

$$) = \int_{\Gamma} G(x, y) \lambda(y) d\Gamma(y)$$

Ingénierie Mathématique & Calcul Scientifique



# Hybrid H-matrix — FMM solvers for wave equations

Toufic Abboud

$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln\left(\frac{1-\alpha}{\alpha}\right) \log(c) + o(\log(c))$$

**Conference in honor of Abderrahmane Bendali**

**Pau, 12-14 December 2017**

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$



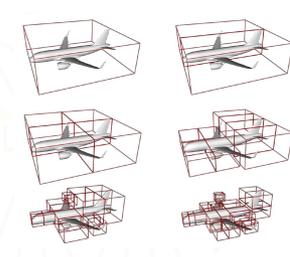
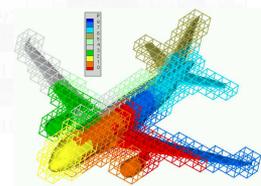
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$$\phi(x) = \int_{\Gamma} G(x, y) \lambda(y) d\Gamma(y)$$

# BEM

- Boundary Element Method, also known as Method of Moment (MoM) is the method of choice for many Electromagnetic problems
- Cost: assembly in  $O(N^2)$ , factorisation/solve in  $O(N^3)$  complexity. In practice  $N = O(f^2)$
- Fast solvers & HPC are mandatory
  - FMM algorithm & iterative solvers (Greengard and Rokhlin, 90's):
    - \* matrix-by-vector product in  $O(N \log N)$  operations
    - \* issues: low frequency or locally refined meshes, ill-conditioned problems, big number of right hand sides (RHS)...
  - H-matrix direct solvers (Hackbusch 00's):
    - \* overall complexity  $O(N (\log N)^2)$  complexity in the case of asymptotically smooth kernel (like Laplace's kernel)
    - \* absence of solid theoretical bases for the oscillating kernel until recently



$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2}$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$

$$) = \int_{\Gamma} G(x, y) \lambda(y) d\Gamma(y)$$

Supported by a grant from **DGA**, the French Government Defence Procurement and Technology Agency, in the framework of **RAPID** program (dual innovation support scheme), under project No. 132906163.

**Aims:** a generic software library implementing state-of-the-art fast direct and iterative solvers for existing BEM-based codes

Leader: **IMACS**. Partners: **Airbus Group** and **Inria** (the French Institute for Research in Computer Science and Automation)



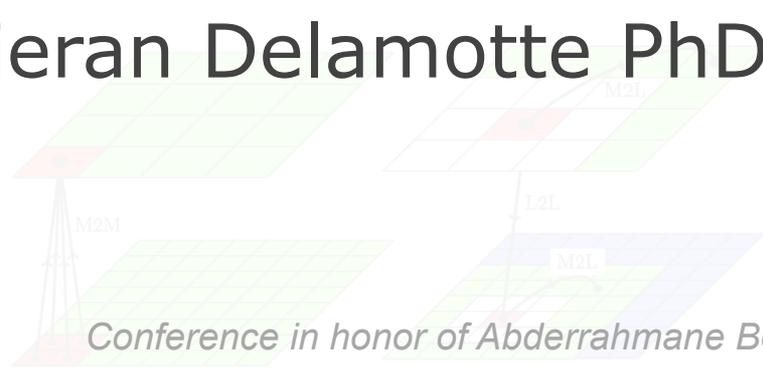
$$G(x, y) = \int_{\Gamma} G(x, y) \lambda(y) d\Gamma(y)$$



## Focus on the local approximate rank of the Helmholtz kernel

$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln \left( \frac{1-\alpha}{\alpha} \right) \log(c) + o(\log(c))$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y) \quad (\text{Kieran Delamotte PhD thesis})$$

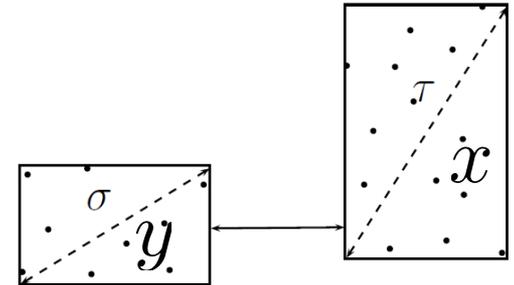




# Compression (analytic method)

$$M_{ij} = \int_{\Gamma} \int_{\Gamma} G(x, y) \Phi_j(y) \Phi_i(x) d\Gamma(y) d\Gamma(x)$$

$$G(x, y) = \frac{e^{ik|x-y|}}{4\pi|x-y|}$$



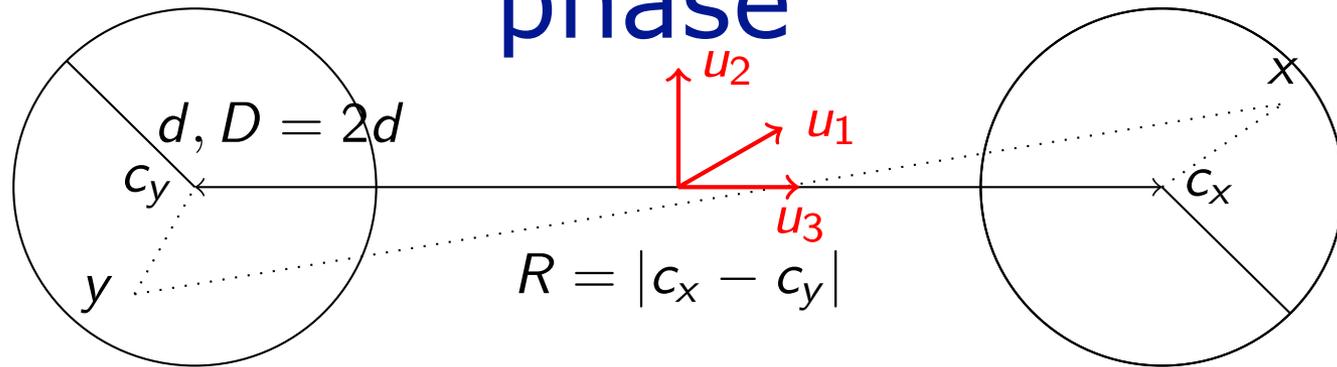
Low rank approximation of the Green's kernel

$$G(x, y) \approx \sum_{q=1}^r u_q(x) v_q(y)$$

→ Low rank approximation of M

Justification for non asymptotically smooth kernels?

## Taylor expansion of the phase



- $D$  is the diameter of the clusters,
- $R$  the distance between clusters

$$x = \xi_1 u_1 + \xi_2 u_2 + \xi_3 u_3,$$

$$y = \eta_1 u_1 + \eta_2 u_2 + \eta_3 u_3$$

In the  $(u_1, u_2, u_3)$  basis the kernel reads as

$$G(x, y) \approx \frac{e^{ikR}}{|x - y|} e^{ik(\xi_3 - \eta_3)} e^{ik \frac{(\xi_1 - \eta_1)^2}{2R}} e^{ik \frac{(\xi_2 - \eta_2)^2}{2R}} e^{i\Phi}$$

with  $\Phi = \mathcal{O}(k \frac{D^3}{R^2})$

# Admissibility condition & rank growth

- Classical static admissibility condition (Hackbusch *et al*)

$$R \geq \alpha D$$

- this works well for ‘low frequencies’, i.e. if combined with :

$$(kD) \leq \beta$$

- the phase term is bounded then the kernel has essentially the same rank as the Laplace one

*otherwise, we show that rank may grow like  $k^2$  in the high frequency regime. The result would lead to a poor asymptotic compression ratio*

# Admissibility condition & rank growth

- Fraunhofer admissibility condition (Engquist, Darve, ...)

$$R \geq \alpha \max((kD), 1) D$$

- 2nd order terms are bounded
- frequency-independant rank (plane waves);
- too restrictive : clusters with big radii, *i.e.*  $(kD) \gg 1$ , have to be far  $\Rightarrow$  small blocks with cst rank: *Divide but NOT Conquer* unless at very high frequencies.

$$N(c, \alpha) = \frac{2c}{\pi} + \dots$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$



