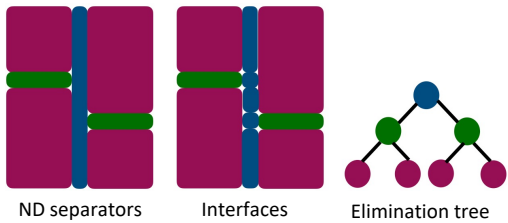


## Overview

- Fast, sparse QR factorization to solve
  - Sparse linear systems:  $Ax = b$ ,  $A \in \mathcal{R}^{N \times N}$
  - Sparse linear LS problems:  $\min_x \|Ax - b\|^2$ ,  $A \in \mathcal{R}^{M \times N}$ ,  $M \geq N$
- Based on a multifrontal QR factorization
- Matrix reordered using Nested Dissection



- Key assumption: Fill-in blocks are low rank
- Sparsification of one interface  $p$ :

$$A_p = \begin{bmatrix} A_{pp} & A_{pn} \\ A_{np} & A_{nn} \end{bmatrix} \rightarrow \begin{bmatrix} I & \bar{A}_{p1n} \\ \bar{A}_{np} & A_{nn} \end{bmatrix} \rightarrow \begin{bmatrix} I_f & & & \\ & I_c & & \\ & & W_{fn}^{(1)T} & W_{cn}^{(1)T} \\ & & W_{fn}^{(2)} & W_{cn}^{(2)} \\ & & W_{fn}' & W_{cn}' \\ & & & A_{nn} \end{bmatrix}$$

using a rank revealing QR on,

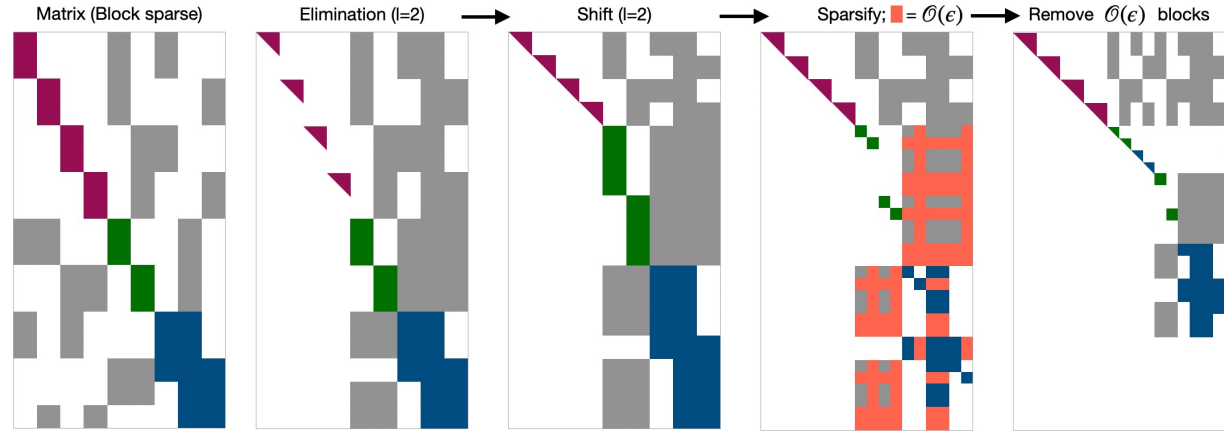
$$A_{p2n} = \begin{bmatrix} Q'_{pc} & Q'_{pf} \end{bmatrix} \begin{bmatrix} W'_{cn} \\ W'_{fn} \end{bmatrix}$$

$$\begin{bmatrix} A_{np}^T & A_{p1n} \end{bmatrix} = \begin{bmatrix} Q_{pf} & Q_{pc} \end{bmatrix} \begin{bmatrix} W_{fn}^{(1)} & W_{fn}^{(2)} \\ W_{cn}^{(1)} & W_{cn}^{(2)} \end{bmatrix}$$

where,  $\|W_{fn}^{(1)}\|_2 \approx \|W_{fn}^{(2)}\|_2 \approx \|W'_{fn}\|_2 = \mathcal{O}(\epsilon)$

## Algorithm: Sparsified QR (spaQR)

**Require:** Matrix  $A \in \mathbb{R}^{M \times N}$ ,  $M \geq N$   
**Require:** Tolerance  $\epsilon$   
**for all**  $l = L - 1, L - 2, \dots, 0$  **do**  
 1. Factorize separators at level  $l$   
 2. Shift rows below diagonal of eliminated separators  
 3. Scale, Sparsify interfaces at all levels  $l' > l$   
 4. Ignore  $\mathcal{O}(\epsilon)$  blocks  
**end for**



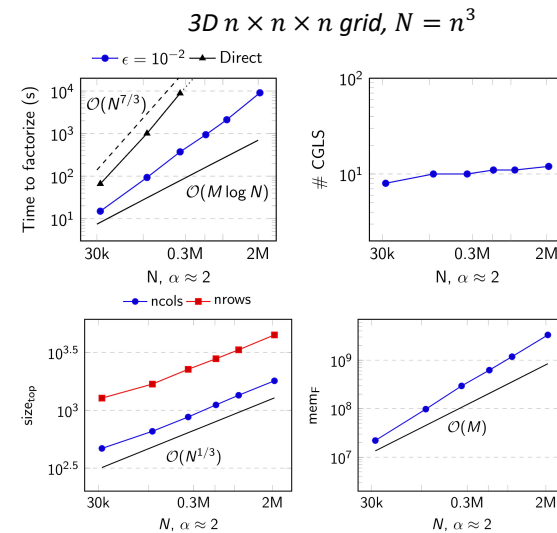
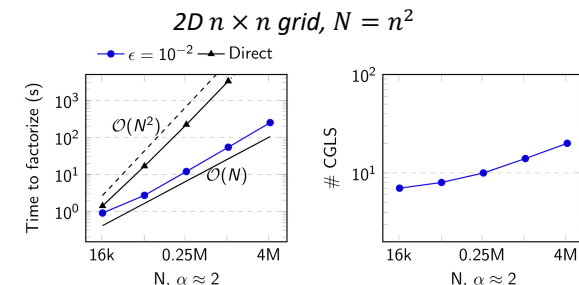
## Results

PDE constrained optimization: Solve  $J^T x = b$

$J$  is the Jacobian of size  $\alpha M \times N$

$$\min_{u,z} \frac{1}{2} \int_{\Omega} (u - u_d)^2 dx + \frac{1}{2} \lambda \int_{\Omega} z^2 dx$$

subject to  $-\nabla \cdot (z \nabla u) = h$  in  $\Omega$   
 $u = 0$  in  $d\Omega$



## Conclusions

- Produces an approx. factorization with tunable cost and accuracy
- Complexity:  $\mathcal{O}(M \log N)$  in time and  $\mathcal{O}(M)$  in space
- Exhibits more parallelism than direct methods
- Applications in CFD, PDE constrained optimization, computer graphics and vision.

References:  
 • A. Gnanasekaran and E. Darve, Hierarchical orthogonal factorization: Sparse Square Matrices, 2020  
 • A. Gnanasekaran and E. Darve, Hierarchical orthogonal factorization: Sparse Least Squares Problems, 202  
 • Code: [https://github.com/Abeynaya/spaQR\\_public](https://github.com/Abeynaya/spaQR_public)