

# High-order Adaptive Mesh Refinement Multigrid Poisson Solver in any Dimension

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# Overview

## 1 Introduction

## 2 AMR geometric multigrid Poisson solver

Discretization

Compact Finite Difference Poisson stencils

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Convergence and accuracy

Computational Cost

## 4 Conclusion

## Poisson solvers in Physics simulation

Poisson Equation in dimension  $d$ :

$$\Delta u = v, \quad \text{with} \quad \Delta = \sum_{i=1}^d \partial_i^2 \quad (1)$$

### Exemple: Reduced MHD

$$\frac{\partial F}{\partial t} + [\varphi, F] = \rho_s^2 [U, \psi] + \eta(\Delta\psi - \Delta\psi_{eq}), \quad (2)$$

$$\frac{\partial U}{\partial t} + [\varphi, U] = [\psi, \Delta\psi] + \nu(\Delta U - \Delta U_{eq}), \quad (3)$$

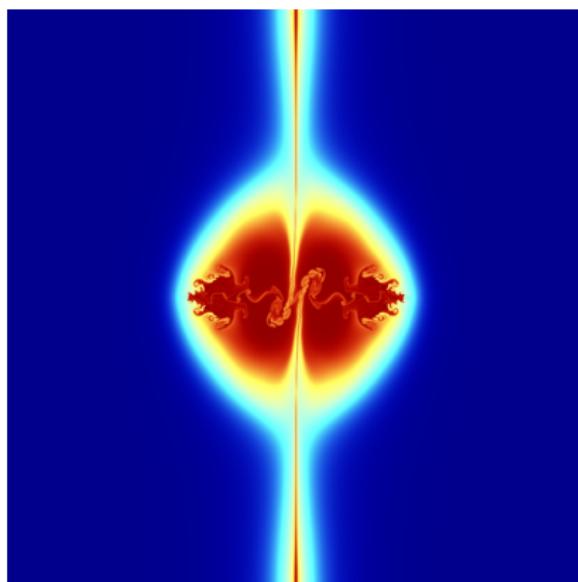
$$(Id - de^2 \Delta) \psi = F \quad \Delta\varphi = U. \quad (4)$$

## 2D Magnetic Reconnection Simulation

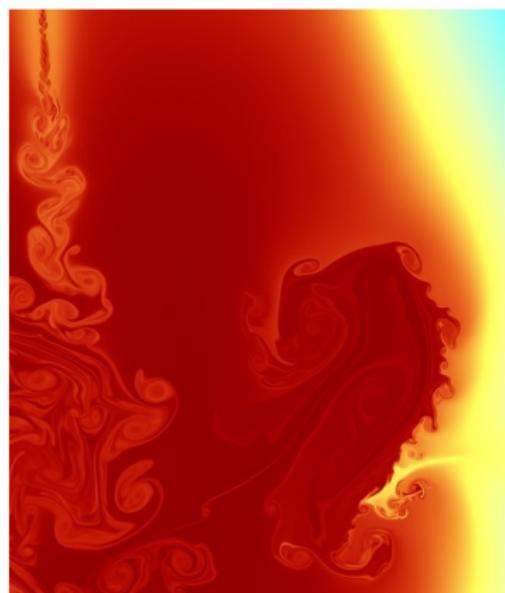
Magnetic stream  $F$

refinement level of the grid

## Turbulence

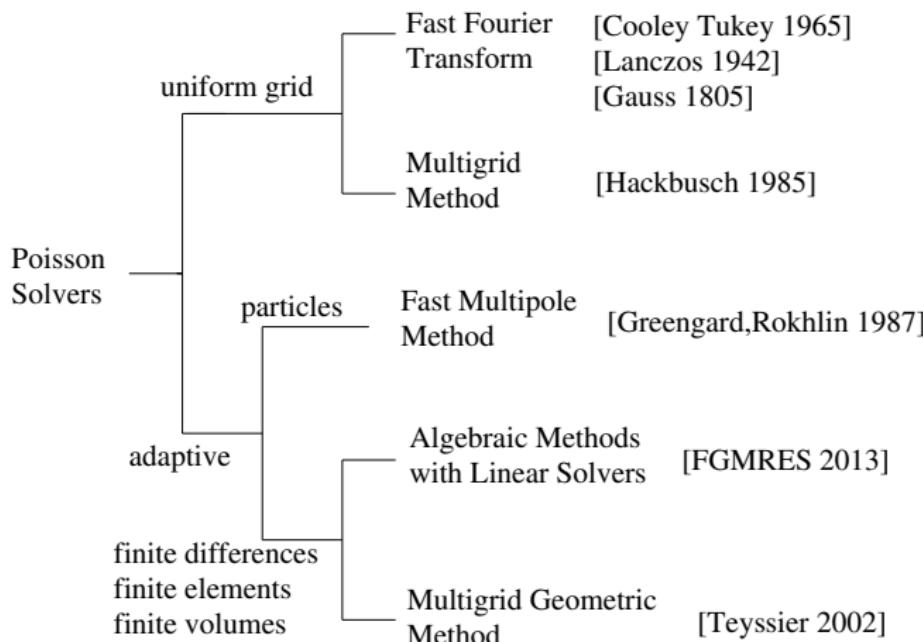


complex structured



multiscaled turbulence

## Poisson Solvers, State-Of-the-Art



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## Adaptive Mesh Refinement Tree structure