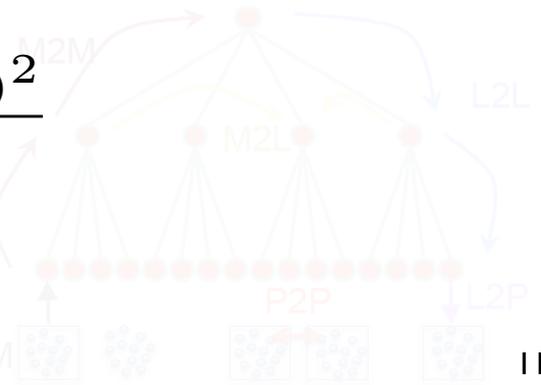
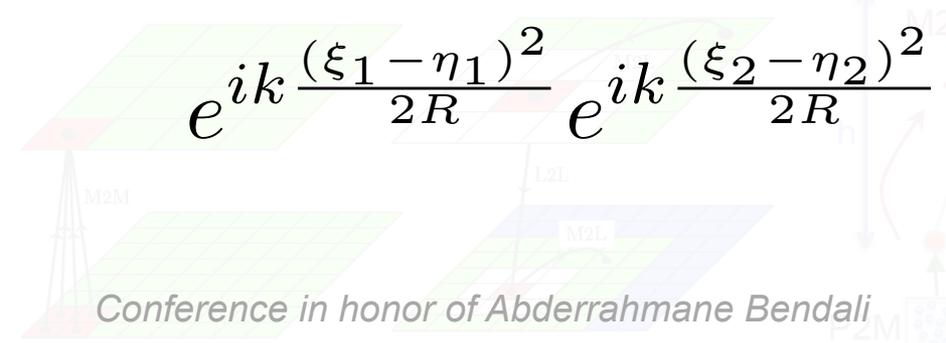
- Fresnel admissibility condition

$$R \geq \alpha \max((kD)^{\frac{1}{2}}, 1)D$$

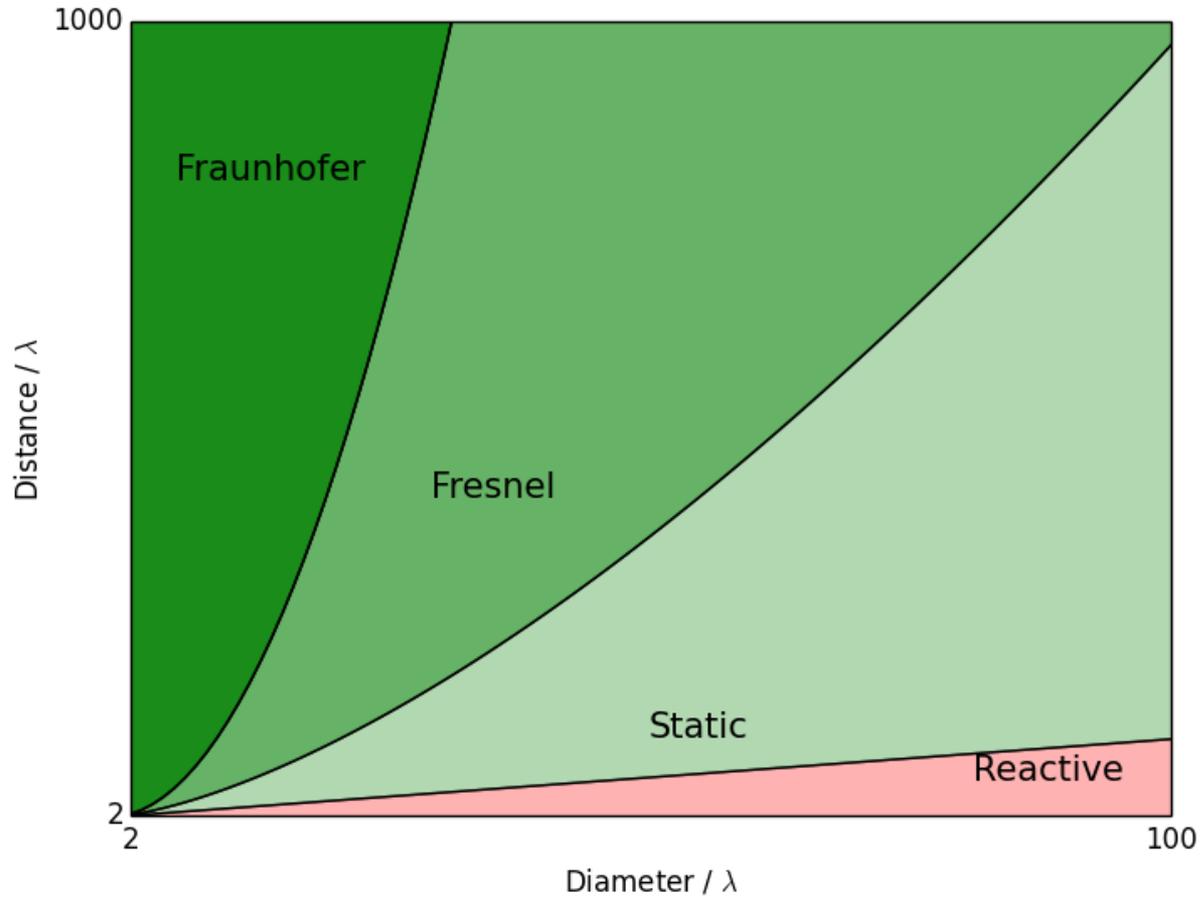
- *3rd order terms are bounded*
- *less restrictive than Fraunhofer condition : bigger blocks with small rank*
- *dominant phase terms are*

$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln\left(\frac{1-\alpha}{\alpha}\right) \log(c) + o(\log(c))$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x)b_{\alpha}(y)$$



# Zoning



# Fresnel or Fox-Li operator

- 2nd order phase terms : product of 1D operators

$$e^{ik \frac{(\xi_1 - \eta_1)^2}{2R}} = e^{ik \frac{\xi_1^2}{2R}} e^{ik \frac{\xi_1 \eta_1}{R}} e^{ik \frac{\eta_1^2}{2R}}$$

- we study the operator

$$F_c : \lambda \mapsto [F_c \lambda](x) = \int_{-1}^1 e^{icxt} \lambda(t) dt \quad x \in [-1, 1]$$

- operator studied in laser context by Fox and Li, in signal theory at Bell Labs by Slepian, Pollak, Landau, connected to results by Szegő and Widom

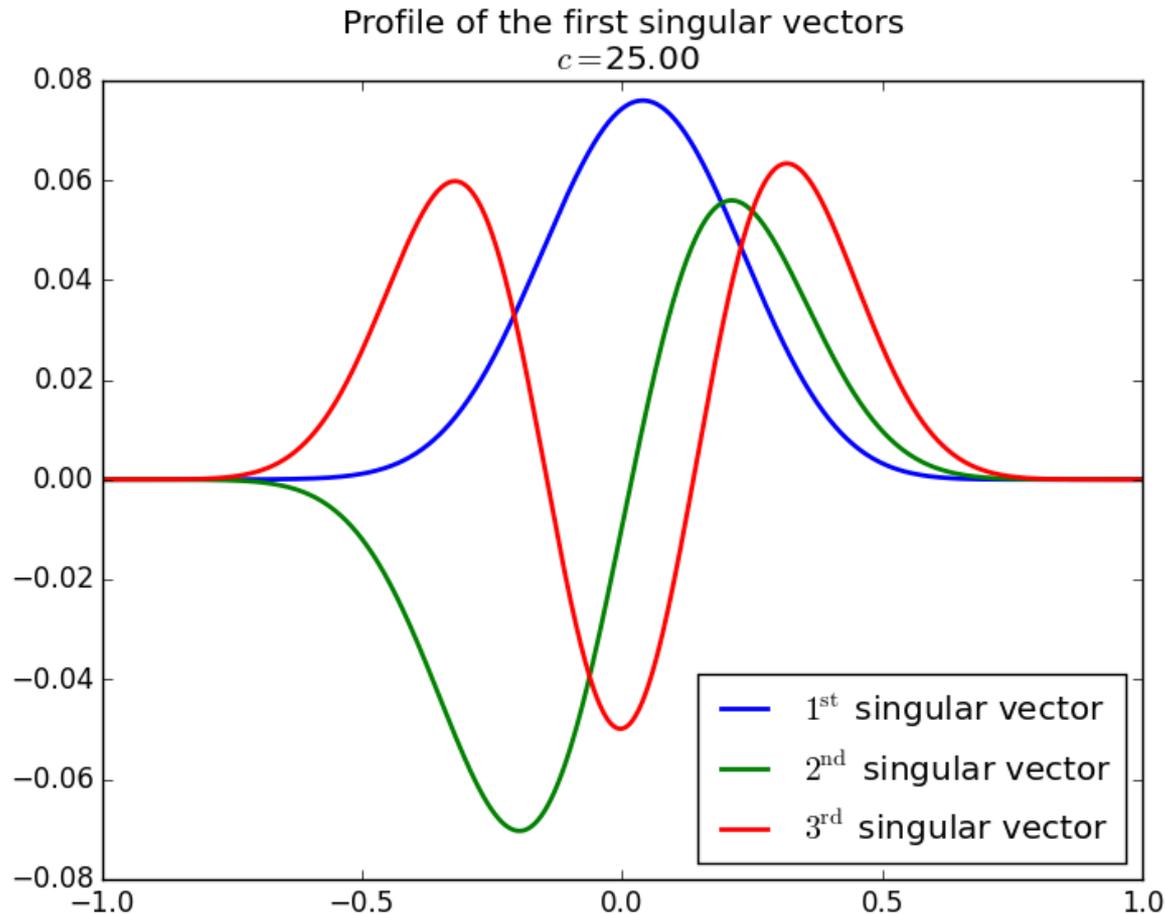
*Singular values of  $F_c$  are the eigenvalues of  $Q_c$ :*

$$Q_c = \frac{c}{2\pi} F_c^* F_c \quad Q_c \phi(x) = \frac{1}{\pi} \int_{-1}^1 \frac{\sin(c(x-t))}{x-t} \phi(t) dt$$

