	$O(sn \lg \frac{n}{s})$
QS $\times$ $s$ Vec	$O(s^3n)$	$O(s^2n \lg \frac{n}{s})$
QS $\times$ QS	$O(s^3n)$	$O(s^2n \lg^2 \frac{n}{s})$
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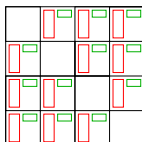
Hierarchically SemiSeparable (HSS) [2]: cf Fast Multipole Method



	SSS [5]	$\mathcal{H}$ [6]	HSS [2]
Size	$O(s^2n)$	$O(sn \lg \frac{n}{s})$	$O(sn)$
Construction	$O(s^2n^2)$	$O(sn^2)$	$O(sn^2)$
QS $\times$ Vec	$O(s^2n)$	$O(sn \lg \frac{n}{s})$	$O(sn)$
QS $\times$ $s$ Vec	$O(s^3n)$	$O(s^2n \lg \frac{n}{s})$	$O(s^2n)$
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LU/QR/LinSys	$O(s^3n)$	$O(s^2n \lg^2 \frac{n}{s})$	$O(s^2n)$

# Structured representation of a quasiseparable matrix

## Block Low Rank (BLR) [1]



	SSS [5]	$\mathcal{H}$ [6]	HSS [2]	BLR [1]
Size	$O(s^2n)$	$O(sn \lg \frac{n}{s})$	$O(sn)$	$O(s^{0.5}n^{1.5})$
Construction	$O(s^2n^2)$	$O(sn^2)$	$O(sn^2)$	$O(sn^2)$
QS $\times$ Vec	$O(s^2n)$	$O(sn \lg \frac{n}{s})$	$O(sn)$	$O(s^{0.5}n^{1.5})$
QS $\times$ $s$ Vec	$O(s^3n)$	$O(s^2n \lg \frac{n}{s})$	$O(s^2n)$	$O(s^{1.5}n^{1.5})$
QS $\times$ QS	$O(s^3n)$	$O(s^2n \lg^2 \frac{n}{s})$	$O(s^2n)$	$O(sn^2)$
LU/QR/LinSys	$O(s^3n)$	$O(s^2n \lg^2 \frac{n}{s})$	$O(s^2n)$	$O(sn^2)$

## Reduction to dense matrix multiplication

- ▶ Improving theoretical complexities complexities:  $O(n^\omega)$ 
  - $\omega = 3$  schoolbook multiplication
  - $\omega \approx 2.8074$  [Strassen 69]
  - $\omega \approx 2.3729$  [Le Gall 14]
- ▶ Efficiency in practice: gather arithmetic in compute intensive kernels



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## Flat representations

- ▶ much better suited for large scale parallelization

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## Flat representations

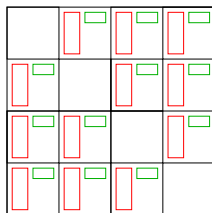
- ▶ much better suited for large scale parallelization

## Contributions

- ▶ speed-up BLR with fast matrix multiplication;
- ▶ Connecting notions of **quasiseparability** and **rank profile matrix** [3];
- ▶ A new flat representation matching the HSS complexities.

# Contributions

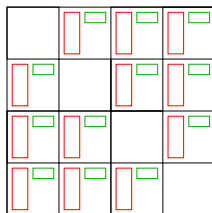
## Block Low Rank (BLR) using fast matrix multiplication



	...	HSS [2]	BLR [1]
Size	...	$O(sn)$	$O(s^{0.5}n^{1.5})$
Construction	...	$O(sn^2)$	$O(sn^2)$
QS $\times$ Vec	...	$O(sn)$	$O(s^{0.5}n^{1.5})$
QS $\times$ $s$ Vec	...	$O(s^2n)$	$O(sn^2)$
QS $\times$ QS	...	$O(s^2n)$	$O(sn^2)$
LU/QR/LinSys	...	$O(s^2n)$	$O(sn^2)$

# Contributions

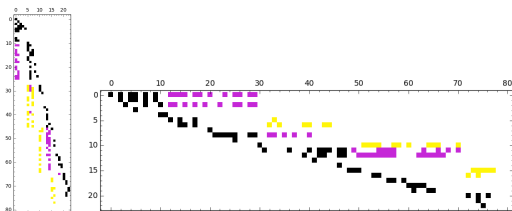
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Construction	...	$O(sn^2)$	$O(sn^2)$	$O(sn^2)$
QS $\times$ Vec	...	$O(sn)$	$O(s^{0.5}n^{1.5})$	$O(s^{0.5}n^{1.5})$
QS $\times$ $s$ Vec	...	$O(s^2n)$	$O(sn^2)$	$O(s^{\omega-1.5}n^{1.5})$
QS $\times$ QS	...	$O(s^2n)$	$O(sn^2)$	$O(s^{\frac{\omega-1}{2}}n^{\frac{\omega+1}{2}})$
LU/QR/LinSys	...	$O(s^2n)$	$O(sn^2)$	$O(s^{\frac{\omega-1}{2}}n^{\frac{\omega+1}{2}})$

