

Final Ph.D. Progress Report

**Integration of *hp*-adaptivity with a Two Grid Solver:
Applications to Electromagnetics.**

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Dissertation Committee: I. Babuska, L. Demkowicz, C. Torres-Verdin, R. Van de Geijn, M. Wheeler.

October 31, 2003

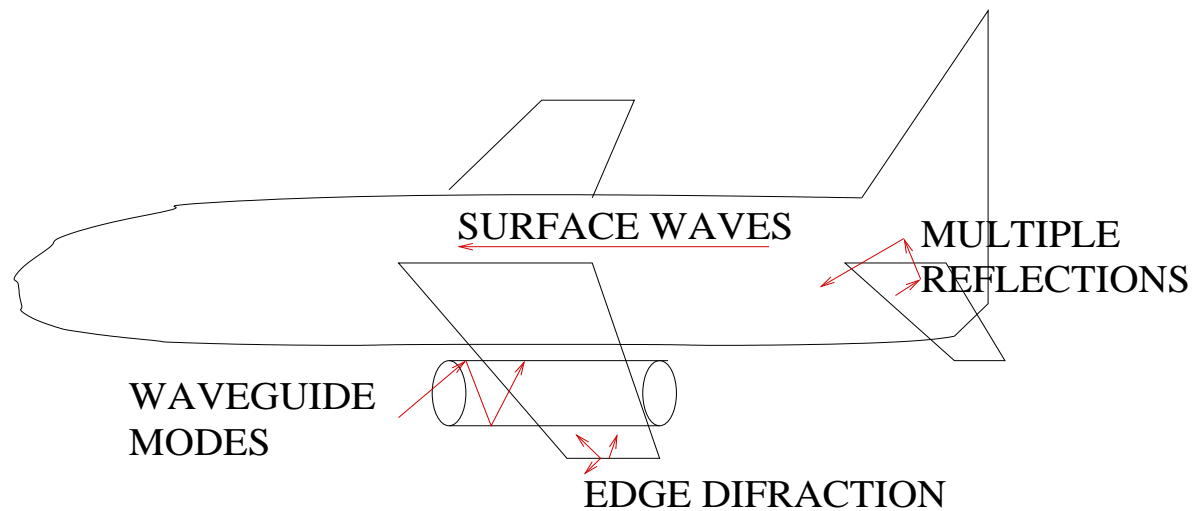
**Institute for Computational Engineering and Sciences (ICES)
The University of Texas at Austin**

OVERVIEW

1. Overview.
2. Motivation.
3. Maxwell's Equations.
4. hp-Adaptivity.
5. The Fully Automatic *hp*-Adaptive Strategy.
6. A Two Grid Solver for SPD Problems.
7. A Two Grid Solver for Electromagnetics.
8. Performance of the Two Grid Solver.
9. Electromagnetic Applications.
10. Conclusions and Future Work.

2. MOTIVATION

Radar Cross Section (RCS) Analysis

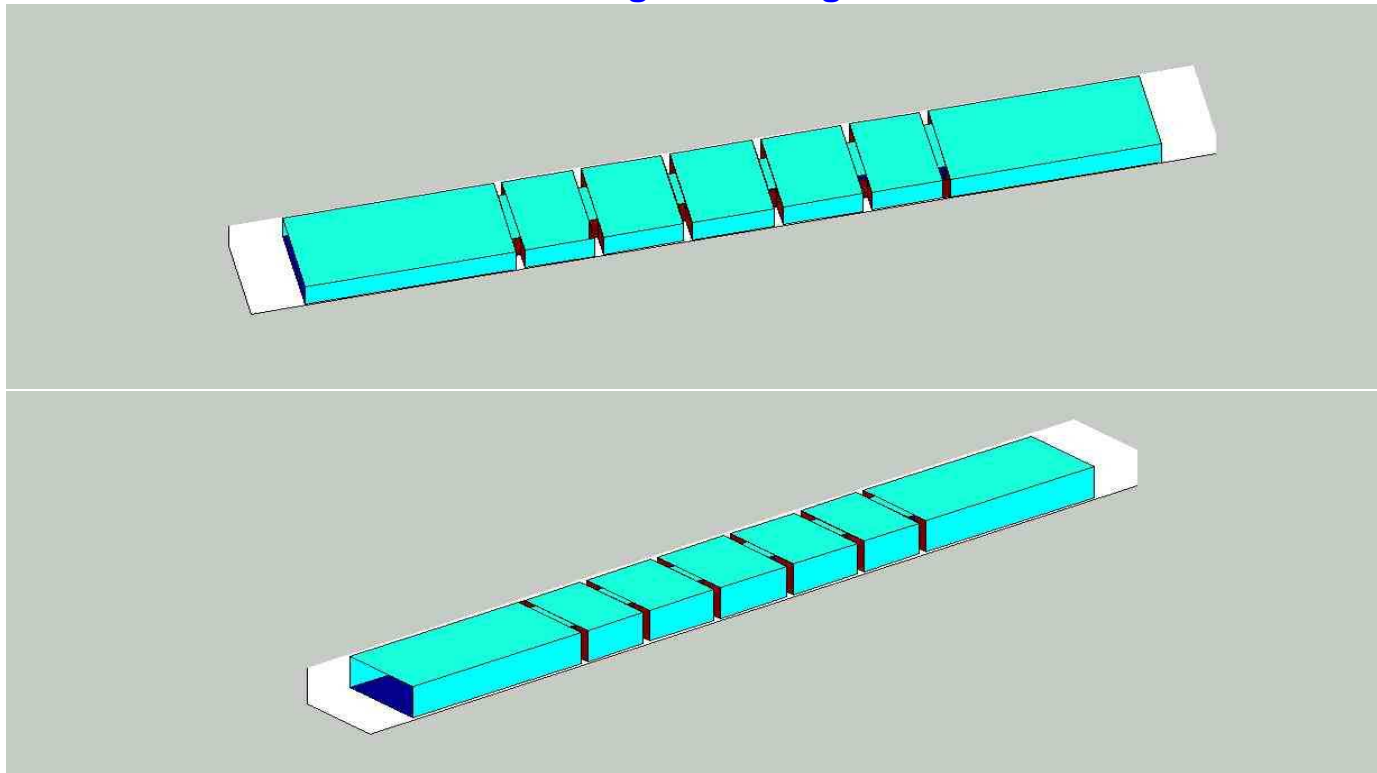


$$\text{RCS} = 4\pi \frac{\text{Power scattered to receiver per unit solid angle}}{\text{Incident power density}} = \lim_{r \rightarrow \infty} 4\pi r^2 \frac{|E^s|}{|E^i|}$$

Goal: Determine the RCS of a plane.

2. MOTIVATION

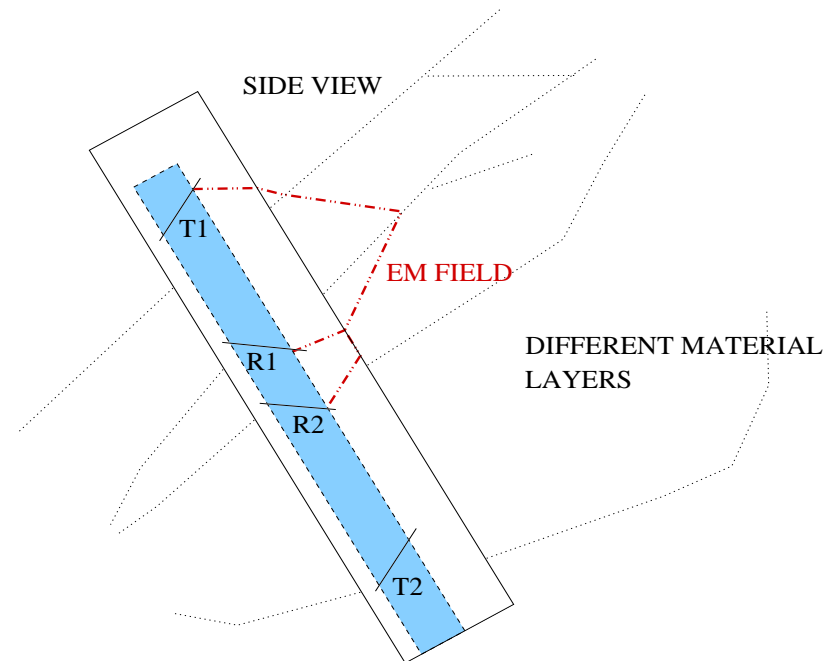
Waveguide Design



Goal: Determine electric field intensity at the ports.

2. MOTIVATION

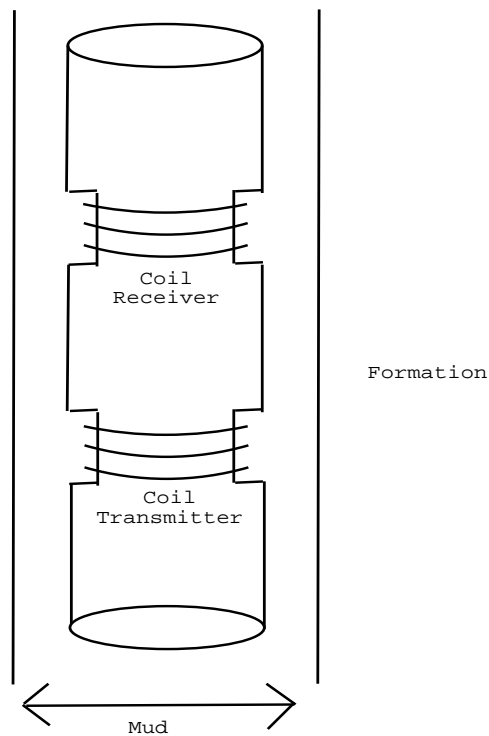
Modeling of a Logging While Drilling (LWD) electromagnetic measuring device



Goal: Determine EM field at the receiver antennas.

2. MOTIVATION

Modeling of a Logging While Drilling (LWD) electromagnetic measuring device



**Simplest case:
ONE COIL TRANSMITTER**

Goal: Determine EM field at the receiver antennas.

3. MAXWELL'S EQUATIONS

Time Harmonic Maxwell's Equations:

$$\nabla \times \mathbf{E} = -j\mu\omega\mathbf{H}$$

$$\nabla \times \mathbf{H} = j\omega\epsilon\mathbf{E} + \sigma\mathbf{E} + \mathbf{J}^{imp}$$

Reduced Wave Equation:

$$\nabla \times \left(\frac{1}{\mu} \nabla \times \mathbf{E} \right) - (\omega^2\epsilon - j\omega\sigma)\mathbf{E} = -j\omega\mathbf{J}^{imp},$$

Boundary Conditions (BC):

- Dirichlet BC at a PEC surface:

$$\mathbf{n} \times \mathbf{E}^s = -\mathbf{n} \times \mathbf{E}^{inc}$$

$$\mathbf{n} \times \mathbf{E} = 0$$

- Neumann continuity BC at a material interface:

$$\mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E}^s = -\mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E}^{inc}$$

$$\mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E} = -j\omega\mathbf{J}_S^{imp}$$

- Silver Müller radiation condition at ∞ :

$$\mathbf{e}_r \times (\nabla \times \mathbf{E}^s) - jk_0 \times \mathbf{E}^s = O(r^{-2})$$

3. MAXWELL'S EQUATIONS

Variational formulation

The reduced wave equation in Ω ,

$$\nabla \times \left(\frac{1}{\mu} \nabla \times E \right) - (\omega^2 \epsilon - j\omega\sigma)E = -j\omega J^{imp},$$

A variational formulation

$$\left\{ \begin{array}{l} \text{Find } E \in H_D(\text{curl}; \Omega) \text{ such that} \\ \int_{\Omega} \frac{1}{\mu} (\nabla \times E) \cdot (\nabla \times \bar{F}) dx - \int_{\Omega} (\omega^2 \epsilon - j\omega\sigma) E \cdot \bar{F} dx = \\ -j\omega \left\{ \int_{\Omega} J^{imp} \cdot \bar{F} dx + \int_{\Gamma_2} J_S^{imp} \cdot \bar{F} dS \right\} \quad \text{for all } F \in H_D(\text{curl}; \Omega). \end{array} \right.$$

A stabilized variational formulation (using a Lagrange multiplier):

$$\left\{ \begin{array}{l} \text{Find } E \in H_D(\text{curl}; \Omega), p \in H_D^1(\Omega) \text{ such that} \\ \int_{\Omega} \frac{1}{\mu} (\nabla \times E) \cdot (\nabla \times \bar{F}) dx - \int_{\Omega} (\omega^2 \epsilon - j\omega\sigma) E \cdot \bar{F} dx - \int_{\Omega} (\omega^2 \epsilon - j\omega\sigma) \nabla p \cdot \bar{F} dx = \\ -j\omega \left\{ \int_{\Omega} J^{imp} \cdot \bar{F} dx + \int_{\Gamma_2} J_S^{imp} \cdot \bar{F} dS \right\} \quad \forall F \in H_D(\text{curl}; \Omega) \\ - \int_{\Omega} (\omega^2 \epsilon - j\omega\sigma) E \cdot \nabla \bar{q} dx = -j\omega \left\{ \int_{\Omega} J^{imp} \cdot \nabla \bar{q} dx + \int_{\Gamma_2} J_S^{imp} \cdot \nabla \bar{q} dS \right\} \quad \forall q \in H_D^1(\Omega). \end{array} \right.$$

3. MAXWELL'S EQUATIONS

De Rham diagram

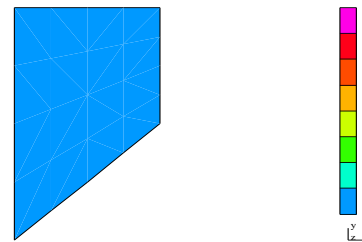
De Rham diagram is critical to the theory of FE discretizations of Maxwell's equations.

$$\begin{array}{ccccccccc}
 \mathbb{R} & \longrightarrow & W & \xrightarrow{\nabla} & Q & \xrightarrow{\nabla \times} & V & \xrightarrow{\nabla \circ} & L^2 & \longrightarrow & 0 \\
 \downarrow id & & \downarrow \Pi & & \downarrow \Pi^{\text{curl}} & & \downarrow \Pi^{\text{div}} & & \downarrow P & & \\
 \mathbb{R} & \longrightarrow & W^p & \xrightarrow{\nabla} & Q^p & \xrightarrow{\nabla \times} & V^p & \xrightarrow{\nabla \circ} & W^{p-1} & \longrightarrow & 0 .
 \end{array}$$

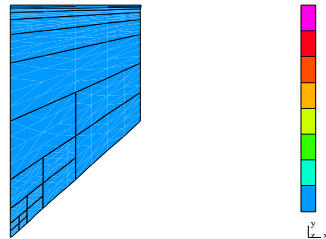
This diagram relates two exact sequences of spaces, on both continuous and discrete levels, and corresponding interpolation operators.

