

# Hierarchical matrix preconditioning for dominant advection problems

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# Plan

1. Introduction
2. Recap on Hierarchical matrices
  - Key ingredients
  - $\mathcal{H}$ -matrix: toy model
3.  $\mathcal{H}$ -matrix applied to dominant advection
  - Theoretical study
  - Numerical experiments
4. Conclusion

# Plan

## 1. Introduction

## 2. Recap on Hierarchical matrices

Key ingredients

$\mathcal{H}$ -matrix: toy model

## 3. $\mathcal{H}$ -matrix applied to dominant advection

Theoretical study

Numerical experiments

## 4. Conclusion

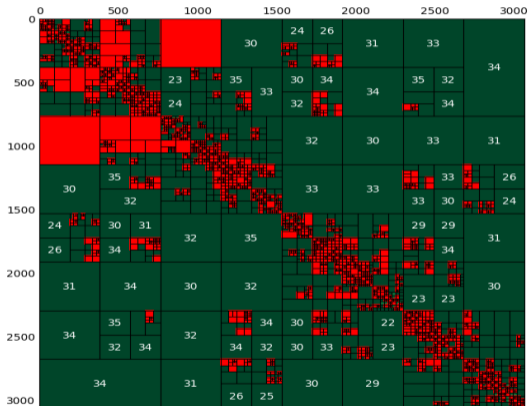
# Introduction

- Advection diffusion:  $-\varepsilon \Delta u + b \cdot \nabla u + cu = f$  on  $\Omega$ ,  $b \in \mathbb{R}^3$ ,  $c, f \in L^2(\Omega)$ ,  $u|_{\partial\Omega} = 0$ .  
**Dominant advection** :  $\varepsilon \rightarrow 0$ .
  - ◆ Numerical resolution of PDEs by discretization methods (FEM:finite elements methods, ...).
  - ◆ Sparse linear system  $Ax = y$ .
  - ◆ Numerical simulations  $\rightarrow$  inversion of large sparse matrices.
- Preconditioned iterative methods:  $M^{-1}Ax = M^{-1}y$  .
- Choice of the preconditioner: *ILU* with  $k$  level of fill-in.
  - ◆ *ILU(0)* same sparsity, convergence issue.
  - ◆ *ILU(k)* lost of the sparsity, convergence improved with  $k$ .
- $\mathcal{H}$ -matrices: approximate the inverse in quasi-linear complexity.
  - ◆ Storage of  $A_{\mathcal{H}}^{-1}$ :  $\mathcal{O}(n \log(n))$  against  $\mathcal{O}(n^2)$ .
  - ◆ Factorisation  $\mathcal{H}$ -LU :  $\mathcal{O}(n \log(n)^3)$  against  $\mathcal{O}(n^3)$ .

# General idea : $\mathcal{H}$ -matrices

- Block representation to approximate/compress a matrix.

- ◆ Compressed blocks  
(approximated blocks).
- ◆ Dense blocks  
(blocks left identical).



Hierarchical representation:  
invert of a FEM matrix  
compressed blocks, dense blocks

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## Two key ingredients

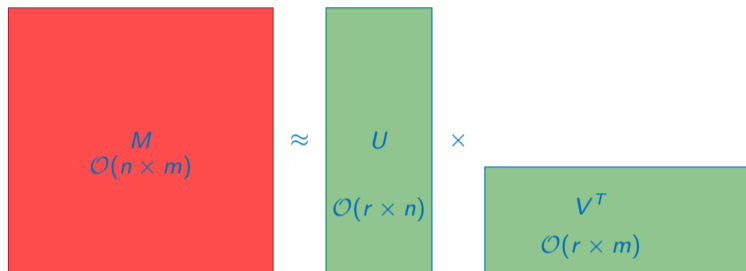
- Compression

- ◆ Compression method → Low rank approximation.

- Partitioning

- ◆ Correspondence DOF  $\longleftrightarrow$  geometrical points.
- ◆ Block representation: Splitting strategy → Cluster tree.
- ◆ Partitioning criterion: Admissibility condition → Block cluster tree.