# Contributions

## Compact Bruhat generator



	...	HSS [2]	BLR [1]	fast BLR	CB
Size	...	$O(sn)$	$O(s^{0.5}n^{1.5})$	$O(s^{0.5}n^{1.5})$	$O(sn)$
Construction	...	$O(sn^2)$	$O(sn^2)$	$O(sn^2)$	$O(s^{\omega-2}n^2)$
QS $\times$ Vec	...	$O(sn)$	$O(s^{0.5}n^{1.5})$	$O(s^{0.5}n^{1.5})$	$O(sn)$
QS $\times$ $s$ Vec	...	$O(s^2n)$	$O(sn^2)$	$O(s^{\omega-1.5}n^{1.5})$	$O(s^{\omega-1}n)$
QS $\times$ QS	...	$O(s^2n)$	$O(sn^2)$	$O(s^{\frac{\omega-1}{2}}n^{\frac{\omega+1}{2}})$	$O(s^{\omega-1}n)$
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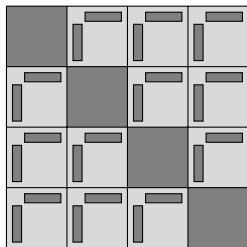
# Outline

- 1 Block Low Rank representation
  - First attempt
  - New approach
- 2 The compact Bruhat representation
  - The rank profile matrix
  - Warm-up : computing the orders of quasiseparability
  - The Bruhat generator
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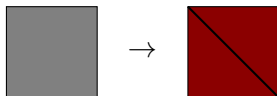
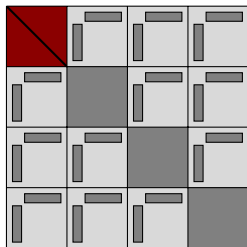
# BLR matrix LU factorization: classical algorithm



Kernel costs:



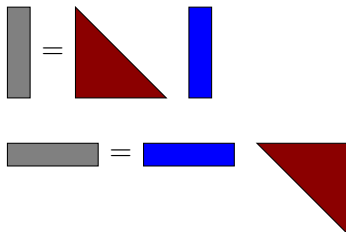
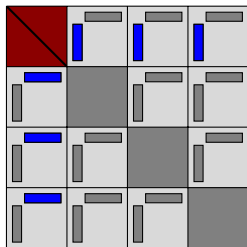
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Kernel costs:

- ▶ **Factor** kernel:  $O(b^3)$

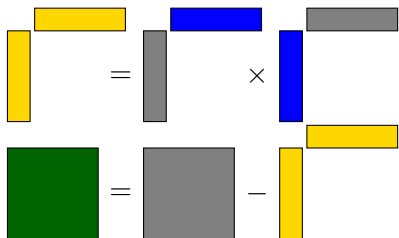
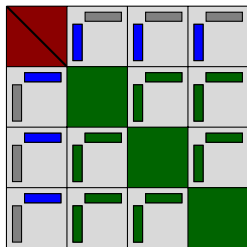
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Kernel costs:

- ▶ **Factor** kernel:  $O(b^3)$
- ▶ **Solve** kernel:  $O(b^2r)$

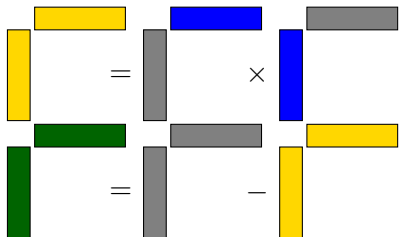
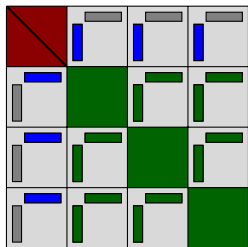
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Kernel costs:

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- ▶ **Solve** kernel:  $O(b^2r)$
- ▶ **Update** kernel:
  - ▷ FR target:  $O(b^2r)$

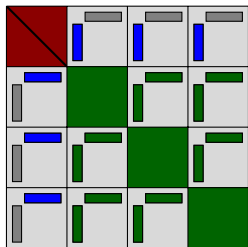
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  - ▷ LR target:  $O(br^2)$

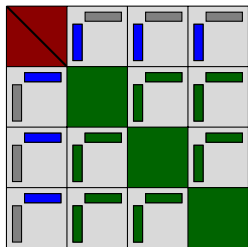
## BLR matrix LU factorization: classical algorithm



Kernel costs:

- ▶ **Factor** kernel:  $O(b^3) \rightarrow O(b^\omega)$
- ▶ **Solve** kernel:  $O(b^2r) \rightarrow O(b^2r^{\omega-2})$
- ▶ **Update** kernel:
  - ▷ FR target:  $O(b^2r) \rightarrow O(b^2r^{\omega-2})$
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## BLR matrix LU factorization: classical algorithm



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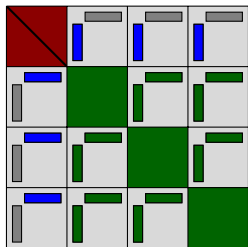
$$O(pb^\omega + p^2b^2r^{\omega-2} + p^3br^{\omega-1}) \\ \subset O(n^2r^{\omega-2}) \quad \text{for } b = \Theta(\sqrt{nr})$$

- ▶ deceptive speed-up ( $r^{3-\omega}$ )
- ▶ granularity of LR is too low

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## BLR matrix LU factorization: classical algorithm



Total:

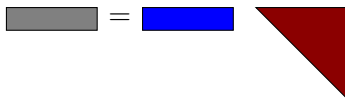
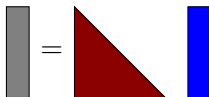
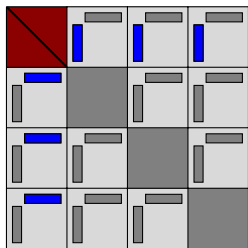
$$O(pb^\omega + p^2b^2r^{\omega-2} + p^3br^{\omega-1}) \\ \subset O(n^2r^{\omega-2}) \quad \text{for } b = \Theta(\sqrt{nr})$$

- ▶ deceptive speed-up ( $r^{3-\omega}$ )
- ▶ granularity of LR is too low
- ▶ How to reduce the exponent in  $n$ ?