*Eigenfunctions are the Prolate Spheroidal Wave Functions of order zero*

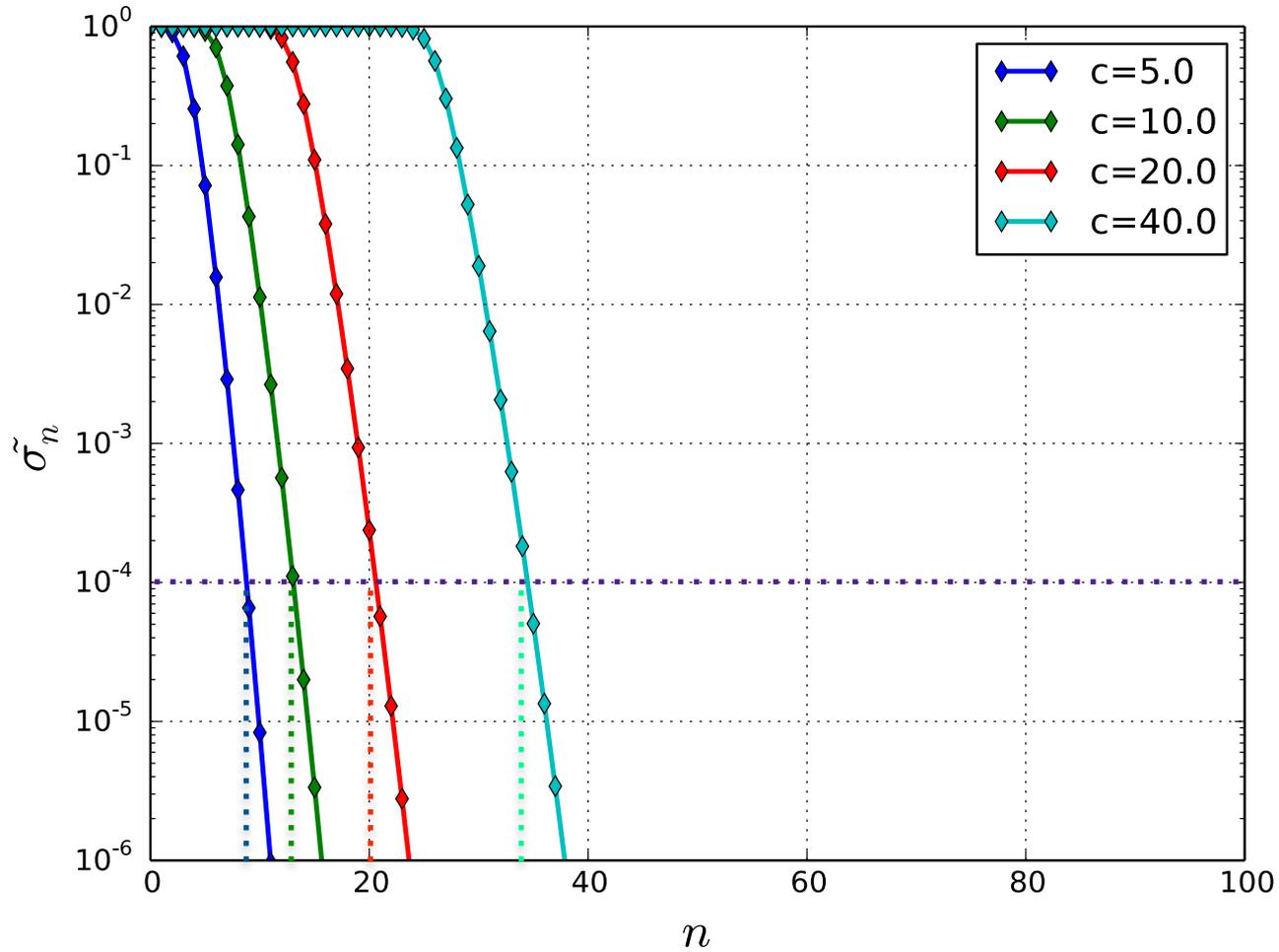
$$(1 - x^2)\psi_n''(x) - 2x\psi_n'(x) + (\chi_n - c^2x^2)\psi_n(x) = 0$$

