

Fatigue lifetime analysis of general 3D crack configurations using \mathcal{H} -matrix accelerated boundary element method

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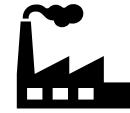
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²POEMs, UMA, ENSTA (Palaiseau, France)

Contents

- I. Industriel context
- II. General scheme for solving crack problem using fast BEM
- III. Integration to fatigue lifespan analysis
- IV. Numerical examples
- V. Conclusion – PhD to be continued...

I. Industrial context



- i. Introduction
- ii. Established tools at Safran
- iii. Goals of the PhD
- iv. Link between lifespan and crack propagation

ARIZE Project – *Aeronautics Research and Industry new horizons finite Elements software*

- Started in 2021
- Funded by the DGAC
- Aims to achieve the environmental objectives set by the European Commission and the French government through innovation
- Partnership between Safran, Onera, Armines, Mines Paris, and Transvalor

How does this thesis fit into this project?

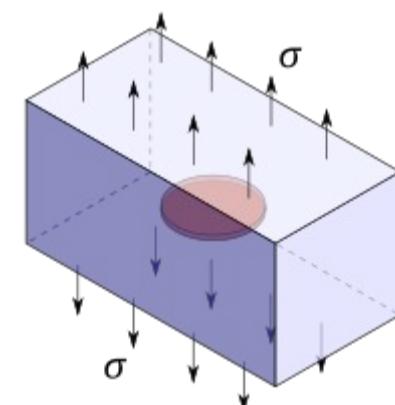
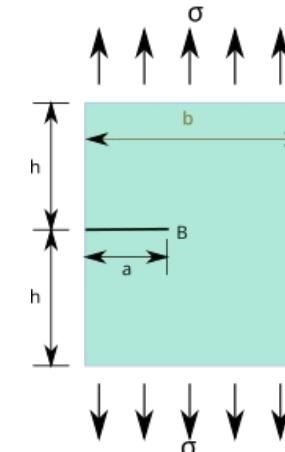
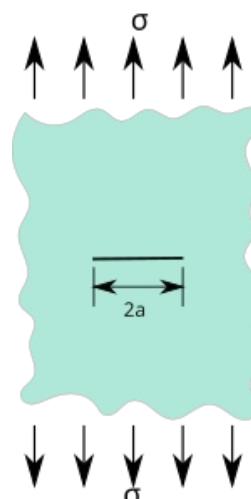
One focus of the ARIZE project involves **accelerating the computational resolution** of partial differential equations in mechanics, particularly in **fracture mechanics**. Safran's current methods are all based on **finite element methods**.

What crack problems models are used at SAE?

Existing Tools at SAE

- 1 Analytical tools
- 2 Semi-analytical tools
- 3 3D crack problems with FEM
- 4 "2.5D" crack problems with FEM

1 Simple 1D, 2D, and 3D **analytical solutions**. Too canonical and therefore not applicable for **industrial** cases.



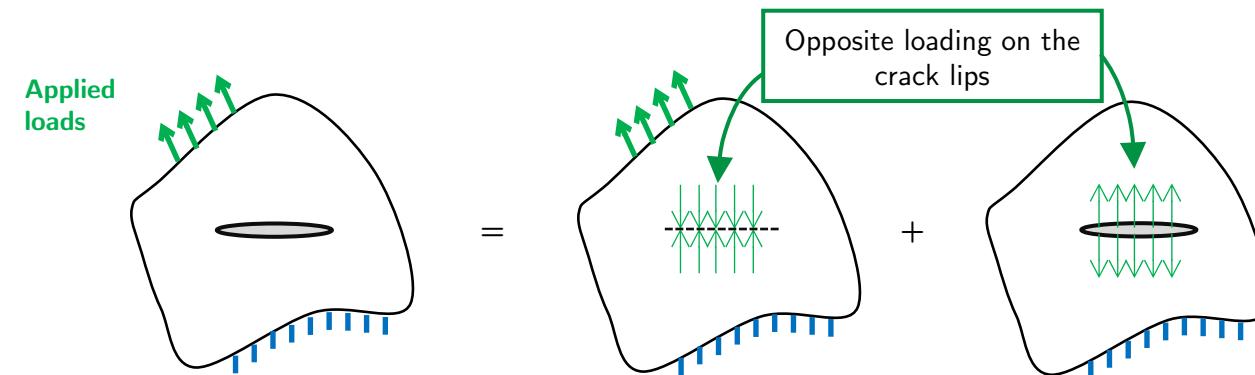
$$K_I = \sigma\sqrt{\pi a} \quad K_I = \sigma\sqrt{\pi a} \left(\frac{1 + \frac{3a}{b}}{2\sqrt{\frac{\pi a}{b}} \left(1 - \frac{a}{b}\right)^{3/2}} \right) \quad K_I = \frac{\sigma\sqrt{\pi}}{E(k)} \sqrt{\frac{b}{a} \left(\frac{a^4 \sin^2 \phi + b^4 \cos^2 \phi}{a^2 \sin^2 \phi + b^2 \cos^2 \phi} \right)^{1/4}}$$

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- 2 Bueckner's superposition principle combined with weight functions.
Disadvantages: over simplified geometry and does not take into account the impact of the presence of the crack on the structure's relaxation.



$$\begin{aligned}
 {}^a) \quad K_I(a) &= {}^b) \quad K_I(b) + {}^c) \quad K_I(c) \\
 K_I(a) &= 0 \\
 K_I(a) &= K_I(c)
 \end{aligned}$$

$$K_I(\mathbf{Q}) = \iint_{\text{Crack surface}} \underbrace{W_{QQ'}(\mathbf{Q}, \mathbf{Q}')}_{\text{Weight function}} \cdot \sigma(\mathbf{Q}') \cdot dS_{Q'}$$

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3

Classical **finite element method**. Disadvantage: very long set up (up to several **months**), as well as for the calculation of crack propagation (up to a few **weeks**).

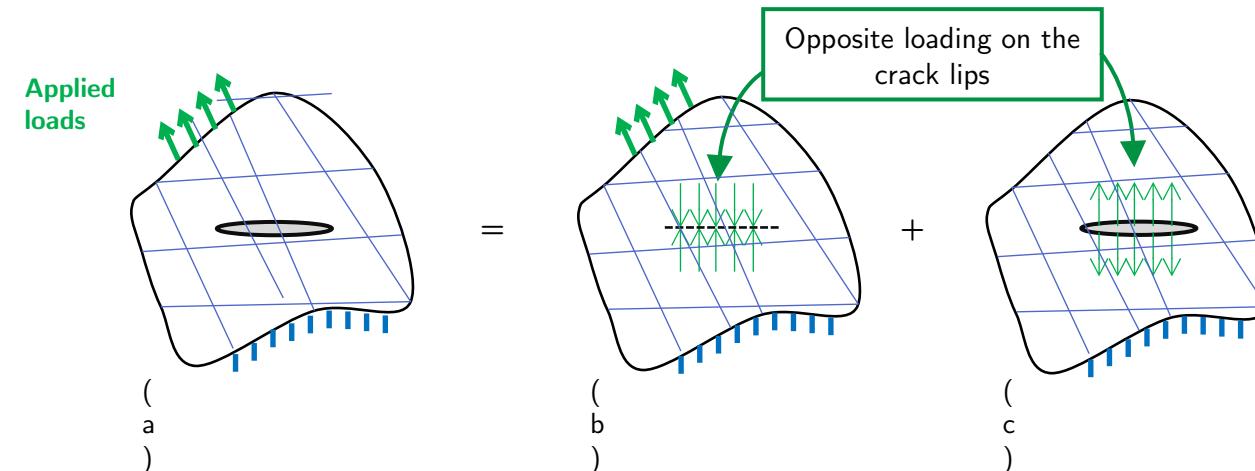
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4 = 2 + 3. Use of the **superposition principle** as in 2 and **finite element method** as in 3.



The SIF is computed by **post-processing** FEM solution

And my thesis?

Existing Tools at SAE

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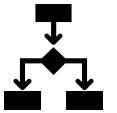


My thesis: solving 3D crack problems (as 3 and 4) but using the **BEM** accepting several hypothesis.

Hypothesis of the framework

- LEFM (Linear Elastic Fracture Mechanics)
- Isotropic, homogeneous material

Final goal : speed up computation time for certain industrial study cases under the assumptions mentioned above.



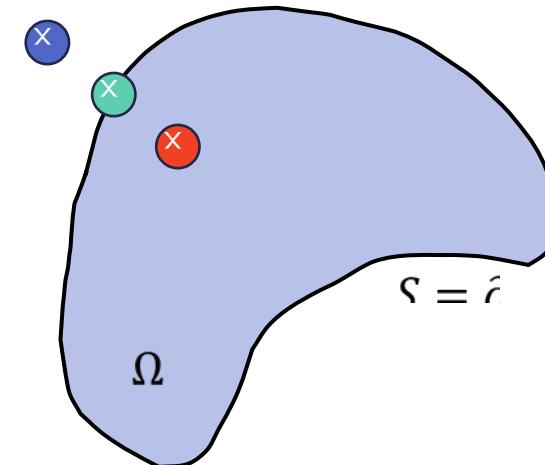
II. General scheme for solving crack problem using fast BEM

- i. Recall : Boundary Integral Equation in Elastostatics
- ii. Displacement Discontinuity Method
- iii. Numerical resolution by the Boundary Element Method
- iv. Computation of (singular) integrals
- v. Focus on enhanced Guiggiani's algorithm
- vi. \mathcal{H} - matrices compression

Boundary integral equation for elastic solids

$$\mathcal{S}\mathbf{t} - \mathcal{D}\mathbf{u} = \begin{cases} \mathbf{u} & \text{in } \Omega \\ \frac{1}{2}\mathbf{u} & \text{over } S \\ \mathbf{0} & \text{ow.} \end{cases}$$

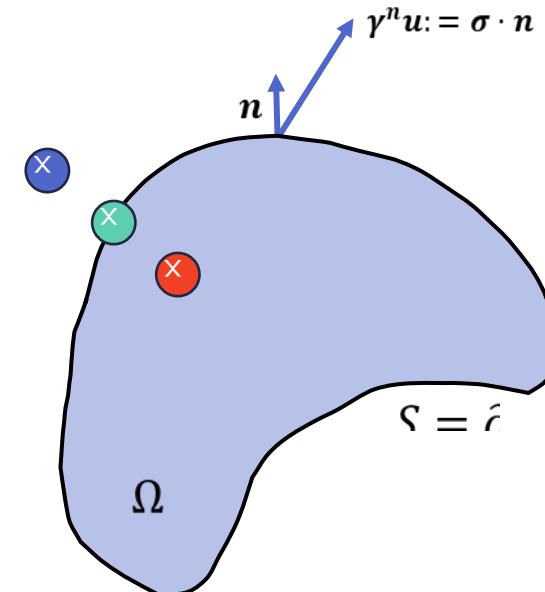
$$\mathcal{D}^*\mathbf{t} - \mathcal{H}\mathbf{u} = \begin{cases} \mathbf{t} & \text{in } \Omega \\ \frac{1}{2}\mathbf{t} & \text{over } S \\ \mathbf{0} & \text{ow.} \end{cases}$$



Boundary integral equation for elastic solids

$$\mathcal{S}\mathbf{t} - \mathcal{D}\mathbf{u} = \begin{cases} \mathbf{u} & \text{in } \Omega \\ \frac{1}{2}\mathbf{u} & \text{on } \Gamma [f](x \in S) = \int G(x, y) \cdot f(y) dT_y \\ \mathbf{0} & \text{ow.} \end{cases}$$

$$\mathcal{D}^*\mathbf{t} - \mathcal{H}\mathbf{u} = \begin{cases} \mathbf{t} & \text{in } \Omega \\ \frac{1}{2}\mathbf{t} & \text{over } S \\ \mathbf{0} & \text{ow.} \end{cases}$$



$$\mathcal{S}\varphi(x) = \int_S G(x, y) \cdot \varphi(y) dT_y$$

$$\mathcal{D}\varphi(x) = \mathbf{p} \cdot \mathbf{v} \int_S (\gamma_y^n G(x, y))^T \cdot \mathbf{f}(y) dT_y$$

$$\mathcal{D}^*\varphi(x) = \mathbf{p} \cdot \mathbf{v} \int_S \gamma_x^n G(x, y) \cdot \mathbf{f}(y) dT_y$$

$$\mathcal{H}\varphi(x) = \mathbf{f} \cdot \mathbf{p} \int_S \gamma_x^n (\gamma_y^n G(x, y))^T \cdot \mathbf{f}(y) dT_y$$

Free space Green's fundamental solution:

$$G(x, y) = \frac{1}{16\pi\mu(1-\nu)r} \left\{ (3-4\nu) \mathbf{I} + \hat{\mathbf{r}} \otimes \hat{\mathbf{r}} \right\}$$

Generalized normal derivative ("trace")

$$\gamma^n u := \sigma \cdot n = (\lambda \operatorname{div} u) n + \mu (\nabla u + \nabla^T u) \cdot n$$

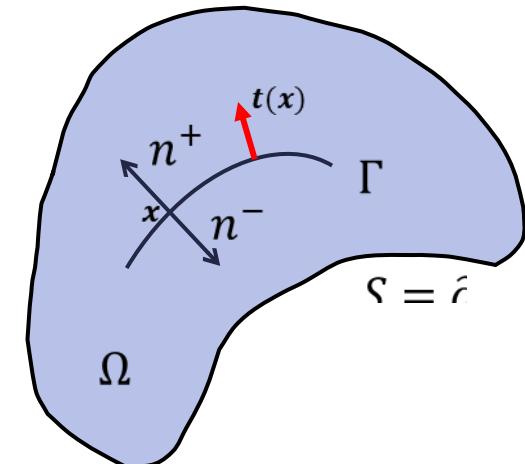
Equilibrium equation (without volumic forces) and Hooke's law :

$$\operatorname{div} \boldsymbol{\sigma} = \vec{0}, \quad \sigma_{ij} = C_{ijkl} \partial u_k / \partial x_l$$

Extension to cracked solids...

Displacement discontinuity method

$$\begin{cases} \mathcal{H}_{\Gamma\Gamma}\phi_{\Gamma} + \mathcal{D}_{\Gamma S}^* \mathbf{t}_S - \mathcal{H}_{\Gamma S} \mathbf{u}_S = -\mathbf{t} & (\text{on } \Gamma) \\ \mathcal{S}_{SS} \mathbf{t}_S - \mathcal{D}_{SS} \mathbf{u}_S + \mathcal{D}_{S\Gamma} \phi_{\Gamma} = \frac{1}{2} \mathbf{u} & (\text{on } S) \end{cases}$$



... And its interior representation formulas.

Interior representation

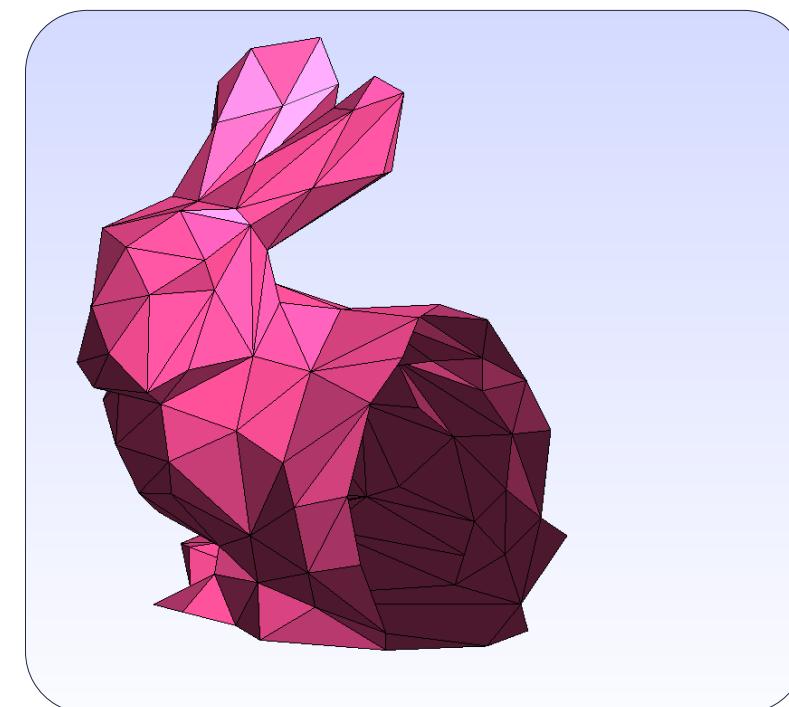
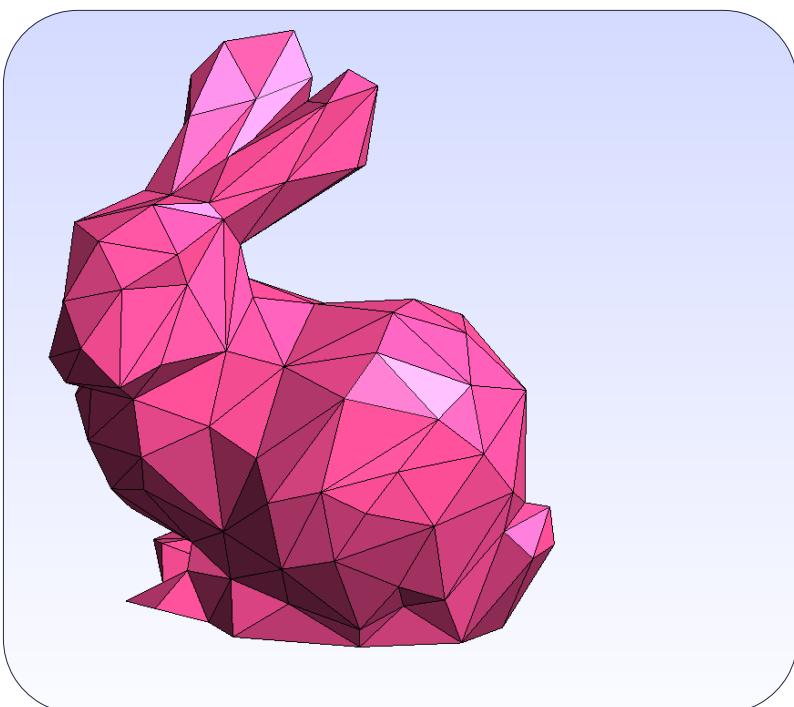
$$\left. \begin{aligned} \mathbf{u}_{\Omega} &= \mathcal{D}_{\Omega\Gamma} \phi_{\Gamma} + \mathcal{D}_{\Omega\Gamma} \mathbf{u}_S \\ \boldsymbol{\sigma}_{\Omega} &= \mathbf{C} :: (\boldsymbol{\epsilon}(\mathcal{D}_{\Omega\Gamma} \phi_{\Gamma}) + \boldsymbol{\epsilon}(\mathcal{D}_{\Omega\Gamma} \mathbf{u}_S)) \end{aligned} \right\} (\text{on } \Omega)$$