Kernel costs:

- ▶ **Factor** kernel:  $O(b^3) \rightarrow O(b^\omega)$
- ▶ **Solve** kernel:  $O(b^2r) \rightarrow O(b^2r^{\omega-2})$
- ▶ **Update** kernel:
  - ▷ FR target:  $O(b^2r) \rightarrow O(b^2r^{\omega-2})$
  - ▷ LR target:  $O(br^2) \rightarrow O(br^{\omega-1})$

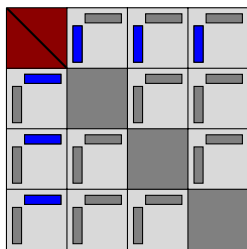
# New Solve kernel



- ▶ Classical:  $O(p^2)$  calls of cost  $O(b^2 r^{\omega-2}) \Rightarrow O(p^2 b^2 r^{\omega-2})$

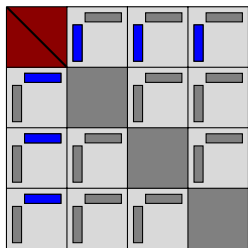


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# New Solve kernel



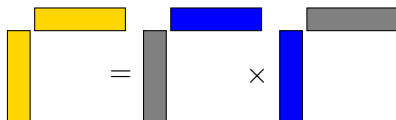
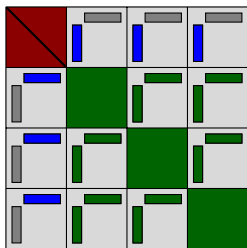
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▶ New:  $O(p)$  calls of cost  $O(b^{\omega-1} p r) \Rightarrow O(p^2 b^{\omega-1} r)$

$\Rightarrow$  Reduction by a factor  $O((b/r)^{3-\omega})$ :

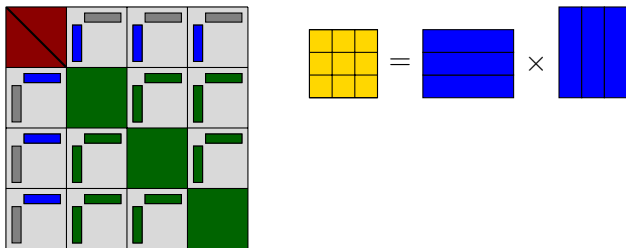
- ▷ no asymptotic gain if  $\omega = 3$  (we only rearranged computations)
- ▷ no gain if  $r \sim b$  (good enough granularity... but  $O(n^\omega)$  complexity)
- ▷ possibly large asymptotic gain in general!

# New Update kernel



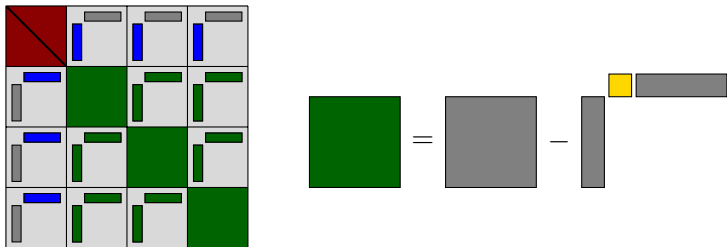
- ▶ Classical product:  $O(p^3)$  calls of cost  $O(br^{\omega-1}) \Rightarrow O(p^3 br^{\omega-1})$

# New Update kernel



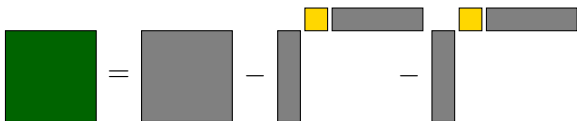
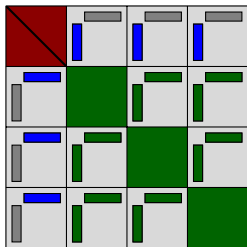
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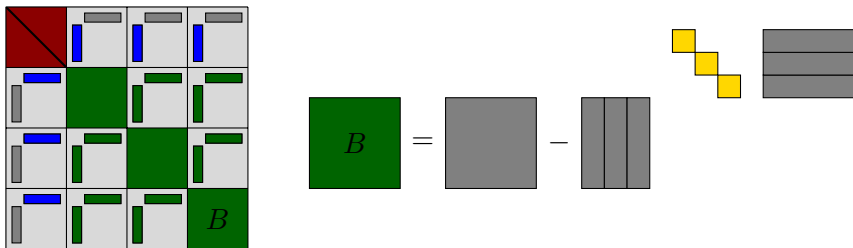
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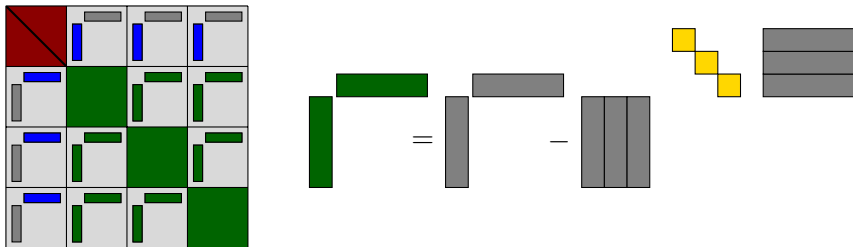
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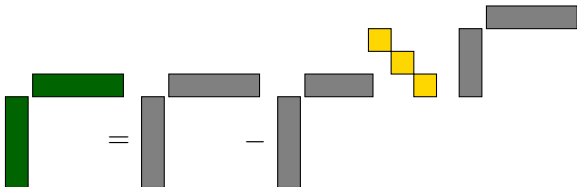
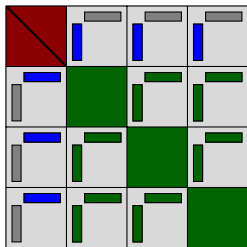
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- $\Rightarrow$  Reduction by a factor  $O((b/r)^{3-\omega})$
- ▶ LR target subtractions:  $O(p^3br^{\omega-1}) \rightarrow O(p^2br^{\omega-1} + p^3r^{\omega})$
- $\Rightarrow$  Reduction by a factor  $O(b/r) \Rightarrow$  gain even for  $\omega = 3!$