4. HP-ADAPTIVITY

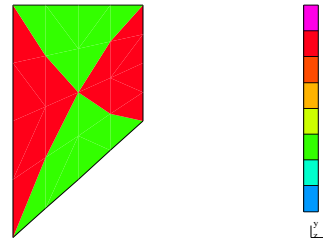
Different refinement strategies for finite elements:



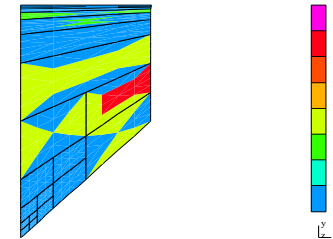
Given initial grid



h-refined grid



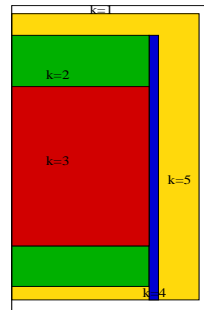
p-refined grid



hp-refined grid

4. HP-ADAPTIVITY

Orthotropic heat conduction example

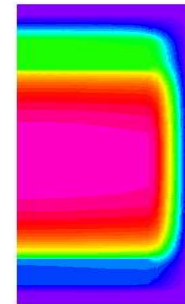


Equation: $\nabla(K\nabla u) = f^{(k)}$

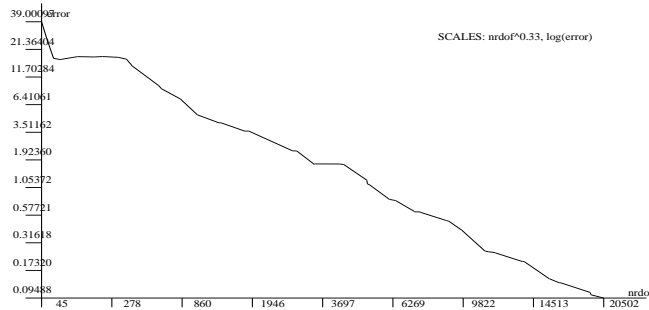
$$K = K^{(k)} = \begin{bmatrix} K_x^{(k)} & 0 \\ 0 & K_y^{(k)} \end{bmatrix}$$

$$K_x^{(k)} = (25, 7, 5, 0.2, 0.05)$$

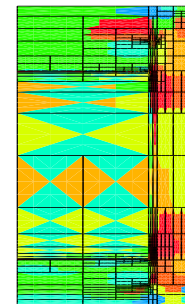
$$K_y^{(k)} = (25, 0.8, 0.0001, 0.2, 0.05)$$



Solution: unknown
 Boundary Conditions:
 $K^{(i)}\nabla u \cdot n = g^{(i)} - \alpha^{(i)}u$



Convergence history
 (tolerance error = 0.1 %)

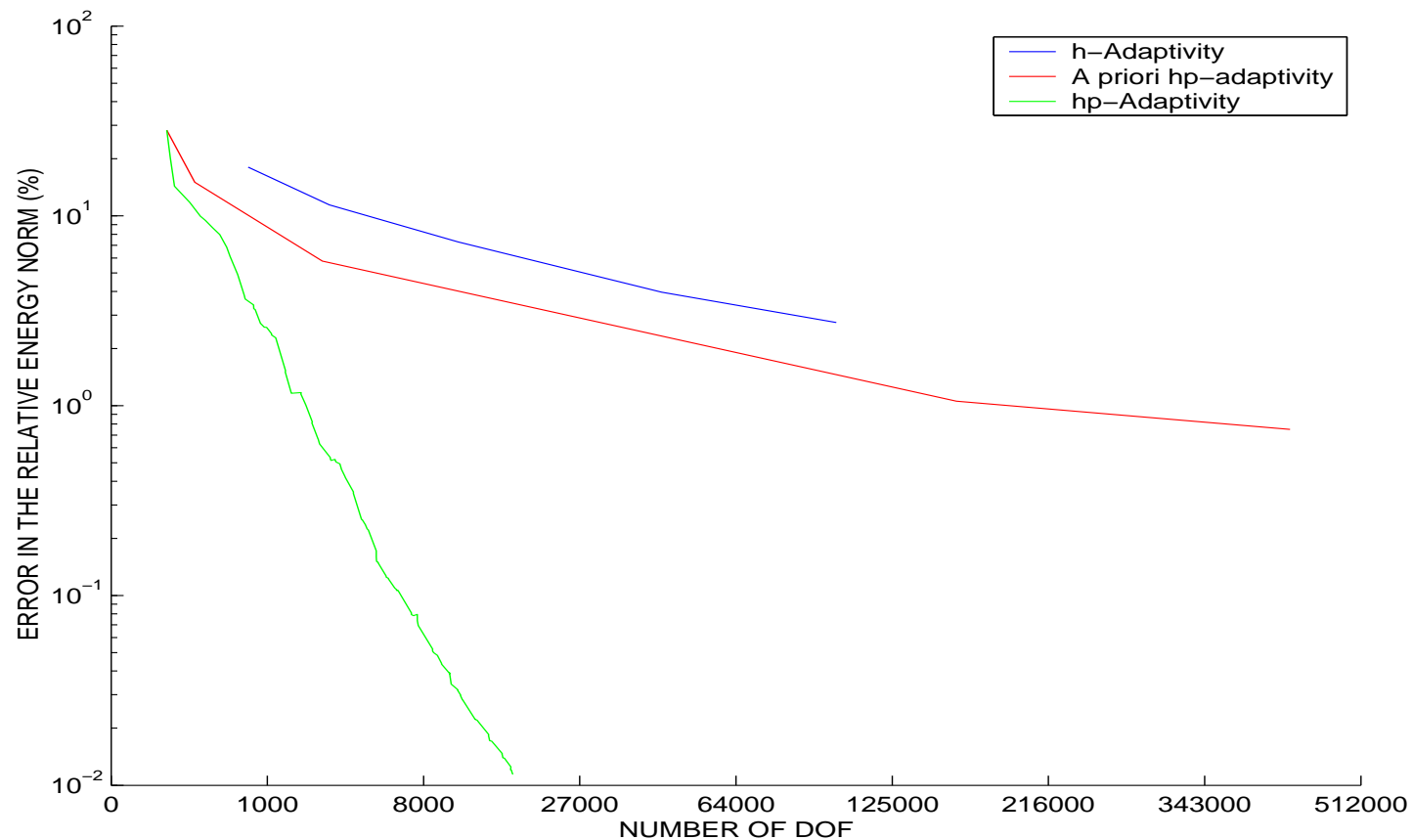


Final *hp* grid

4. HP-ADAPTIVITY

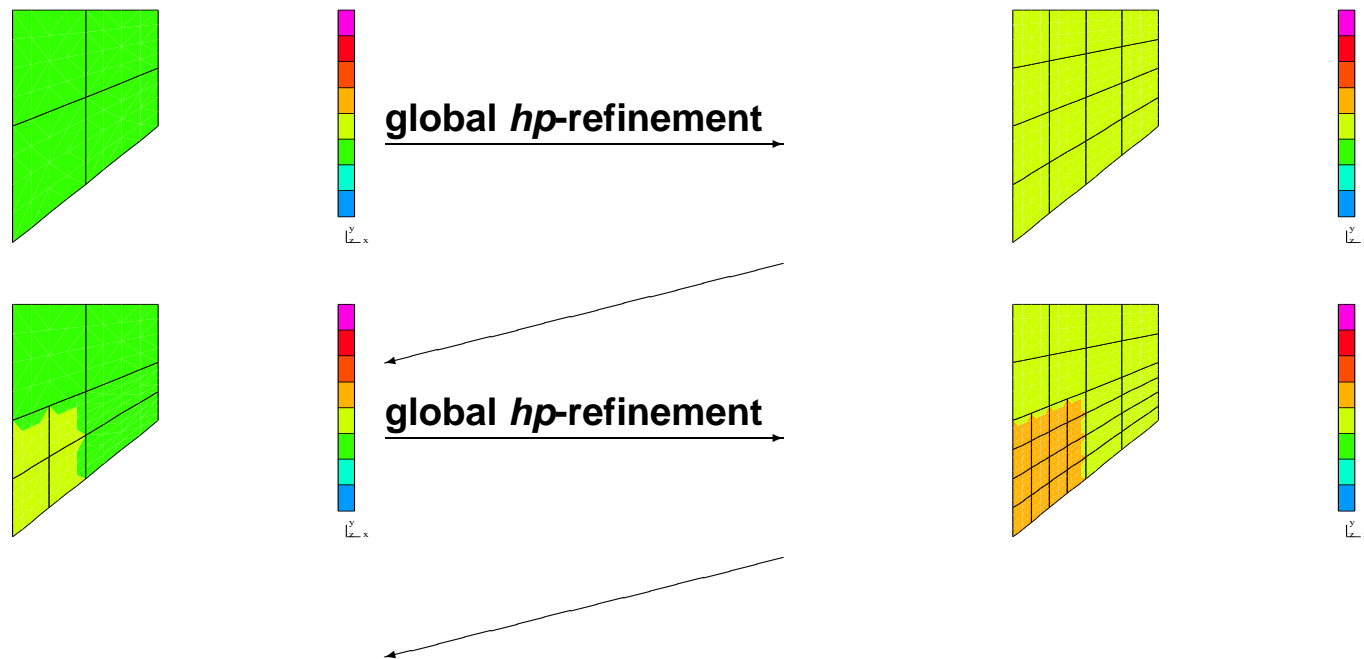
Convergence comparison

Orthotropic heat conduction example



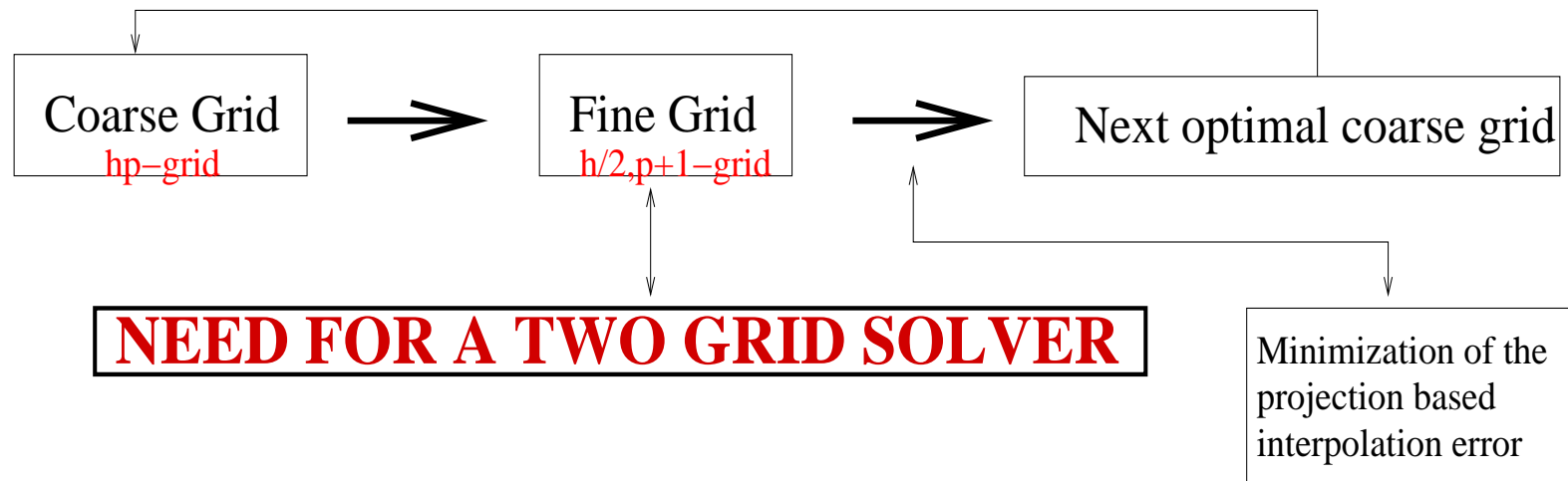
5. THE FULLY AUTOMATIC *HP*-ADAPTIVE STRATEGY

Fully automatic *hp*-adaptive strategy



5. THE FULLY AUTOMATIC *HP*-ADAPTIVE STRATEGY

Automatic *hp*-adaptivity delivers exponential convergence and enables solution of challenging EM problems