$$\phi = \mathbf{u}^+ - \mathbf{u}^-$$

Crack Opening Displacement

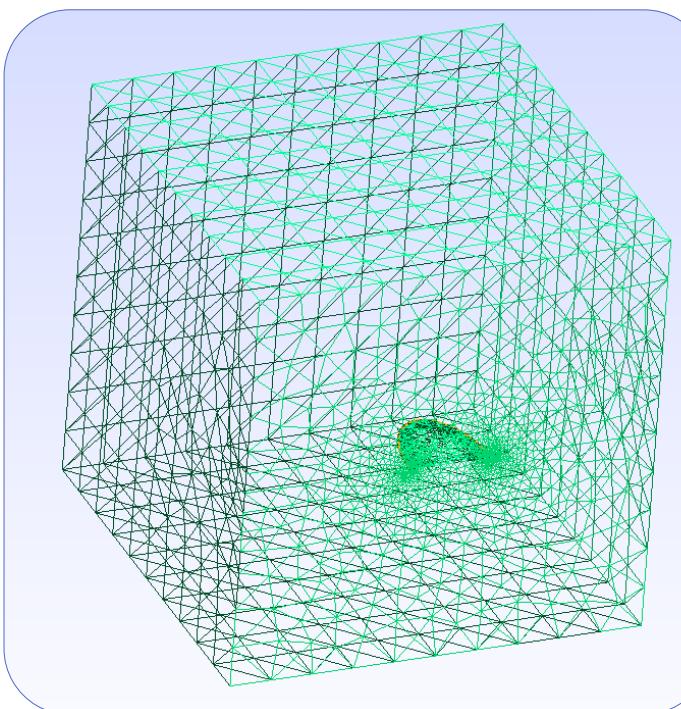
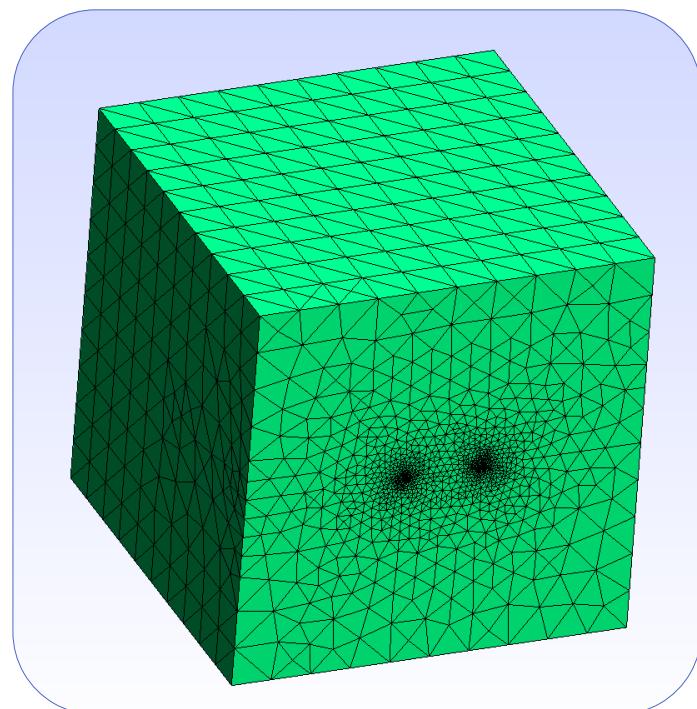
Main characteristics of the BEM

- **Only the boundary** (3D surface – including the crack) is meshed : purely 2D éléments in 3D framework
- We describe each 2D element by a reference element and its polynomial interpolant (**like in the FEM**)
- The **numerical difficulty** occurs when integrating singular kernels
- **Nyström method** : the quadrature nodes encode the unknown discrete values

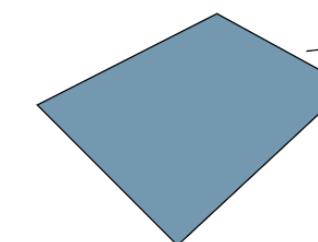


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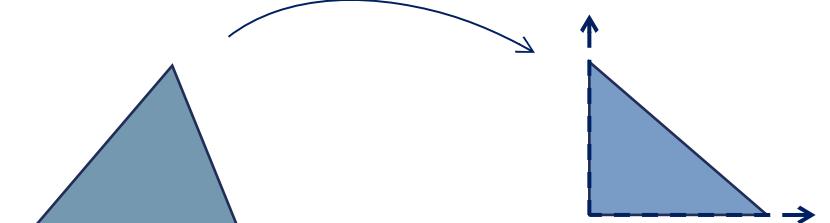
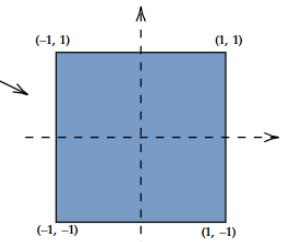
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Physical element



Reference element



Elementary integration on a reference element τ_e

- 1) x is far enough from τ_e (**regular integration** – Gauss-Legendre quadrature)
- 2) x is on τ_e (**singular integration**) → Enhanced Guiggiani's algorithm
- 3) Intermediate case : x is close to τ_e (**nearly-singular integration**) but not in

(un)known function \uparrow singular kernel \rightarrow
 $\mathcal{K}[\varphi](x) = \int_{\tau_e} \mathbf{K}(x, y) \cdot \varphi(y) dS_y$
 \downarrow source point \longrightarrow integration element

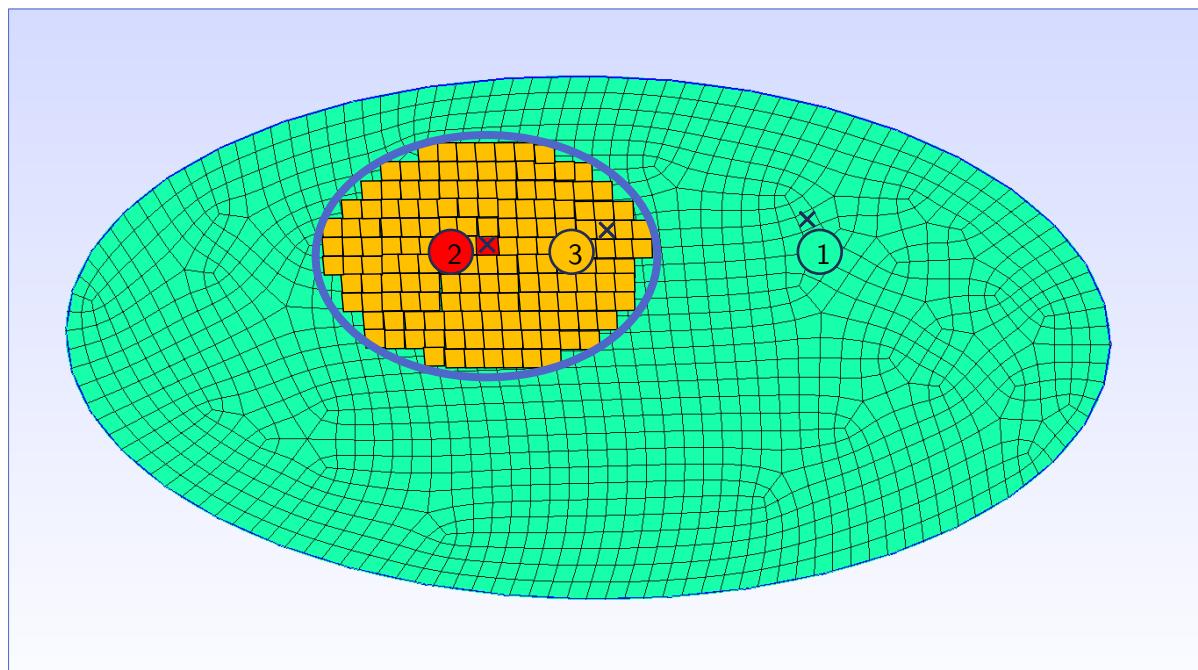


Illustration on an elliptic crack surface

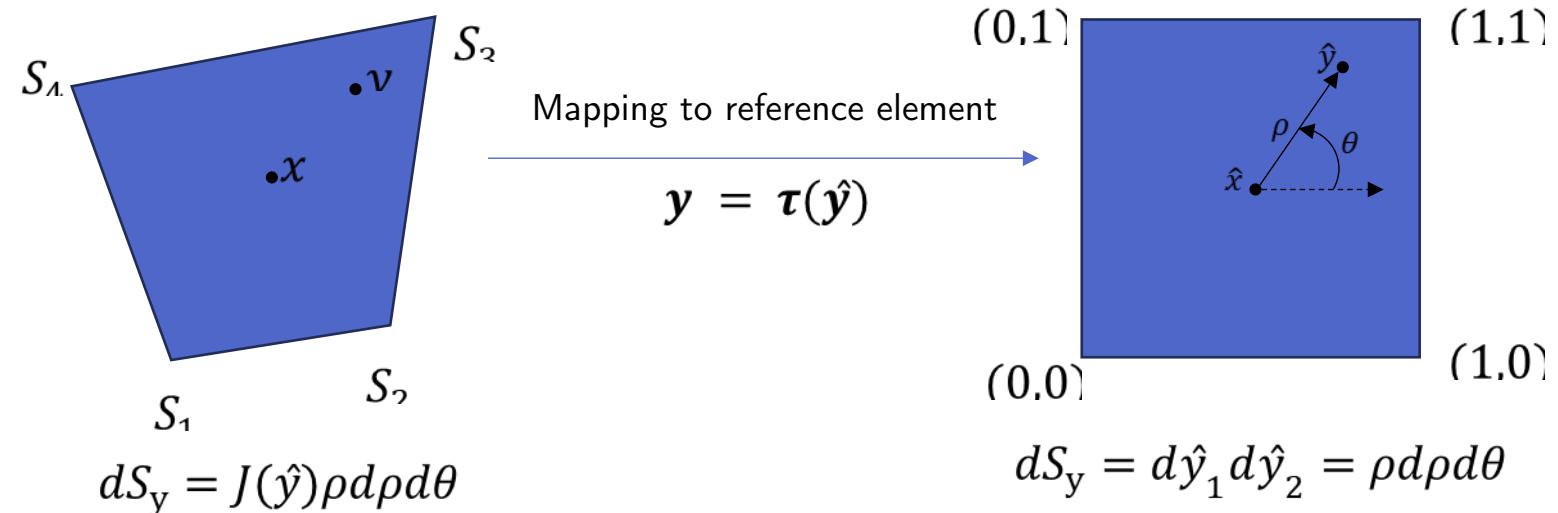
- Integration element
- ✗ Source point
- ✗ Singular case
- ✗ Nearly-singular case
- ✗ Regular case

Guiggiani (1992) direct algorithm for computing singular elementary integrals

Core idea

$$K(\hat{x}, \hat{y}) := \rho J(\hat{y}) N(\hat{y}) K(x, y) = \left\{ K(\rho, \theta) - \frac{K_{-2}(\theta)}{\rho^2} - \frac{K_{-1}(\theta)}{\rho} \right\} + \left\{ \frac{K_{-2}(\theta)}{\rho^2} + \frac{K_{-1}(\theta)}{\rho} \right\}$$

Regularized part = $\mathcal{O}(1)$ Remainder, analytically handled



Main goal : get the expression of the Laurent coefficients $K_{-1}(\theta)$ and $K_{-2}(\theta)$

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Direct approach, described in
Guiggiani's original article

Main goal : get the expression of the Laurent coefficients $K_{-1}(\theta)$ and $K_{-2}(\theta)$

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Laurent coefficient	<table border="1"> <thead> <tr> <th>Laurent coefficient</th><th>Laplace $\Delta u + f = 0$</th></tr> </thead> <tbody> <tr> <td>Original hypersingular kernel</td><td>$\frac{1}{4\pi r^3} \{(\mathbf{I} - 3\hat{\mathbf{r}} \otimes \hat{\mathbf{r}}) \cdot \mathbf{n}(x)\} \cdot \mathbf{n}(y)$</td></tr> <tr> <td>$K_{-2}(\theta)$</td><td>$\frac{J(\eta)N(\eta)}{4\pi \mathbf{f}(\eta) \cdot \mathbf{u}(\theta) }$</td></tr> <tr> <td>$K_{-1}(\theta)$</td><td></td></tr> </tbody> </table>	Laurent coefficient	Laplace $\Delta u + f = 0$	Original hypersingular kernel	$\frac{1}{4\pi r^3} \{(\mathbf{I} - 3\hat{\mathbf{r}} \otimes \hat{\mathbf{r}}) \cdot \mathbf{n}(x)\} \cdot \mathbf{n}(y)$	$K_{-2}(\theta)$	$\frac{J(\eta)N(\eta)}{4\pi \mathbf{f}(\eta) \cdot \mathbf{u}(\theta) }$	$K_{-1}(\theta)$	
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$K_{-1}(\theta)$									



Direct approach, described in
Guiggiani's original article

Hypersingular elastostatics kernel :

$$\begin{aligned}
 \gamma_{1,x} (\gamma_{1,y} G(\mathbf{x}, \mathbf{y}))^T &= \frac{\mu}{4\pi(1-\nu)r^3} \left[3\hat{\mathbf{r}} \cdot \mathbf{n}_y \left\{ (1-2\nu)\mathbf{n}_x \otimes \hat{\mathbf{r}} + \nu((\hat{\mathbf{r}} \cdot \mathbf{n}_x)\mathbf{I} + \hat{\mathbf{r}} \otimes \mathbf{n}_x) - 5(\hat{\mathbf{r}} \cdot \mathbf{n}_x)\hat{\mathbf{r}} \otimes \hat{\mathbf{r}} \right\} \right. \\
 &\quad + 3\nu \left\{ (\hat{\mathbf{r}} \cdot \mathbf{n}_x)\mathbf{n}_y \otimes \hat{\mathbf{r}} + (\mathbf{n}_x \cdot \mathbf{n}_y)\hat{\mathbf{r}} \otimes \hat{\mathbf{r}} \right\} + (1-2\nu) \left\{ 3(\hat{\mathbf{r}} \cdot \mathbf{n}_x)\hat{\mathbf{r}} \otimes \mathbf{n}_y + (\mathbf{n}_x \cdot \mathbf{n}_y)\mathbf{I} \right. \\
 &\quad \left. + \mathbf{n}_y \otimes \hat{\mathbf{r}} \right\} - (1-4\nu)\mathbf{n}_x \otimes \mathbf{n}_y \right] \quad (1.15)
 \end{aligned}$$

Main goal : get the expression of the Laurent coefficients $K_{-1}(\theta)$ and $K_{-2}(\theta)$

Improvement of the Guiggiani's direct approach from a numerical aspect

$$K_{-2}(\theta) = \lim_{\rho \rightarrow 0} K(\rho, \theta) \rho^2 \quad 1$$

$$K_{-1}(\theta) = \lim_{\rho \rightarrow 0} \left\{ \rho K(\rho, \theta) - \frac{K_{-2}(\theta)}{\rho} \right\} \quad 2$$

Main goal : get the expression of the Laurent coefficients $K_{-1}(\theta)$ and $K_{-2}(\theta)$

Improvement of the Guiggiani's direct approach from a numerical aspect

$$K_{-2}(\theta) = \lim_{\rho \rightarrow 0} K(\rho, \theta) \rho^2 \quad 1$$

$$K_{-1}(\theta) = \lim_{\rho \rightarrow 0} \left\{ \rho K(\rho, \theta) - \frac{K_{-2}(\theta)}{\rho} \right\} \quad 2$$