# Complexity of the new algorithm

- ▶ Putting everything together, the complexity of the new algorithm is:  
$$O(pb^\omega + p^2b^{\omega-1}r + p^3b^{\omega-2}r^2)$$
$$\subset O(n^{(\omega+1)/2}r^{(\omega-1)/2}) \quad \text{for } b = \Theta((nr)^{1/2})$$

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$$\subset O(n^{(\omega+1)/2}r^{(\omega-1)/2}) \quad \text{for } b = \Theta((nr)^{1/2})$$
- ▶  $\omega = 3 \Rightarrow O(n^2r)$
- ▶  $\omega = 2 \Rightarrow O(n^{3/2}r^{1/2}) \equiv \text{BLR storage complexity} \Rightarrow \text{nice!}$
- ▶  $\omega = \log_2 7 \Rightarrow O(n^{1.9}r^{0.9})$

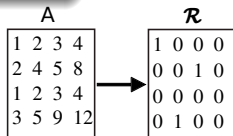
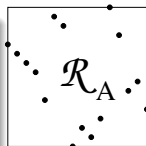
# Outline

- 1 Block Low Rank representation
  - First attempt
  - New approach
- 2 The compact Bruhat representation
  - The rank profile matrix
  - Warm-up : computing the orders of quasiseparability
  - The Bruhat generator
  - The Compact Bruhat generator

# The rank profile matrix [Dumas, P. and Sultan'15]

## Definition (Rank Profile matrix)

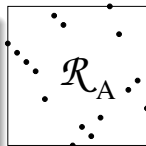
The unique  $\mathcal{R}_A \in \{0, 1\}^{m \times n}$  such that any pair of  $(i, j)$ -leading sub-matrix of  $\mathcal{R}_A$  and of  $A$  have the same rank.



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## Theorem

- ▶ *RowRP and ColRP read directly on  $\mathcal{R}_A$*
- ▶ *Same holds for any  $(i, j)$ -leading submatrix.*

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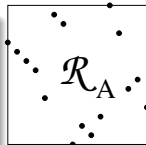
$$\text{RowRP} = \{1\}$$

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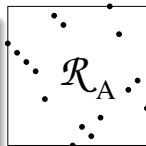
$$\text{RowRP} = \{1, 2\}$$

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$$\text{RowRP} = \{1, 4\}$$

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# Computing the rank profile matrix

From a PLUQ decomposition

$$A = PLUQ = P \begin{bmatrix} L & 0 \\ M & I_{m-r} \end{bmatrix} \begin{bmatrix} I_r & \\ & 0 \end{bmatrix} \begin{bmatrix} U & V \\ & I_{n-r} \end{bmatrix} Q$$

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$$A = PLUQ = P \begin{bmatrix} L & 0 \\ M & I_{m-r} \end{bmatrix} P^T P \begin{bmatrix} I_r & \\ & 0 \end{bmatrix} QQ^T \begin{bmatrix} U & V \\ & I_{n-r} \end{bmatrix} Q$$

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## Theorem (Dumas, P. and Sultan 15)

- ▶ *The RPM can be recovered from a PLUQ decomposition computed by any Gaussian elimination algorithm under some conditions on its pivoting strategy.*
- ▶ *in time  $O(mnr^{\omega-2})$*
- ▶ *in-place*

# Finding the quasiseparability orders

## Left triangular matrices

Of the form: 
$$\begin{bmatrix} * & * & * & * & 0 \\ * & * & * & 0 & \\ * & * & 0 & & \\ * & 0 & & & \\ 0 & & & & \end{bmatrix}$$

Examples: 
$$\begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix} \times \text{Lower or Upper} \times \begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix}$$

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

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

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Examples: 
$$\begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix} \times \text{Lower or Upper} \times \begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix}$$

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

# Finding the quasiseparability orders

## Left triangular matrices

Of the form: 
$$\begin{bmatrix} * & * & * & * & 0 \\ * & * & * & 0 \\ * & * & 0 \\ * & 0 \\ 0 \end{bmatrix}$$

Examples: 
$$\begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix} \times \text{Lower or Upper} \times \begin{bmatrix} & & & & 1 \\ & & & & \\ & & & & \\ & & & & \\ 1 & & & & \end{bmatrix}$$

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



# Finding the quasiseparability orders

## Left triangular matrices

Of the form: 
$$\begin{bmatrix} * & * & * & * & 0 \\ * & * & * & 0 \\ * & * & 0 \\ * & 0 \\ 0 \end{bmatrix}$$

Examples: 
$$\begin{bmatrix} & & & 1 \\ & & \cdot & \\ & & & \\ 1 & & & \end{bmatrix} \times \text{Lower or Upper} \times \begin{bmatrix} & & & 1 \\ & & \cdot & \\ & & & \\ 1 & & & \end{bmatrix}$$

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

# Finding the quasiseparability orders

## Left triangular matrices

Of the form: 
$$\begin{bmatrix} * & * & * & * & 0 \\ * & * & * & 0 \\ * & * & 0 \\ * & 0 \\ 0 \end{bmatrix}$$