# Singular vectors of $F_c$

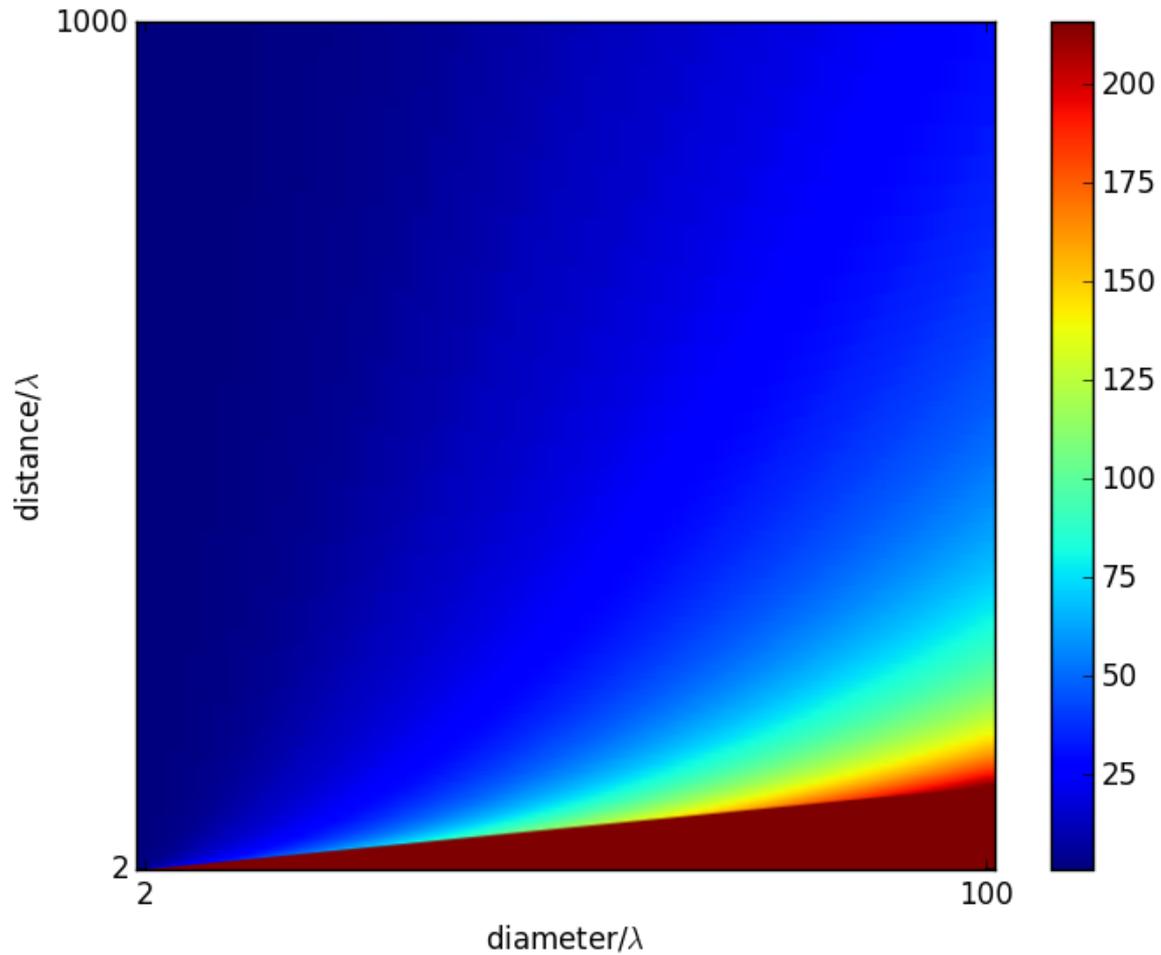


Prolate spheroidal wave functions of order zero

# Singular values & $\epsilon$ -rank of $F_c$



# $\epsilon$ -rank of $F_c$



*Theorem[Landau-Widom ('80)]*

*[Conjectured by Slepian ('65)]*

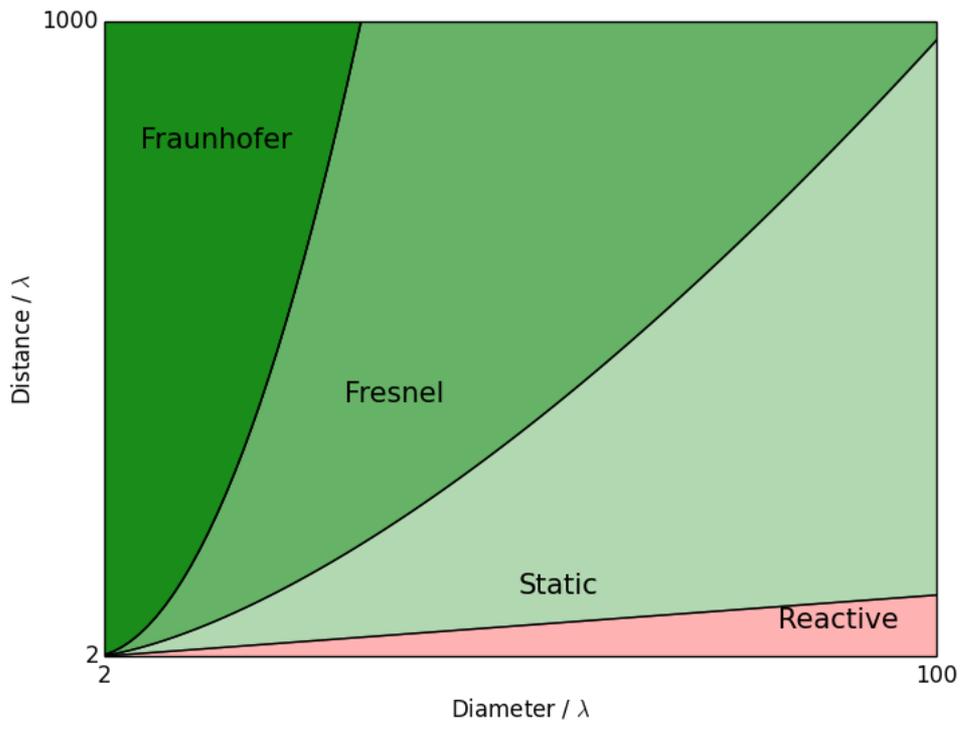
$$N(c, \alpha) = \#\{n \geq 1 : \mu_n > \alpha\}$$
$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \log\left(\frac{1-\alpha}{\alpha}\right) \log(c) + o(\log(c))$$

That provides the  $\varepsilon$ -rank of the Fresnel kernel:

$$\varepsilon - \text{rank}(F_c) \simeq \frac{2c}{\pi} + \frac{2}{\pi^2} |\log(\varepsilon)| \log(c)$$

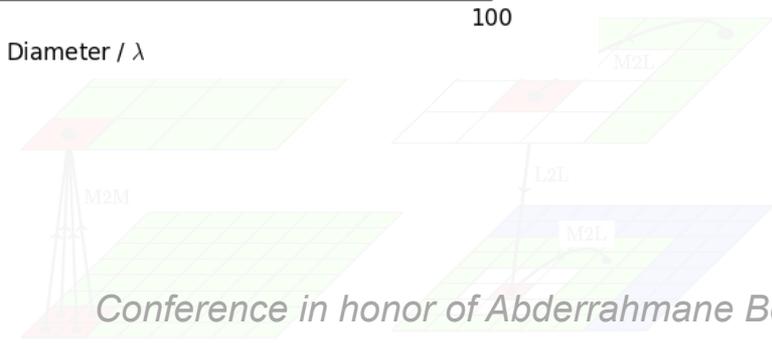
$$G(x, y) = \int_{\Gamma} \lambda(y) d\Gamma(y)$$

# Admissibility - rank



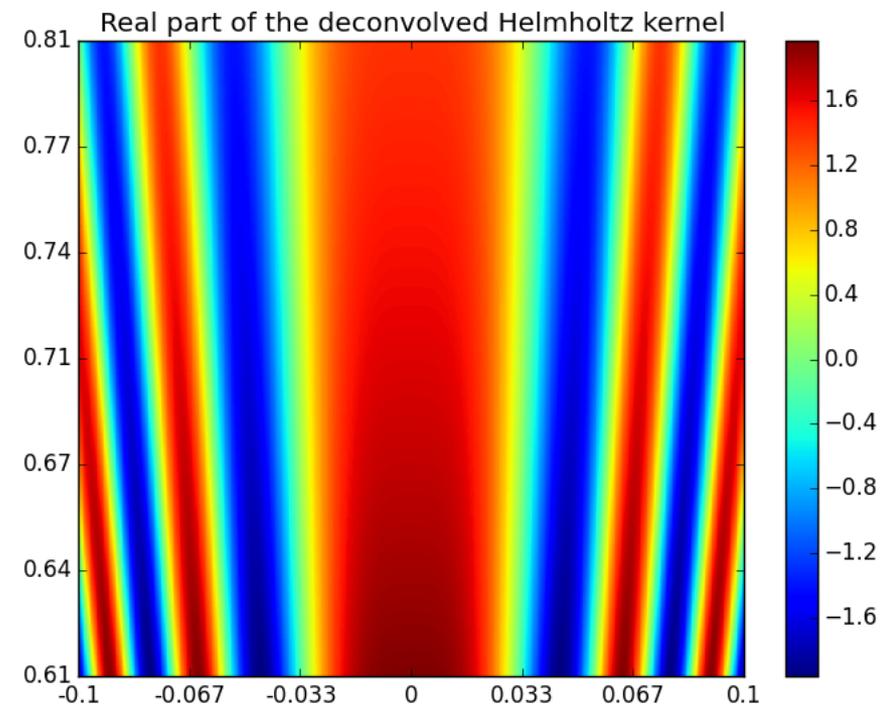
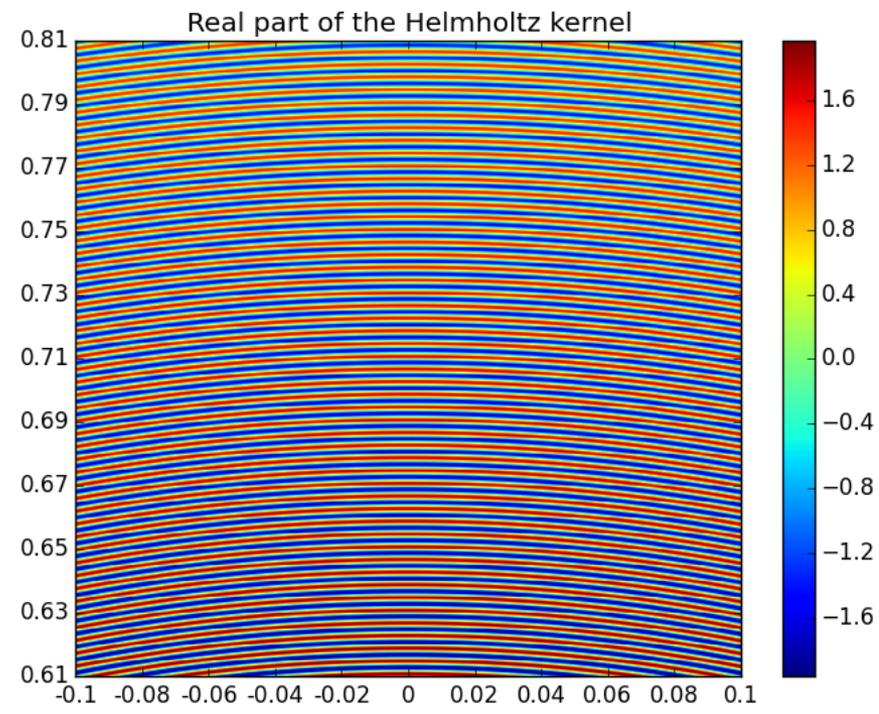
$$c = \frac{(kd)^2}{kR}$$

Hackbusch	$R \sim d$	$c \sim kd$
Fresnel	$R \sim k^{\frac{1}{2}} d^{\frac{3}{2}}$	$c \sim (kd)^{\frac{1}{2}}$
Fraunhofer	$R \sim kd^2$	$c \sim 1$



# HCA2 method

- This provides an alternative to ACA algorithm
- HCA2: interpolation based technique (functional equivalent of ACA)
- Fresnel admissibility condition
- *Oriented boxes*
- *Number of interpolation points* chosen according to Landau-Widom theorem in  $(u_1, u_2)$  directions and like  $O(\log \epsilon)$  along  $u_3$
- block rank may grow like  $k^{1/2}$  or  $k$  depending on the cross section