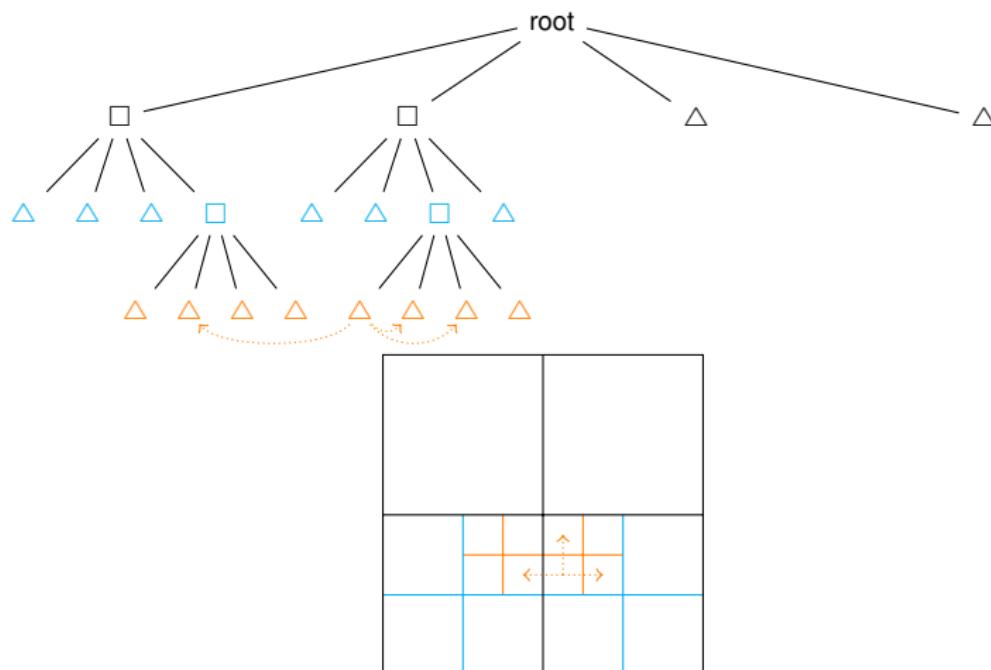
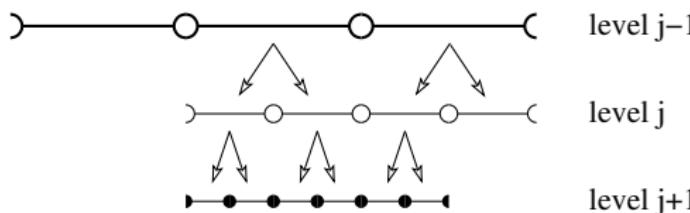


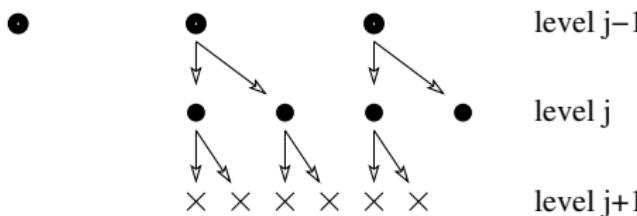
Figure: Fully-threaded tree (top) and its associated non-uniform grid (bottom).