6. A TWO GRID SOLVER FOR SPD PROBLEMS

We seek x such that $Ax = b$. Consider the following iterative scheme:

$$\begin{aligned} r^{(n+1)} &= [I - \alpha^{(n)} AS]r^{(n)} \\ x^{(n+1)} &= [I - \alpha^{(n)} S]r^{(n)} \end{aligned}$$

where S is a matrix, and $\alpha^{(n)}$ is a relaxation parameter. $\alpha^{(n)}$ *optimal* if:

$$\alpha^{(n)} = \arg \min \| x^{(n+1)} - x \|_A = \frac{(A^{-1}r^{(n)}, Sr^{(n)})_A}{(Sr^{(n)}, Sr^{(n)})_A}$$

Then, we define our two grid solver as:

$$\begin{aligned} &1 \text{ Iteration with } S = S_F = \sum A_i^{-1} \quad + \\ &1 \text{ Iteration with } S = S_C = PA_C^{-1}R \end{aligned}$$

6. A TWO GRID SOLVER FOR SPD PROBLEMS

Error reduction and stopping criteria

Let $e^{(n)} = x^{(n)} - x$ the error at step n , $\tilde{e}^{(n)} = [I - S_C A]e^{(n)} = [I - P_C]e^{(n)}$. Then:

$$\frac{\|e^{(n+1)}\|_A^2}{\|e^{(n)}\|_A^2} = 1 - \frac{|(\tilde{e}^{(n)}, S_F A \tilde{e}^{(n)})_A|^2}{\|\tilde{e}^{(n)}\|_A^2 \|S_F A \tilde{e}^{(n)}\|_A^2} = 1 - \frac{|(\tilde{e}^{(n)}, (P_C + S_F A)\tilde{e}^{(n)})_A|^2}{\|\tilde{e}^{(n)}\|_A^2 \|S_F A \tilde{e}^{(n)}\|_A^2}$$

Then:

$$\frac{\|e^{(n+1)}\|_A^2}{\|e^{(n)}\|_A^2} \leq \sup_e \left[1 - \frac{|(e, (P_C + S_F A)e)_A|^2}{\|e\|_A^2 \|S_F A e\|_A^2} \right] \leq C < 1 \quad \text{(Error Reduction)}$$

For our stopping criteria, we want: Iterative Solver Error \approx Discretization Error. That is:

$$\frac{\|e^{(n+1)}\|_A}{\|e^{(0)}\|_A} \leq 0.01 \quad \text{(Stopping Criteria)}$$

A TWO GRID SOLVER FOR ELECTROMAGNETICS

We seek x such that $Ax = b$. Consider the following iterative scheme:

$$\begin{aligned} r^{(n+1)} &= [I - \alpha^{(n)} AS]r^{(n)} \\ x^{(n+1)} &= [I - \alpha^{(n)} S]r^{(n)} \end{aligned}$$

where S is a matrix, and $\alpha^{(n)}$ is a relaxation parameter. $\alpha^{(n)}$ *optimal* if:

$$\alpha^{(n)} = \arg \min \| x^{(n+1)} - x \|_B = \frac{(A^{-1}r^{(n)}, Sr^{(n)})_B}{(Sr^{(n)}, Sr^{(n)})_B} \quad \text{(NOT COMPUTABLE)}$$

Then, we define our two grid solver for **Electromagnetics** as:

$$\begin{aligned} &1 \text{ Iteration with } S = S_F = \sum A_i^{-1} \quad + \\ &1 \text{ Iteration with } S = S_\nabla = \sum G_i^{-1} \quad + \\ &1 \text{ Iteration with } S = S_C = PA_C^{-1}R \end{aligned}$$

A TWO GRID SOLVER FOR ELECTROMAGNETICS

A two grid solver for discretization of Maxwell's equations
using *hp*-FE

Consider the following two problems:

Problem I: $\nabla \times \nabla \times \mathbf{E} - k^2 \mathbf{E} = \mathbf{J}$

Matrix form: $Au = v$

Two grid solver V-cycle:

$$TG = (I - \alpha_1 S_F A)(I - \alpha_2 S_\nabla A)(I - S_C A_C)$$

Problem II: $\nabla \times \nabla \times \mathbf{E} + \mathbf{E} = \mathbf{J}$

Matrix form: $\hat{A}u = v$

Two grid solver V-cycle:

$$\widehat{TG} = (I - \alpha_1 \hat{S}_F \hat{A})(I - \alpha_2 \hat{S}_\nabla \hat{A})(I - \hat{S}_C \hat{A}_C)$$

Theorem: If h is small enough, then:

$$\| TGe^{(n)} \| \leq \| \widehat{TGe}^{(n)} \| + Ch$$

Notice that C is independent of h and p .

A TWO GRID SOLVER FOR ELECTROMAGNETICS

A two grid solver for discretization of Maxwell's equations using *hp*-FE

Helmholtz decomposition:

$$H_D(\text{curl}; \Omega) = (\text{Ker}(\text{curl})) \oplus (\text{Ker}(\text{curl}))^\perp$$

We define the following subspaces (T = grid, K = element, v = vertex, e = edge):

$$\begin{aligned} \Omega_{k,i}^v &= \text{int}(\cup\{\bar{K} \in T_k : v_{k,i} \in \partial K\}) ; & \Omega_{k,i}^e &= \text{int}(\cup\{\bar{K} \in T_k : e_{k,i} \in \partial K\}) & \text{Domain decomposition} \\ M_{k,i}^v &= \{u \in M_k : \text{supp}(u) \subset \Omega_{k,i}^v\} ; & M_{k,i}^e &= \{u \in M_k : \text{supp}(u) \subset \Omega_{k,i}^e\} & \text{Nedelec's elements decomposition} \\ W_{k,i}^v &= \{u \in W_k : \text{supp}(u) \subset \Omega_{k,i}^v\} ; & W_{k,i}^e &= \{u \in W_k : \text{supp}(u) \subset \Omega_{k,i}^e\} = \emptyset & \text{Polynomial spaces decomposition} \end{aligned}$$

Hiptmair proposed the following decomposition of M_k :

$$M_k = \sum_e M_{k,i}^e + \sum_v \nabla W_{k,i}^v$$

Arnold *et. al* proposed the following decomposition of M_k :

$$M_k = \sum_v M_{k,i}^v$$

8. PERFORMANCE OF THE TWO GRID SOLVER

Numerical Studies

2002

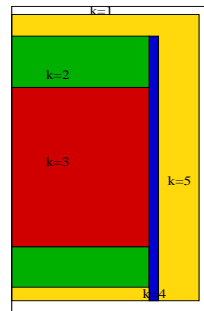
- Importance of the choice of shape functions.
- Importance of the relaxation parameter.
- Selection of patches for the block Jacobi smoother.
- Effect of averaging.
- Error estimation.
- Smoothing vs two grid solver.
- Guiding *hp*-adaptivity with a partially converged fine grid solution.

2003

- **Guiding *hp*-adaptivity** with a partially converged fine grid solution for EM problems.
- **Efficiency** of the two grid solver.
- **Number of elements per wavelength** required by the two grid solver to converge.
- **Control** of the **dispersion** error.
- **Applications** to real world problems.

8. PERFORMANCE OF THE TWO GRID SOLVER

Orthotropic heat conduction example

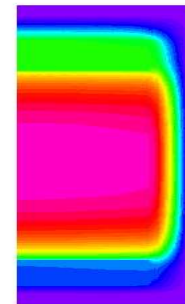


Equation: $\nabla(K\nabla u) = f^{(k)}$

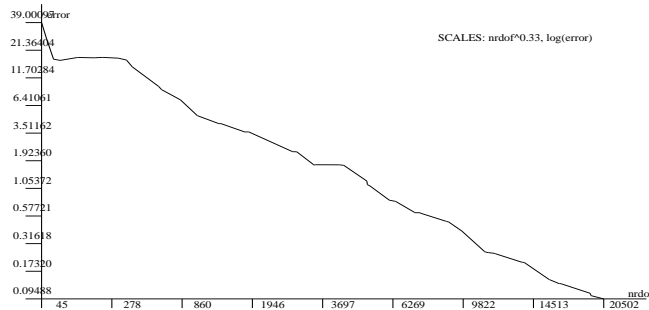
$$K = K^{(k)} = \begin{bmatrix} K_x^{(k)} & 0 \\ 0 & K_y^{(k)} \end{bmatrix}$$

$$K_x^{(k)} = (25, 7, 5, 0.2, 0.05)$$

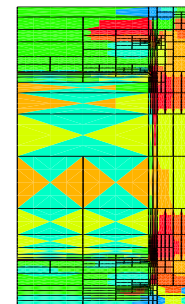
$$K_y^{(k)} = (25, 0.8, 0.0001, 0.2, 0.05)$$



Solution: unknown
 Boundary Conditions:
 $K^{(i)}\nabla u \cdot n = g^{(i)} - \alpha^{(i)}u$



Convergence history
 (tolerance error = 0.1 %)

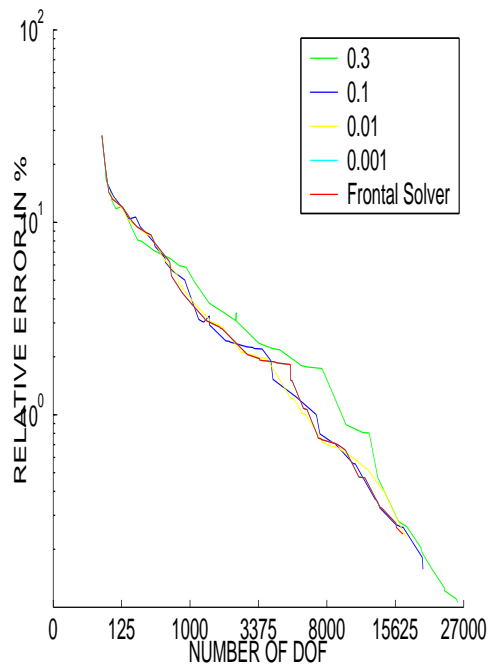


Final hp grid

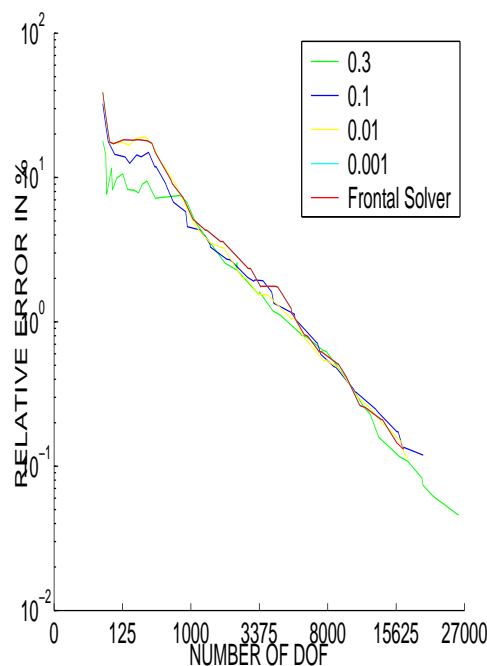
8. PERFORMANCE OF THE TWO GRID SOLVER

Guiding automatic *hp*-refinements

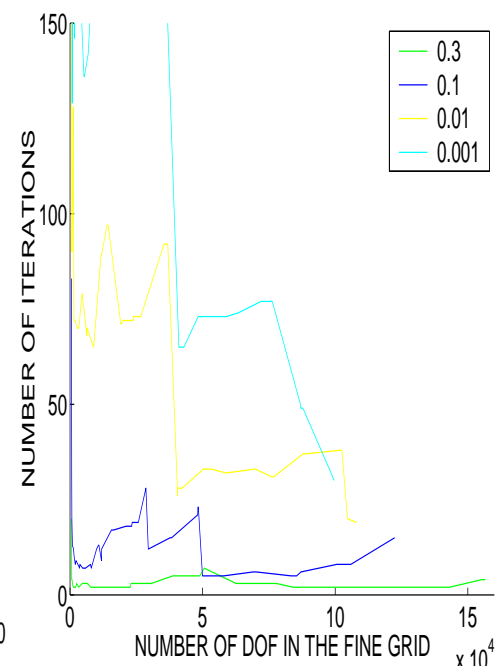
Orthotropic heat conduction. Guiding *hp*-refinements with a partially converged solution.