## First key ingredient: Low rank approximation

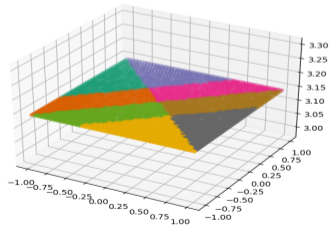
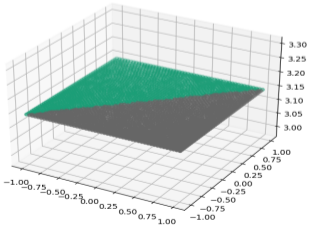
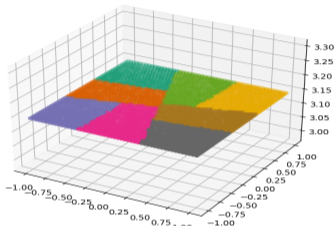
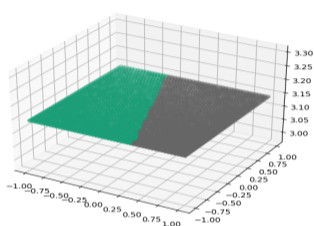


- Tolerance  $tol > 0$  ,  $\|M - UV^T\|_F \leq tol\|M\|_F$  with  $r \ll n$ .
- Low-rank format: Arithmetic in linear complexity.
- Best low rank approximation: truncated SVD in  $\mathcal{O}(n^3)$ .
- Factorisation up to rank  $r_0$ /precision  $tol$  in  $\mathcal{O}(r_0 n)$  (Adaptive Cross Approximation).



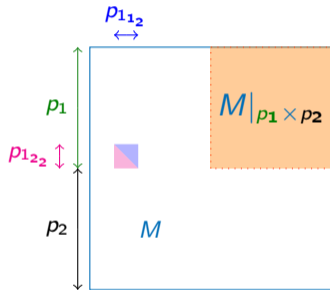
# Block representation and Cluster tree

Cluster tree = Recursive partitioning of the cluster of points given a splitting strategy.



depth = 1

depth = 3



Interaction:

$p_2 \times p_1$  (depth = 1),

$p_{1_2_2} \times p_{1_2}$  (depth = 3).

## Second key ingredient: Block Cluster Tree

- Cluster of points  $\mathcal{P}$  + Splitting strategy  $\Rightarrow$  Cluster tree.

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## Second key ingredient: Block Cluster Tree

- Cluster of points  $\mathcal{P}$  + Splitting strategy  $\Rightarrow$  Cluster tree.
- Recursively build a partition of  $\mathcal{P} \times \mathcal{P}$  with the cluster tree: **Block cluster tree**.
- Node  $M|_{\tau \times \sigma} \leftrightarrow$  interaction between two nodes  $\tau$  and  $\sigma$  of the cluster tree.

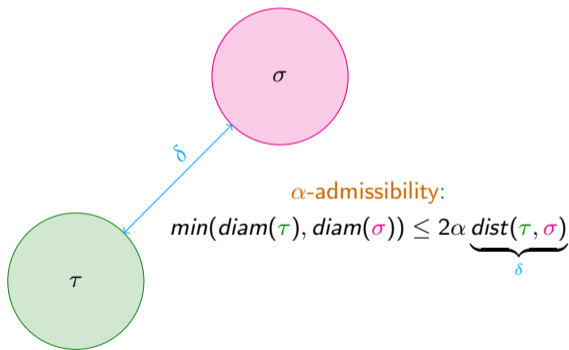
◆  $(\tau, \sigma)$  **admissible**

$\Rightarrow M|_{\tau \times \sigma}$  is a leaf.

◆ Else

$$M|_{\tau \times \sigma} = \begin{pmatrix} M|_{\tau_1 \times \sigma_1} & M|_{\tau_1 \times \sigma_2} \\ M|_{\tau_2 \times \sigma_1} & M|_{\tau_2 \times \sigma_2} \end{pmatrix}$$

- $\tau \times \tau$  cannot be **admissible**.



# $\mathcal{H}$ -matrix

Recursively build the  $\mathcal{H}$ -matrix with the block cluster tree by approximating its leaves.