## Cell-based/vertex-based

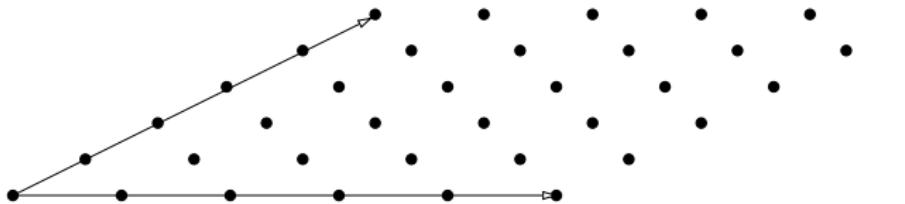
- refinement centered on the cell



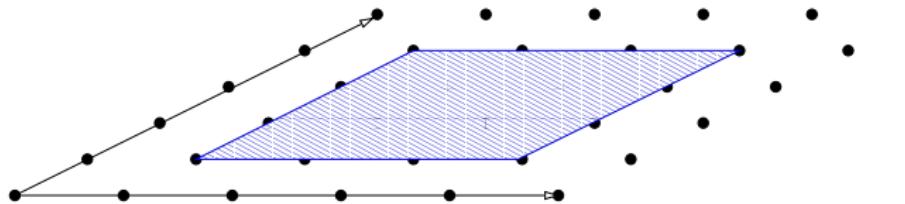
- refinement centered on the point



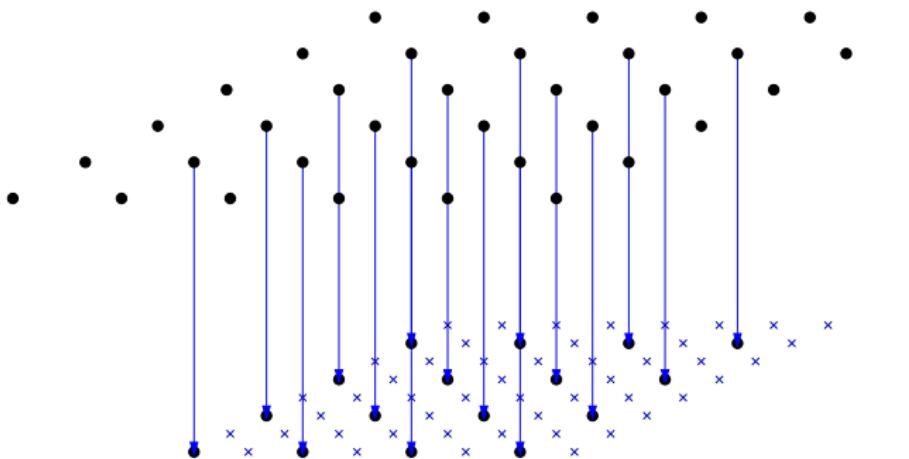
## Point activation



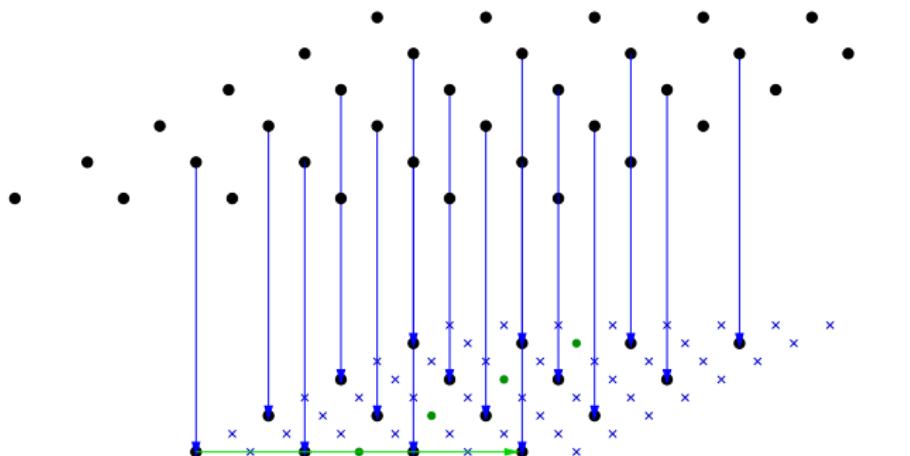
## Point activation



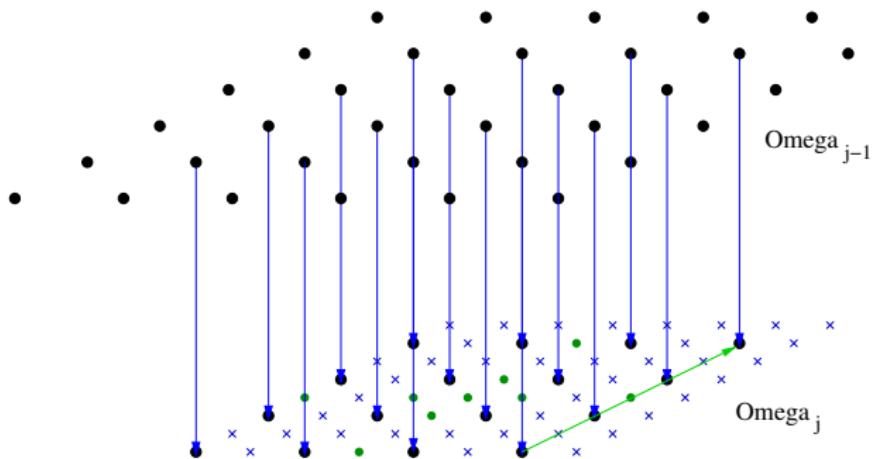
## Point activation



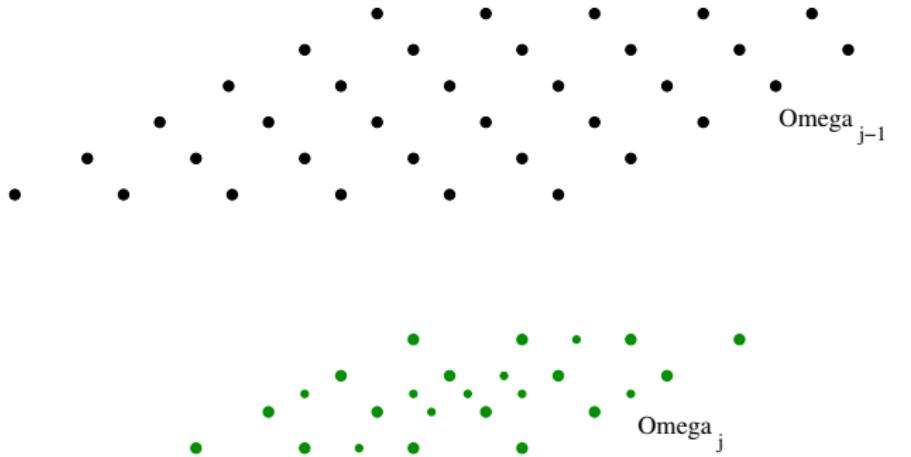
## Point activation



## Point activation



## Point activation



## Boundary of a refined subdomain

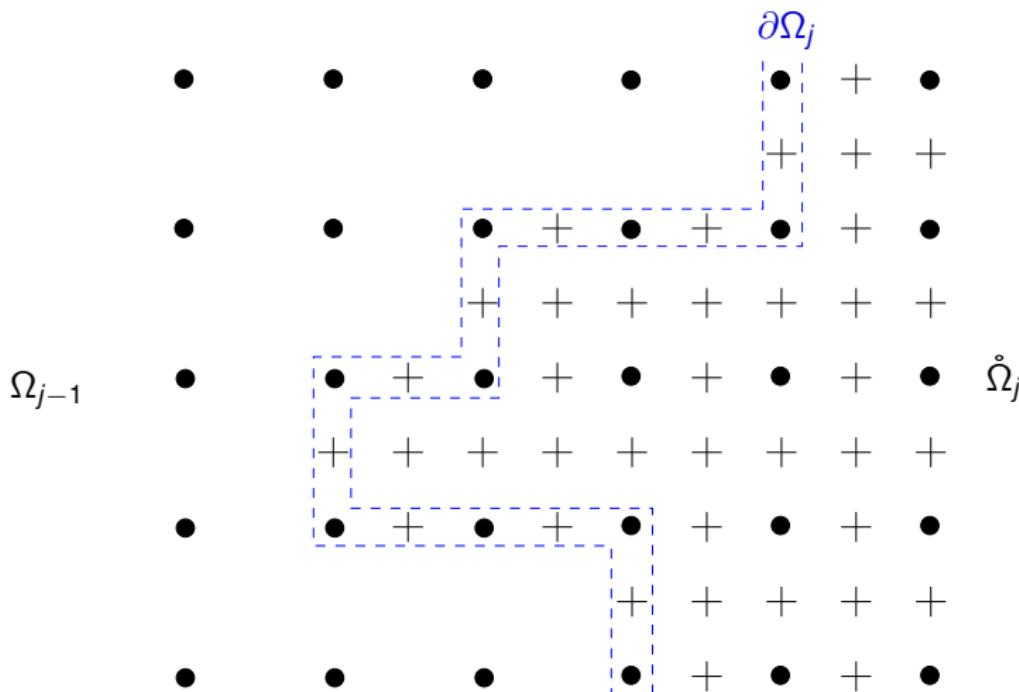


Figure: Example of domain  $\Omega_j$  divided between its boundary  $\partial\Omega_j$  and its interior  $\mathring{\Omega}_j$ .