Hybrid approach (direct / Richardson) for the Guiggiani's method :

$$1 \quad K_{-2}(\theta) = \frac{1}{A^3(\theta)} \underbrace{\hat{K}(\hat{\mathbf{A}}(\theta))}_{\text{regular part of the kernel}} N(\eta) J(\eta), \quad A(\theta) = \left\| \mathbf{D}\boldsymbol{\tau}(\boldsymbol{\eta}) \cdot \begin{pmatrix} \cos\theta \\ \sin\theta \end{pmatrix} \right\|, \quad \hat{\mathbf{A}}(\theta) = \mathbf{D}\boldsymbol{\tau}(\boldsymbol{\eta}) \cdot \begin{pmatrix} \cos\theta \\ \sin\theta \end{pmatrix} / A(\theta)$$

Richardson extrapolation

Guiggiani algorithm

Main goal : get the expression of the Laurent coefficients $K_{-1}(\theta)$ and $K_{-2}(\theta)$

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$$2 \quad \begin{aligned} (a) \quad K_{-1}(\theta) &= \rho^{-1} [\rho^2 K(\rho, \theta) - K_{-2}(\theta)] + \underset{\rho \rightarrow 0}{\mathcal{O}}(\rho) \\ (b) \quad K_{-1}(\theta) &= (t\rho)^{-1} [(t\rho)^2 K(t\rho, \theta) - K_{-2}(\theta)] + \underset{\rho \rightarrow 0}{\mathcal{O}}(\rho) \end{aligned}$$

same with $t\rho$ instead of ρ

$$(b) - t(a) \Rightarrow K_{-1}(\theta) = \frac{1}{1-t} \{ t\rho(K(t\rho, \theta) - \rho K(\rho, \theta)) + (t - t^{-1}) K_{-2}(\theta) \} + \underset{\rho \rightarrow 0}{\mathcal{O}}(\rho^2)$$

and so on... $\tilde{K}_{-1}(\theta)$

Richardson extrapolation

Guiggiani algorithm

Problem
BEM matrices are dense, non-symmetrical and non-definite-positive

	+	-
BEM	Surface mesh, very less dofs	Dense matrices, non-symmetrical
FEM	Sparse matrices, symmetrical and semi-definite-positive	Huge number of dofs

- **High storage** requirement $\mathcal{O}(n^2)$
- Long **assembling** $\mathcal{O}(n^2)$ and **solving**

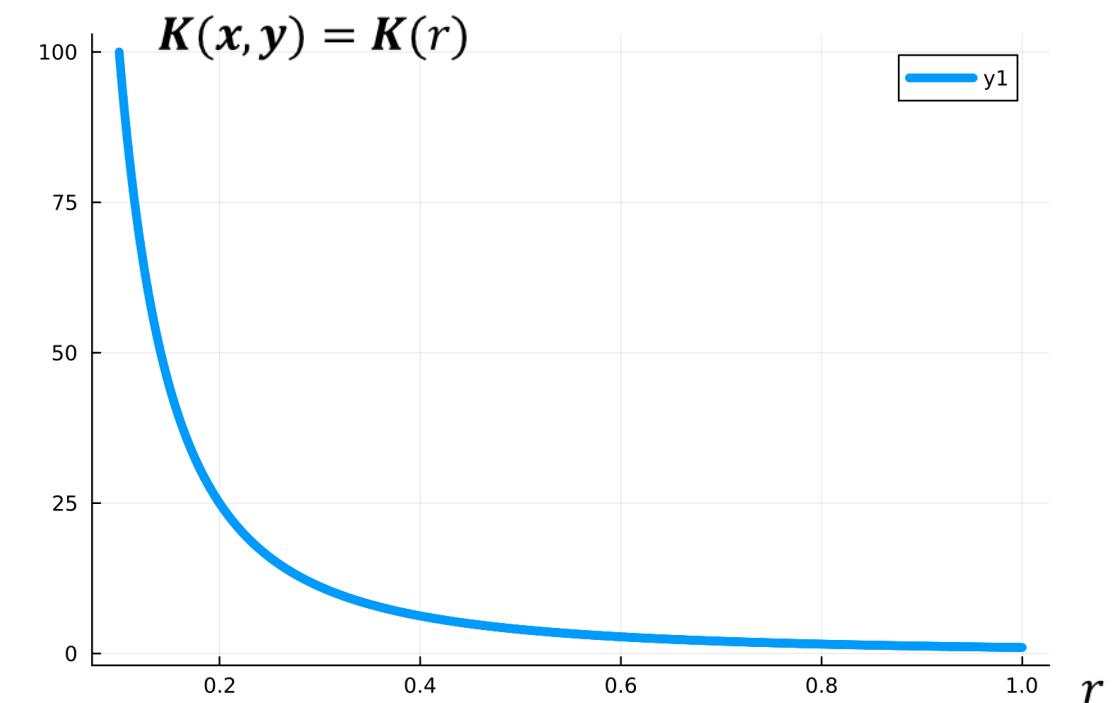


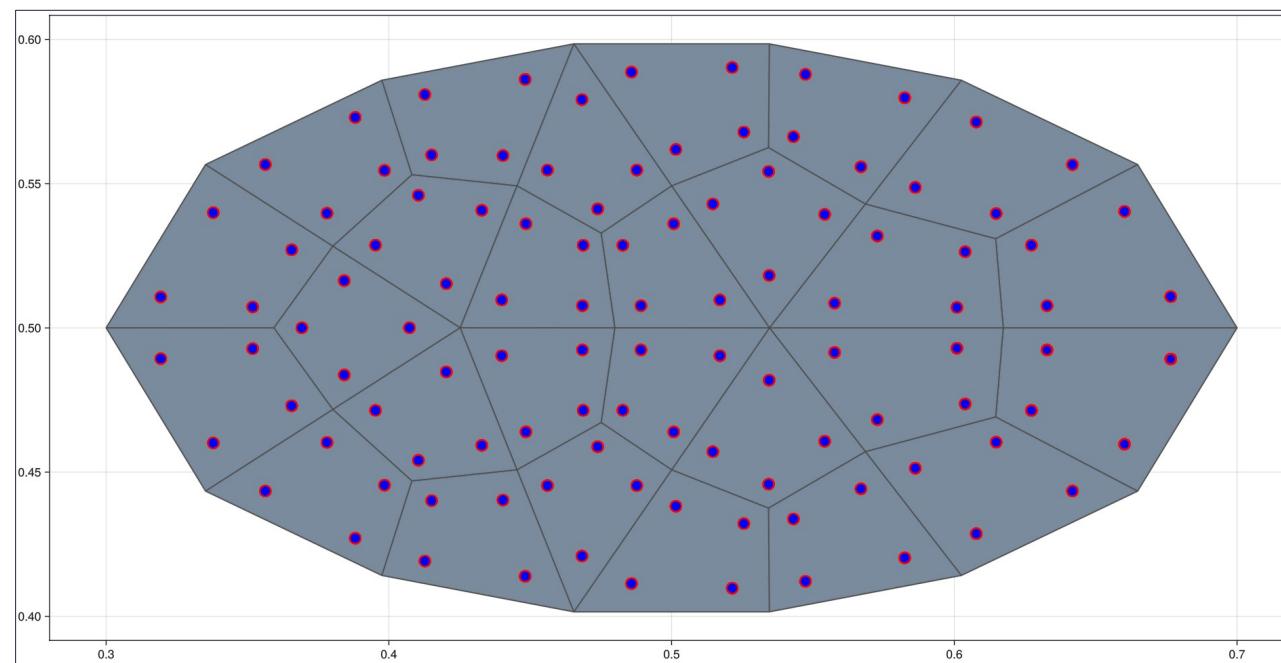


$$\mathbf{K}(\mathbf{x}, \mathbf{y}) = \underset{r \rightarrow +\infty}{o}\left(\frac{1}{r^s}\right), \quad s \in \{1, 2, 3\}$$

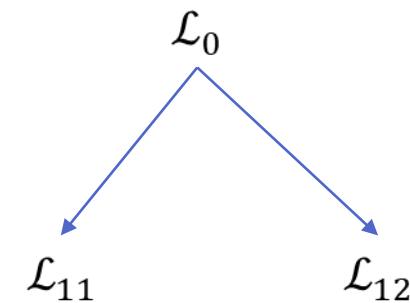
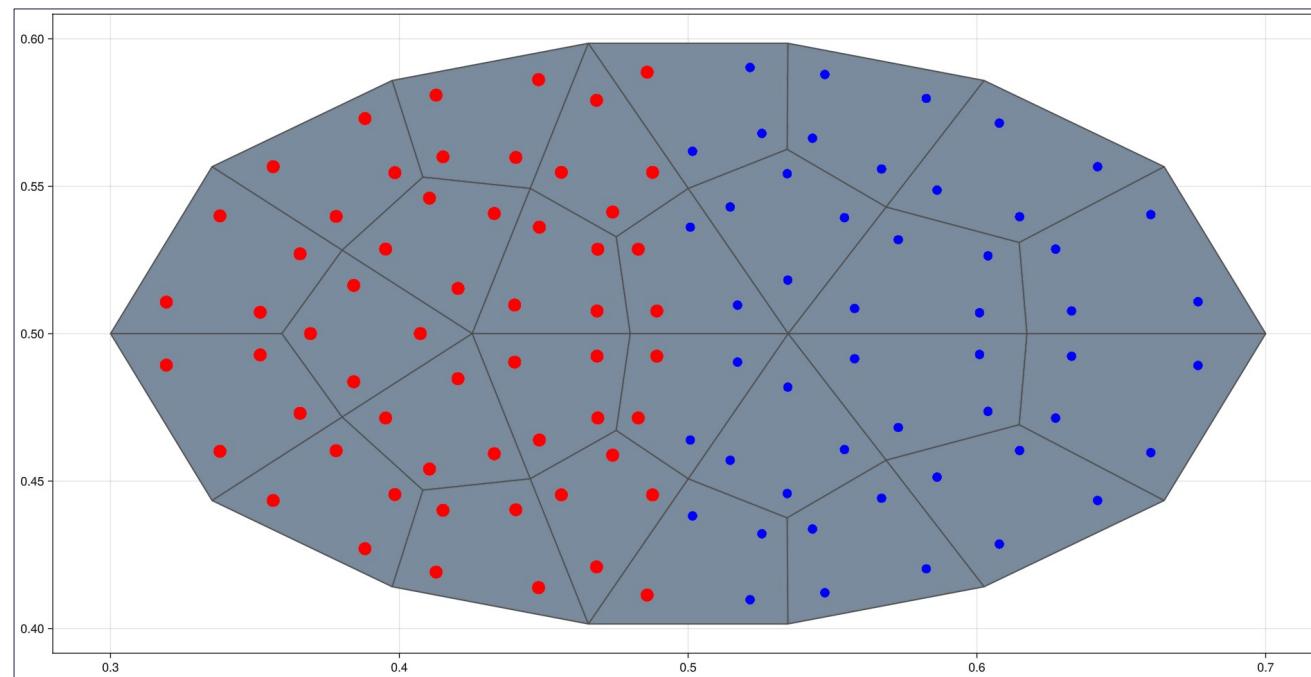
$$\forall \alpha, \beta \left| \frac{\partial^{|\alpha|+|\beta|} \mathbf{K}(\mathbf{x}, \mathbf{y})}{\partial x_\alpha \partial y_\beta} \right| \leq (|\alpha| + |\beta|) C \|\mathbf{x} - \mathbf{y}\|^{-(|\alpha|+|\beta|+s)}$$

Smoothness condition





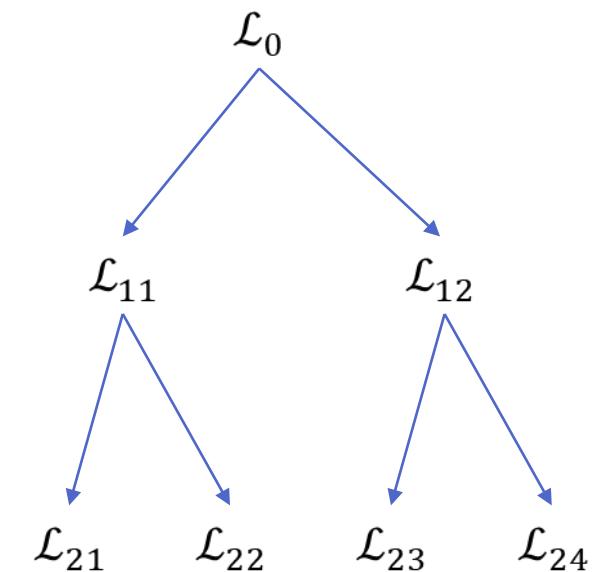
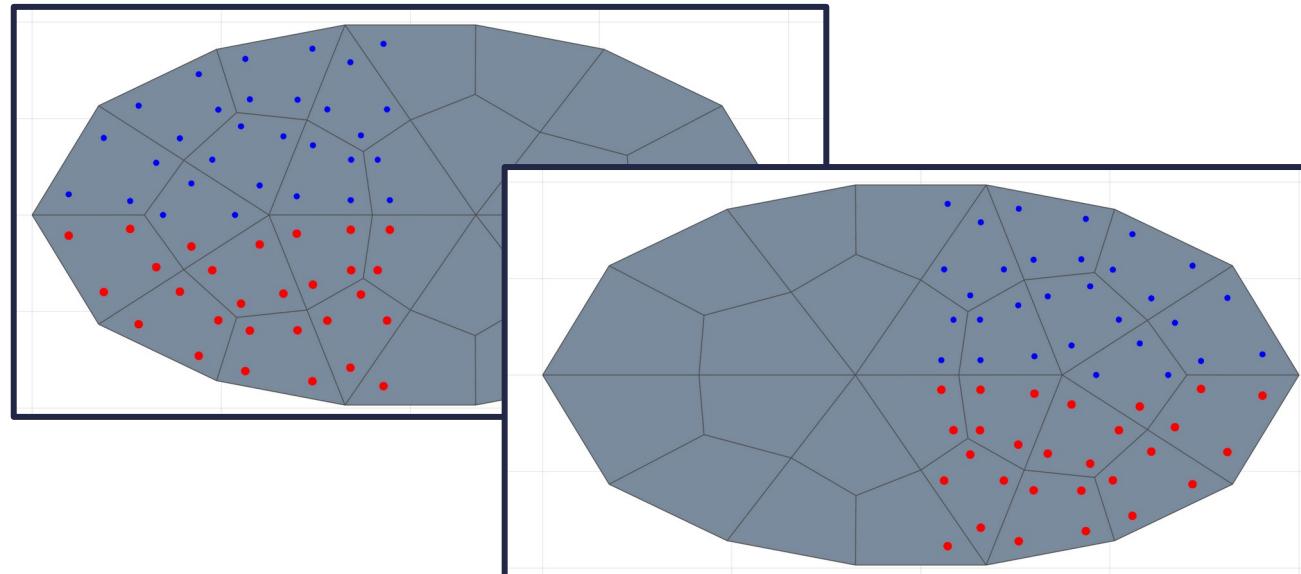
$$\mathcal{L}_0$$



Asymptotically Smooth kernels

Clustering

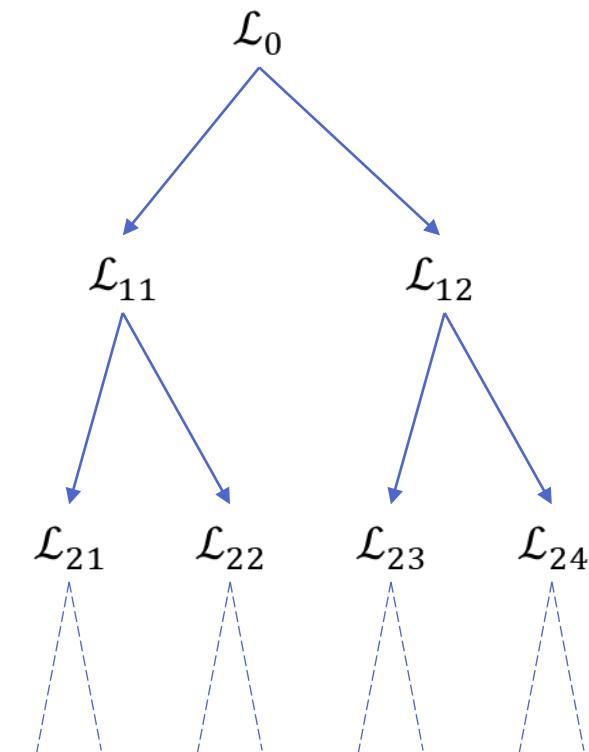
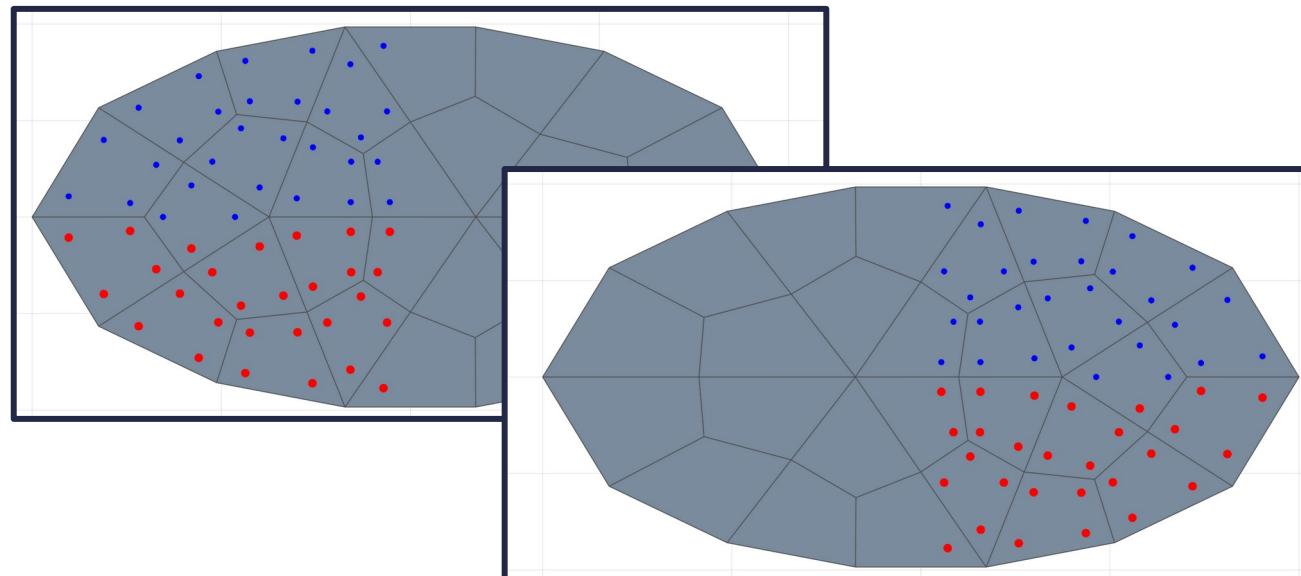
PACA algorithm