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## Algorithm 0: Hierarchical formatting

---

build-Hmatrix ( $\tau \times \sigma$ ):

if  $\tau \times \sigma$  is *admissible* then

|  $M_{\mathcal{H}}|_{\tau \times \sigma} = \text{LowRank}(M|_{\tau \times \sigma})$

else

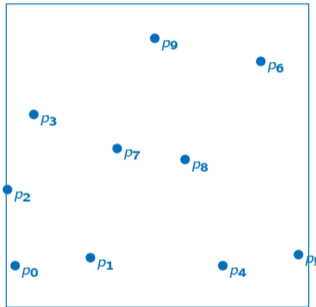
| for ( $\tau' \in \text{sons}(\tau), \sigma' \in \text{sons}(\sigma)$ ) do

| | build-Hmatrix ( $\tau' \times \sigma'$ )

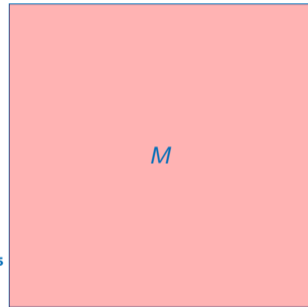
| end

end

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Cluster tree  
depth = 0



$\mathcal{H}$ -matrix  
depth = 0

# $\mathcal{H}$ -matrix

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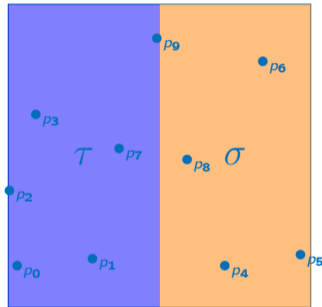
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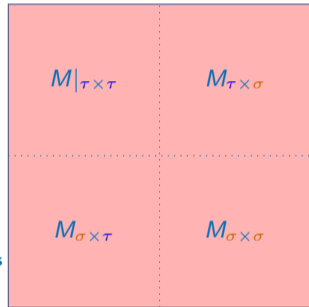
end

end

---



Cluster tree  
depth = 1



Block matrix  
depth = 1

# $\mathcal{H}$ -matrix

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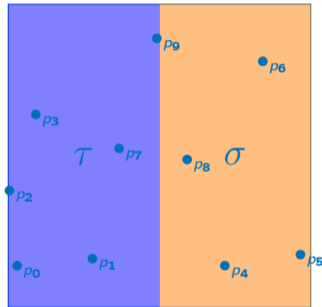
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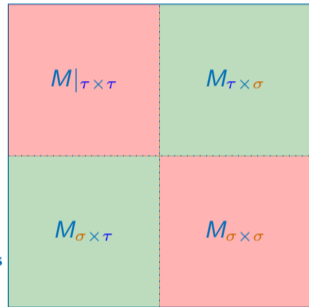
  end

end

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Cluster tree  
depth = 1



$\mathcal{H}$ -matrix  
depth = 1

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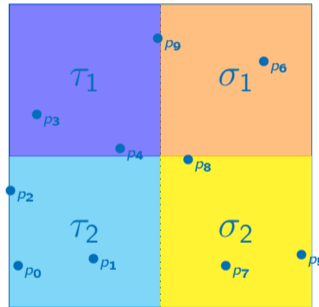
    for  $(\tau' \in \text{sons}(\tau), \sigma' \in \text{sons}(\sigma))$  do

        | build-Hmatrix ( $\tau' \times \sigma'$ )

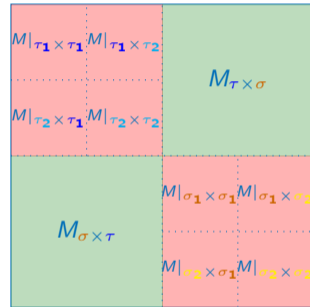
    end

end

---



Cluster tree  
depth = 2



Block matrix  
depth = 2



# $\mathcal{H}$ -matrix

Recursively build the  $\mathcal{H}$ -matrix with the block cluster tree by approximating its leaves.

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## Algorithm 0: Hierarchical formatting

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else

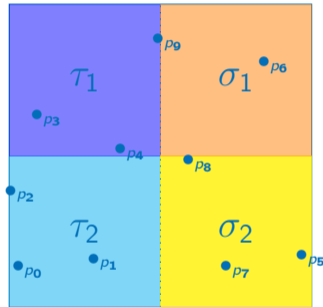
    for  $(\tau' \in \text{sons}(\tau), \sigma' \in \text{sons}(\sigma))$  do