Energy error estimate



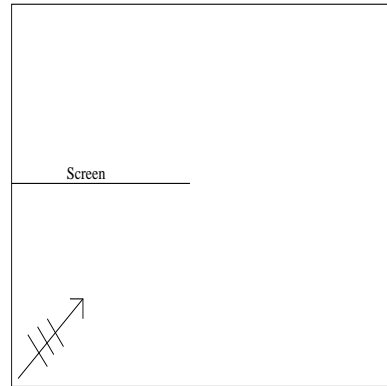
Discretization error estimate



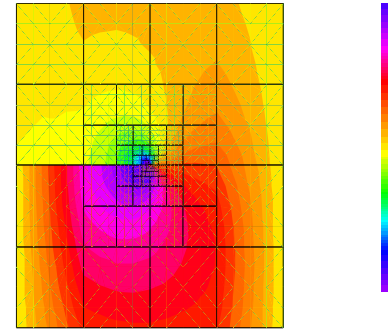
Number of iterations

8. PERFORMANCE OF THE TWO GRID SOLVER

Plane Wave incident into a screen (diffraction problem)



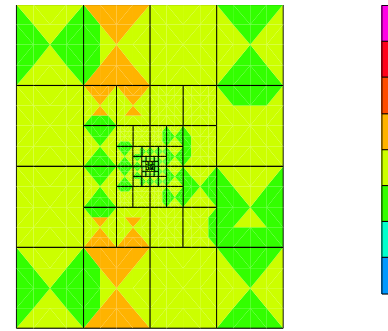
Geometry



Second component of electric field



Convergence history
(tolerance error = 0.1 %)

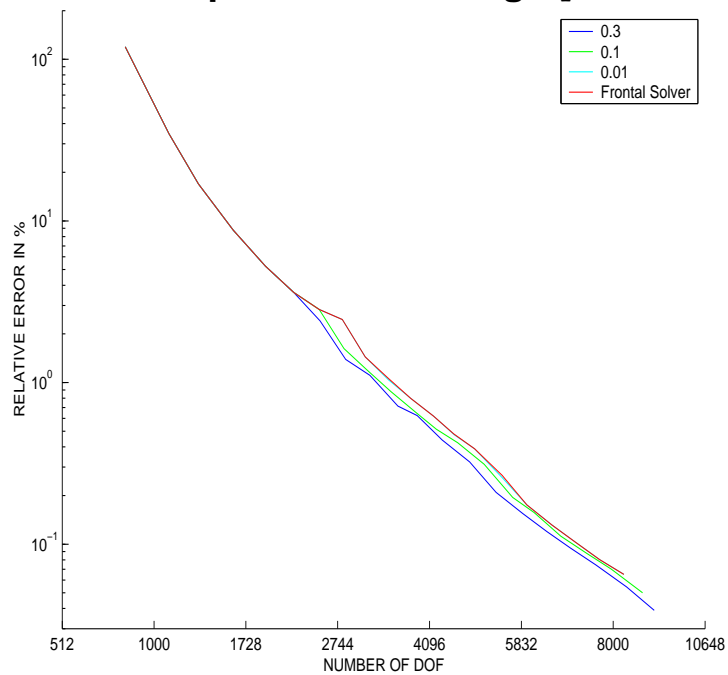


Final *hp*-grid

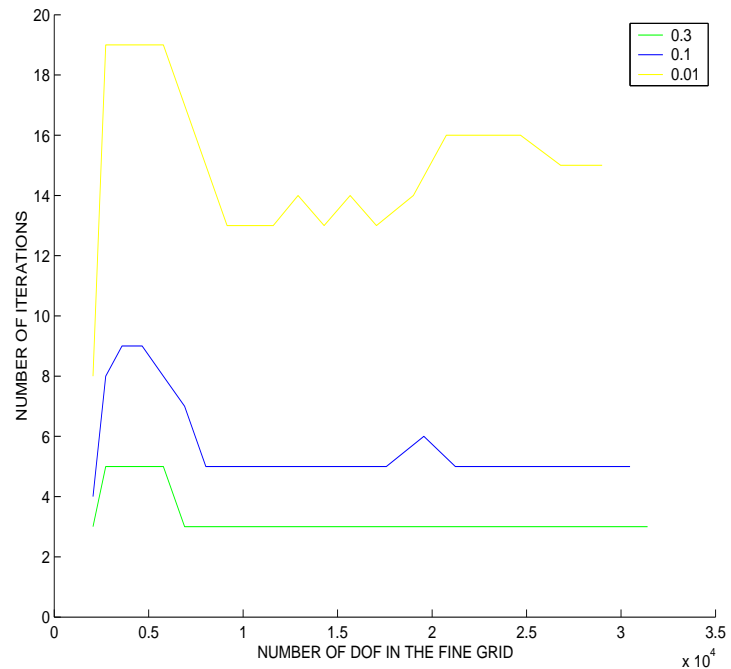
Numerical Results

Guiding automatic *hp*-refinements

Diffraction problem. Guiding *hp*-refinements with a partially converged solution.



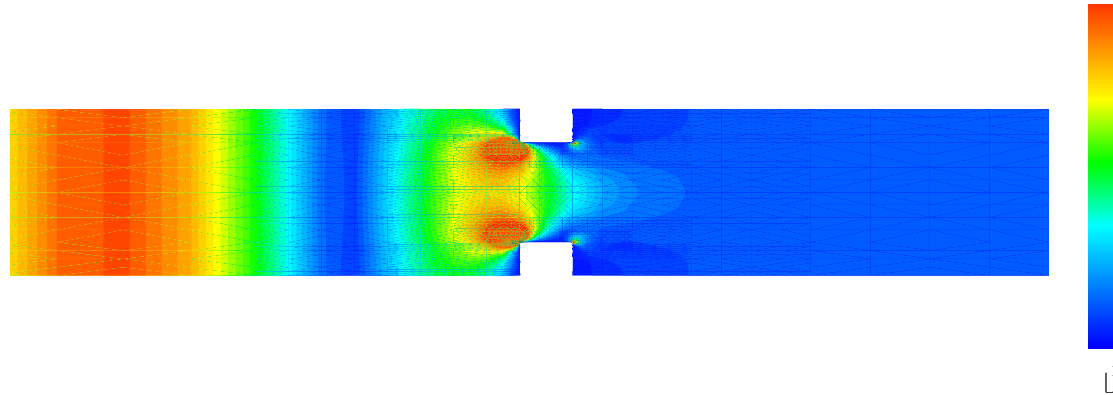
Discretization error estimate



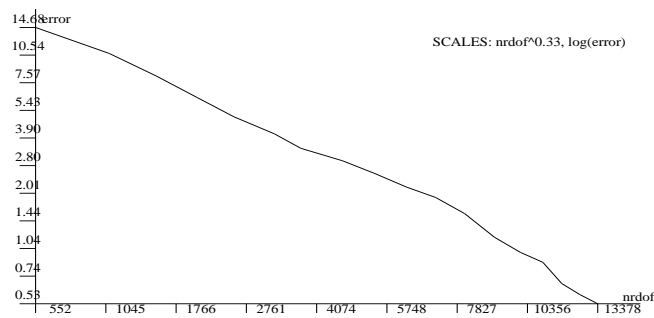
Number of iterations

8. PERFORMANCE OF THE TWO GRID SOLVER

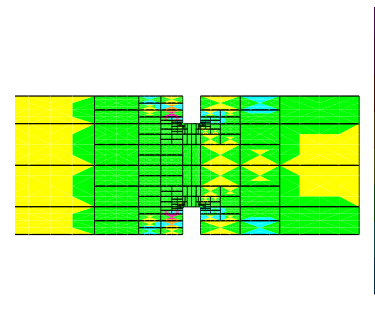
Waveguide example



Module of Second Component of Magnetic Field



Convergence history
(tolerance error = 0.5 %)

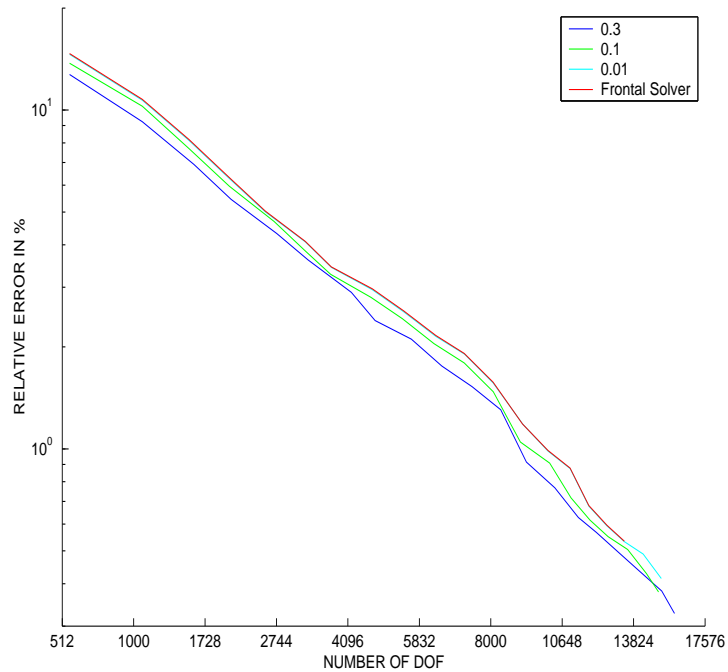


Final hp -grid

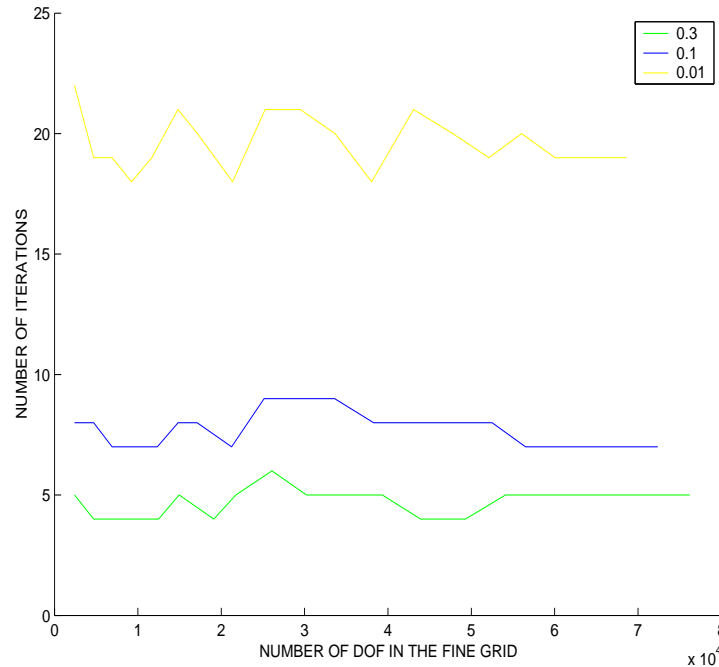
8. PERFORMANCE OF THE TWO GRID SOLVER

Guiding automatic *hp*-refinements

Waveguide example. Guiding *hp*-refinements with a partially converged solution.



Discretization error estimate



Number of iterations

8. PERFORMANCE OF THE TWO GRID SOLVER

Efficiency of the two grid solver

We studied scalability of the solver with respect h and p .

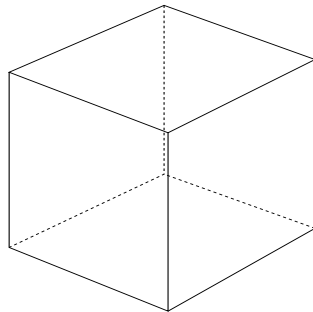
$$\text{Speed} = \text{Coarse grid solve} + \mathcal{O}(p^9 N)$$

We implemented an efficient solver.

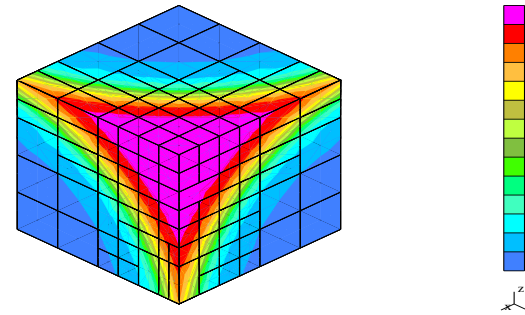
- Fast integration rules.
- Fast matrix vector multiplication.
- Fast assembling.
- Fast patch inversion.
- Fast construction of prolongation/restriction operator.

8. PERFORMANCE OF THE TWO GRID SOLVER

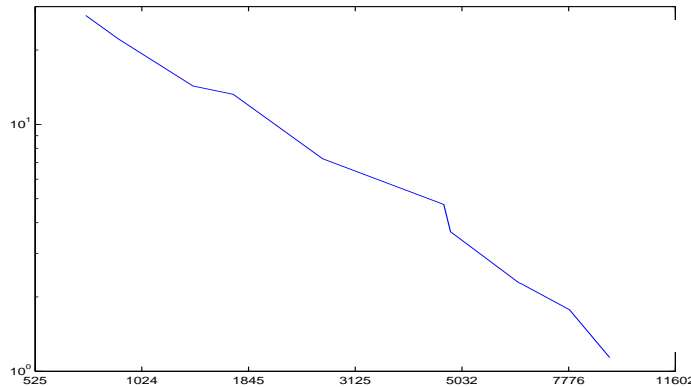
3D shock like solution example



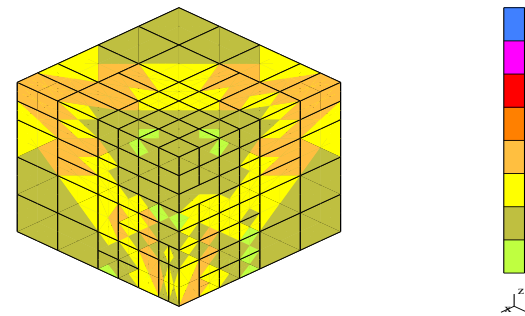
Equation: $-\Delta u = f$
 Geometry: unit cube



Solution: $u = \text{atan}(20 * \sqrt{r} - \sqrt{3})$
 $r = (x - .25) ** 2 + (y - .25) ** 2 + (z - .25) ** 2$
 Dirichlet Boundary Conditions



Convergence history
 (tolerance error = 1%)