## Boundaries

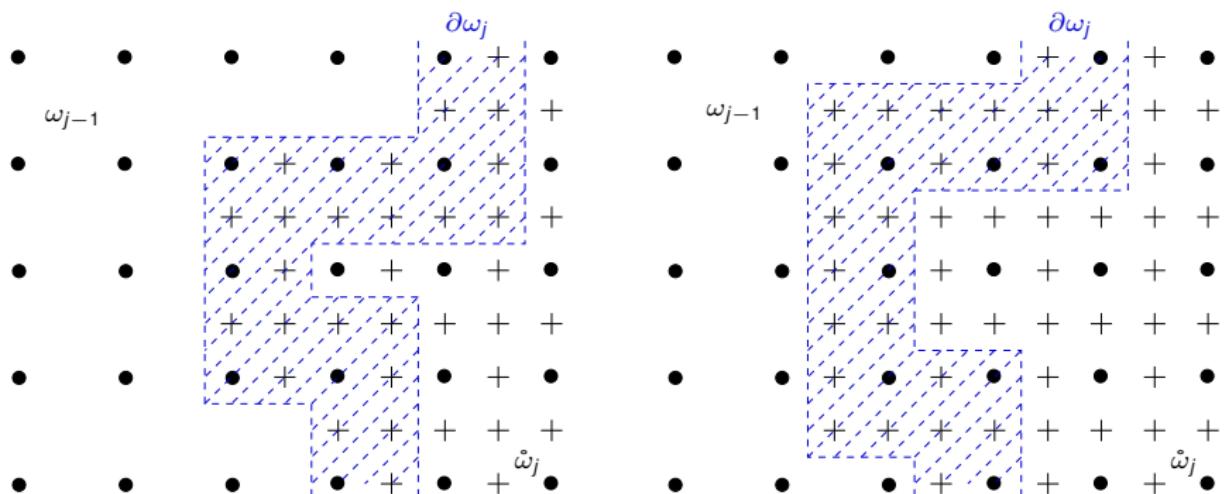


Figure: Two different ways of fixing the boundary.

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## 2D usual non-compact stencil

$$\frac{1}{12h^2} \begin{array}{ccccccc} & & (-1) & & 16 & & \\ & & \downarrow & & \downarrow & & \\ & (-1) & - & 16 & - & 60 & - & 16 & - & (-1) & u = v \\ & \downarrow & & \downarrow & & \downarrow & & \downarrow & & \downarrow \\ & & 16 & & & 16 & & & & (-1) \end{array} \quad (5)$$

## “Mehrstellenverfahren” scheme

$$\frac{1}{6h^2} \begin{array}{c} \text{---} \\ | \quad | \quad | \\ \text{---} \\ | \quad | \quad | \\ \text{---} \\ | \quad | \quad | \\ \text{---} \end{array} \begin{matrix} 1 & 4 & 1 \\ 4 & -20 & 4 \\ 1 & 4 & 1 \end{matrix} u = \frac{1}{12} \begin{array}{c} \text{---} \\ | \quad | \quad | \\ \text{---} \\ | \quad | \quad | \\ \text{---} \\ | \quad | \quad | \\ \text{---} \end{array} \begin{matrix} 1 \\ 1 & 8 & 1 \\ 1 \end{matrix} v \quad (6)$$

## 6th order any-dimensional case

$$A = \alpha [p \ 1 \ p]^d + \begin{array}{c} \gamma \\ | \\ \gamma - \beta - \gamma \\ | \\ \gamma \end{array} \quad (7)$$

namely

$$a_0 = \alpha + \beta, \quad a_{(1,0,\dots,0)} = \alpha p + \gamma, \quad \text{and} \quad a_{(\underbrace{1, \dots, 1}_{k \text{ times}}, 0, \dots, 0)} = \alpha p^k \quad \text{for } k \geq 2.$$

To reach the sixth-order accuracy, computations lead to:

$$p = \frac{1}{3}, \quad \alpha = \frac{3}{2} \left( \frac{3}{5} \right)^{d-2}, \quad \beta = -\frac{25+2d}{6} \quad \text{and} \quad \gamma = \frac{1}{6}.$$

## 6h order any-dimensional case

$$\begin{array}{c}
 b_{01} \\
 | \\
 0 \\
 | \\
 B = \omega[q \ 1 \ q]^d + b_{01} - 0 - \lambda - 0 - b_{01} \\
 | \\
 0 \\
 | \\
 b_{01}
 \end{array}$$