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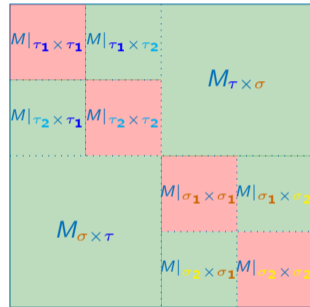
    end

end

---



Cluster tree  
depth = 2



$\mathcal{H}$ -matrix  
depth = 2

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Theoretical study

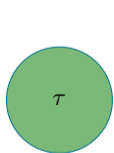
Numerical experiments

## 4. Conclusion

## Theoretical study: $-\varepsilon \Delta u + b \cdot \nabla u + cu = f$

- Solution  $u$  with  $u = 0$  on  $\partial\Omega$ ,  $b \in \mathbb{R}^3$ ,  $c \in L^\infty(\Omega)$ ,  $f \in L^2(\Omega)$  and  $\varepsilon > 0$ ,  
**dominant advection:**  $\varepsilon \rightarrow 0$ .
- Variational formulation:  $a(u, v) = \int_\Omega \varepsilon \nabla u \cdot \nabla v + b \cdot \nabla u v + cu \, dx = \int_\Omega f v \, dx \quad \forall v \in H_0^1(\Omega)$
- Discrete problem  $AU = F$  (example: finite elements methods).
- Approximate  $A^{-1}$  with  $\mathcal{H}$ -matrix?
  - ◆ Elliptic problems: [W.Hackbush], [M.Bebendorf].
- Compression really low with the  $\mathcal{H}$ -matrix format for  $\varepsilon \rightarrow 0$ .
  - ◆ Dominant advection: [S.Le Borne] heuristic reasoning on structured meshes.
  - ◆ Extend the proofs of [M.Faustmann, J.Melenk, D.Praetorius] and [S.Börm] to the case  $\varepsilon \rightarrow 0$ .

Idea: Use the "orthogonality" of  $a$ :



$$a(u|_\tau, v) = \int_\Omega f v = 0 \\ \text{supp}(v) \subset \tau, \text{supp}(f) \subset \sigma$$

Separation of the support  
 $\Rightarrow$  "orthogonality"

## How it works: **admissibility condition** $\Rightarrow$ $\mathcal{H}$ -matrix approximability

- What **admissibility condition** on  $(\tau, \sigma)$  would ensure a good hierarchical approximation ?
- **Corollary:** Assume  $(\tau, \sigma)$  **admissible**, then  $\exists q \in (0, 1)$ ,  $C, C_{dim} > 0$  such that  $\forall p \in \mathbb{N}_{\geq 2}$  we can find  $U \in \mathbb{R}^{|\tau| \times k}$ ,  $V \in \mathbb{R}^{|\sigma| \times k}$  (with  $k \leq C_{dim} p^{dim(\Omega)+1}$ ) satisfying  $\forall x \in \mathbb{R}^{|\sigma|}$

$$\|(A^{-1}|_{\tau \times \sigma} - UV^T)x\|_{Frobenius} \leq Cq^p \|x\|$$

- **Theorem:** Assume  $(\tau, \sigma)$  **admissible**, then  $\exists q \in (0, 1)$ ,  $p \in \mathbb{N}^2$ ,  $C, C_1, C_2 > 0$  and  $v \in V$ , where  $dim(V) < Cp^{dim(\Omega)+1}$ , locally approximating the solution on  $\tau$  with the estimates

$$\|\nabla(u|_{\tau} - v)\|_{L^2(\tau)} \leq C_1 q^p \|f\|_{L^2(\Omega)}$$

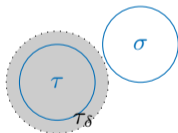
$$\|u|_{\tau} - v\|_{L^2(\tau)} \leq C_2 q^{p+1} \|f\|_{L^2(\Omega)}$$

- Idea of the proof: Successive approximations on narrower sets.

## Key ingredient of the proof: Caccioppoli inequality

- $\|\nabla u\|_{L^2(\tau)} \leq \frac{C}{\text{dist}(\tau, \partial\tau^+)} \|u\|_{L^2(\tau^+)}$  where  $\tau \subsetneq \tau^+$ .
- "orthogonality" of  $a$  for two clusters  $\tau$  and  $\sigma$  with  $\text{dist}(\tau, \sigma) = \delta > 0$  and  $\text{supp}(f) \subset \sigma$ .  
Separation  $\Rightarrow \tau_\delta = \{x \in \Omega \mid \text{dist}(x, \tau) < \delta\}$ .