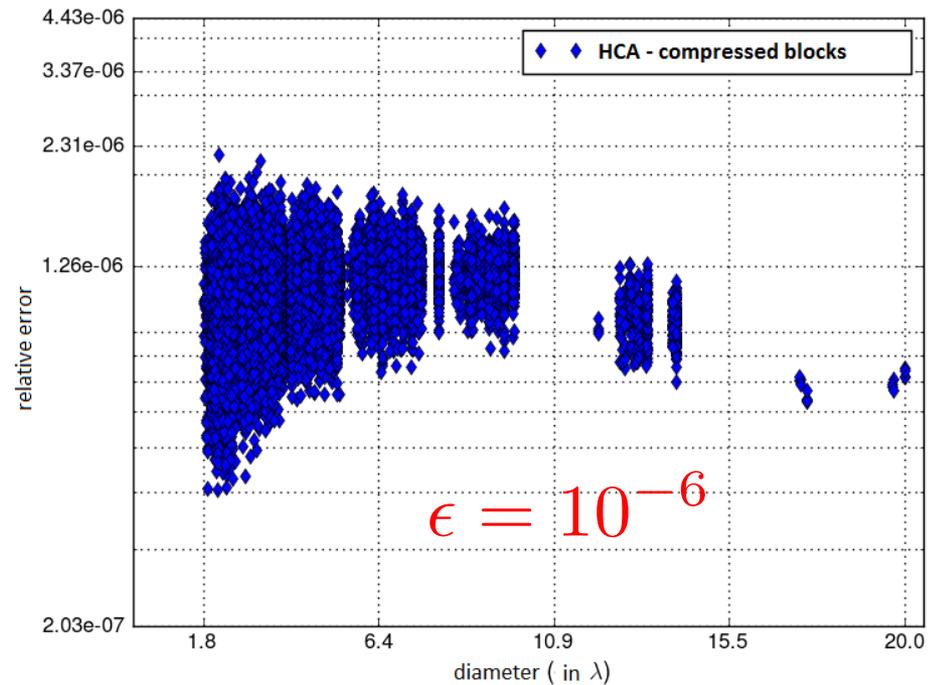
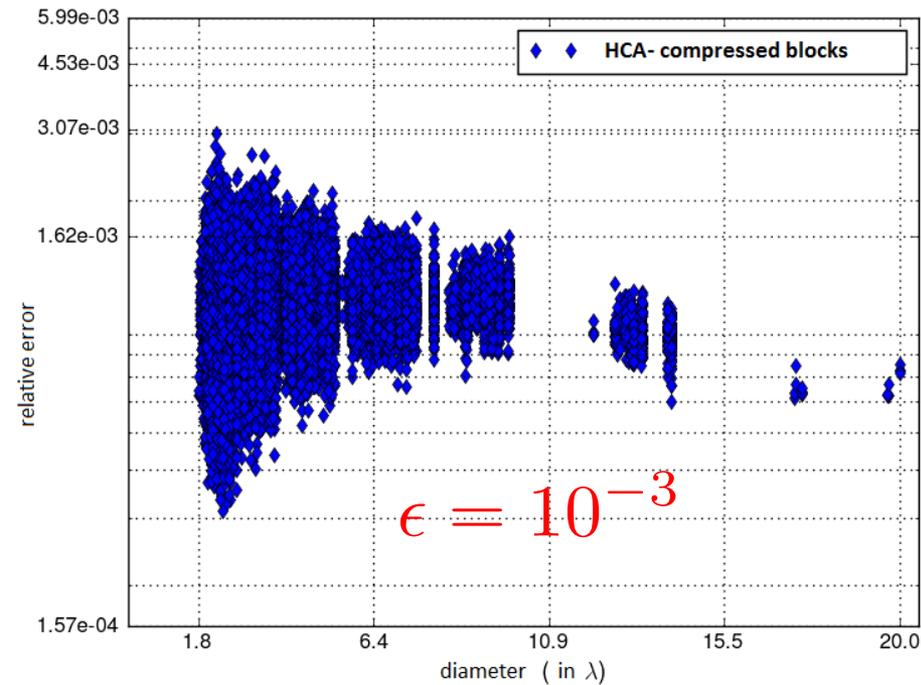


$$G(x, y) \approx \sum_{\alpha=1}^r \frac{a_{\alpha}(x)b_{\alpha}(y)}{4\pi|x-y|} = \frac{e^{ik|x-y|}}{4\pi|x-y|}$$

$$G_{\hat{u}_3}(h) = \frac{e^{ik(|x-y| - \hat{u}_3 \cdot (x-y))}}{4\pi|x-y|}$$

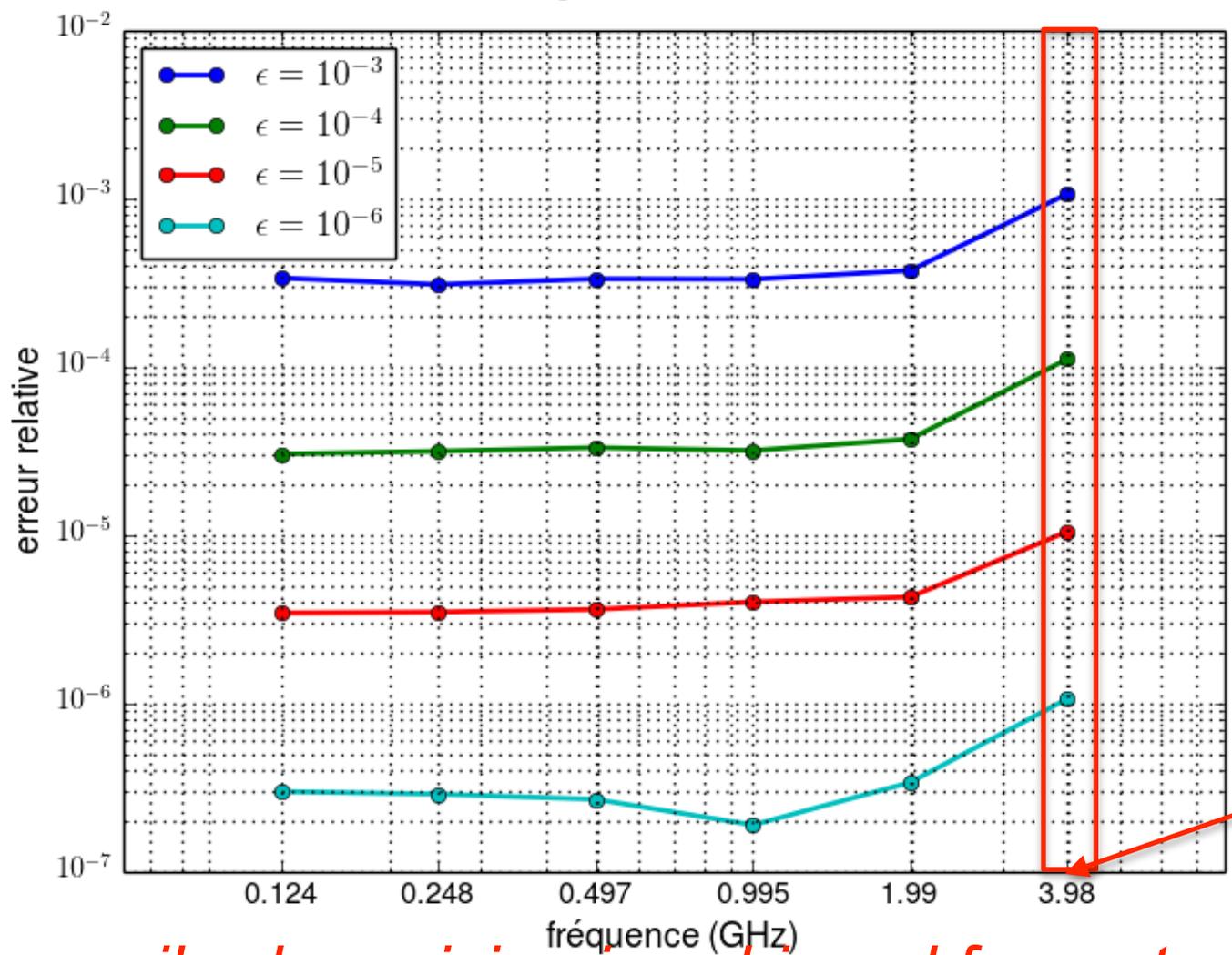
- Fresnel admissibility condition
- Interpolation nodes chosen according to Landau-Widom theorem