namely

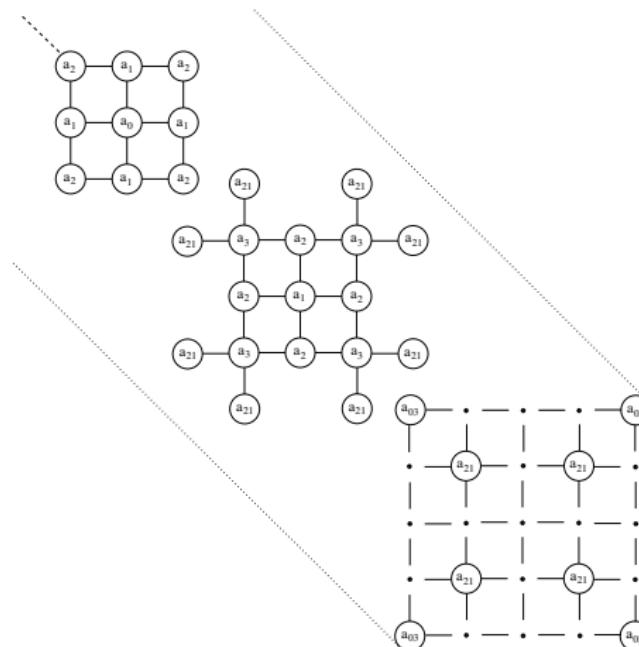
$$b_0 = \omega + \lambda, \quad b_{\underbrace{(1, \dots, 1, 0, \dots, 0)}_{k \text{ times}}} = \omega q^k \quad \text{for } k \geq 1, \quad \text{and} \quad b_{(2, 0, \dots, 0)} = b_{01},$$

with

$$q = \frac{1}{7}, \quad \omega = \frac{9}{10} \left(\frac{7}{9}\right)^d, \quad \lambda = \frac{1}{10} + \frac{d}{120} \quad \text{and} \quad b_{01} = -\frac{1}{240}.$$

## 10th order 3-dimensional case

Bottom half of the symmetric three-dimensional stencil:



There are 113 points for the  $B$  stencil:  $b_0, b_1, b_2, b_3, b_{01}, b_{11}, b_{02}, b_{03}, b_{001}, b_{101}, b_{0001}$ .

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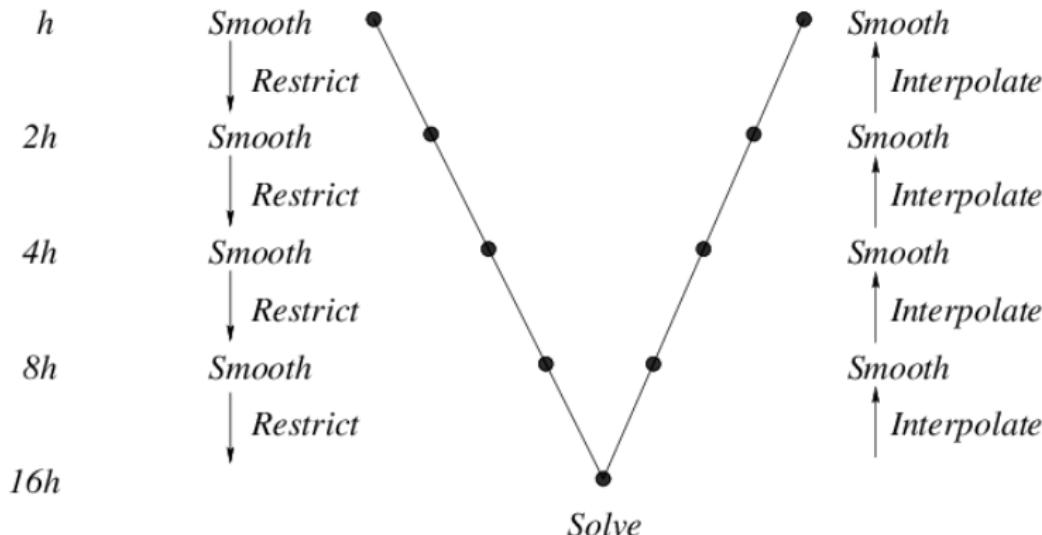
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## In the uniform case, the V–cycle

*Grid Spacing*



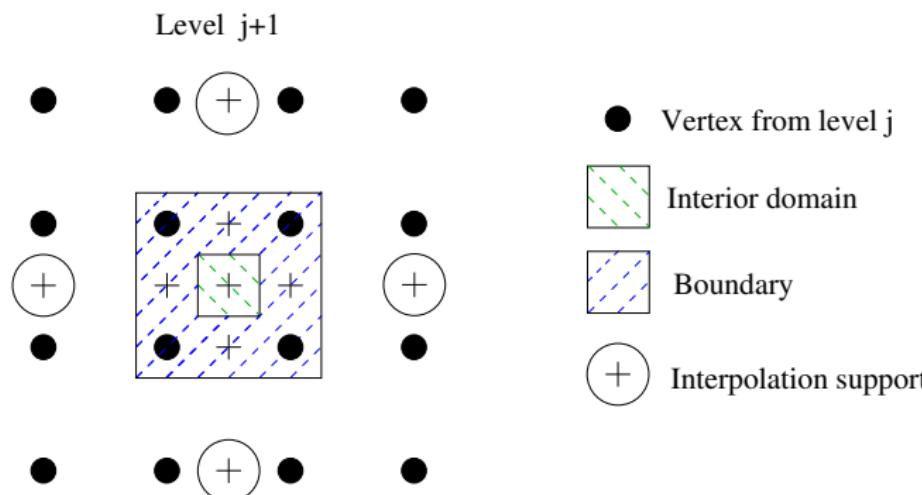
[Colin Fox]

## Gauss-Seidel iterations

*Gauss-Seidel iteration (Smoothening):*      for  $x_\lambda \in \mathring{\omega}_j$  do

$$\text{tmp}_{n+1}(x_\lambda) = \text{tmp}_n(x_\lambda) + \frac{h^2}{a_0} \left( \text{res}(x_\lambda) - \frac{1}{h^2} A \text{ tmp}_n(x_\lambda) \right)$$

## Interpolation of the boundary



**Figure:** Instance of fourth-order consistent distribution of vertices with their types. Only one among the horizontal couple of interpolatory points or the vertical one is necessary.