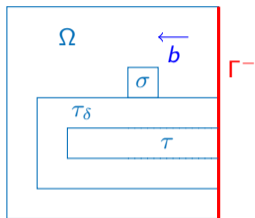
$\forall \lambda > 1, \exists \eta$  with  $\text{supp}(\eta) \subset \tau_\delta, \eta|_\tau = 1$  and  
 $\|\nabla \eta\|_{L^\infty(\Omega)} \leq \frac{\lambda}{\delta}$



Support of the cut-off  $\eta$

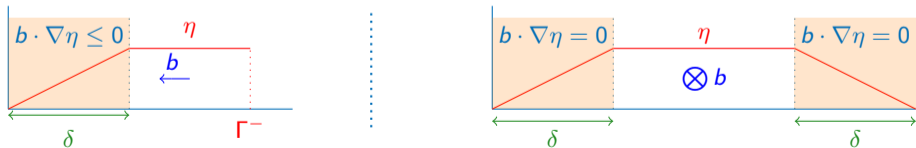
$$\begin{aligned}
 \|\nabla u\|_{L^2(\tau)}^2 &\leq \|\nabla(\eta u)\|_{L^2(\tau_\delta)}^2 \\
 &= \langle \nabla u, \nabla(\eta^2 u) \rangle_{L^2(\Omega)} + \langle u \nabla \eta, u \nabla \eta \rangle_{L^2(\Omega)} \\
 &= \varepsilon^{-1} \underbrace{a(u, \eta^2 u)}_{=0} - \varepsilon^{-1} \int_{\Omega} b \cdot \nabla(u) \eta^2 u dx \\
 &\quad - \varepsilon^{-1} \int_{\Omega} c \eta^2 u^2 dx + \langle u \nabla \eta, u \nabla \eta \rangle_{L^2(\Omega)} \\
 &\leq \frac{\lambda^2}{\delta^2} \|u\|_{L^2(\tau_\delta)}^2 + \int_{\Omega} \eta u^2 \frac{b}{\varepsilon} \cdot \nabla \eta dx.
 \end{aligned}$$

# Choice of the geometry: vanishing of the term $b \cdot \nabla \eta$



Cluster  $\tau$  aligned on advection streams and which reach the incident border  $\Gamma^- = \{\mathbf{n}_{\partial\Omega} \cdot \mathbf{b} < 0\}$ .

Support of the cut-off in  $\tau_\delta$ .



Variation of  $\eta$  in the directions  $b$  and  $b^\perp$ .

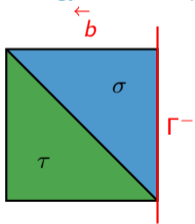
# Admissible partitionning

- Caccioppoli  $\Rightarrow$   $\alpha$ -admissibility, explicit admissibility condition:

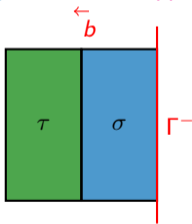
$(\tau, \sigma)$  is admissible if  $\tau$  is aligned with  $b$  which reaches  $\Gamma^-$  and

$$\text{dist}(\tau, \sigma) > 2\alpha \text{diam}_\infty(\tau)$$

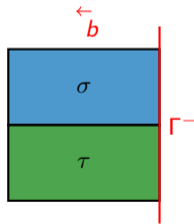
- Clustering strategy  $\rightarrow$  suited partitioning for the Caccioppoli frame.



Not suited partitionning  
 $\tau$  and  $\sigma$  not in the frame



ill suited partitionning  
 $\tau$  not in the frame

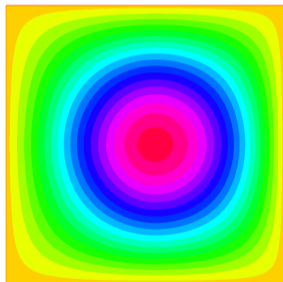


Well suited partitionning  
 $\tau$  and  $\sigma$  in the frame

## Test case: matrices obtained with FreeFem

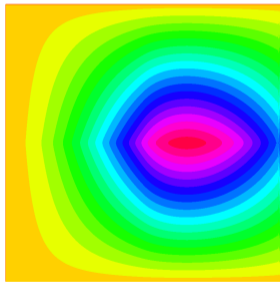
On  $[0, 1] \times [0, 1]$ , with an **unstructured triangular mesh** of 3057 nodes:

$$-\varepsilon \Delta u + \begin{pmatrix} 1 \\ 0 \end{pmatrix} \cdot \nabla u + 2u = 10(1-x)(1-y)$$