- Control of the precision & of the growth of the block rank



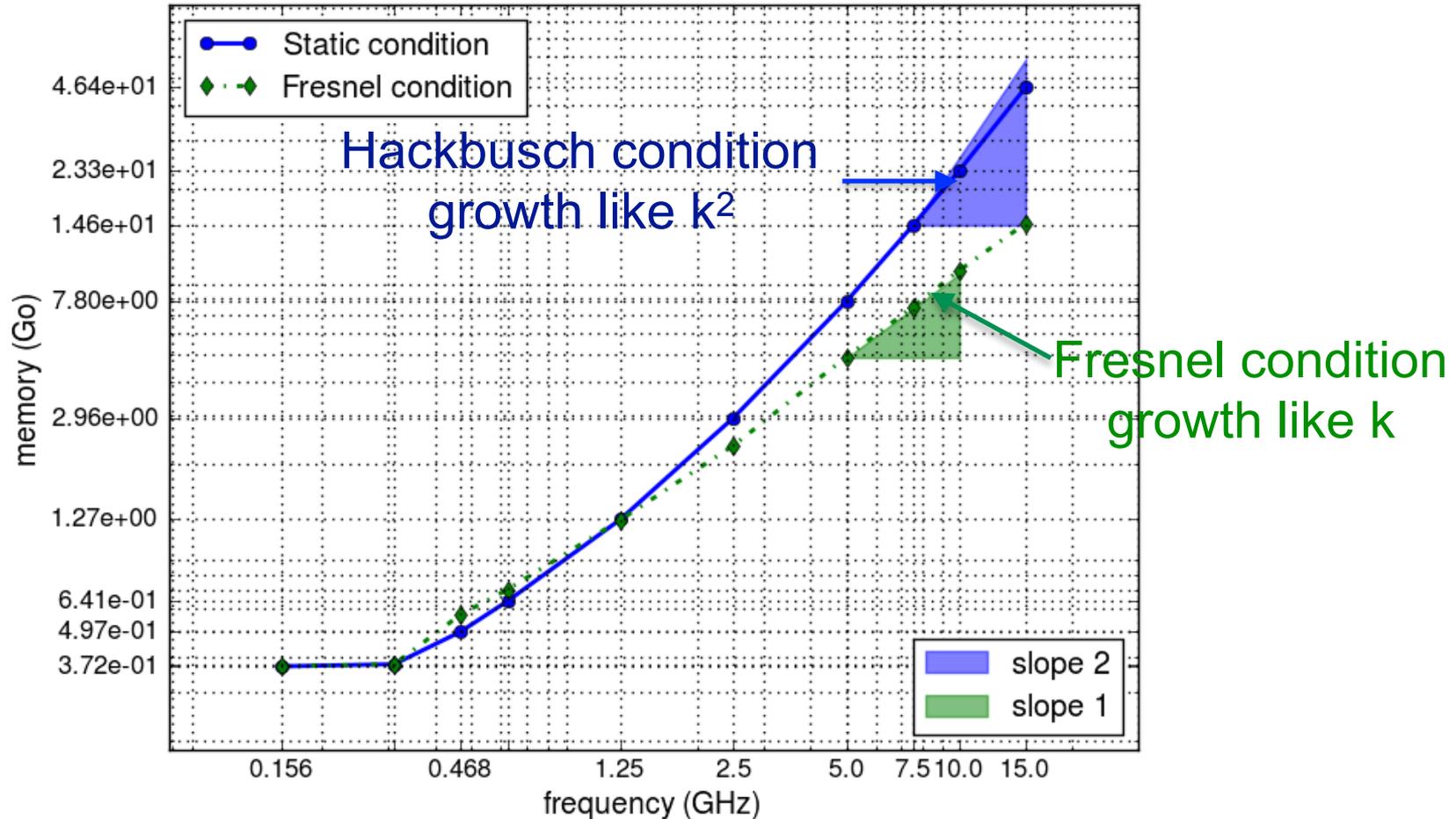
*prescribed precision is achieved blockwise*

# Frequency stable error



*prescribed precision is achieved for matvec*

# Influence of admissibility condition on the memory growth



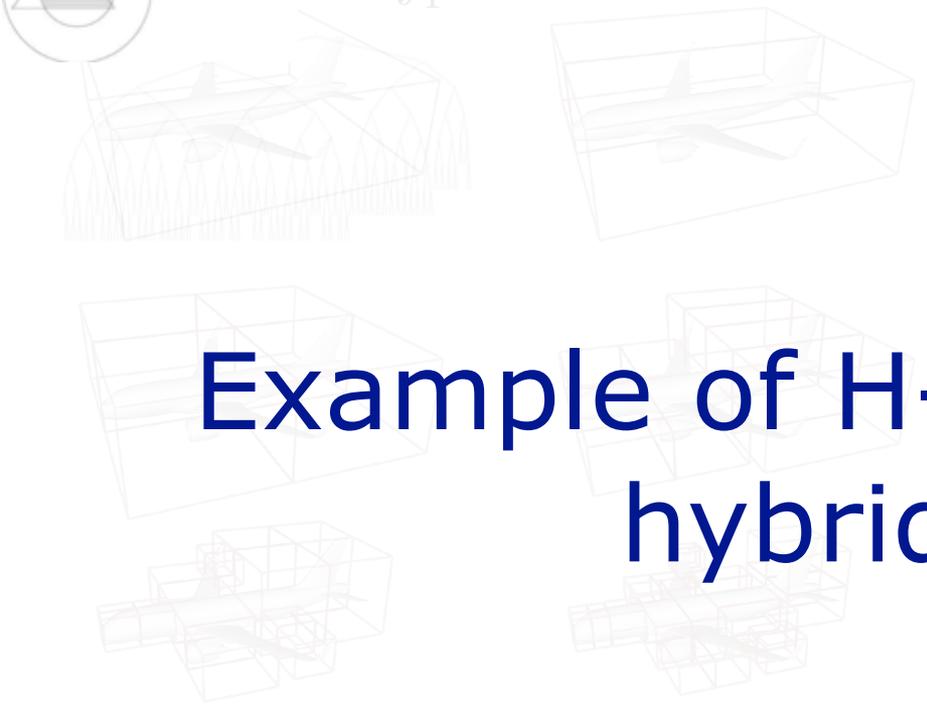
# Conclusion & perspectives

- Fresnel admissibility: frequency-dependent condition
- block rank growth control
- HCA2 method for oscillating kernel: robust, precise and efficient
- Better clustering for wave problems exploiting rank estimates
- Improving parallelism:
  - Distributing computation tasks
  - Memory management.
- Hybridisation of solvers for the biggest cases

$$N(c, \alpha) = \frac{2c}{1 - \ln(1 - \alpha)} \log(c) + \ln(\log(c))$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$

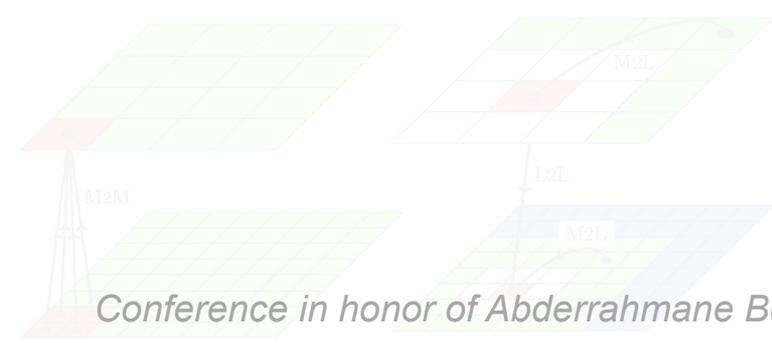
$$G(x, y) = \int_{\Gamma} G(x, y) \lambda(y) d\Gamma(y)$$