Asymptotically Smooth kernels

Clustering

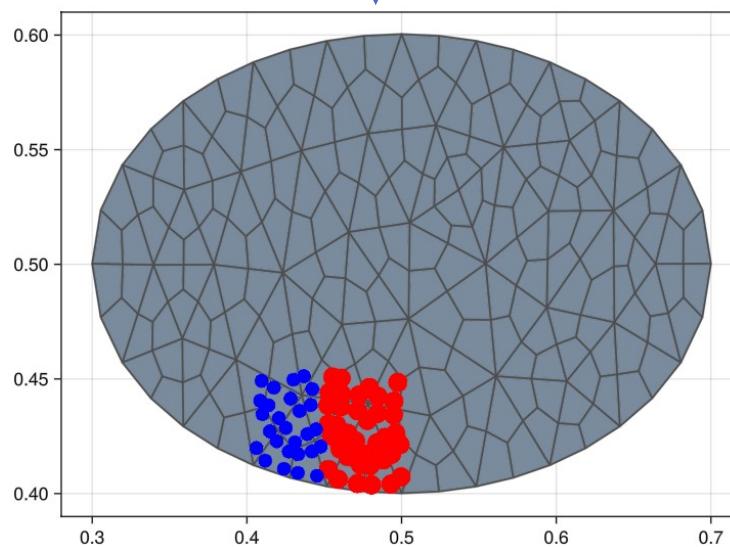
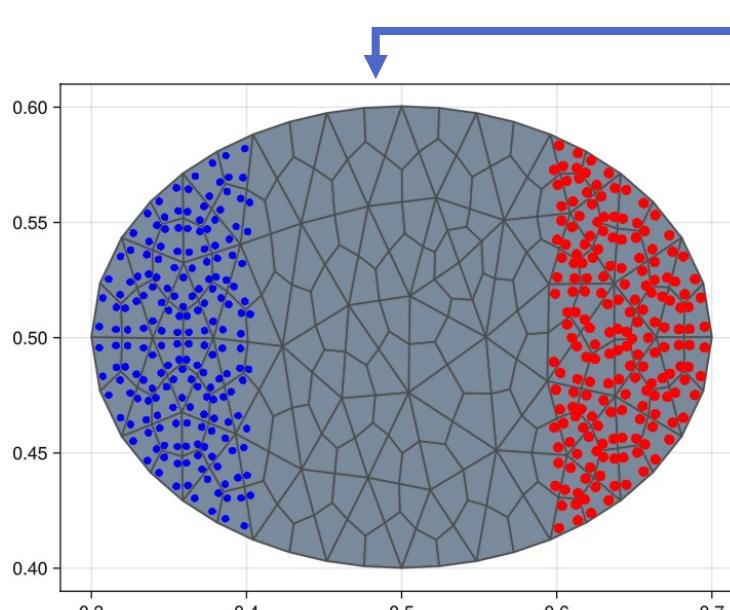
PACA algorithm



and so on...

II. General scheme for crack problem using fast BEM

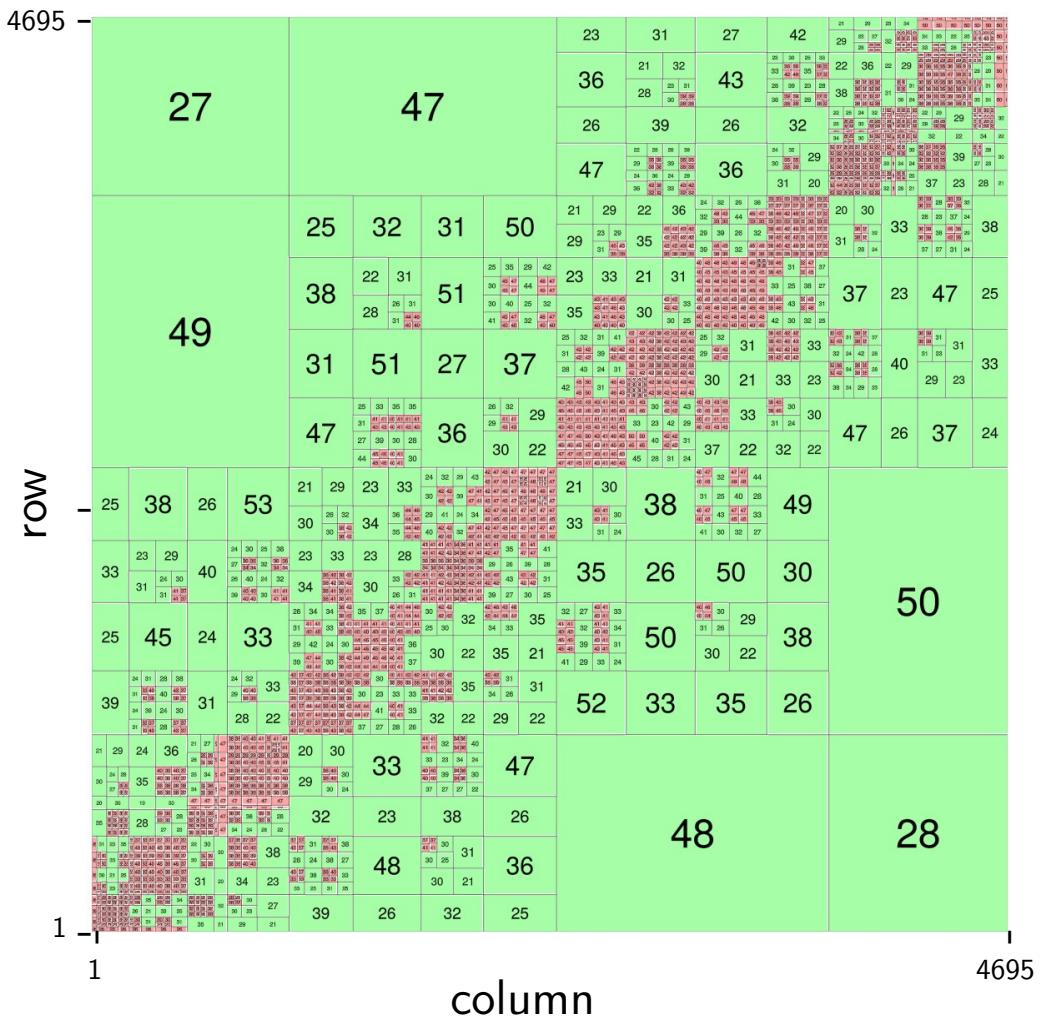
6. \mathcal{H} – matrices compression



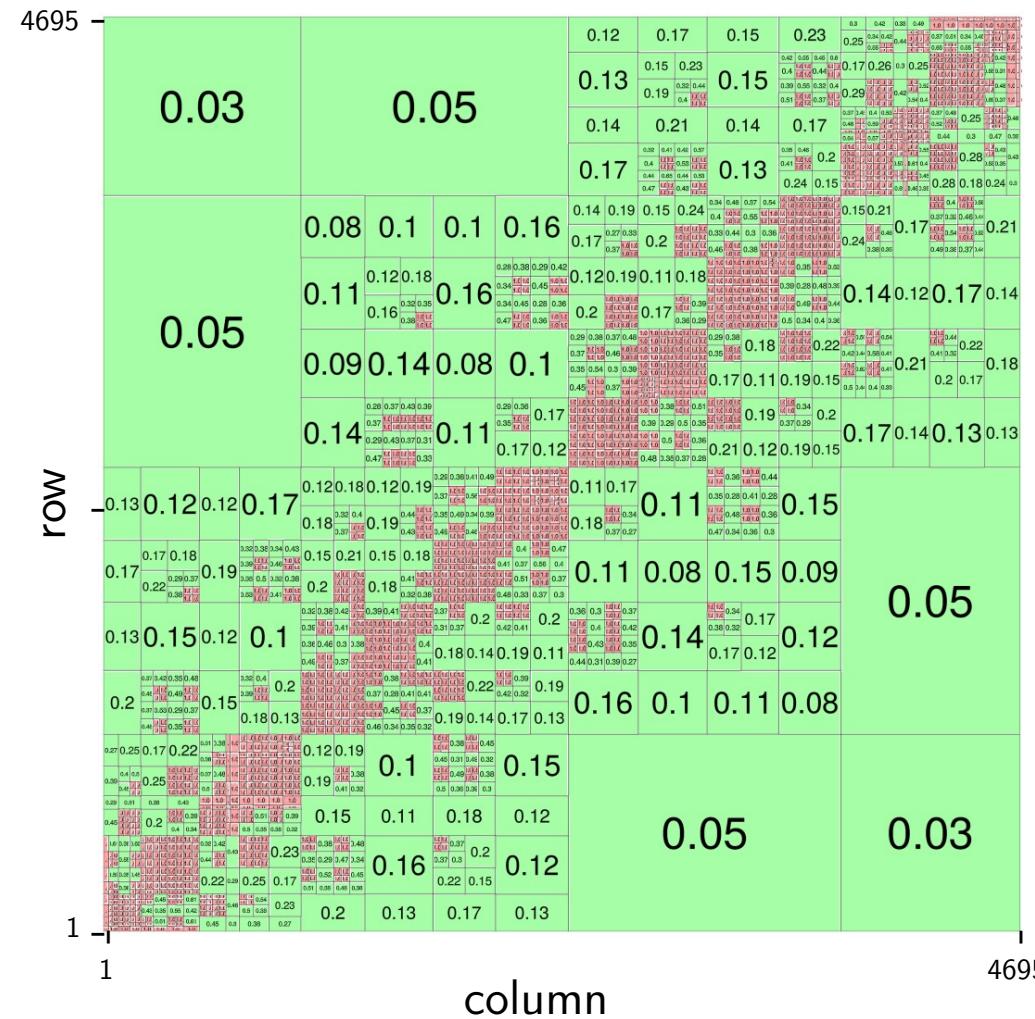
The figure shows a 10x10 grid of numbers ranging from 250 to 750. The grid is color-coded: green for values 250-350, pink for 350-450, and blue for 450-550. A blue border highlights the first three columns of the top-left section (250-350) and the first three rows of the bottom-left section (450-550). A green border highlights the last three columns of the top-right section (350-450) and the last three rows of the bottom-right section (550-650). Large black numbers are placed at the intersections of the highlighted areas: 26 at (250, 250), 44 at (350, 250), 46 at (250, 450), 45 at (550, 450), 28 at (450, 550), 44 at (550, 550), and 28 at (650, 550).

	250	350	450	550	650	750
250	26	44	46	45	28	44
350	35	35	35	35	35	35
450	35	35	35	35	35	35
550	35	35	35	35	35	35
650	35	35	35	35	35	35
750	35	35	35	35	35	35