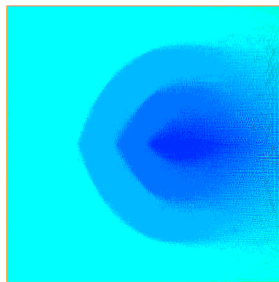
$\varepsilon = 1$

Advection diffusion



$\varepsilon = 10^{-3}$

Transitory case



$\varepsilon = 10^{-6}$

Dominant advection



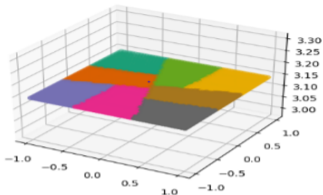
## Approximation of $A^{-1}$

- C++ library HTOOL for  $\mathcal{H}$ -matrices (Pierre Marchand INRIA).
- Standard approach: Principal Component Analysis +  $\alpha$ -admissibility.
- New approach: Partitioning into tubes aligned with the convection + new condition.
- $M_{\mathcal{H}}$  hierarchical approximation of  $M$ ,  $\mathcal{F} = \{\text{leaves of } M_{\mathcal{H}}\}$ ,  $\mathcal{R} = \{\text{admissible leaves of } M_{\mathcal{H}}\}$ ,

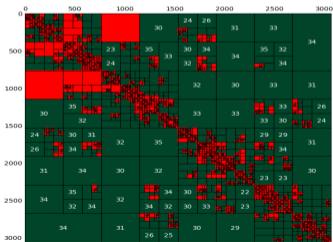
$$\text{Compression}(M_{\mathcal{H}}) = \frac{\sum_{R \in \mathcal{R}} \text{rank}(R)(\text{line}(R) + \text{col}(R)) + \sum_{F \in \mathcal{F} \setminus \mathcal{R}} \text{line}(F) \times \text{col}(F)}{\text{line}(M) \times \text{col}(M)}$$

$$\text{error}(M_{\mathcal{H}}) = \frac{\|M - M_{\mathcal{H}}\|_F}{\|M\|_F}$$

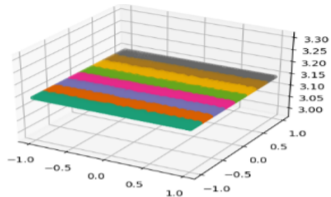
# Approximation of $A^{-1}$ : C++ library HTOOL



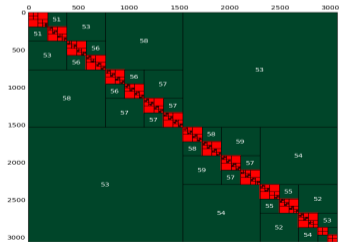
Splitting PCA



Usual condition + PCA  
(State of the art)

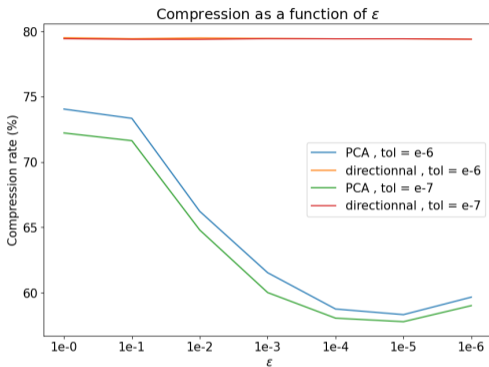


New splitting



New condition + new splitting  
(Our method)

# Quality of approximations: matrices $3057 \times 3057$



Compression VS  $\epsilon$

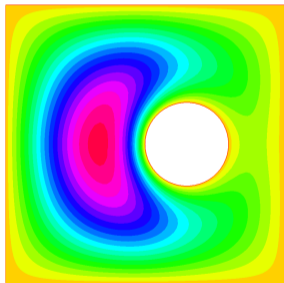


Relative error VS of  $\epsilon$

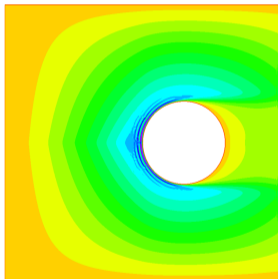
Compression+20%, error  $10^9$  times lower!

## Test case: domain with a hole

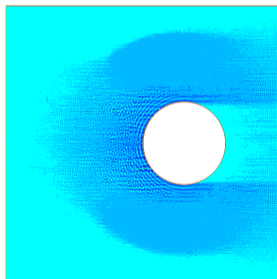
The same equation on a domain with a hole, mesh with 3227 nodes.