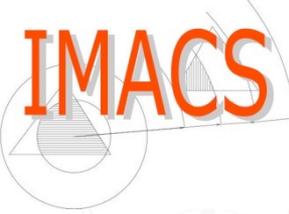


## Example of H-matrix — FMM hybridisation

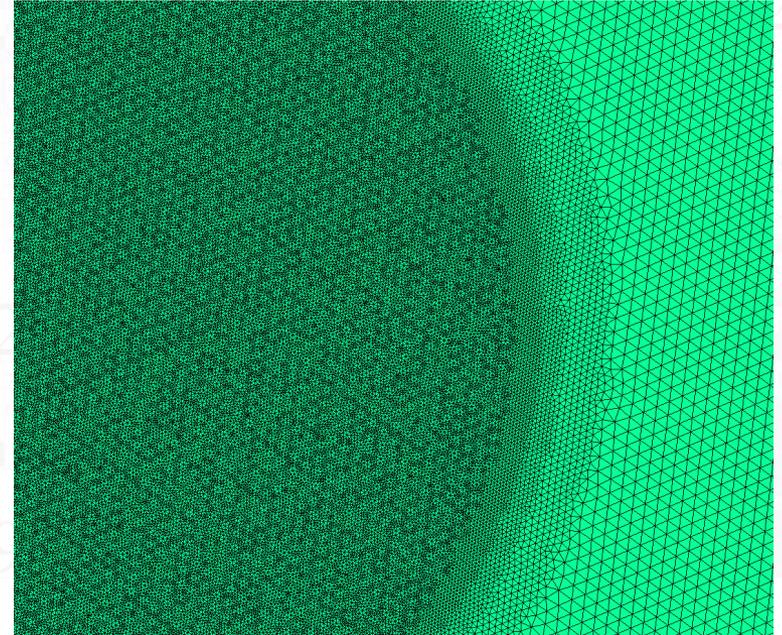
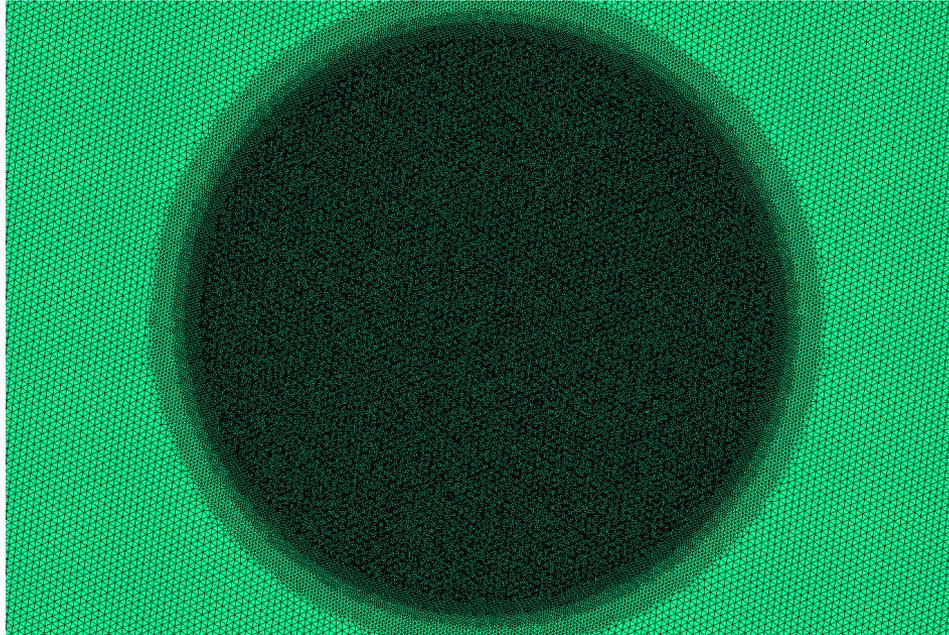
$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln \left( \frac{1-\alpha}{\alpha} \right) \log(c) + o(\log(c))$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$





# Locally refined meshes: plate refined in its centre



non refined plate: 235 kdof,  $h_{\max} = 2.7\text{mm}$

2 different refinements:

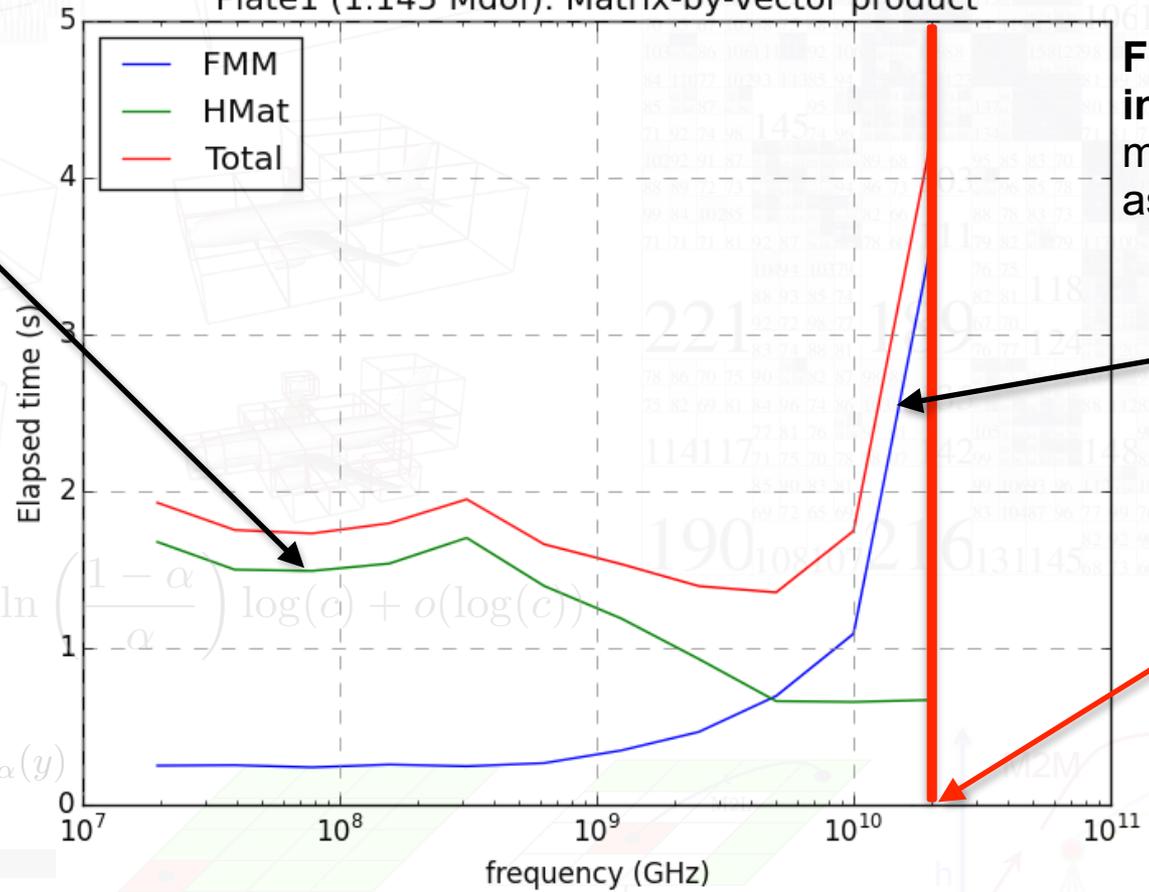
plate1: 1.1 Mdof —  $h_{\max}/h_{\min} \sim 13$

plate2: 1.7 Mdof —  $h_{\max}/h_{\min} \sim 22$

$$G(x, y) = \int_{\Gamma} G(\dots)$$

# Plates: elapsed time (6core proc)

Plate1 (1.145 M dof): Matrix-by-vector product



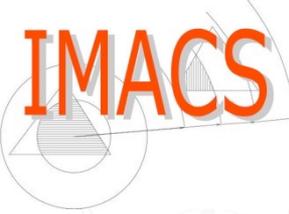
**Hmat for near field interactions:** more important for low frequencies

**FMM for far field interactions:** more multipoles and levels as frequency grows

$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln \left( \frac{1-\alpha}{\alpha} \right) \log(\alpha) + o(\log(c))$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$