## Interpolets

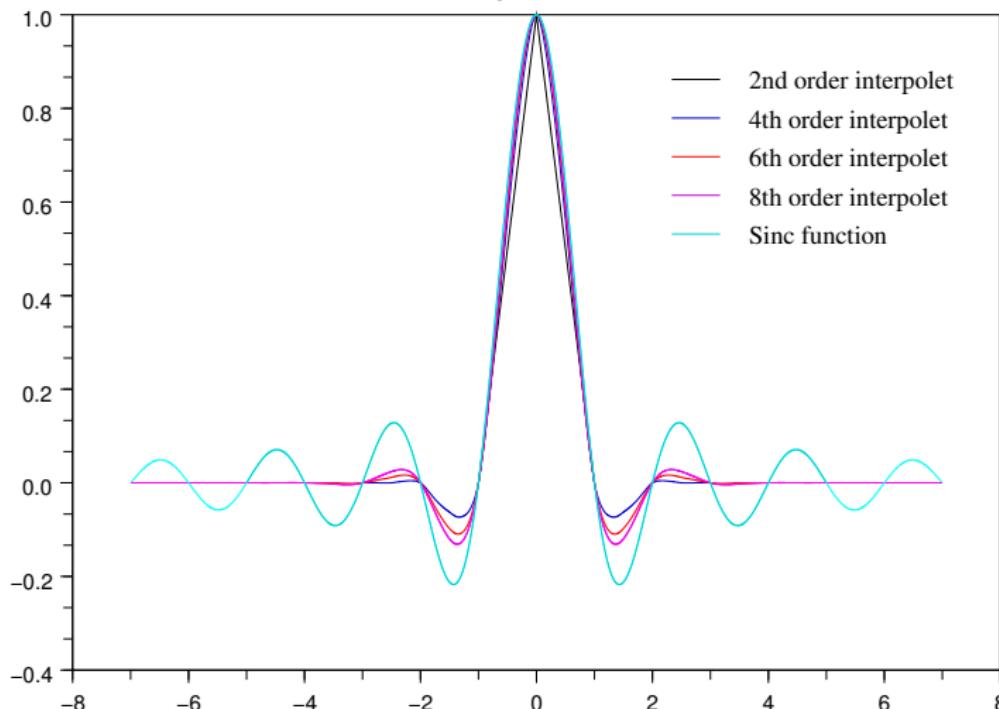


Figure: Interpolants of 2nd, 4th, 6th and 8th orders tending to the sinc function.

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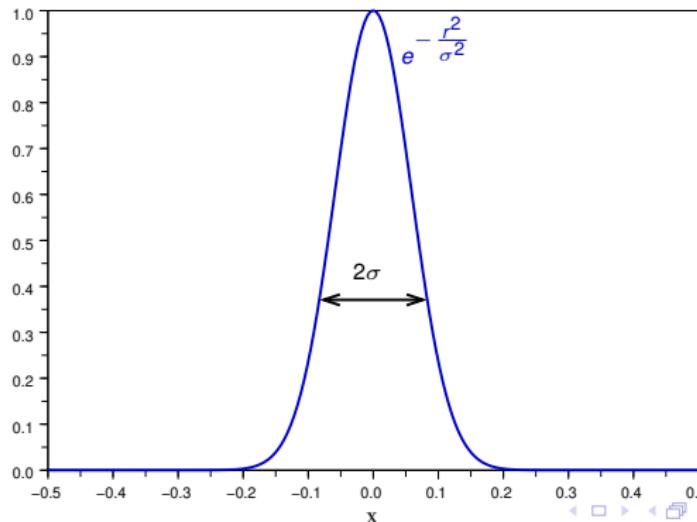
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## Test case: the Gaussian

$$u(\mathbf{x}) = \exp\left(-\frac{r^2}{\sigma^2}\right) \quad (8)$$

with  $r = |\mathbf{x}|$  and the parameter  $\sigma$  small enough.

$$v(\mathbf{x}) = \left(\frac{4|\mathbf{x}|^2}{\sigma^4} - \frac{2d}{\sigma^2}\right) \exp\left(-\frac{|\mathbf{x}|^2}{\sigma^2}\right)$$



## Convergence of the algorithm

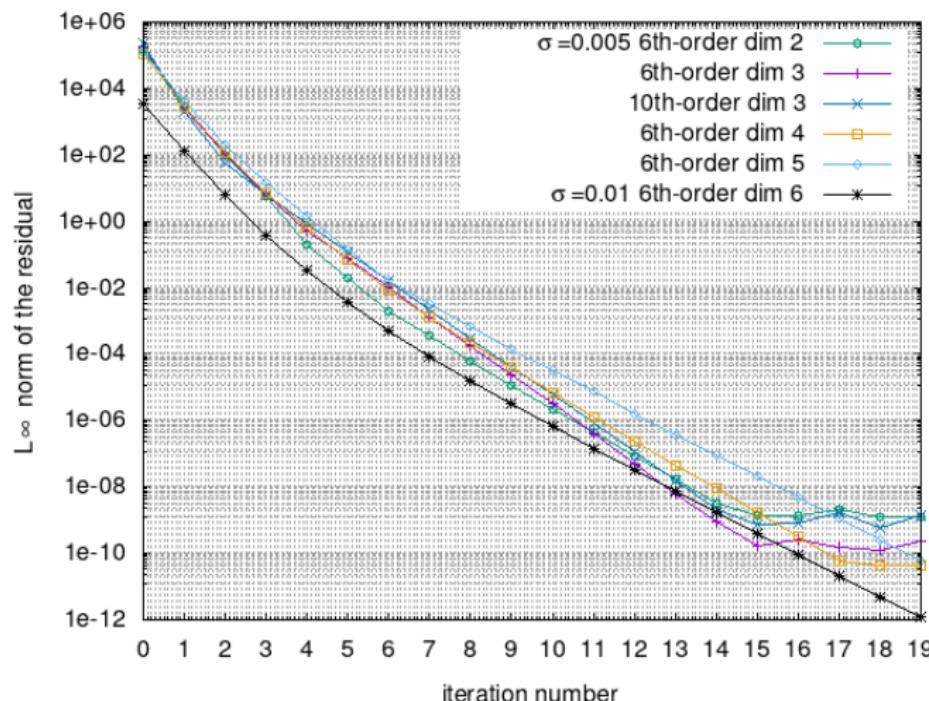


Figure: Convergence process of the multigrid algorithm. The first iterations achieve the largest reductions of the residual.