Absolute rank of all \mathcal{H} –Matrices sub-blocks



Relative rank of all \mathcal{H} –Matrices sub-blocks

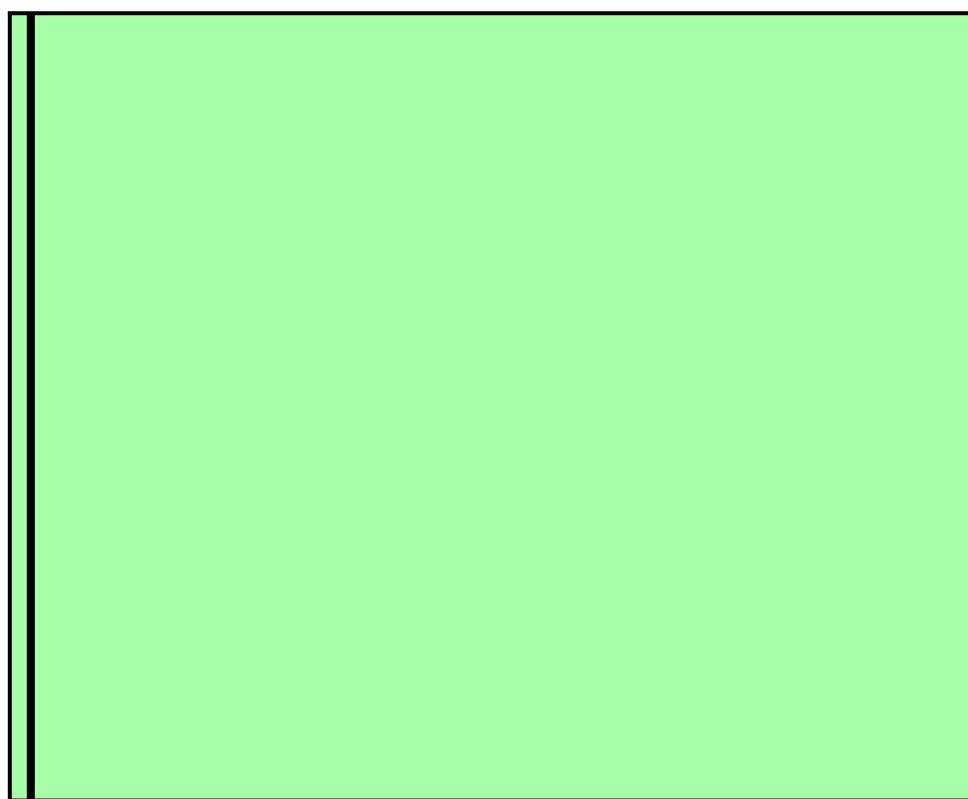


compressed

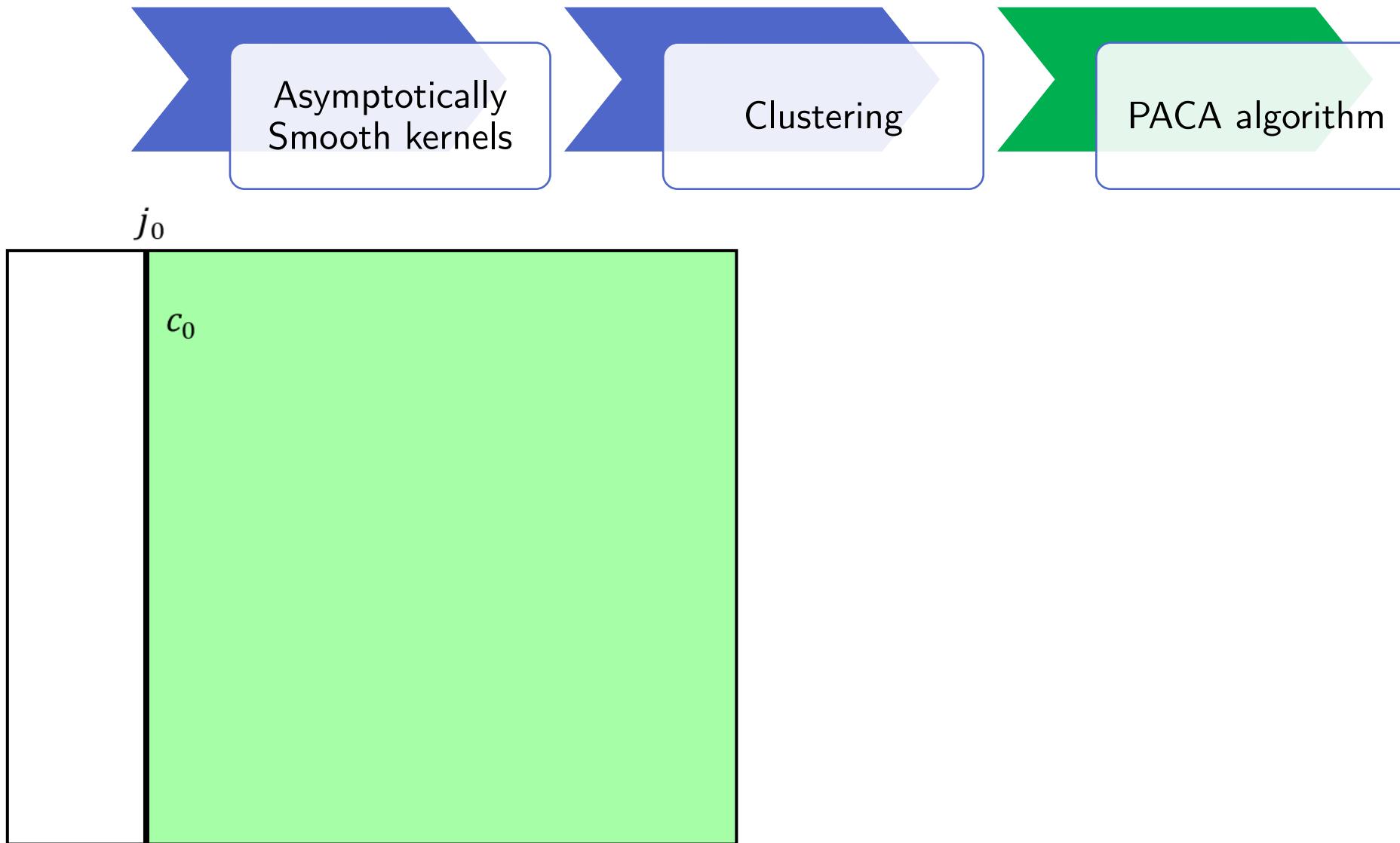
non- compressed



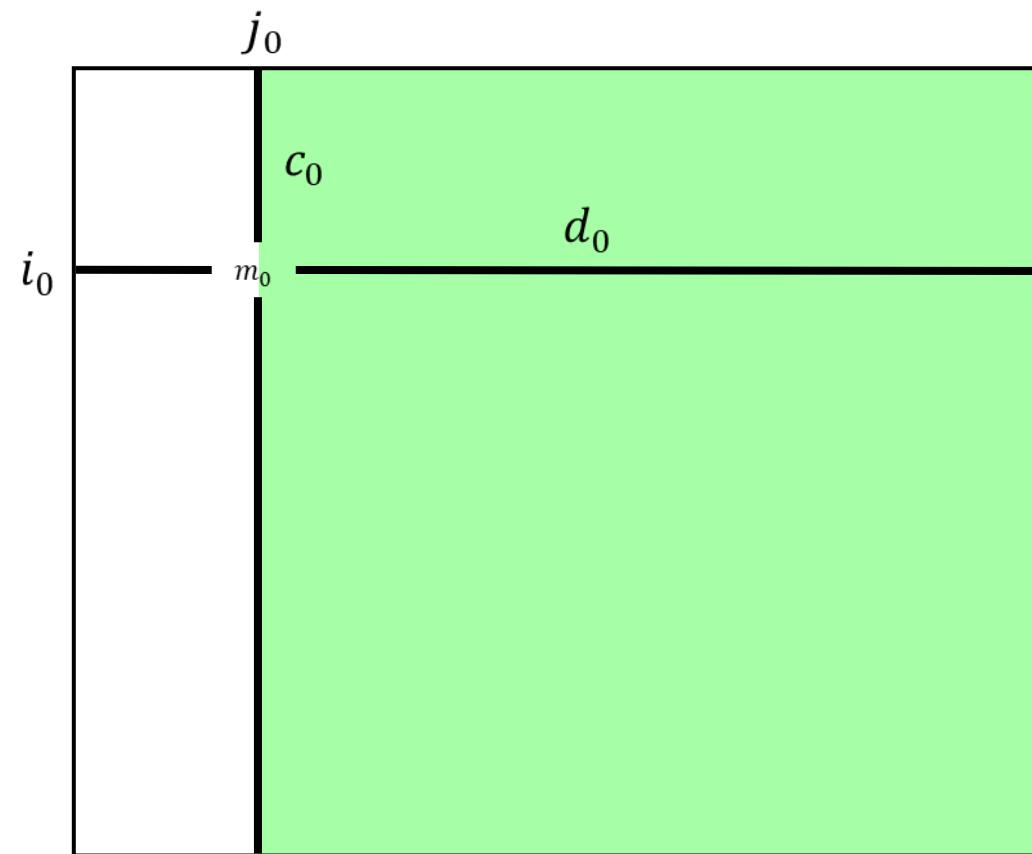
1



Admissible block B

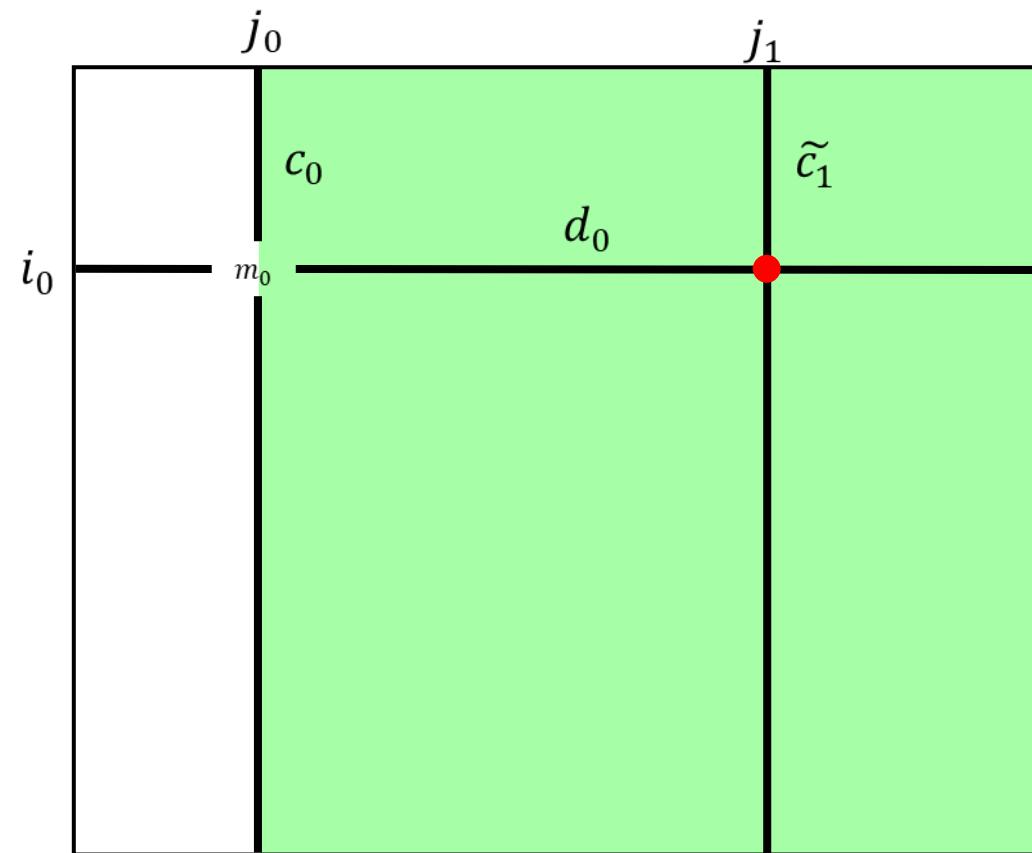


Admissible block B



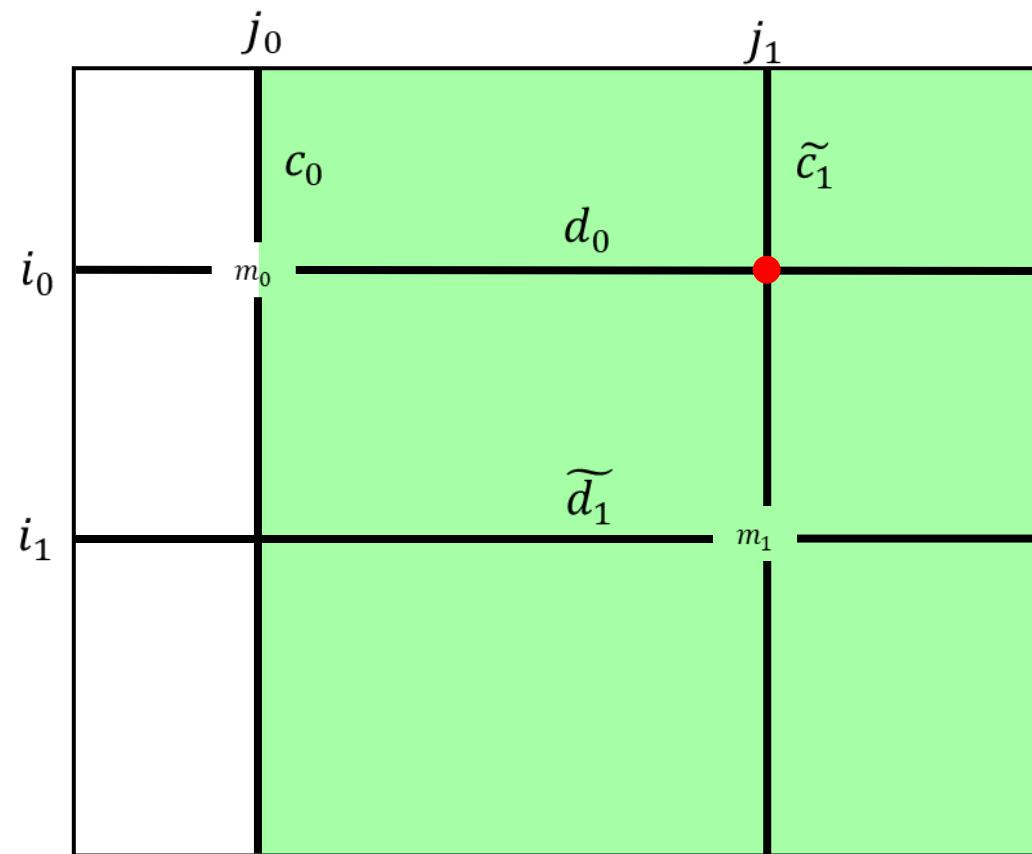
$$R_0 = \frac{1}{m_0} c_0 d_0^T \longrightarrow \text{rank 1}$$

Admissible block B



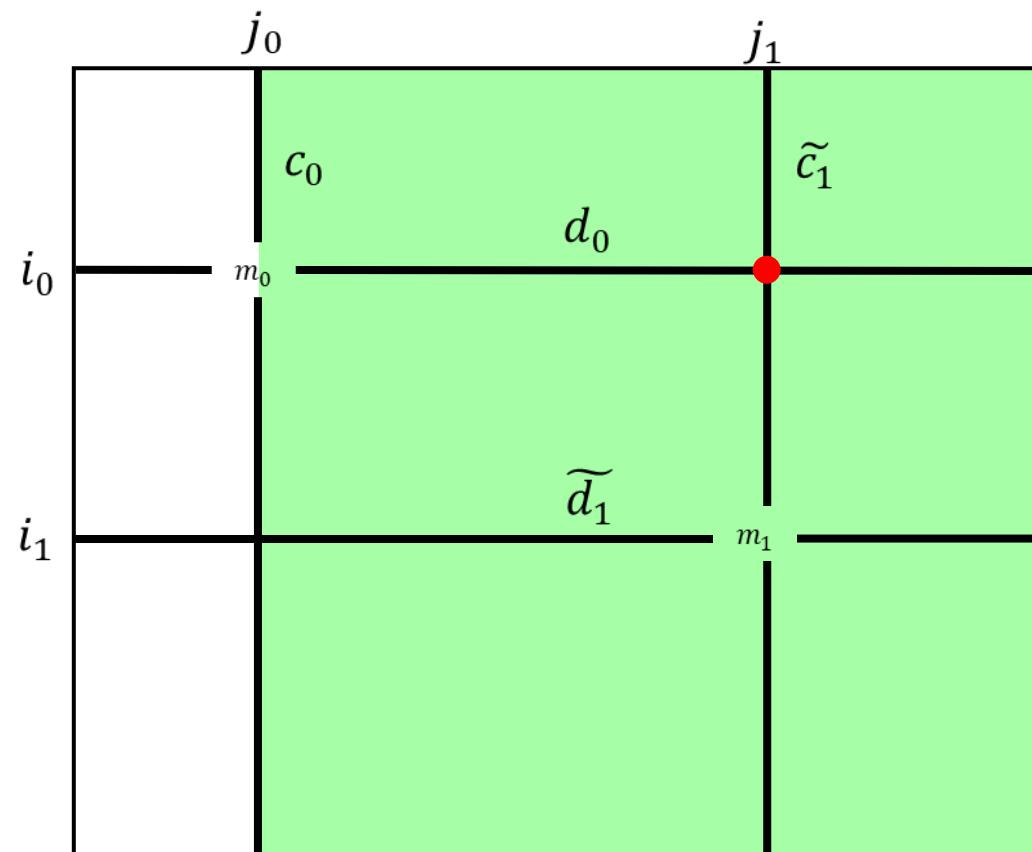
Admissible block B

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$$R_0 = \frac{1}{m_0} c_0 d_0^T \longrightarrow \text{rank 1}$$

$$R_1 = \frac{1}{m_1} c_1 d_1^T \longrightarrow \text{rank 1} \quad (c_1 := \tilde{c}_1 - d_0[j_1]c_0) \\ (d_1 := \tilde{d}_1 - c_0[i_1]d_0)$$

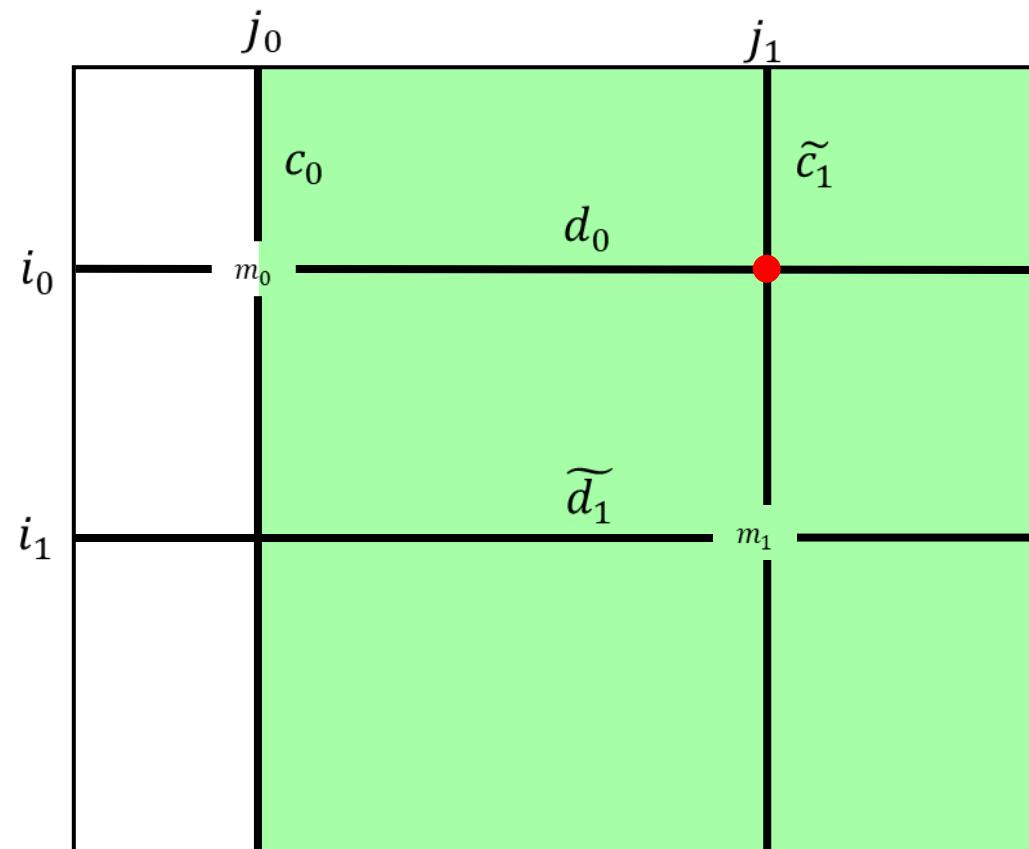


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and so on...

$$R_r = \frac{1}{m_r} c_r d_r^T \longrightarrow \text{rank 1} \quad (c_r := \tilde{c}_r - d_{r-1}[j_r]c_{r-1}) \\ (d_r := \tilde{d}_r - c_{r-1}[i_r]d_{r-1})$$



$$R_0 = \frac{1}{m_0} c_0 d_0^T \longrightarrow \text{rank 1}$$

$$R_1 = \frac{1}{m_1} c_1 d_1^T \longrightarrow \text{rank 1 } (c_1 := \tilde{c}_1 - d_0[j_1]c_0) \\ (d_1 := \tilde{d}_1 - c_0[i_1]d_0)$$

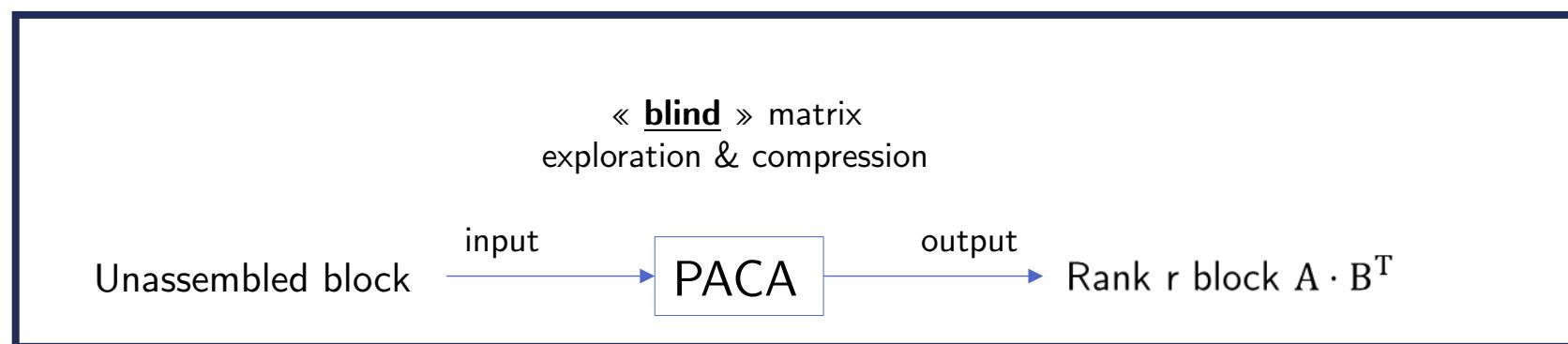
and so on...