$\varepsilon = 1$   
Advection diffusion

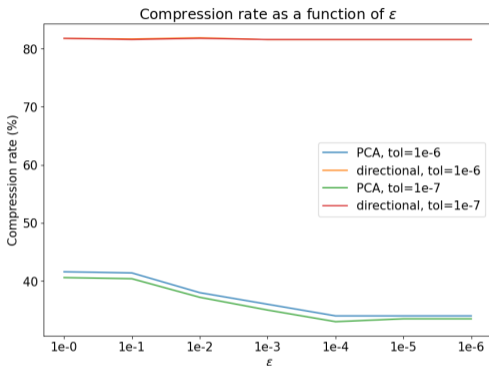


$\varepsilon = 10^{-3}$   
Transitory case

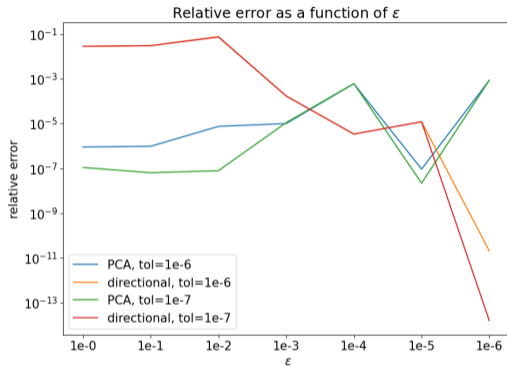


$\varepsilon = 10^{-6}$   
Dominant advection

# Quality of approximations: matrices $3227 \times 3227$



Compression VS  $\epsilon$



Relative error VS  $\epsilon$

Compression +40% and error less than  $10^{-5}$  for  $\epsilon \leq 10^{-4}$ .

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## Conclusion

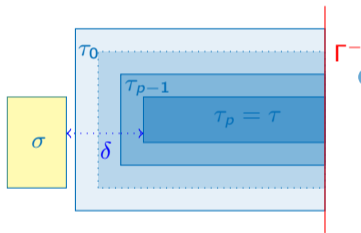
- Proposal for a new admissibility condition and a new partitioning suited for dominant advection.
- The compression and error results obtained in our tests significantly surpass those of the current state of the art.

# THANK YOU!

- [1] M.Bebendorf. *Hierarchical matrices : a means to efficiently solve elliptic boundary value problems*, volume 63 of *Lecture notes in computational science and engineering*. 2008.
- [2] L.Grasedyck W.Hackbusch and S.Börm. *An introduction to hierarchical matrices*, 2001.
- [3] S.Börm. *Efficient numerical methods for non-local operators: H2-matrix compression, algorithms and analysis*, volume 14. European Mathematical Society, 2010.
- [4] J.Melenk M.Faustmann and D.Praetorius.  $\mathcal{H}$ -matrix approximability of the inverses of FEM matrices. *Numerische Mathematik*, 131(4), 2015.
- [5] S.Le Borne and L.Grasedyck. H-matrix preconditioners in convection-dominated problems. *SIAM J. Matrix Anal. Appl.*, 27(4), 2006.



## Local approximation in low dimension



Concentric bounding boxes

- We impose  $\text{diam}_\infty(\tau) \leq 2\alpha\delta$ ,  $\delta > \text{dist}(\tau_0, \sigma) = \gamma\delta > 0$
- Poincaré Wirtinger:  $v_1 \in V_1$ ,  $\dim(V_1) = l^{\dim(\Omega)}$ ,  $l \in \mathbb{N}$  with  $\|u - v_1\|_{L^2(\tau_0)} \leq \frac{\text{diam}_\infty(\tau_0)}{l} \|\nabla u\|_{L^2(\tau_0)}$ .
- Cacciopoli :  $\tau_1 \rightarrow \tau_0$

$$\begin{aligned} \|\nabla(u - v_1)\|_{L^2(\tau_1)} &\leq \frac{C}{\text{dist}(\tau_1, \partial\tau_0)} \|u - v_1\|_{L^2(\tau_0)} \\ &\leq \underbrace{C' \frac{p}{l(1-\gamma)}}_{q < 1} \|\nabla u\|_{L^2(\tau_0)} \end{aligned} \quad \begin{array}{l} \text{PW} \\ + \text{condition} \end{array}$$

- $\forall k \leq p$  we find  $v_k \in V_k$  approximation of  $u - \sum_{i \leq k-1} v_i$  in  $L^2(\tau_{k-1})$  (PW) then in  $H^1(\tau_k)$  (Cacciopoli).