## Accuracy 6th order stencils

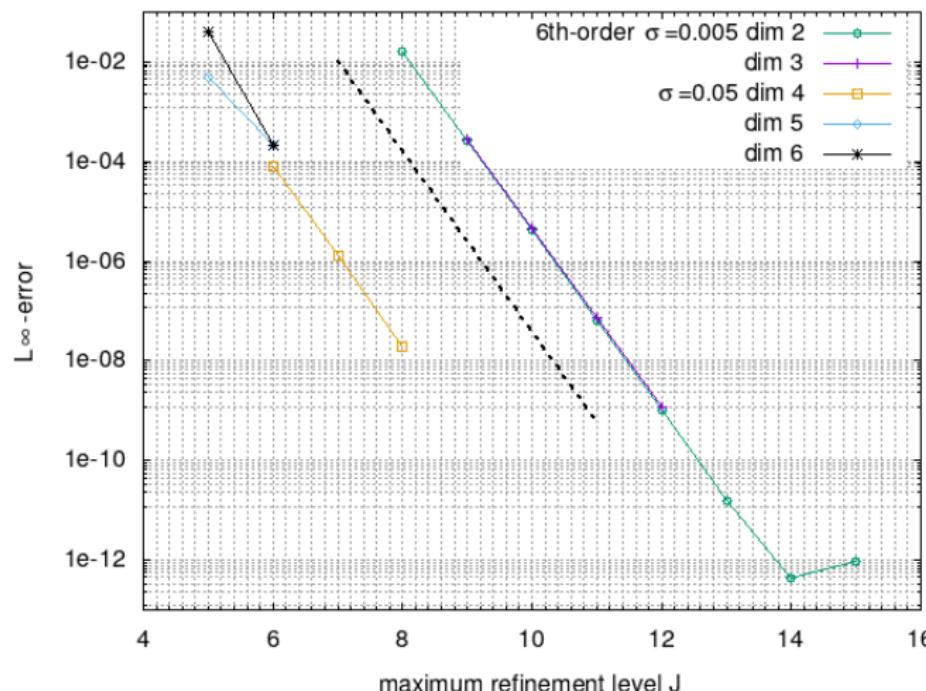


Figure: Order of convergence for the 6th-order scheme in various dimensions. The dashed line represents the 6th-order slope.

## Accuracy 10th order stencil

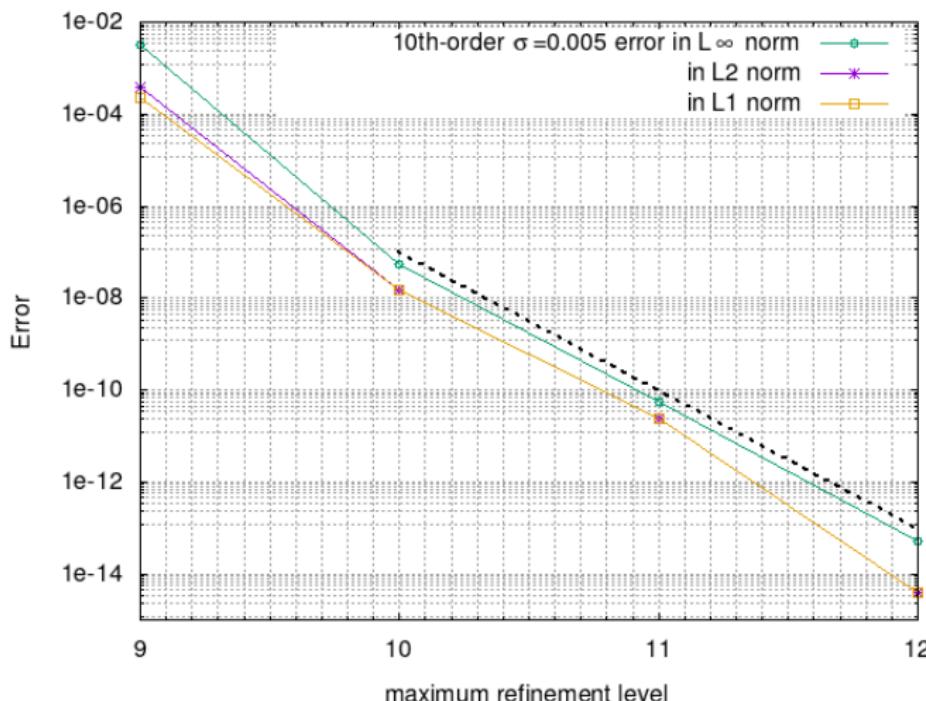
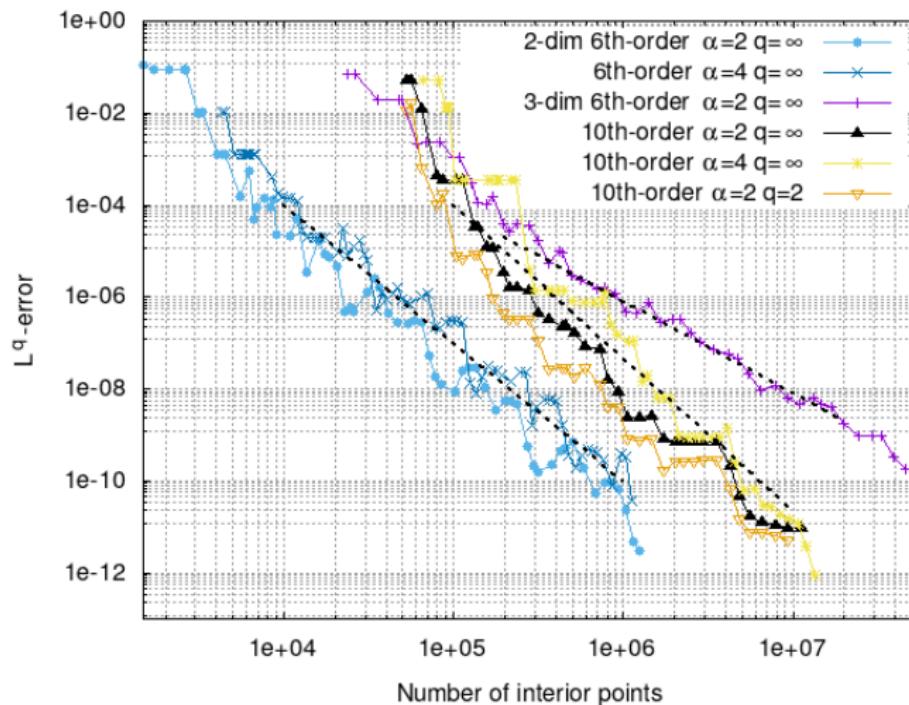