Intel®Core™ i7-3930K processor with 64Go RAM

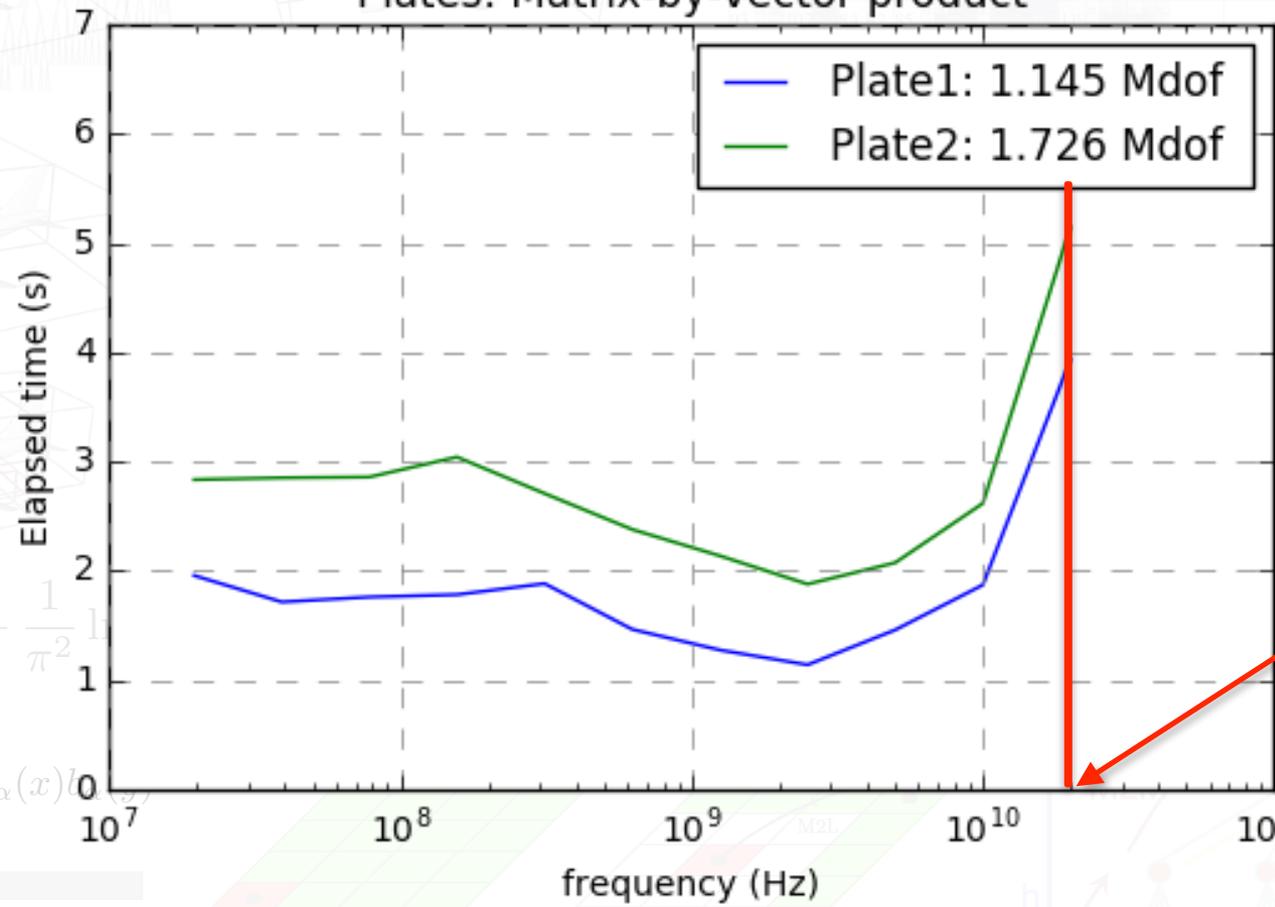


$$G(x, y) = \int_{\Gamma} G(x, y)$$

# Plates: elapsed time (6core proc)



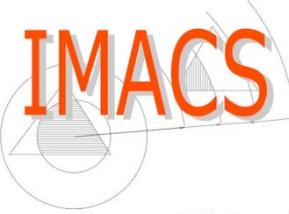
Plates: Matrix-by-vector product



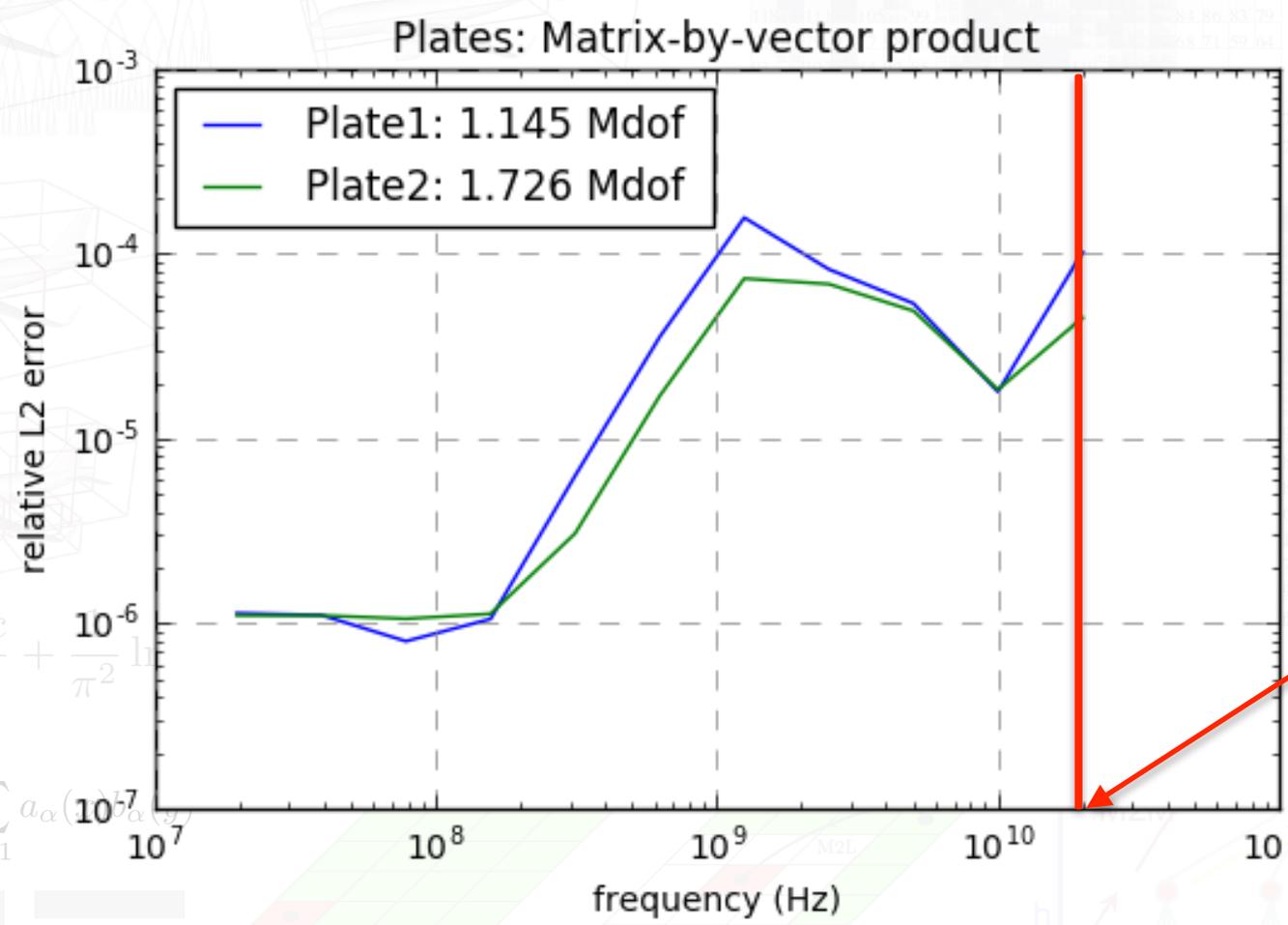
$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2}$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) l_{\alpha}(y)$$

Intel®Core™ i7-3930K processor with 64Go RAM



# Plates: precision



$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln \dots$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha} \phi_{\alpha}(x) \phi_{\alpha}(y)$$



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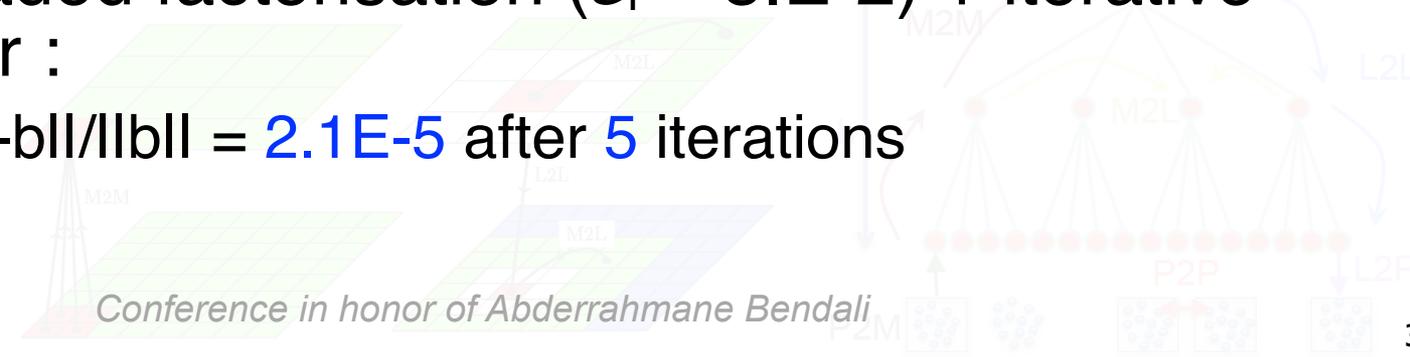
# H-matrix preconditioner

# Approximate factorisation

- assemble the H-matrix with the desired precision
- factorise the matrix with a degraded accuracy
- use this factorised matrix to provide an approximate inverse => preconditioner
- example with a 160640 dof :
  - accurate factorisation ( $\epsilon_f = 1.E-4$ ) :
    - $\|Ax-b\|/\|b\| = 1.2E-4$
  - degraded factorisation ( $\epsilon_f = 5.E-2$ ) + iterative solver :
    - $\|Ax-b\|/\|b\| = 2.1E-5$  after 5 iterations

$$N(c, \alpha) = \frac{2c}{\pi} + \frac{1}{\pi^2} \ln \left( \frac{1}{1 - \alpha} \right) \log(c) + \alpha \log(c)$$

$$G(x, y) \approx \sum_{\alpha=1}^r a_{\alpha}(x) b_{\alpha}(y)$$



- Advanced linear algebra algorithm & HPC:
  - \* H-matrix, FMM & hybridisation
  - \* handling of ooc, multithreading and distribution
  - \* block GMRES iterative solver
  - \* parallel computing using runtime system
- Library already used by several aeronautic and defence industrials
- Benchmarks showed a significant gain compared to existing solutions