$$R_r = \frac{1}{m_r} c_r d_r^T \longrightarrow \text{rank 1 } (c_r := \tilde{c}_r - d_{r-1}[j_r]c_{r-1}) \\ (d_r := \tilde{d}_r - c_{r-1}[i_r]d_{r-1})$$

Error estimate

$$\epsilon_{PACA,k} = \frac{\|c_k\|_F \cdot \|d_k\|_F}{\|\mathbf{B}_k\|_F}$$

$$B \approx B_r = \sum_{k=1}^r R_k \longrightarrow \text{rank r}$$



GMRES

Matrix-vector multiplication algorithm

Doing that for each sub-block



III. Integration to lifespan analysis

- i. Post-processing BEM results : computing the stress intensity factor
- ii. Integration to the global pipeline at Safran for lifespan assessment

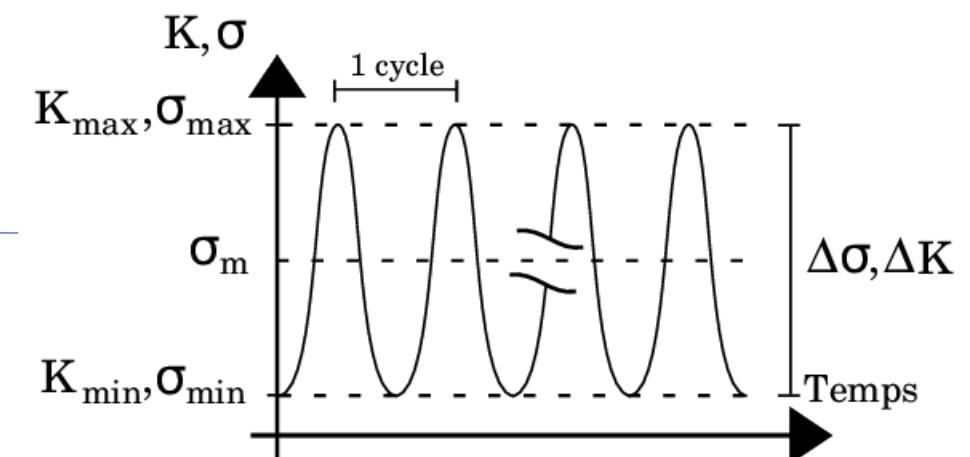
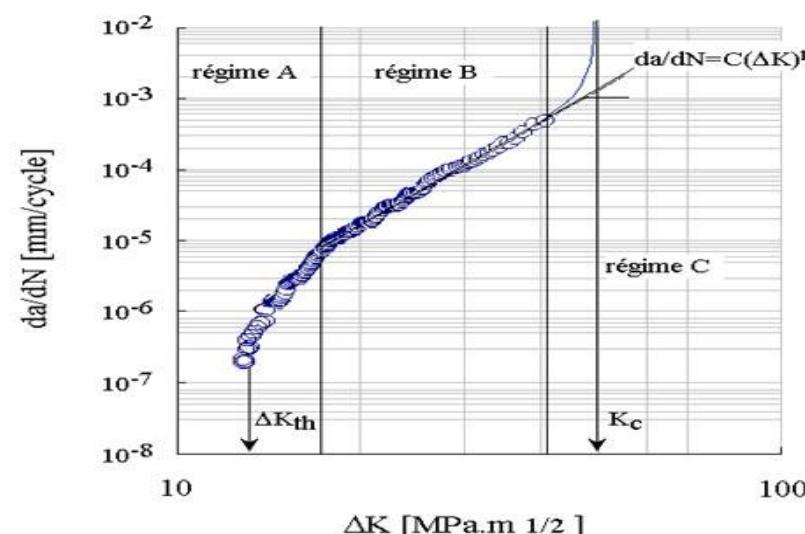
What does it mean to « solve » a crack problem at Safran ?

Irwin criterion (1957)

The piece breaks $\Leftrightarrow K > K_c$

Paris' law – fatigue approach (1963)

$$\frac{da}{dN} = C(\Delta K)^m$$



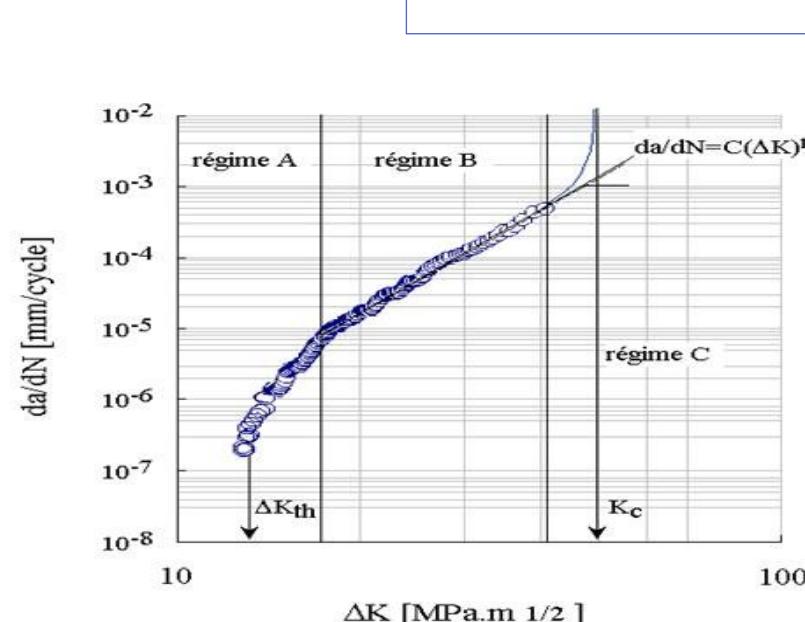
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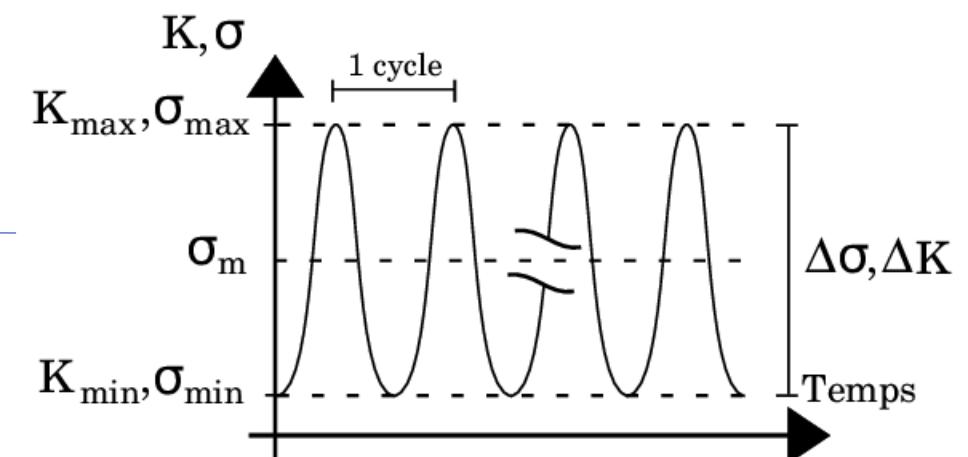
The piece breaks $\Leftrightarrow K > K_c$

Paris' law – fatigue approach (1963)

$$\frac{da}{dN} = C(\Delta K)^m$$



$$N_{max} = \dots$$



Direct kinematic extrapolation**Theoretical SIF**

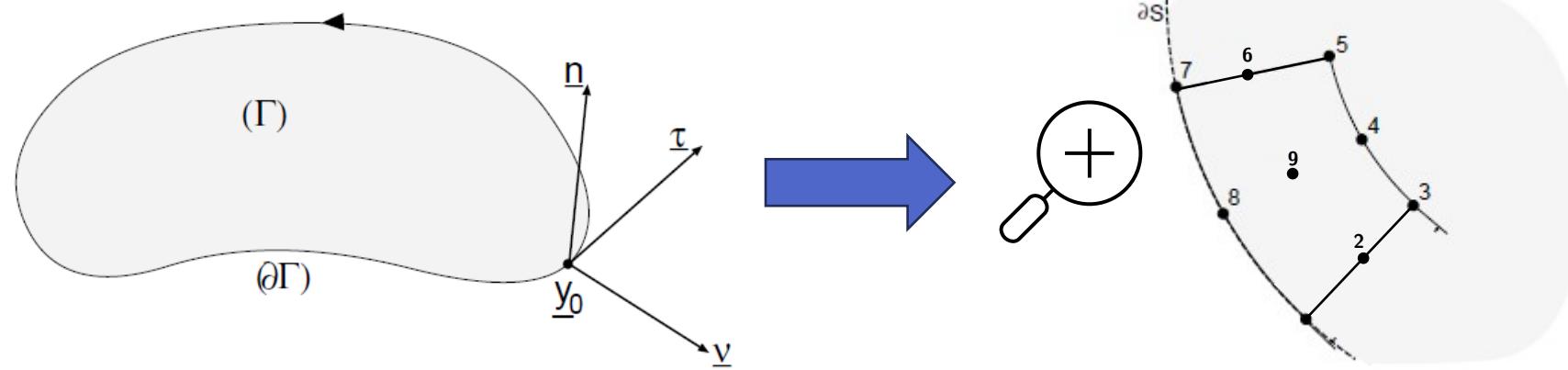
$$K = \lim_{d \rightarrow 0} \frac{\mu}{4(1-\nu)} \sqrt{\frac{2\pi}{d}} \phi_n$$

Numerical SIF

$$K = \frac{\mu}{4(1-\nu)} \sqrt{\frac{2\pi}{d_{6,9,2}}} \phi_n^{6,9,2}$$

$$u(d) = K\sqrt{d}f(\theta) + O(d)$$

Williams' asymptotic expansion



Quarter-node éléments (a.k.a Barsoum elements)

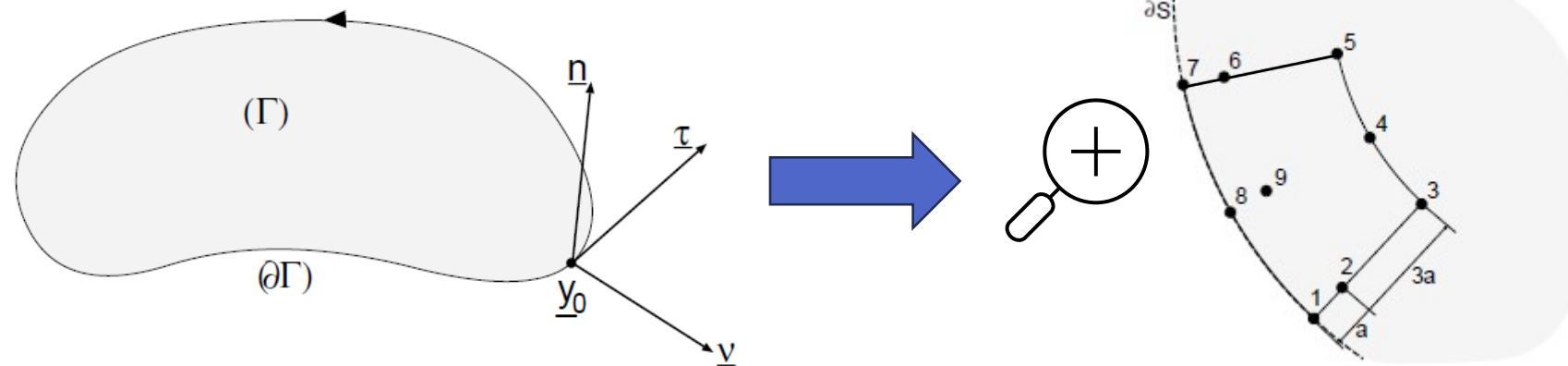
Theoretical SIF

$$K = \lim_{d \rightarrow 0} \frac{\mu}{4(1-\nu)} \sqrt{\frac{2\pi}{d}} \phi_n$$

$$\phi(y) = \sqrt{\frac{d}{a}} \left(2\phi^2 - \frac{1}{2}\phi^3 + \sqrt{\frac{d}{a}} \left(\frac{1}{2}\phi^3 - \phi^2 \right) \right)$$

Numerical SIF

$$K^1 = \frac{\mu}{4(1-\nu)} \sqrt{2\pi} a (2\phi^2 - \phi^3)$$



Weighting function

Change of variable : seeking *a priori* the COD ϕ as

$$\phi = w \cdot \psi$$

- w must be **asymptotically as the square root** of the crack front distance:

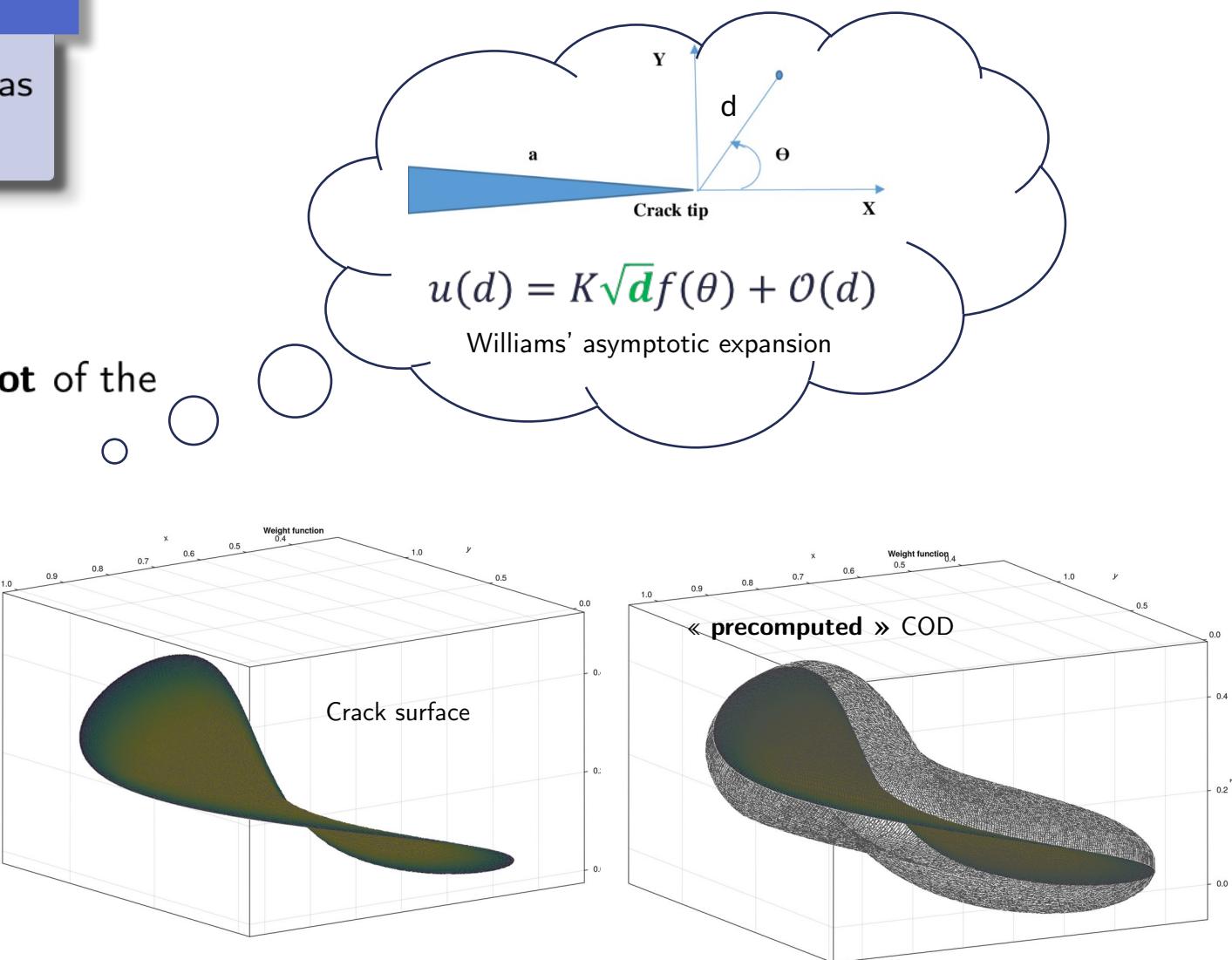
$$w \underset{y \rightarrow \text{crack front}}{\sim} \sqrt{d}$$

Consequences:

- New **weighted kernels**

$$K_w = w \cdot K$$

- Better SIF approximation**



Weighting function

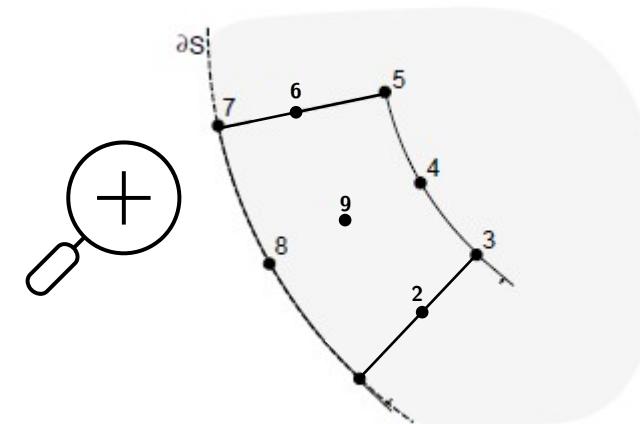
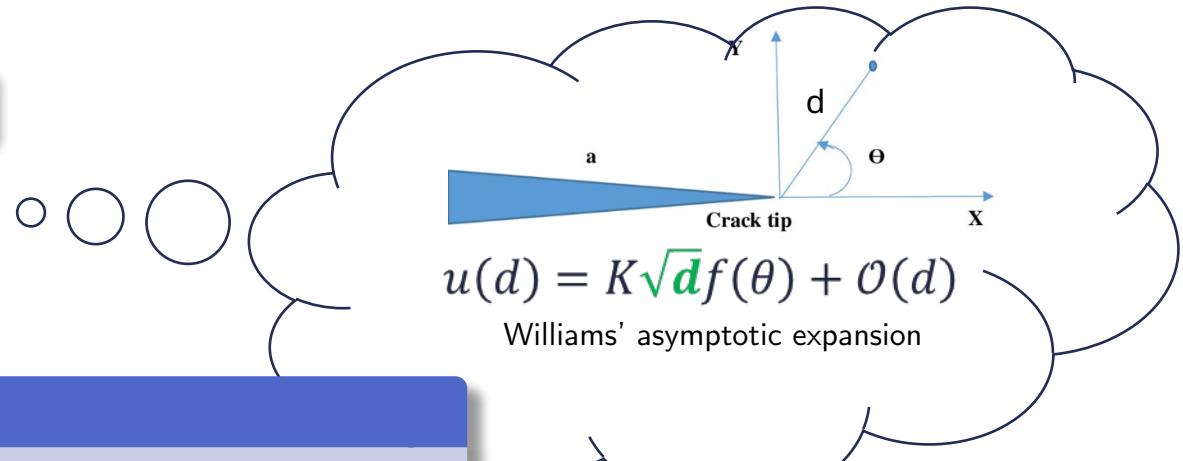
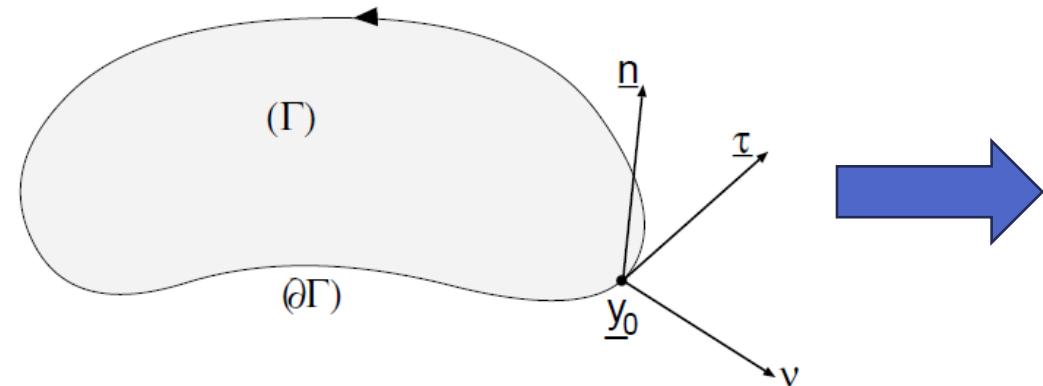
Change of variable : seeking *a priori* the COD ϕ as
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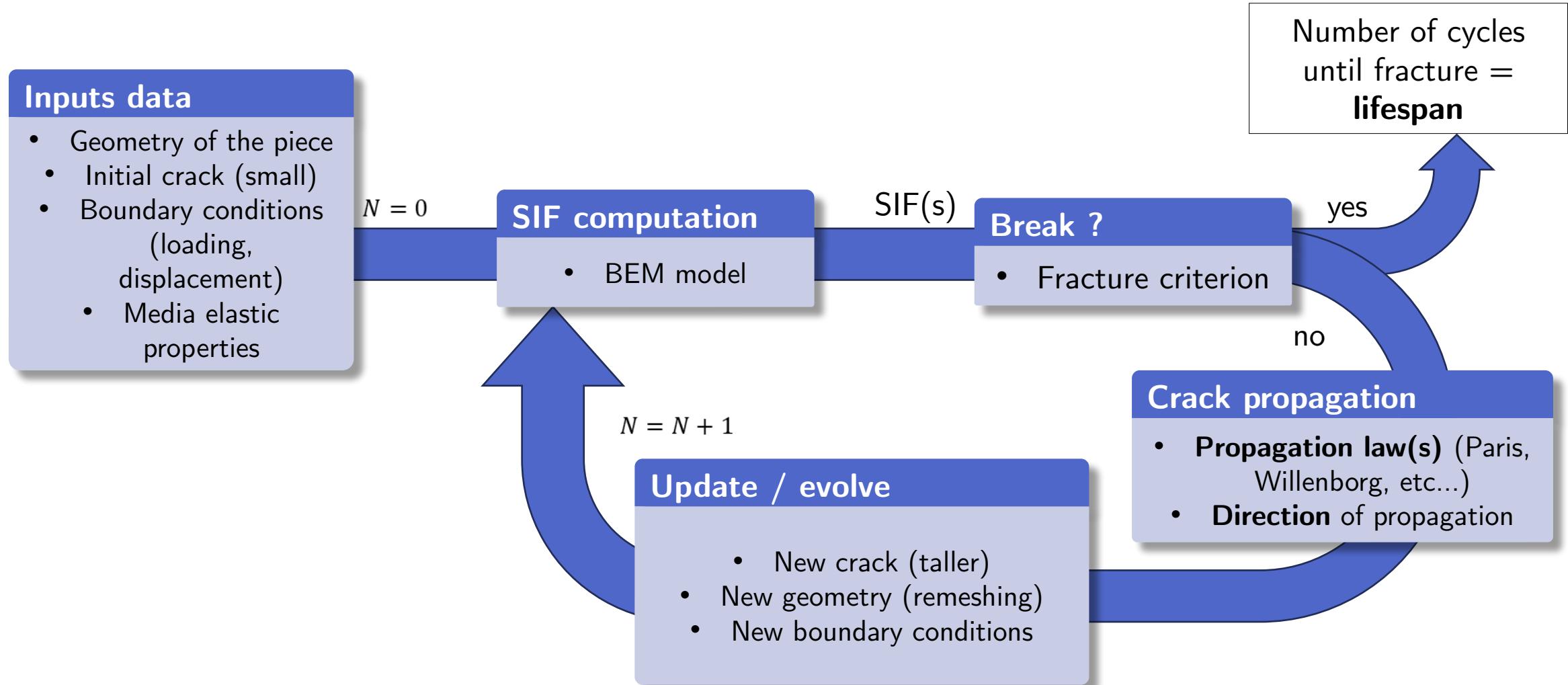
Theoretical SIF

$$K = \lim_{d \rightarrow 0} \frac{\mu}{4(1-\nu)} \sqrt{\frac{2\pi}{d}} \phi_n$$

Numerical SIF

$$K^{6,9,2} = \frac{\mu}{4(1-\nu)} \sqrt{2\pi} \psi_n^{6,9,2}, \quad \phi = w \cdot \psi$$

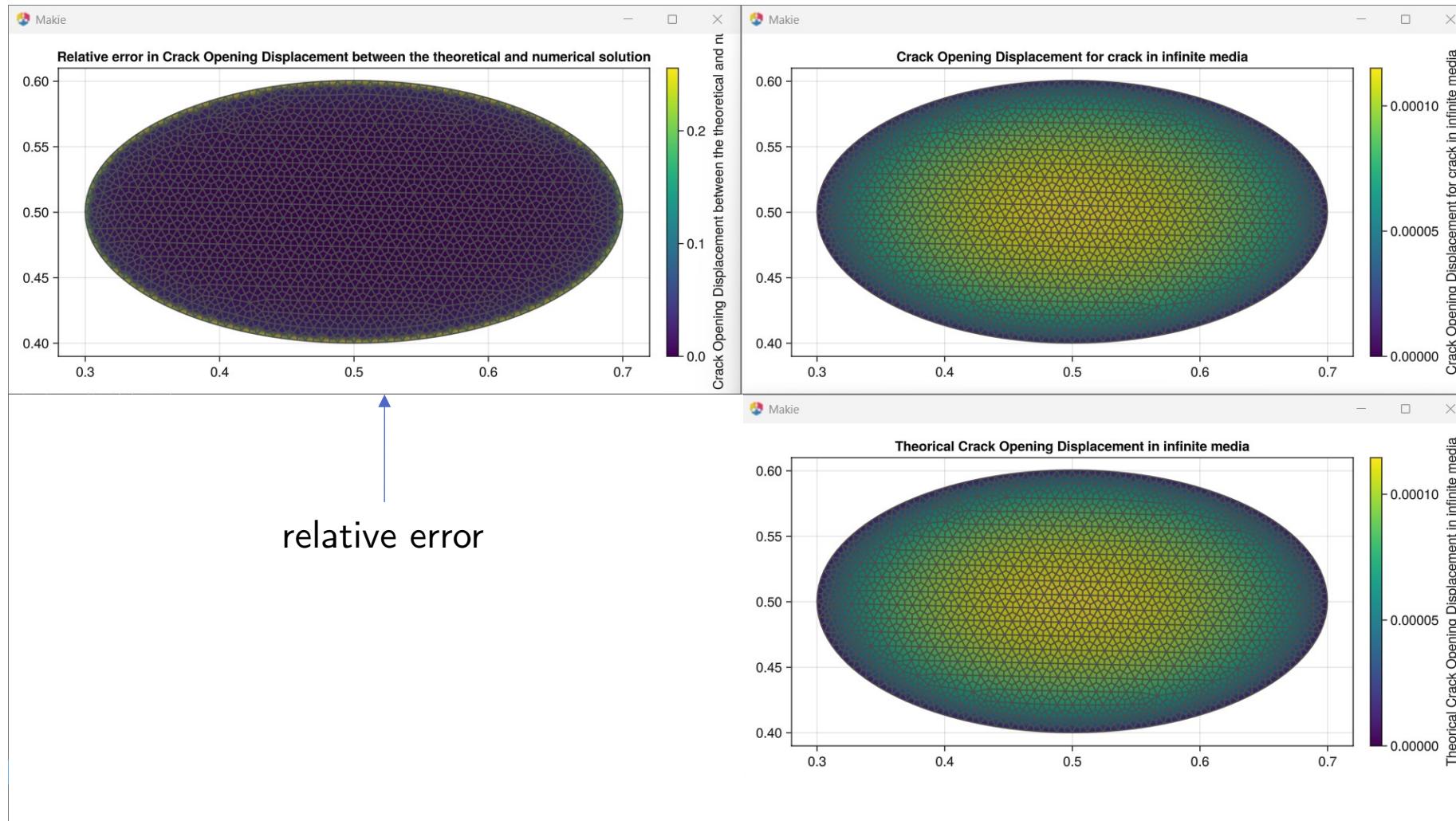






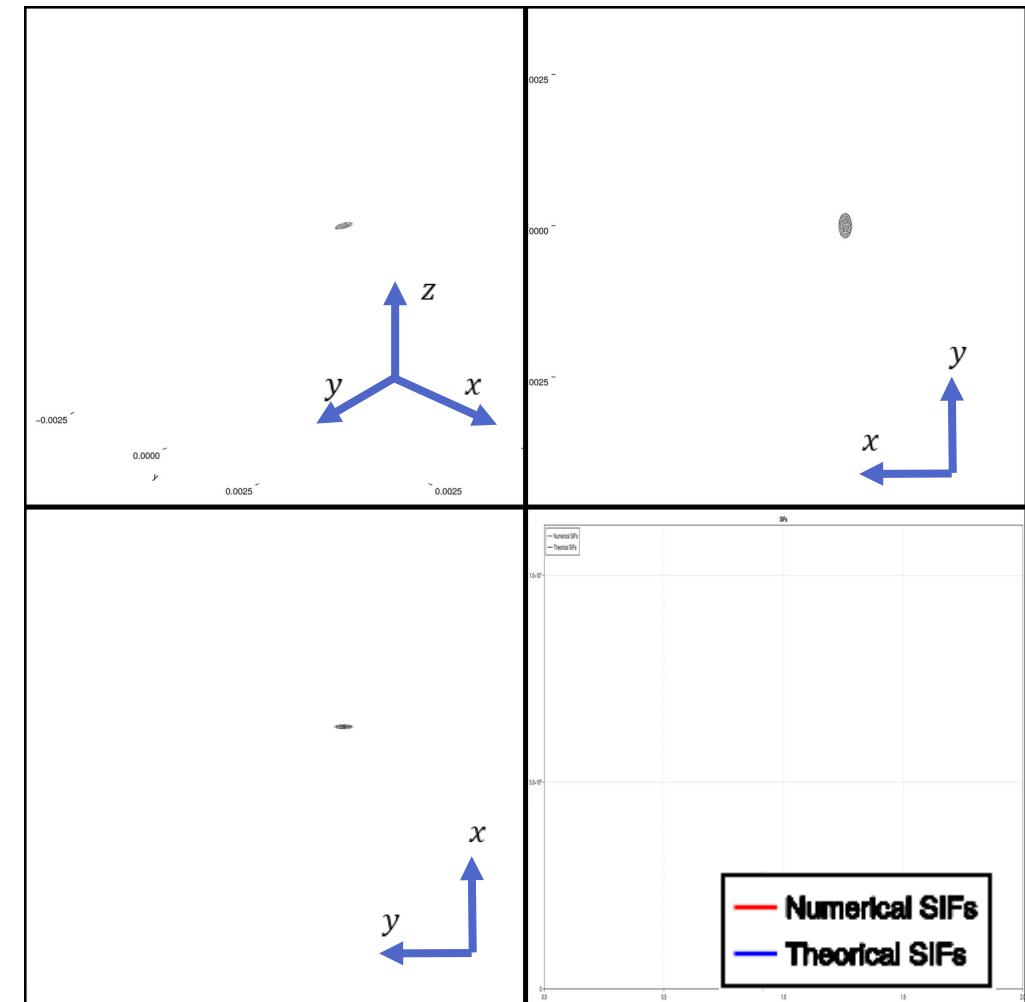
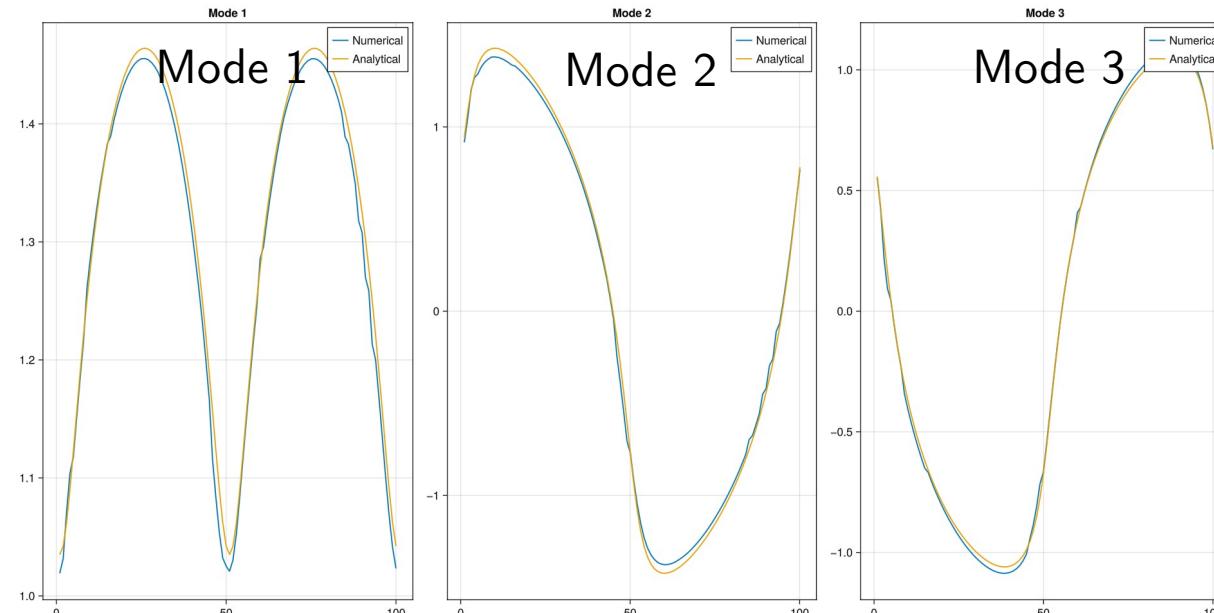
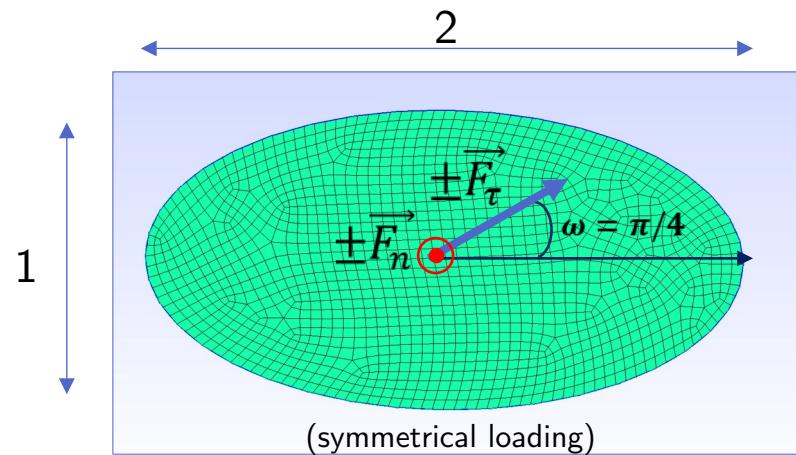
IV. Numerical examples / validation