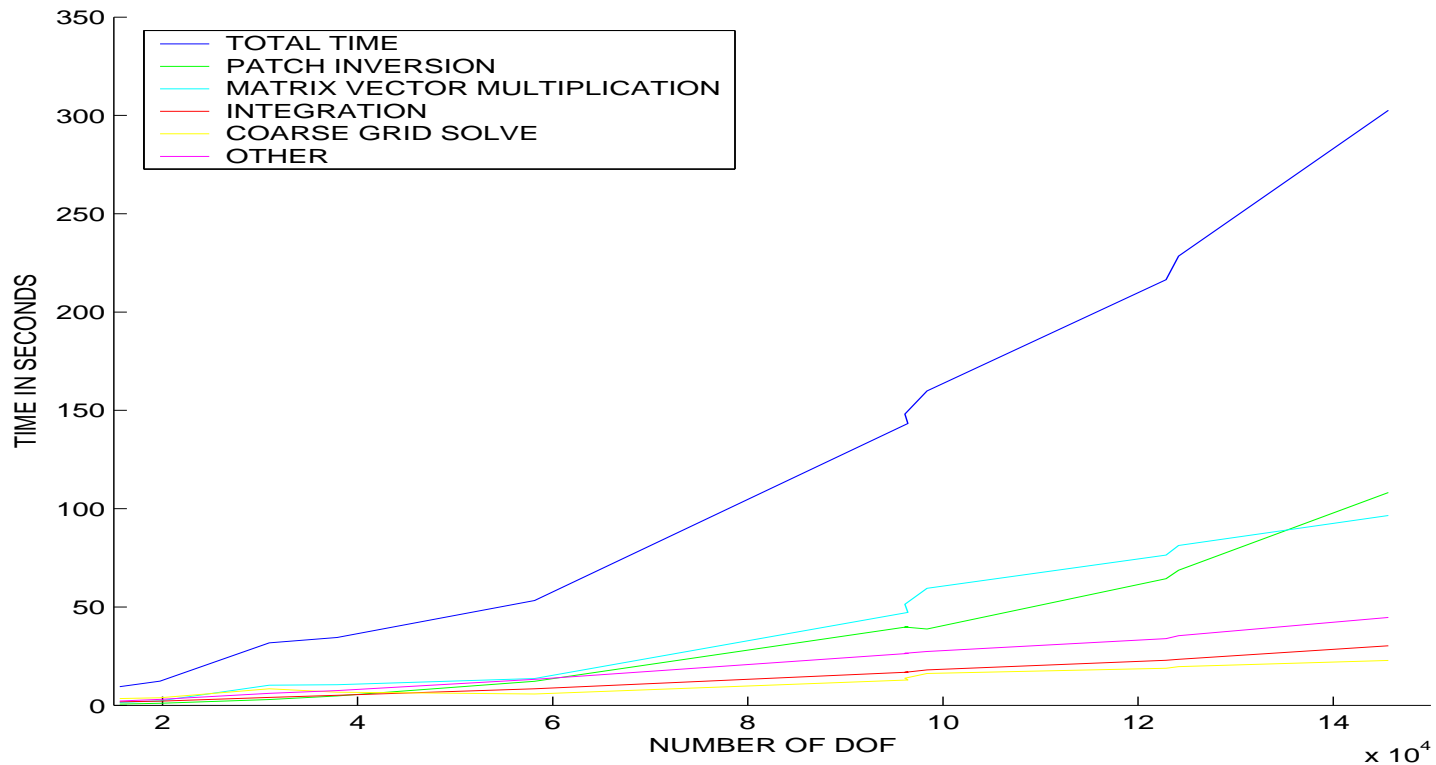


Final *hp* grid

8. PERFORMANCE OF THE TWO GRID SOLVER

Performance of the two grid solver

3D shock like solution example

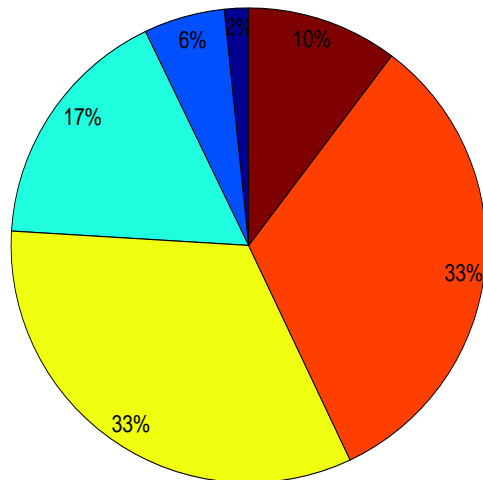
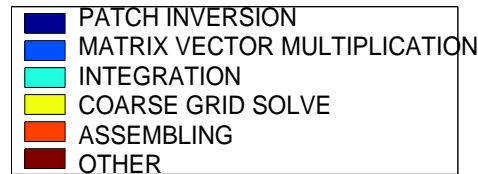


In core computations, AMD Athlon 1 Ghz processor.

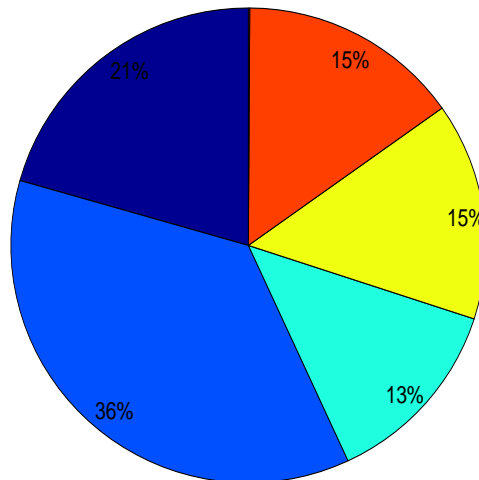
8. PERFORMANCE OF THE TWO GRID SOLVER

Performance of the two grid solver

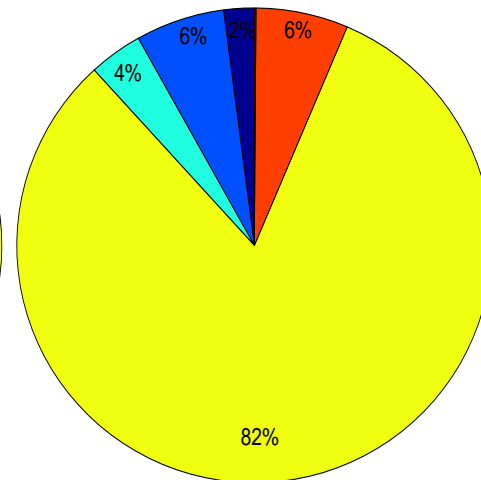
3D shock like solution problem



Nrdofs \approx 2.15 Million
 Total time \approx 8 minutes
 Memory* \approx 1.0 Gb
 p=2



Nrdofs \approx 0.27 Million
 Total time \approx 10 minutes
 Memory* \approx 2.0 Gb
 p=8



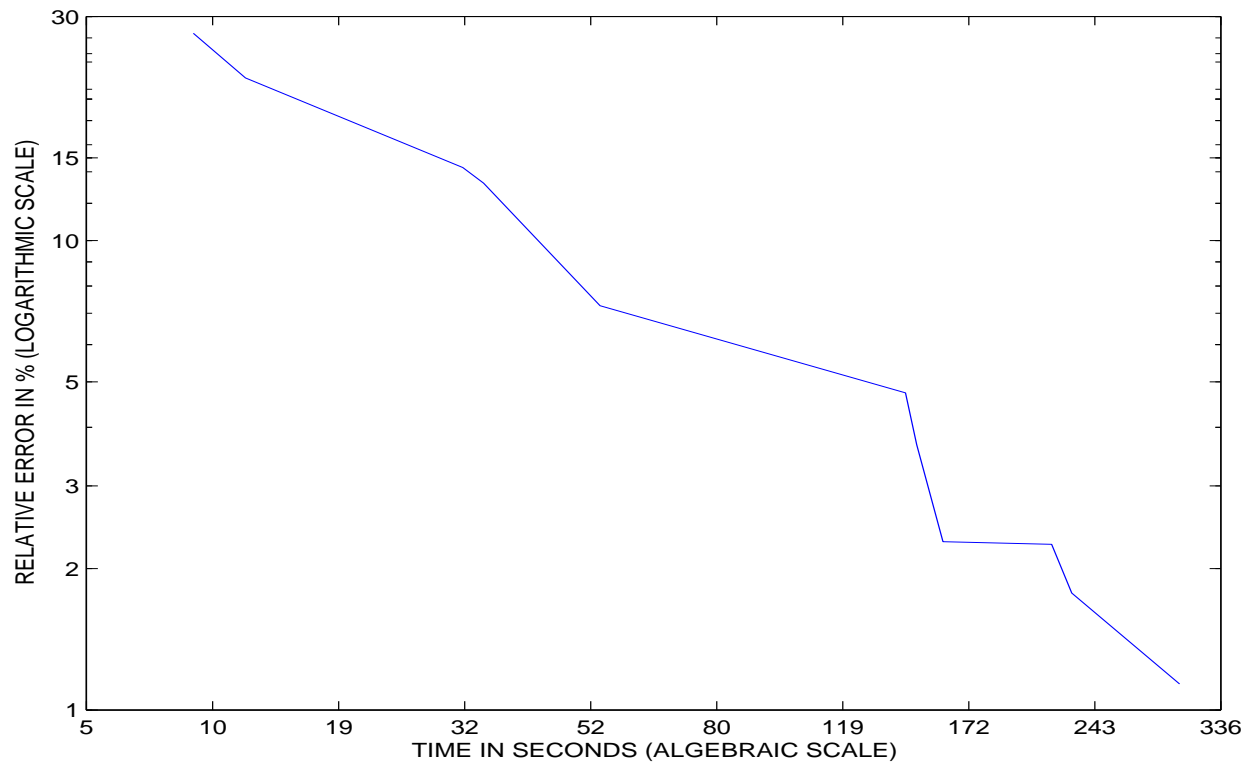
Nrdofs \approx 2.15 Million
 Total time \approx 50 minutes
 Memory* \approx 3.5 Gb
 p=4

*Memory = memory used by nonzero entries of stiffness matrix
 In core computations, IBM Power4 1.3 Ghz processor.

8. PERFORMANCE OF THE TWO GRID SOLVER

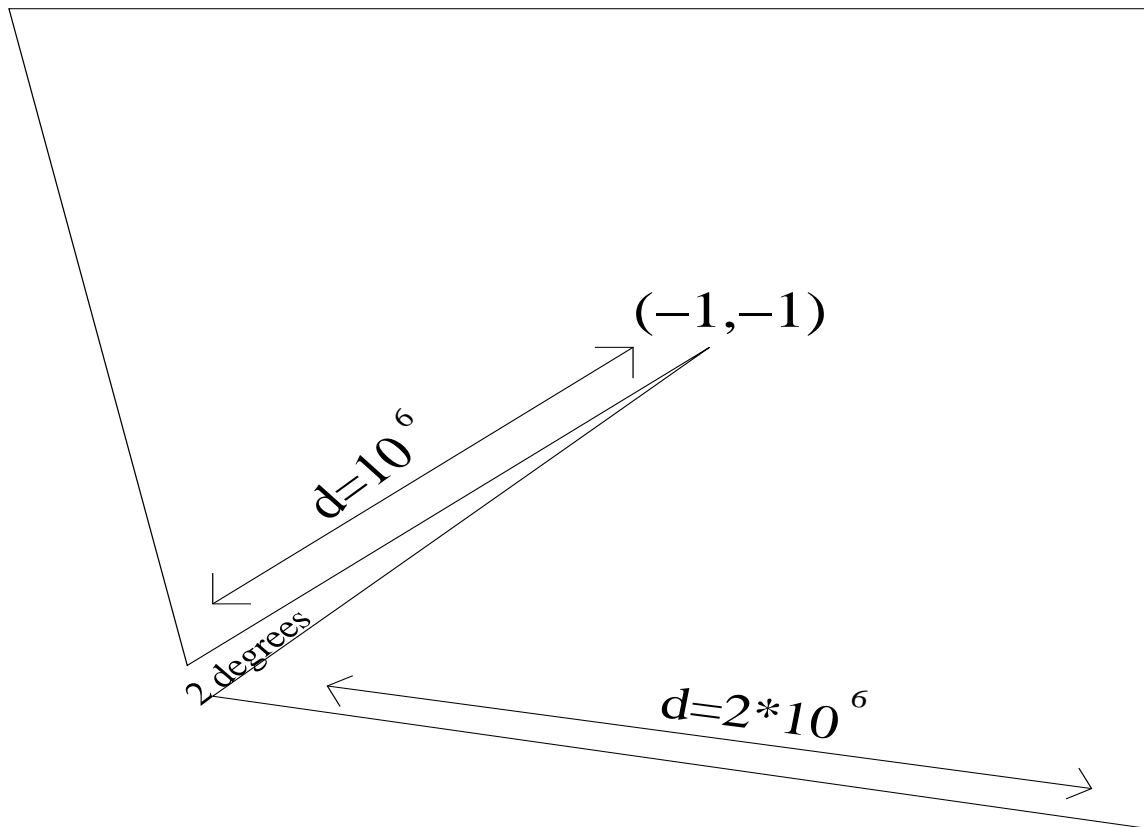
Convergence history

3D shock like solution example.
Scales: ERROR VS TIME.