Figure: Convergence for the 3-dimensional 10th-order scheme in various norms. The dashed line represents the 10th-order slope.

## Convergence with a continuous refinement



**Figure:** Experimental  $L^q$ -error as a function of the number of points. The theoretical slopes  $\varepsilon_{L^q} = KN_{pt}^{-p/d}$  ( $\frac{p}{d} = \frac{6}{2}, \frac{10}{3}, \frac{6}{3}$  from left to right) are represented by dashed lines.

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## Cost comparison with FMM

Total cost  $C$  of the multigrid Poisson solver:

$$C = (\#B + N_{it}(4\#A + 2 \times 3 + 2(2p + 2))) N_{pt} \quad (9)$$

$\#B$  and  $\#A$  stand for the numbers of non zero elements of the stencils  $B$  and  $A$ ,  $N_{it}$  the number of V-cycle iterations,  $N_{pt}$  the number of points and  $2p$  the order of the method.

For instance, the 3-dimensional 6th-order HOC stencil verifies  $\#A = 27$ ,  $\#B = 25$  and only needs 12 iterations to converge to the computer rounding error so its cost is given by  $C = 1585N_{pt}$ . Of course it is possible to decrease this cost by fixing a larger error tolerance and taking fewer iterations. For instance converging to  $10^{-6}$  only takes 5 iterations then the cost is given by  $C = 675N_{pt}$ .

This compares advantageously to the costs given for the  $p$ th-order Fast Multipole Method whose optimal implementation in 3 dimensions yields

$$C = 200N_{pt}p + 3.5N_{pt}p^2.$$

Which means  $C = 1326N_{pt}$  for the 6th-order  $p = 6$  case.

## Experimental computational time with OpenMP

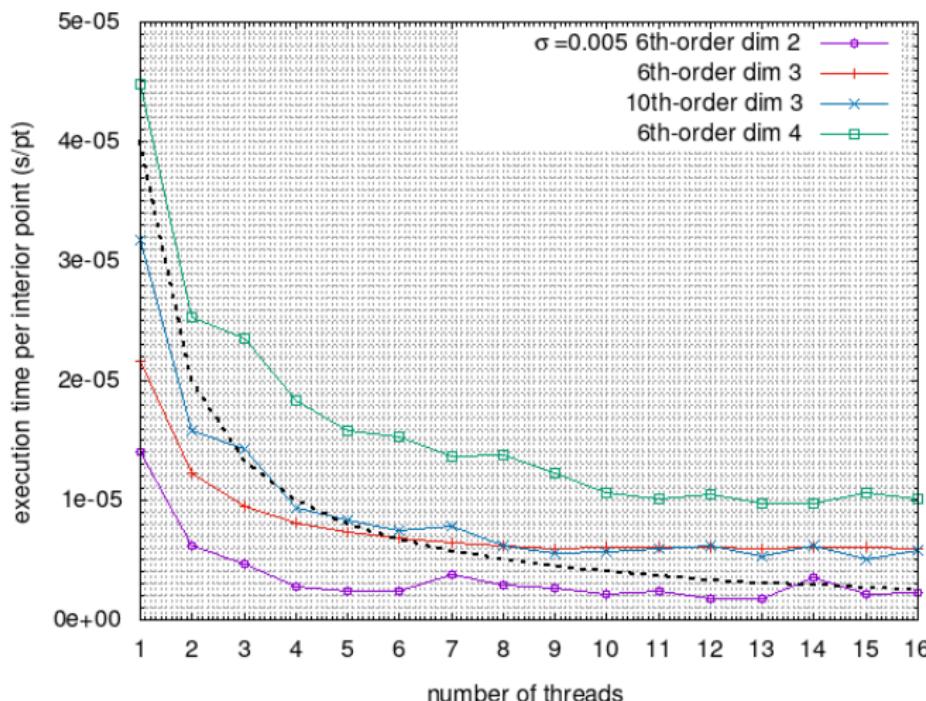


Figure: Time per interior point to converge to  $\|\Delta u_n - \Delta u\| < 10^{-10} \|\Delta u\|$  when varying the number of threads.

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## Conclusion – Perspectives

Original vertex-centered AMR multigrid Poisson solver of any orders and for any dimensions.

Perspectives:

- direct computation of the gradient of the potential from the density function as it is done in Fast Multipole Methods,
- MPI-parallelization,
- introduction of immersed boundary to provide an AMR immersed boundary scheme