Examples: 
$$\begin{bmatrix} & & & & & & 1 \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ 1 & & & & & & \end{bmatrix} \times \text{Lower or Upper} \times \begin{bmatrix} & & & & & & 1 \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ 1 & & & & & & \end{bmatrix}$$

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

# Quasiseparability orders on the rank profile matrix

$$\begin{bmatrix}
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{1} & \boxed{1} & \boxed{1} & \boxed{0} & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 1 & 2 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 2 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix},
 \begin{bmatrix}
 \boxed{1} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix}$$

Rank = 1

Quasiseparability order = 1

# Quasiseparability orders on the rank profile matrix

$$\begin{bmatrix}
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{1} & \boxed{1} & \boxed{1} & 0 & 0 \\
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{0} & \boxed{1} & \boxed{2} & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 2 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix},
 \begin{bmatrix}
 \boxed{1} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & 0 & 0 \\
 \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{1} & \boxed{0} & \boxed{0} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix}$$

Rank = 2

Quasiseparability order = 2

# Quasiseparability orders on the rank profile matrix

$$\begin{bmatrix}
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{1} & \boxed{1} & 1 & 0 & 0 \\
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{0} & \boxed{1} & 2 & 0 & 0 \\
 \boxed{2} & \boxed{2} & \boxed{1} & \boxed{0} & \boxed{2} & \boxed{2} & \boxed{1} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix},
 \begin{bmatrix}
 \boxed{1} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & 0 & 0 & 0 \\
 \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{1} & \boxed{0} & 0 & 0 & 0 \\
 \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & \boxed{0} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix}$$

Rank = 2

Quasiseparability order = 2

# Quasiseparability orders on the rank profile matrix

$$\left[ \begin{array}{cccccc|cccc}
 2 & 2 & 1 & 0 & 2 & 1 & 1 & 1 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 1 & 2 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 2 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right], \quad
 \left[ \begin{array}{cccccc|cccc}
 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right]$$

Rank = 1

Quasiseparability order = 2

# Quasiseparability orders on the rank profile matrix

$$\left[ \begin{array}{cccccc|cccc}
 2 & 2 & 1 & 0 & 2 & 1 & 1 & 1 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 1 & 2 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 2 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right], \quad
 \left[ \begin{array}{cccccc|cccc}
 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right]$$

Rank = 1

Quasiseparability order = 2

## Major difficulty

Complexity in  $O(n^2 r^{\omega-2})$  where  $r = \text{rank}(A) \gg r_L, r_U$ . But

# Quasiseparability orders on the rank profile matrix

$$\left[ \begin{array}{cccccc|cccc}
 2 & 2 & 1 & 0 & 2 & 1 & 1 & 1 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 1 & 2 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 2 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\
 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right], \quad
 \left[ \begin{array}{cccccc|cccc}
 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\
 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{array} \right]$$

Rank = 1

Quasiseparability order = 2

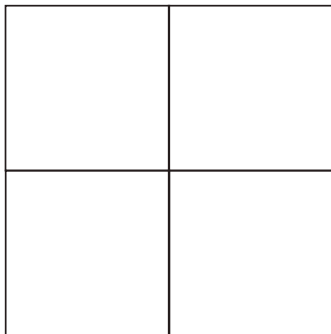
## Major difficulty

Complexity in  $O(n^2 r^{\omega-2})$  where  $r = \text{rank}(A) \gg r_L, r_U$ . But

- ▶ only a few pivots ( $O(r_L, r_U)$ ) near the top left corner
- ▶ when numerous, most pivots must be near the anti-diagonal  
 $\rightsquigarrow$  cheaper elimination