- i. Elliptic crack in infinite media
- ii. Elliptic crack in cube under mixed boundary conditions
- iii. Quick presentation of the Julia library for solving 3D crack problems : CrackFastBEM



IV. Numerical examples

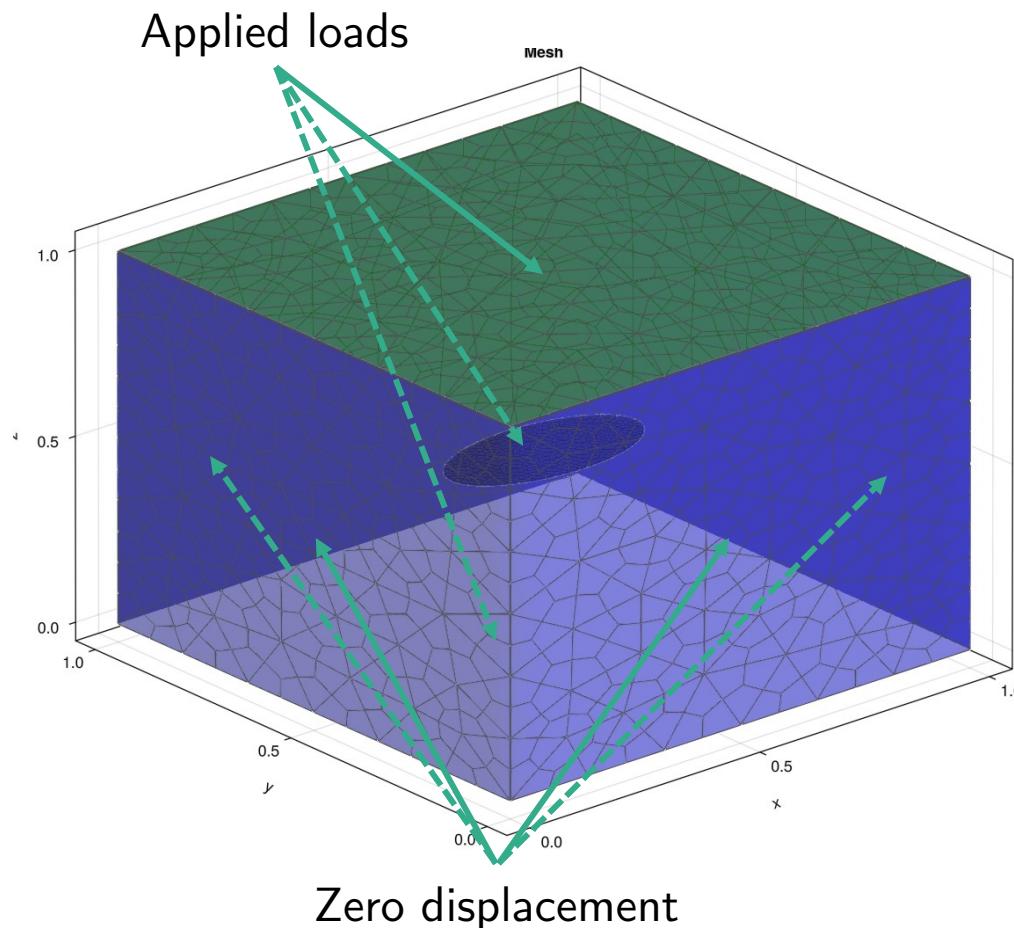
1. Elliptic crack in infinite media



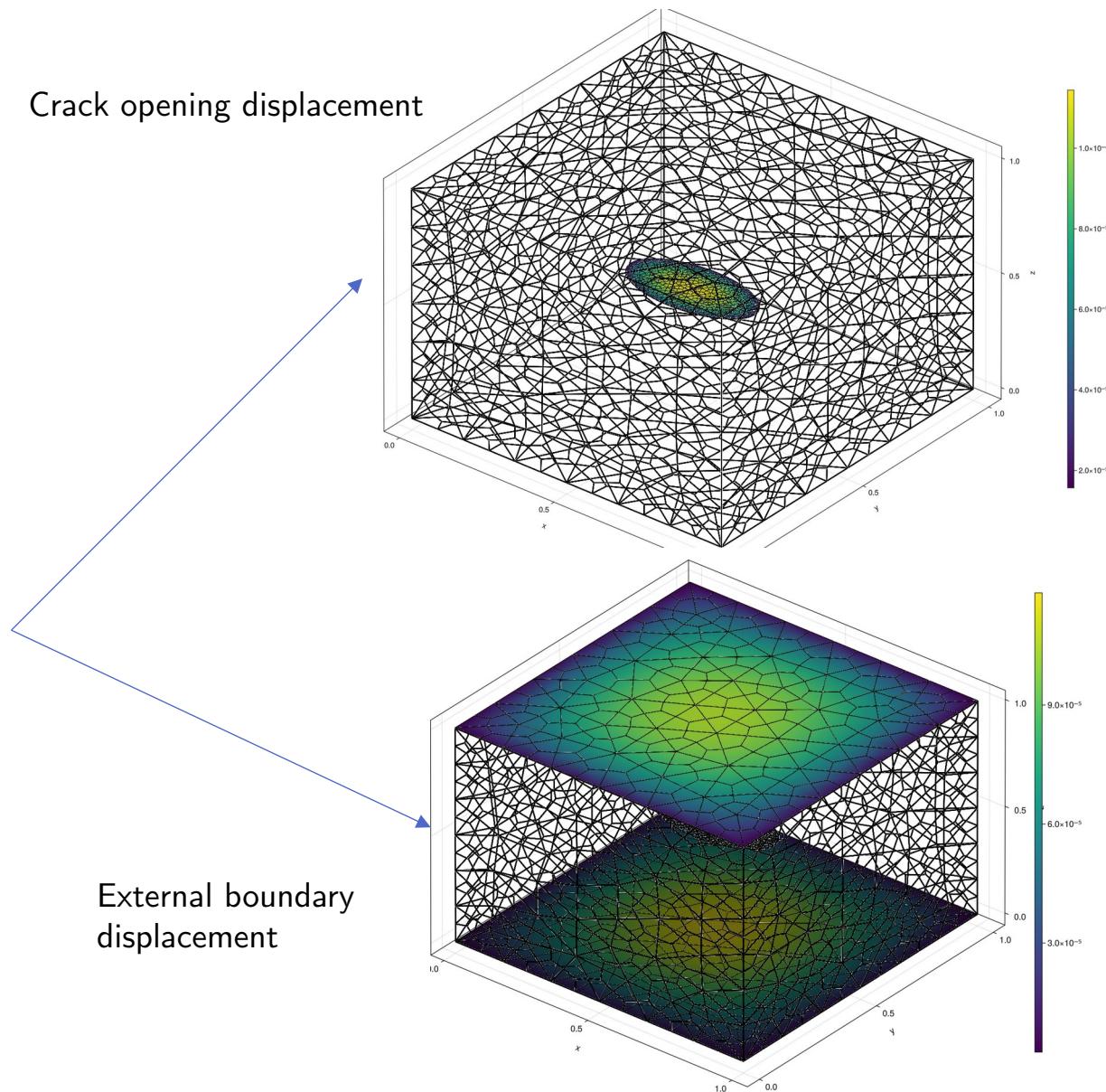
Last iteration = fatigue lifespan

IV. Numerical examples

2. Elliptic crack in cube under mixed boundary conditions

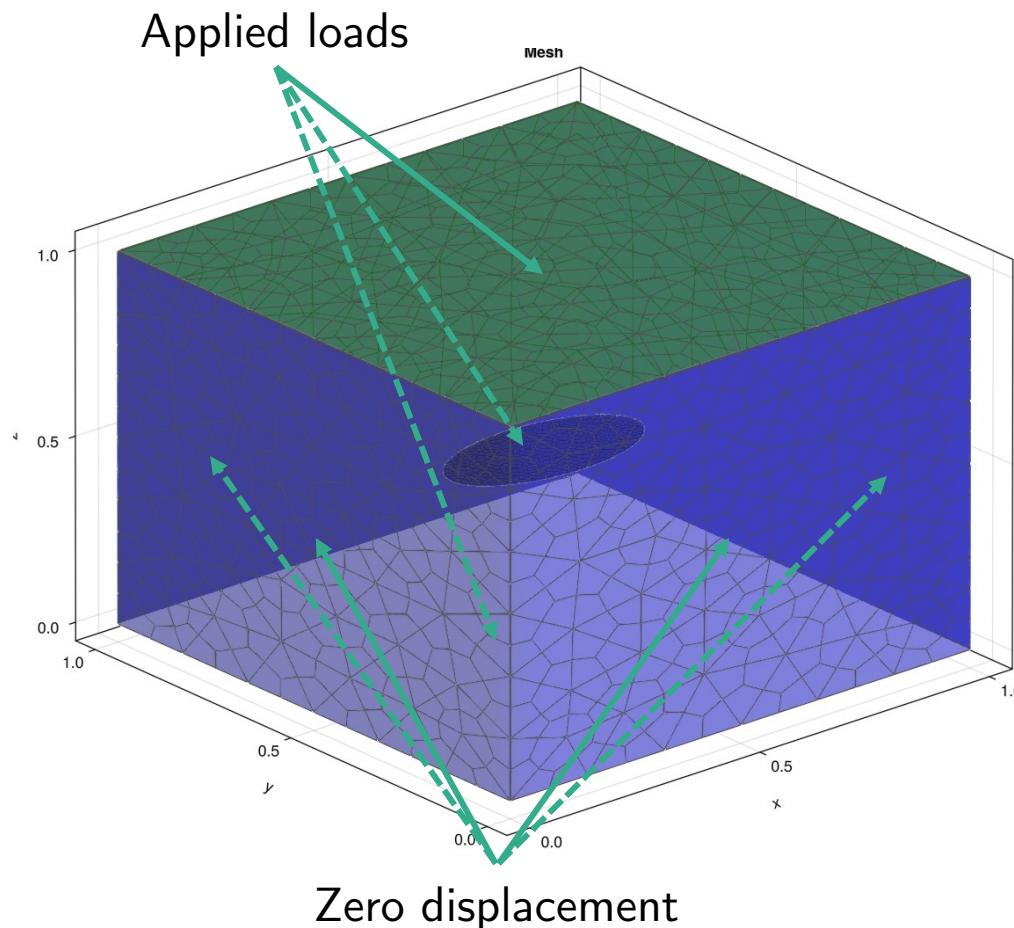


Crack opening displacement

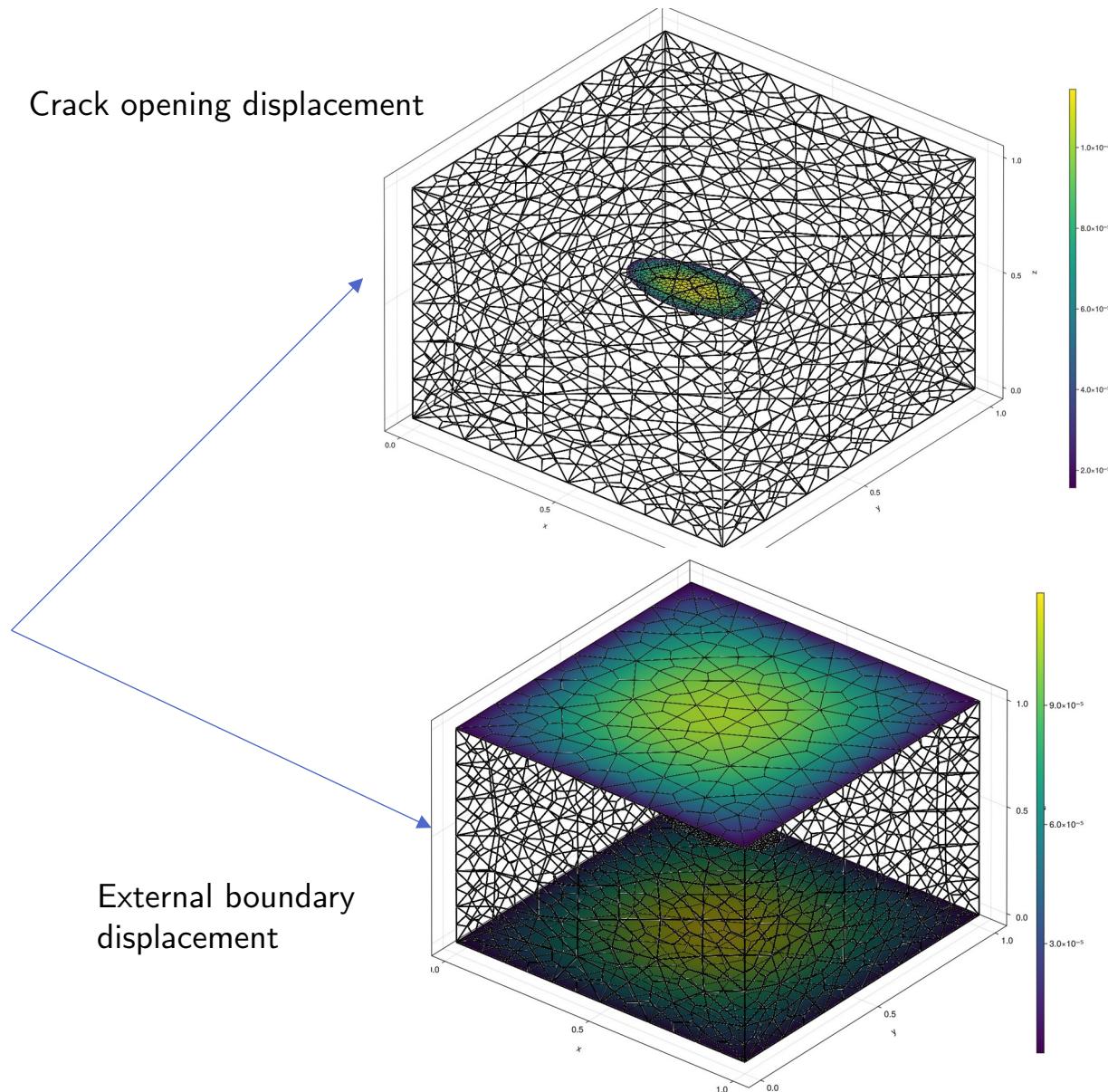


IV. Numerical examples

2. Elliptic crack in cube under mixed boundary conditions



Crack opening displacement



Library under development in **julia**

- Julia package for fatigue lifespan estimation for a 3D general crack configuration :
CrackFastBEM.jl

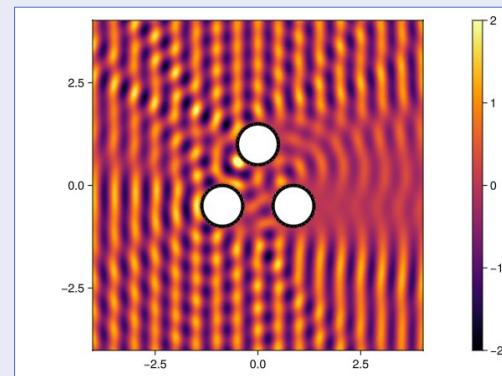
Features : under development, but the goals are to deal with :

- 3D crack configurations herited from a BEM mesh
- Mixed or simple boundary conditions (Dirichlet + Neumann)
 - Crack in infinite solid
 - Crack in finite solid
 - Surface breaking cracks
- SIF computation : 1) naive approach, 2) with weighting function
- Compatible with the Bueckner's superposition principle
- « user-friendly » API

Inti.jl

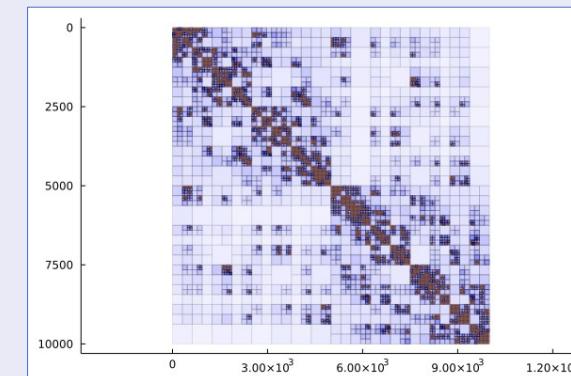


Julia **library** for solving boundary and volume integral equations using Nyström discretization method



HMatrices.jl

Julia **library** for assembling hierarchical matrices.



Forthcoming goals...

- FEM / BEM coupling
- Industrialization with Safran : FEM model → Coupling with BEM → Bueckner superposition
- Thermal gradient : thermal dilatation term
- T & Tz stress

Thank you for your attention