## Local approximation in low dimension

- $\tau_p = \tau$ ,  $v = \sum_{i \leq p} v_i|_{\tau} \in V = \sum_{i \leq p} V_i$  where  $\dim(V) \leq pl^d$  and we have the estimates:

$$\|\nabla(u - v)\|_{L^2(\tau)} \leq Cq^p \|\nabla u\|_{L^2(\tau_0)} \lesssim? Cq^p \|f\|_{L^2(\Omega)}$$

$$\|u - v\|_{L^2(\tau)} \leq C' q^{p+1} \|\nabla u\|_{L^2(\tau_0)} \lesssim? C' q^{p+1} \|f\|_{L^2(\Omega)}$$

- $\text{dist}(\tau_0, \partial\tau_\delta) = \gamma\delta > 0 \Rightarrow$  **Cacciopoli**:  $\|\nabla u\|_{L^2(\tau_0)} \leq \frac{C}{\gamma\delta} \|u\|_{L^2(\tau_\delta)}$ .

- $c_0 = \inf(c - \text{div}(\frac{b}{2})) > 0 \Rightarrow \|u\|_{L^2(\Omega)} \|f\|_{L^2(\Omega)} \geq a(u, u) \geq c_0 \|u\|_{L^2(\Omega)}^2$ .

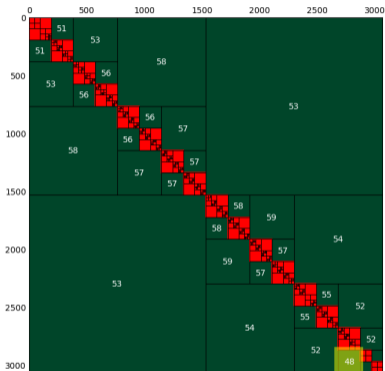
- QED and explicit admissibility condition  $(\tau, \sigma)$ :

Bound independent of  $u$

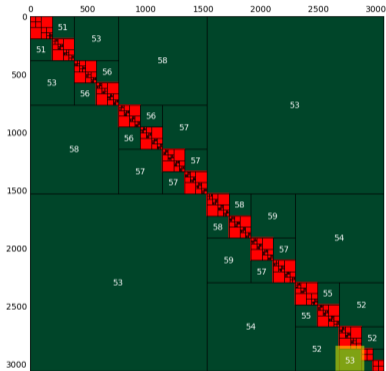
The couple  $(\tau, \sigma)$  is said to be admissible if  $\tau$  is an aligned cluster on  $b$  reaching  $\Gamma^-$

$$\text{and } \text{dist}(\tau, \sigma) > 2\alpha \text{diam}_\infty(\tau)$$

# Importance of the tolerance of ACA

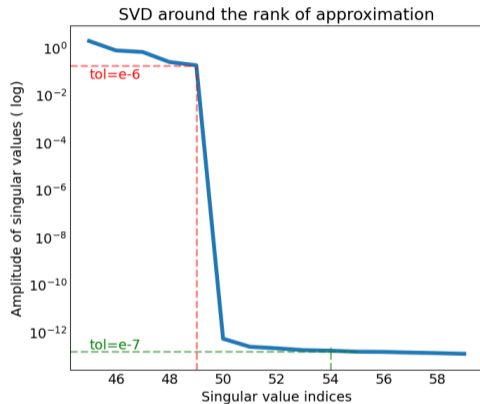
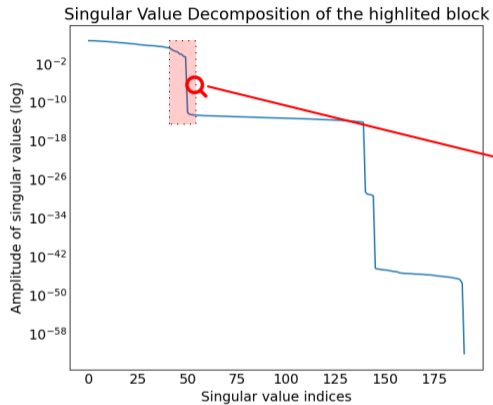


tolerance =  $10^{-6}$ ,  $\epsilon = 10^{-3}$



tolerance =  $10^{-7}$ ,  $\epsilon = 10^{-3}$

# Shape of the block's SVD



Error of a rank  $k$  approximation is proportional to the  $k + 1^{\text{th}}$  singular value.