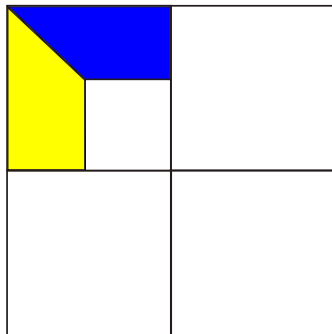


# Computing the left-triangular part of a rank profile matrix



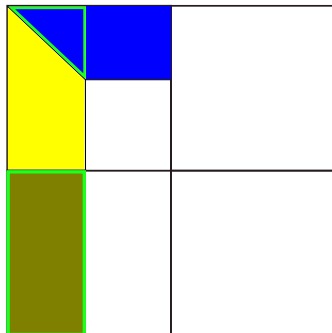
$2 \times 2$  block splitting

# Computing the left-triangular part of a rank profile matrix



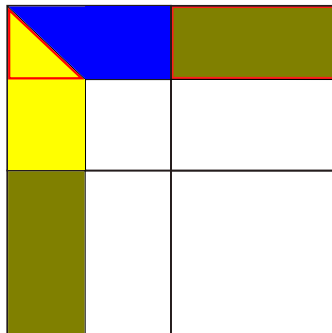
Recursive call

# Computing the left-triangular part of a rank profile matrix



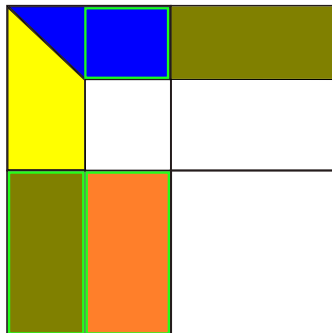
$$\text{TRSM: } B \leftarrow BU^{-1}$$

# Computing the left-triangular part of a rank profile matrix



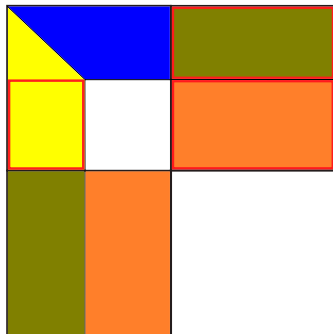
$$\text{TRSM: } B \leftarrow L^{-1}B$$

# Computing the left-triangular part of a rank profile matrix



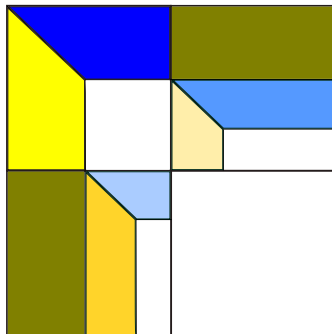
MatMul:  $C \leftarrow C - A \times B$

# Computing the left-triangular part of a rank profile matrix



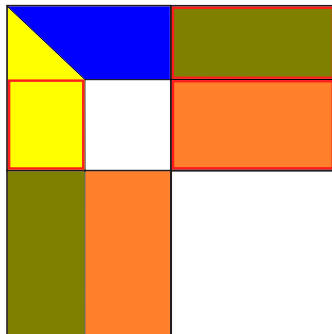
$$\text{MatMul: } C \leftarrow C - A \times B$$

# Computing the left-triangular part of a rank profile matrix



2 recursive calls : not possible as no longer left-triangular

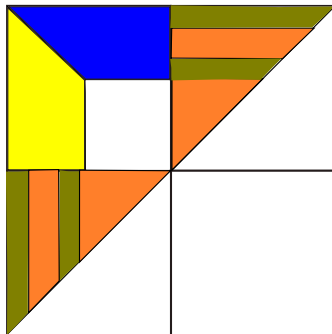
# Computing the left-triangular part of a rank profile matrix



2 recursive calls : not possible as no longer left-triangular

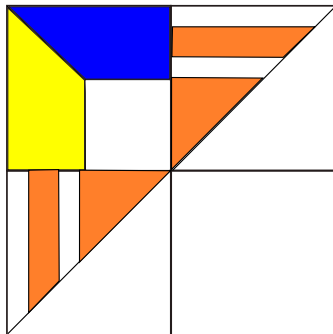


# Computing the left-triangular part of a rank profile matrix



2 recursive calls : not possible as no longer left-triangular  
 $\rightsquigarrow$  permute back to original position

# Computing the left-triangular part of a rank profile matrix

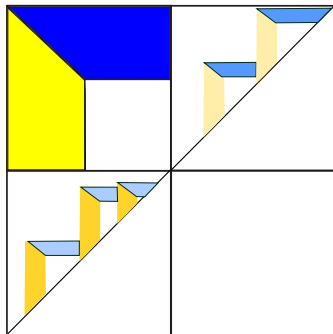


2 recursive calls : not possible as no longer left-triangular

↪ permute back to original position

↪ clear out cols/rows already processed

# Computing the left-triangular part of a rank profile matrix



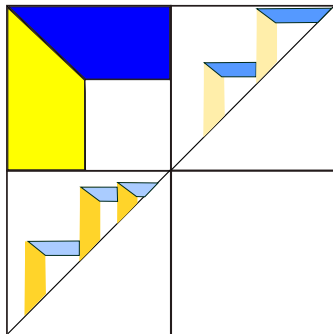
2 recursive calls : not possible as no longer left-triangular

↪ permute back to original position

↪ clear out cols/rows already processed

recursive call

# Computing the left-triangular part of a rank profile matrix



2 recursive calls : not possible as no longer left-triangular

↪ permute back to original position

↪ clear out cols/rows already processed

recursive call

- ▶  $O(s^{\omega-2}n^2)$
- ▶ in place

# The Bruhat generator

Idea: *When numerous, most pivots must lie near the anti-diagonal.*

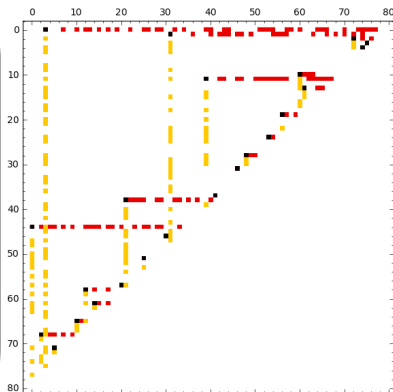
## Definition (Bruhat Generator)

From a PLUQ decomposition revealing the rank profile matrix  $E$  : the triple  $(\mathcal{L}, \mathcal{E}, \mathcal{U})$

$$\mathcal{L} = \text{Left}(P \begin{bmatrix} L & 0 \end{bmatrix} Q),$$

$$\mathcal{E} = \text{Left}(E),$$

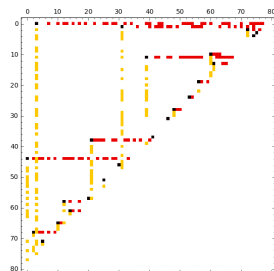
$$\mathcal{U} = \text{Left}(P \begin{bmatrix} U \\ 0 \end{bmatrix} Q).$$



# The Bruhat Generator

For a quasiseparability order  $s$ :

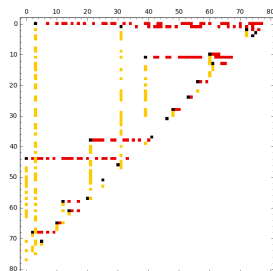
- ▶ ✓ Generator:  $A = \text{Left}(\mathcal{L}\mathcal{E}^T\mathcal{U})$