9. ELECTROMAGNETIC APPLICATIONS

Edge diffraction example (Baker-Hughes): Electrostatics

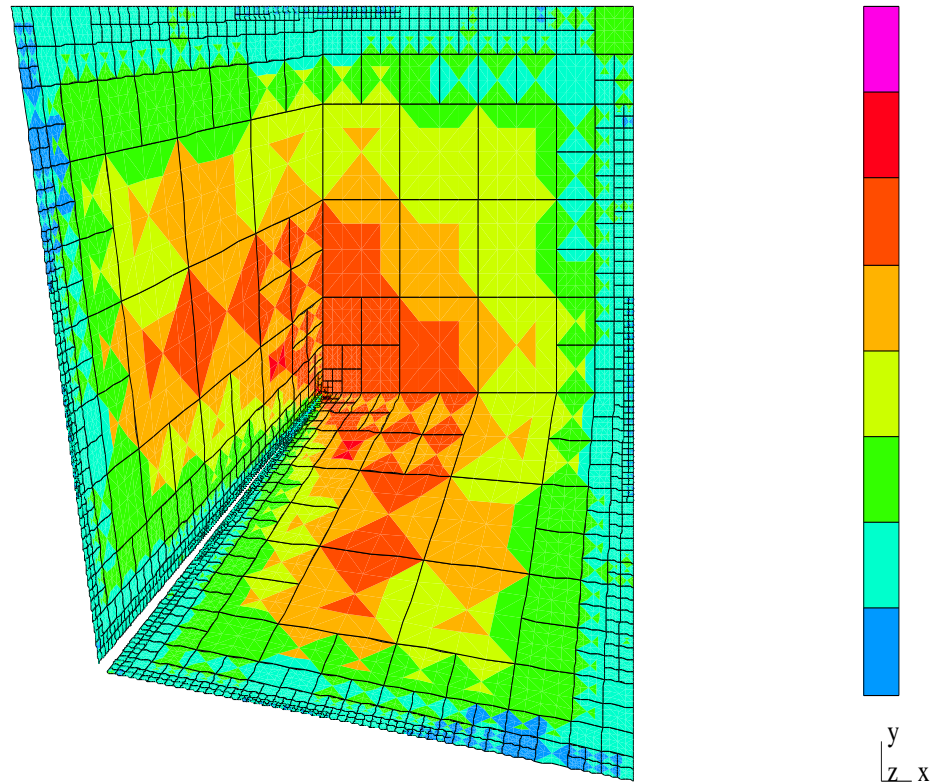


Dirichlet Boundary Conditions
 $u(\text{boundary}) = -\ln r, r = \sqrt{x^2 + y^2}$

9. ELECTROMAGNETIC APPLICATIONS

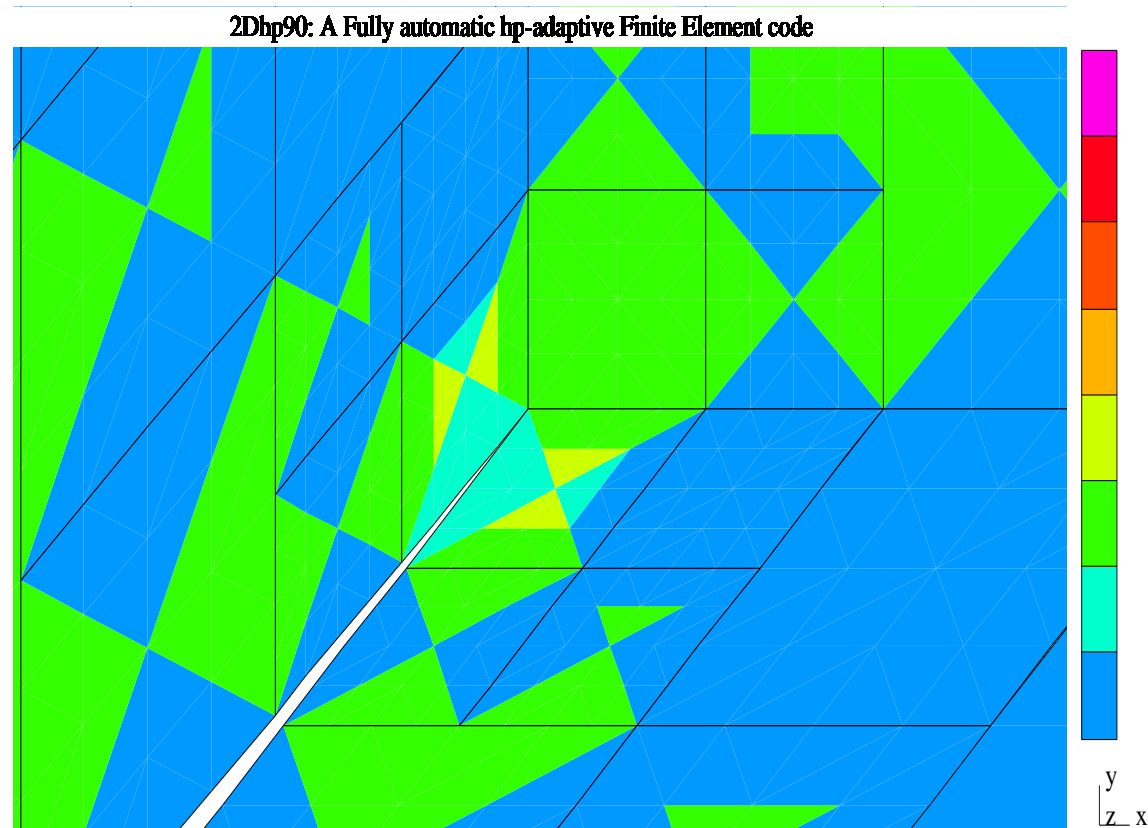
Edge diffraction example: final *hp*-grid, Zoom = 1

2Dhp90: A Fully automatic *hp*-adaptive Finite Element code



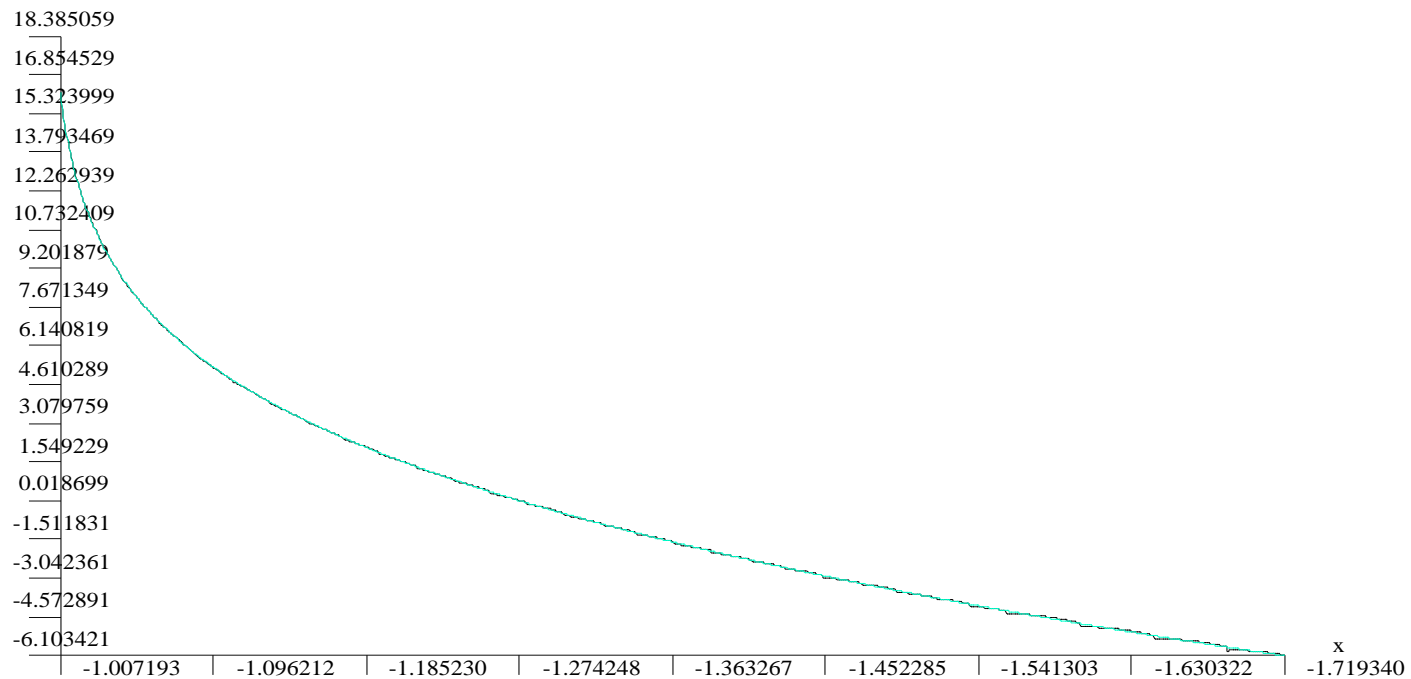
9. ELECTROMAGNETIC APPLICATIONS

Edge diffraction example: final *hp*-grid, Zoom = 10^{13}



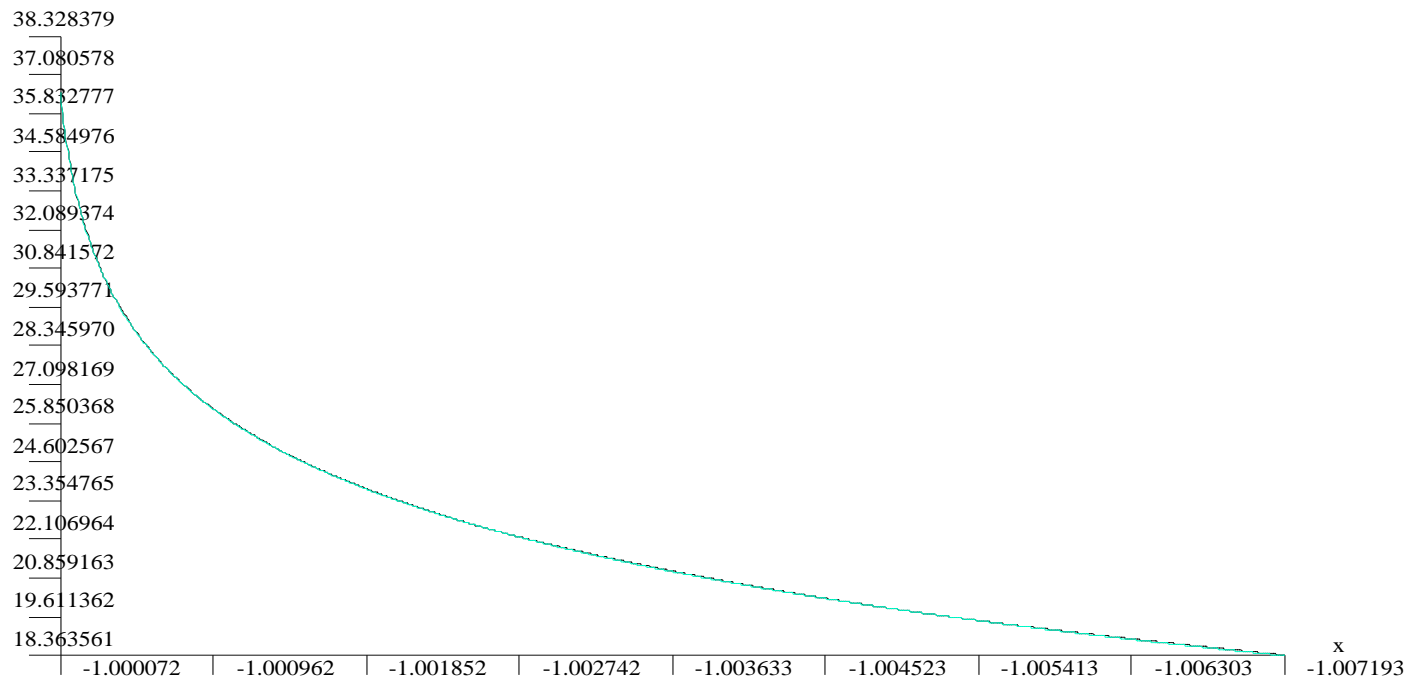
9. ELECTROMAGNETIC APPLICATIONS

Edge diffraction example: Comparison between exact and approximate solution at distances 0.01-1 from the singularity



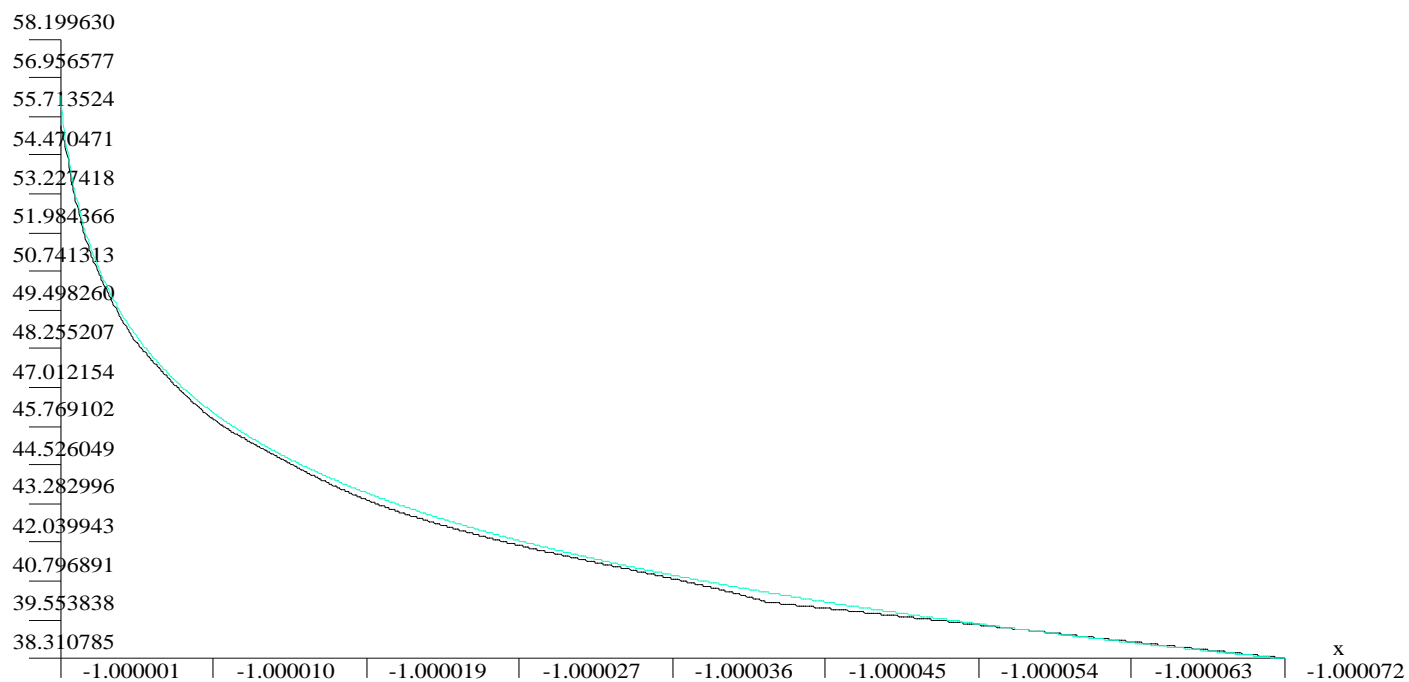
9. ELECTROMAGNETIC APPLICATIONS

Edge diffraction example: Comparison between exact and approximate solution at distances 0.0001-0.01 from the singularity