# The Bruhat Generator

For a quasiseparability order  $s$ :

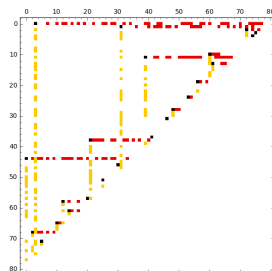
- ▶ ✓ Generator:  $A = \text{Left}(\mathcal{L}\mathcal{E}^T\mathcal{U})$
- ▶ ✓ Size:  $O(ns)$



# The Bruhat Generator

For a quasiseparability order  $s$ :

- ▶ ✓ Generator:  $A = \text{Left}(\mathcal{L}\mathcal{E}^T\mathcal{U})$
- ▶ ✓ Size:  $O(ns)$
- ▶ ✓ Computed in  $O(n^2 s^{\omega-2})$   
 $\rightsquigarrow$  Adapted from the tile rec. PLUQ algo  
 (same as for the QS orders)

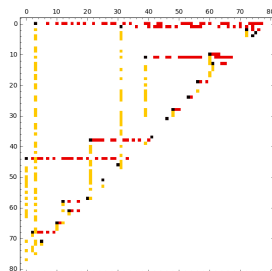




# The Bruhat Generator

For a quasiseparability order  $s$ :

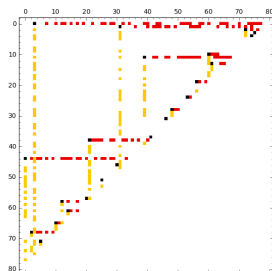
- ▶ ✓ Generator:  $A = \text{Left}(\mathcal{L}\mathcal{E}^T\mathcal{U})$
- ▶ ✓ Size:  $O(ns)$
- ▶ ✓ Computed in  $O(n^2 s^{\omega-2})$   
 $\rightsquigarrow$  Adapted from the tile rec. PLUQ algo  
 (same as for the QS orders)
- ▶ ✓ QuasiSep  $\times$  Vector in  $O(\text{Size}) = O(ns)$



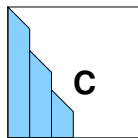
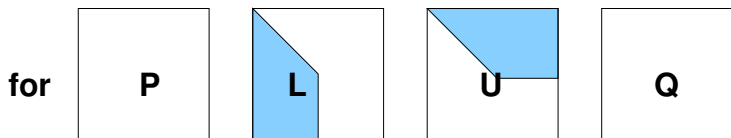
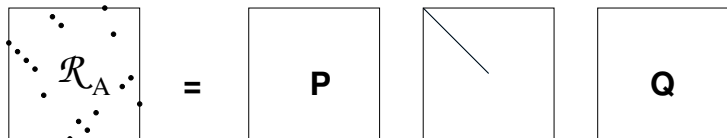
# The Bruhat Generator

For a quasiseparability order  $s$ :

- ▶ ✓ Generator:  $A = \text{Left}(\mathcal{L}\mathcal{E}^T\mathcal{U})$
- ▶ ✓ Size:  $O(ns)$
- ▶ ✓ Computed in  $O(n^2 s^{\omega-2})$   
 $\rightsquigarrow$  Adapted from the tile rec. PLUQ algo  
 (same as for the QS orders)
- ▶ ✓ QuasiSep  $\times$  Vector in  $O(\text{Size}) = O(ns)$
- ▶ ✗ Scattered  $\rightsquigarrow$  no fast matrix multiplication

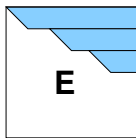


# Row and column echelon forms from PLUQ



$$\mathbf{C} = \mathbf{P}\mathbf{L}\mathbf{P}_s$$

sort



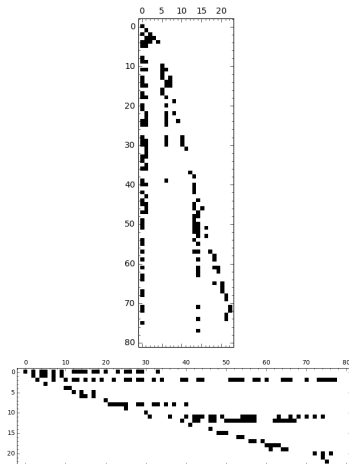
$$\mathbf{Q}_s \mathbf{U} \mathbf{Q} = \mathbf{E}$$

# The compact Bruhat Generator

- ▶  $C = \mathcal{L}Q$ : Col. permutation to column echelon form
- ▶  $E = \mathcal{P}U$ : Row permutation to row echelon form

$$A = \text{Left}(\text{CRE})$$

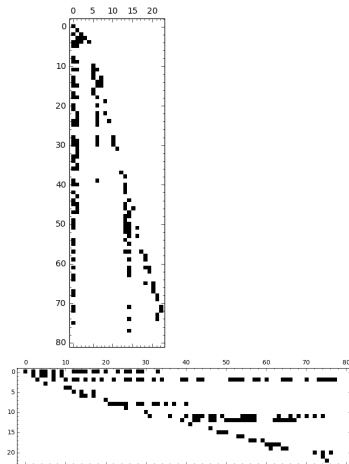
for an  $r \times r$  permutation  $R$ .  
 (generalized Bruhat decomposition  
 [Manthey Helmke 07][7]).



# The compact Bruhat Generator

## Major difficulty

Again  $r = \text{rank}(A) \gg s$

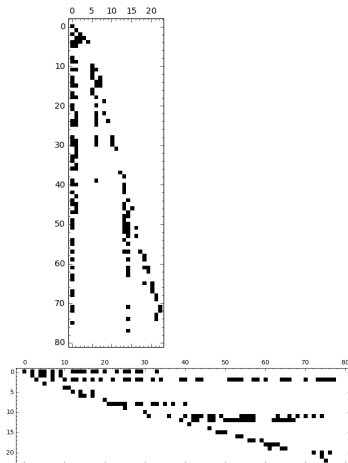


# The compact Bruhat Generator

## Major difficulty

Again  $r = \text{rank}(A) \gg s$

- ▶ But double structure: either echelon or Left-triangular

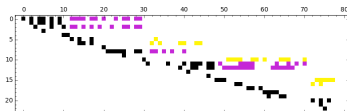
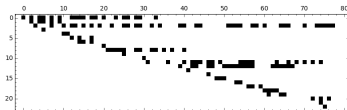
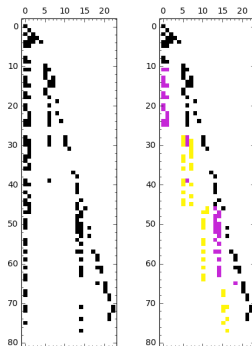


# The compact Bruhat Generator

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Again  $r = \text{rank}(A) \gg s$

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- ▶ Fold them in blocks of size  $s$

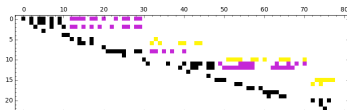
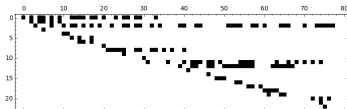
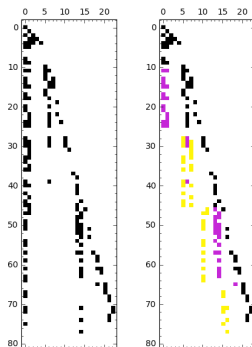


# The compact Bruhat Generator

## Major difficulty

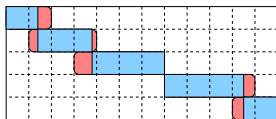
Again  $r = \text{rank}(A) \gg s$

- ▶ But double structure: either echelon or Left-triangular
- ▶ Fold them in blocks of size  $s$
- ▶ forming a **dense**  $O(ns)$  block bi-diagonal structure



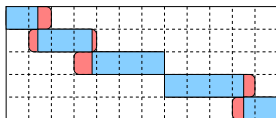


# Complexities



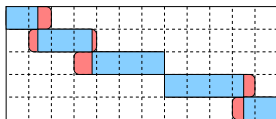
- ▶  $QS \times \text{Vector}$ :  $O(ns)$

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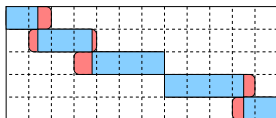
- ▶  $QS \times \text{Vector}$ :  $O(ns)$
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- ▶  $LU(QS)$ :  $O(ns^{\omega-1})$  using randomize recompression [in preparation]
- ▶  $\text{Solve}(QS)$ :  $O(ns^{\omega-1})$  using randomize recompression [in preparation]

# Perspectives

## Faster construction

- ▶ [Storjohann Yang 15][9] computes RowRP in Las-Vegas  $\mathcal{O}(|A| + r^\omega)$
- ▶ [Dumas P. Sultan 16][4] extends it to RPM

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- ▶ Does it applies to computing
  - ▷ the QS orders ?
  - ▷ a CB generator ?

## Numerical stability

- ▶ related to the use of subcubic matrix multiplication
- ▶ related to the pivoting strategy relieving the RPM

# References



P. Amestoy et al. "Improving Multifrontal Methods by Means of Block Low-Rank Representations". In: *SIAM Journal on Scientific Computing* 37.3 (Jan. 2015), A1451–A1474. DOI: 10.1137/120903476.



S. Chandrasekaran, M. Gu, and T. Pals. "A Fast ULV Decomposition Solver for Hierarchically Semiseparable Representations". In: *SIAM Journal on Matrix Analysis and Applications* 28.3 (Jan. 2006), pp. 603–622. DOI: 10.1137/S0895479803436652.



J.-G. Dumas, C. Pernet, and Z. Sultan. "Computing the Rank Profile Matrix". In: *Proc. ISSAC'15*. Bath, United Kingdom: ACM, 2015, pp. 149–156. DOI: 10.1145/2755996.2756682.



J.-G. Dumas, C. Pernet, and Z. Sultan. "Fast Computation of the Rank Profile Matrix and the Generalized Bruhat Decomposition". In: *Journal of Symbolic Computation* (2016). arXiv:1601.01798 [cs.SC].



Y. Eidelman and I. Gohberg. "On a new class of structured matrices". en. In: *Integral Equations and Operator Theory* 34.3 (Sept. 1999), pp. 293–324. DOI: 10.1007/BF01300581.



W. Hackbusch. *Hierarchical Matrices: Algorithms and Analysis*. Vol. 49. Dec. 2015. DOI: 10.1007/978-3-662-47324-5.



W. Mantey and U. Helmke. "Bruhat canonical form for linear systems". In: *Linear Algebra and its Applications* 425.2–3 (2007). Special Issue in honor of Paul Fuhrmann, pp. 261–282. DOI: 10.1016/j.laa.2007.01.022.



Z. Sheng, P. Dewilde, and S. Chandrasekaran. "Algorithms to Solve Hierarchically Semi-separable Systems". en. In: *System Theory, the Schur Algorithm and Multidimensional Analysis*. Ed. by D. Alpay and V. Vinnikov. Operator Theory: Advances and Applications 176. DOI: 10.1007/978-3-7643-8137-0\_5. Birkhäuser Basel, 2007, pp. 255–294.