9. ELECTROMAGNETIC APPLICATIONS

Edge diffraction example: Comparison between exact and approximate solution at distances 0.000001-0.0001 from the singularity

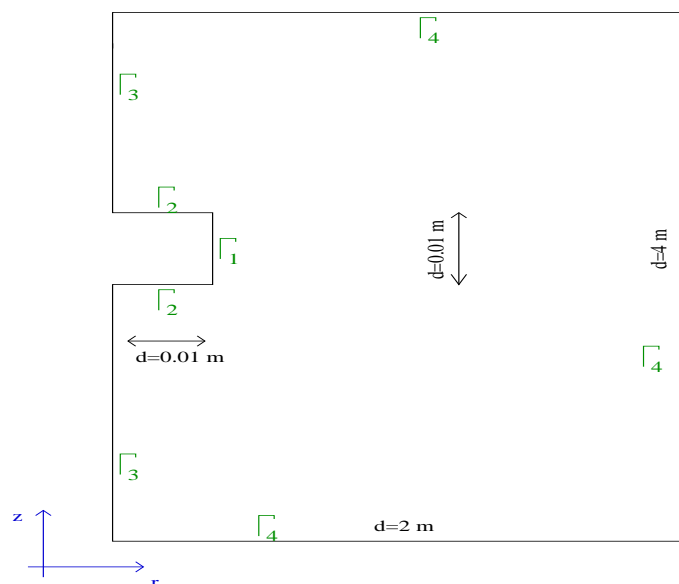


9. ELECTROMAGNETIC APPLICATIONS

Time Harmonic Maxwell's Equations

$$\nabla \times \mathbf{E} = -j\mu\omega\mathbf{H}$$

$$\nabla \times \mathbf{H} = j\omega\epsilon\mathbf{E} + \sigma\mathbf{E}$$



Reduced Wave Equation:

$$\nabla \times \left(\frac{1}{\mu} \nabla \times \mathbf{E} \right) - (\omega^2\epsilon - j\omega\sigma)\mathbf{E} = -j\omega\mathbf{J}^{imp}$$

Boundary Conditions (BC):

Dirichlet BC at a PEC surface:

$$\mathbf{n} \times \mathbf{E} = 0 \text{ on } \Gamma_2 \cup \Gamma_4$$

Neumann BC's:

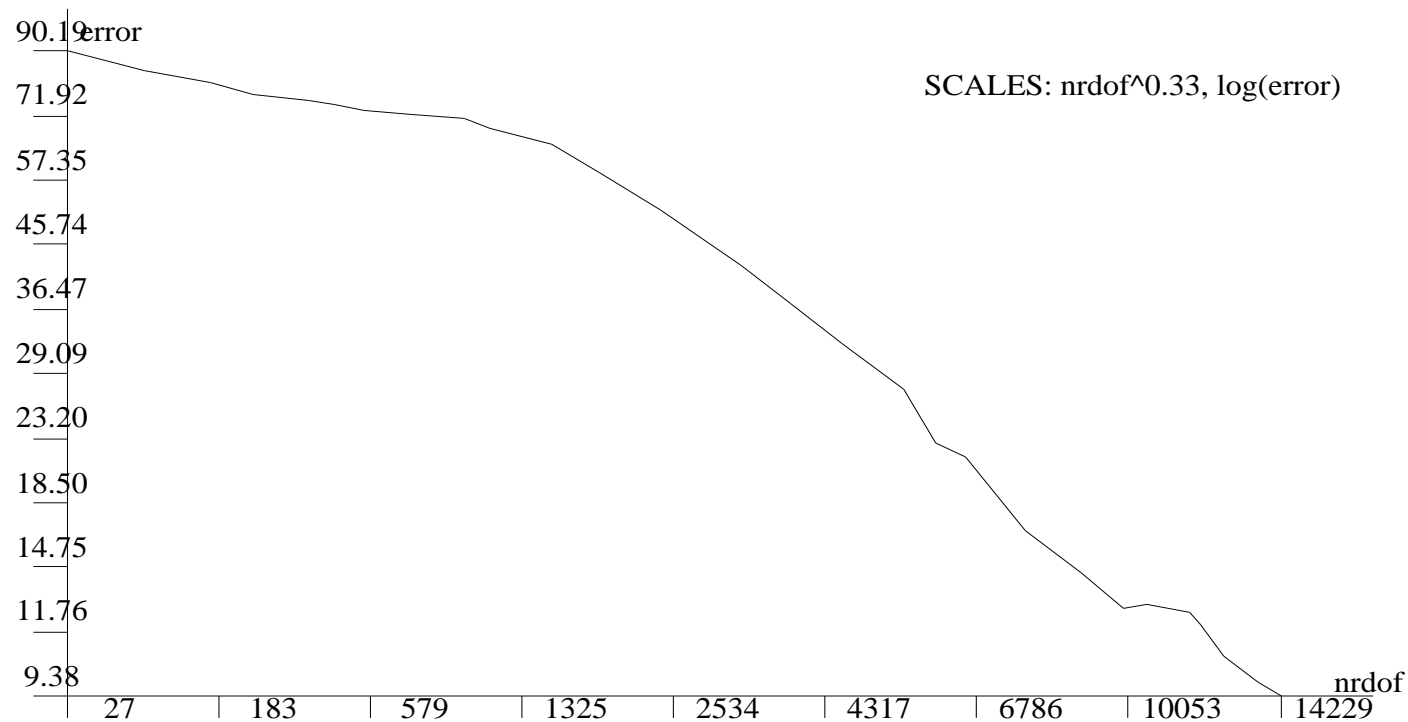
$$\mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E} = -j\omega \text{ on } \Gamma_1$$

$$\mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E} = 0 \text{ on } \Gamma_3$$

9. ELECTROMAGNETIC APPLICATIONS

Battery example: Convergence history

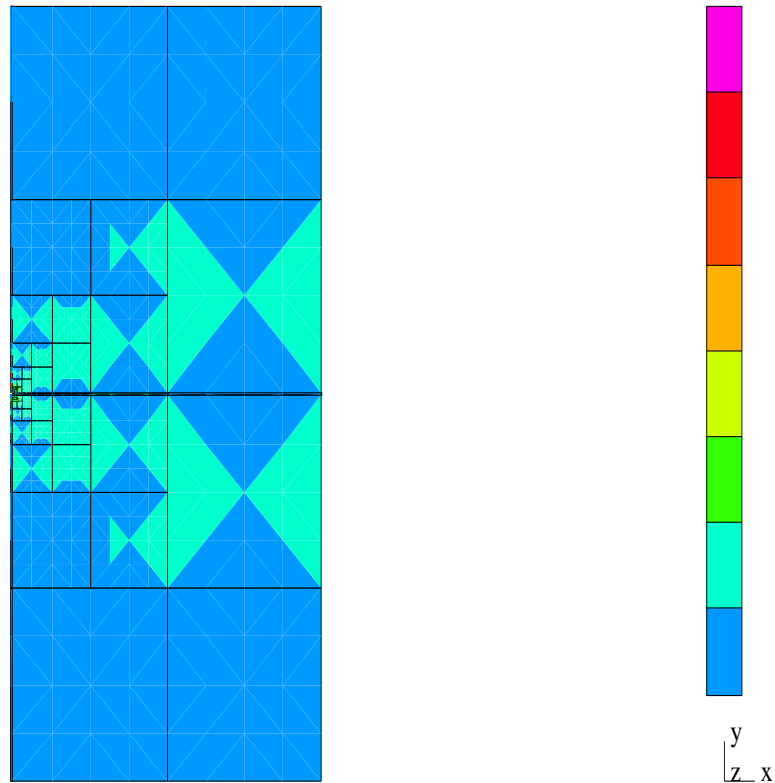
2Dhp90: A Fully automatic hp-adaptive Finite Element code



9. ELECTROMAGNETIC APPLICATIONS

Battery example: final *hp*-grid, Zoom = 1

2Dhp90: A Fully automatic *hp*-adaptive Finite Element code



9. ELECTROMAGNETIC APPLICATIONS

Why the optimal grid is so bad?

Optimization is based on minimization of the ENERGY NORM of the error, given by:

$$\| error \|^2 = \int | error |^2 + \int | \nabla \times error |^2$$

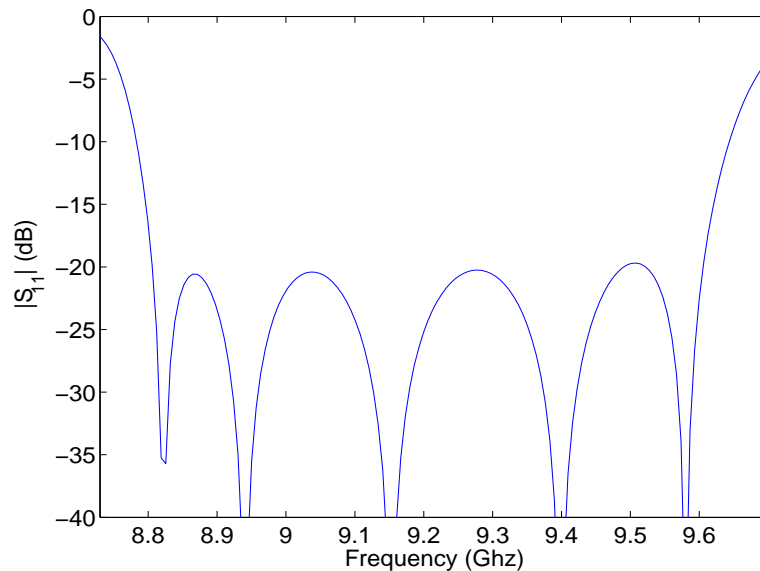
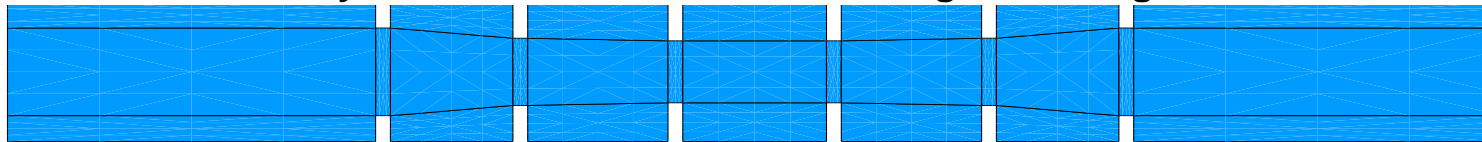
Interpretation of results:

- The grid is optimal for the selected refinement criteria,
- but our **refinement criteria is inadequate** for our purposes.

9. ELECTROMAGNETIC APPLICATIONS

Waveguide example with five iris

Geometry of a cross section of the rectangular waveguide



Return loss of the waveguide structure

H-plane five resonant iris filter.

Dominant mode (source): TE_{10} —mode.

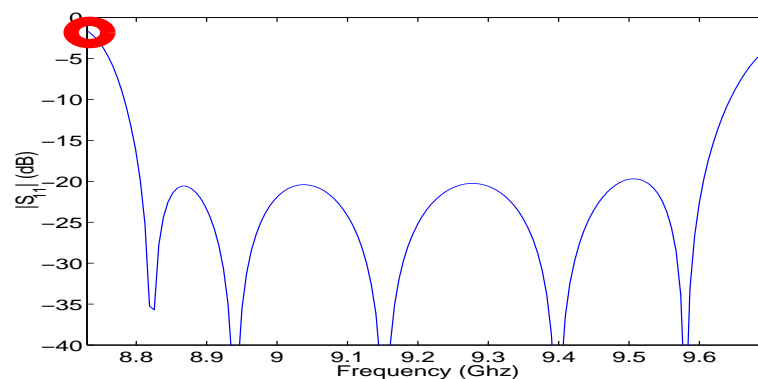
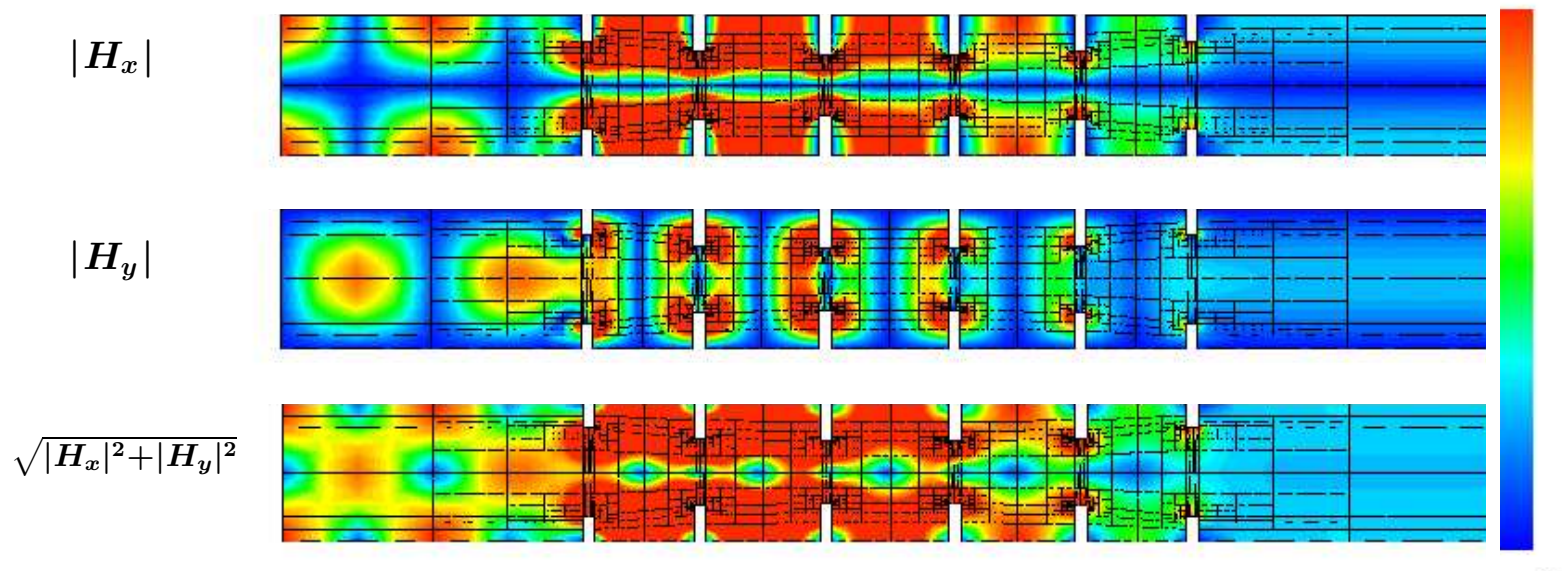
Dimensions $\approx 20 \times 2 \times 1$ cm.

Operating Frequency $\approx 8.8 - 9.6$ GHz

Cutoff frequency ≈ 6.56 GHz

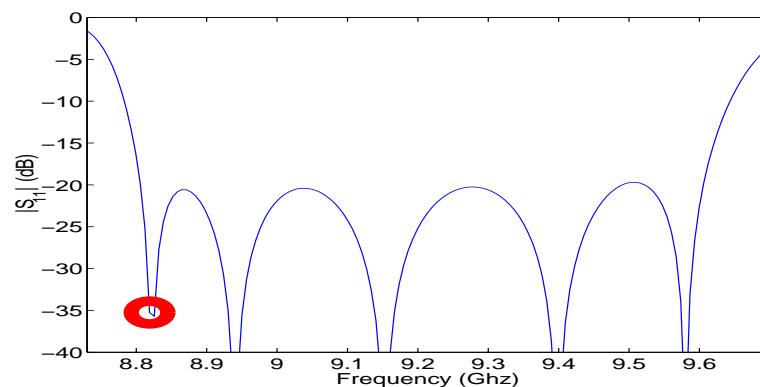
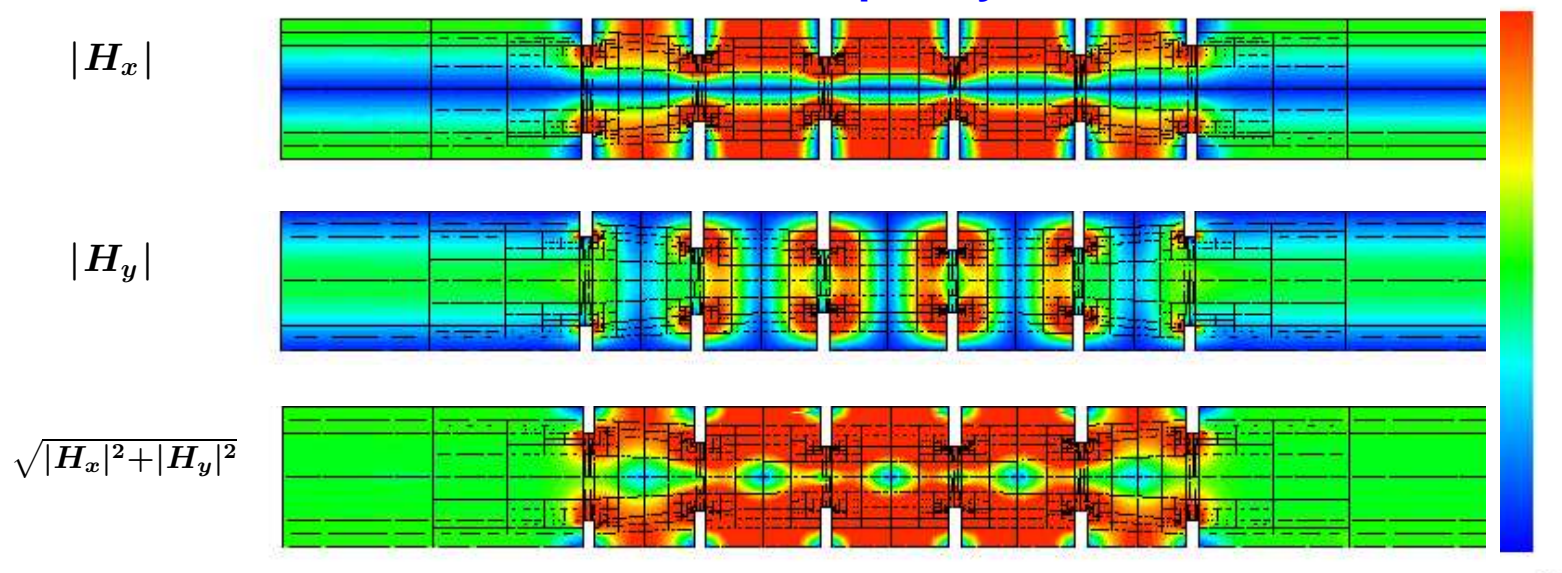
9. ELECTROMAGNETIC APPLICATIONS

FEM solution for frequency = 8.72 Ghz



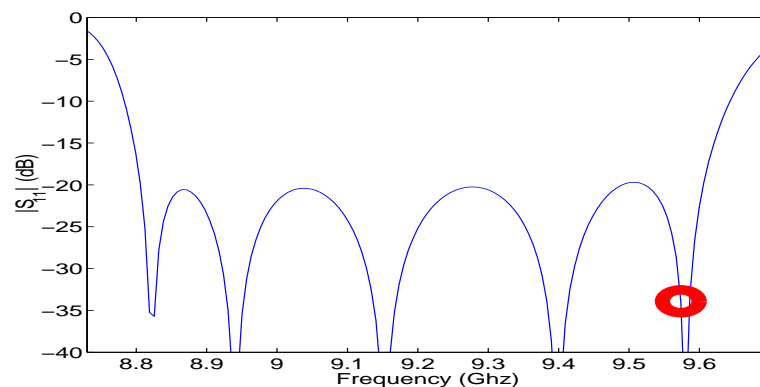
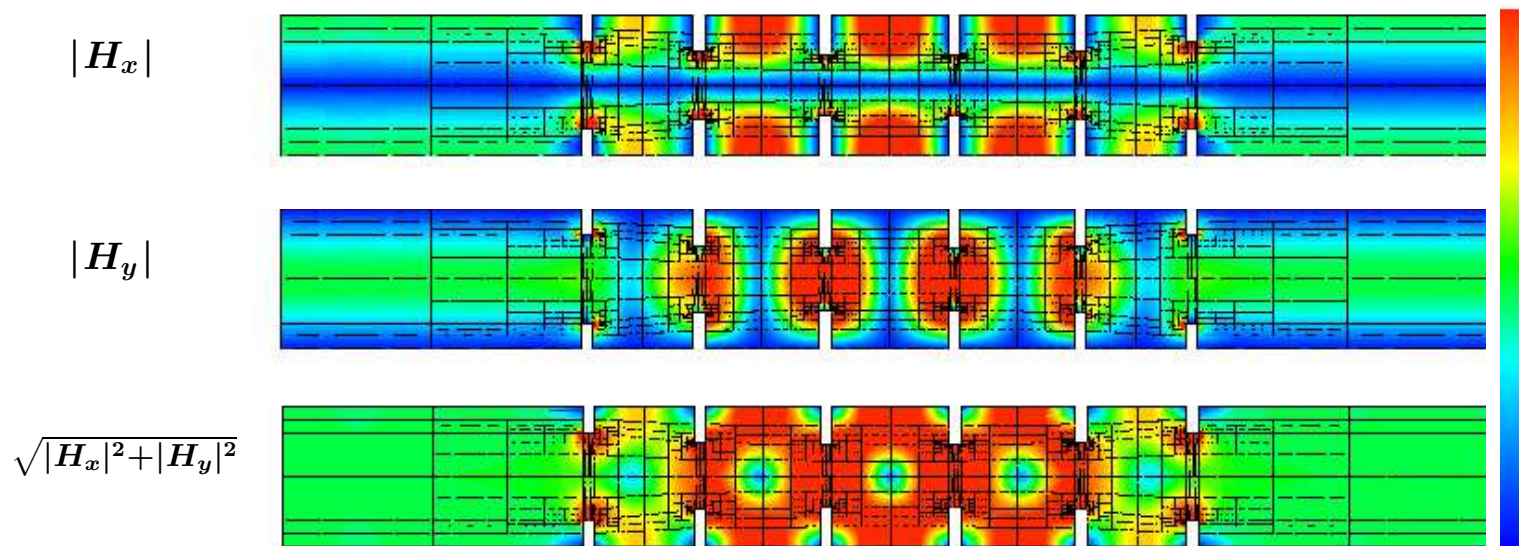
9. ELECTROMAGNETIC APPLICATIONS

FEM solution for frequency = 8.82 Ghz



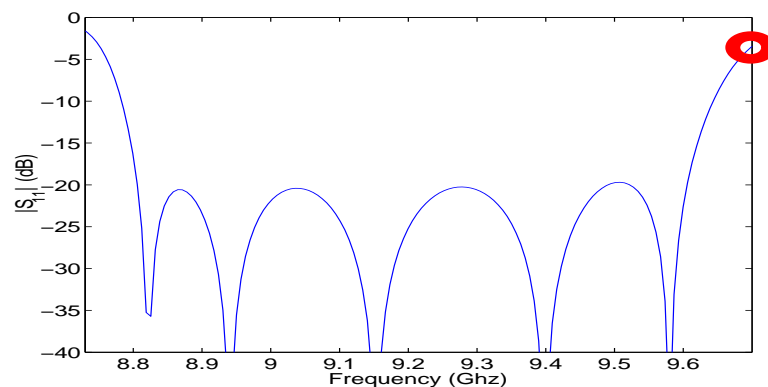
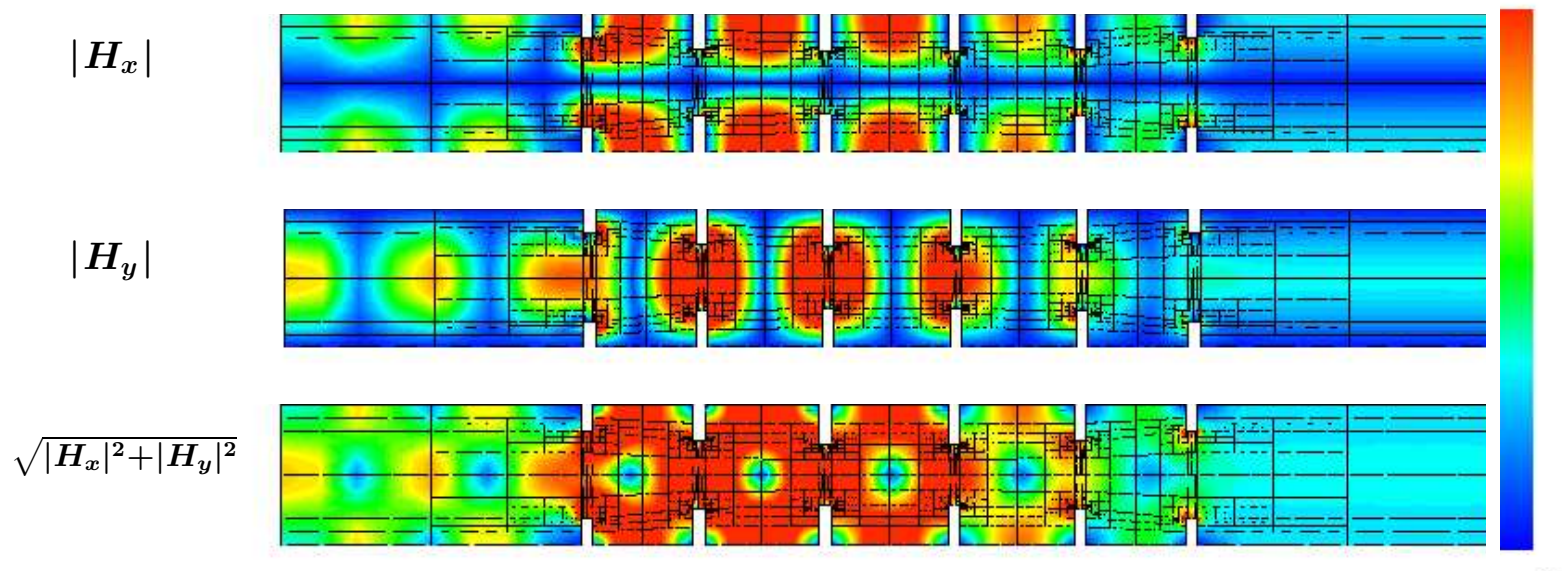
9. ELECTROMAGNETIC APPLICATIONS

FEM solution for frequency = 9.58 Ghz



9. ELECTROMAGNETIC APPLICATIONS

FEM solution for frequency = 9.71 Ghz



9. ELECTROMAGNETIC APPLICATIONS

Gridding Techniques for the Waveguide Problem

Our refinement technology incorporates:

An hp -adaptive algorithm

Low dispersion error

Small h is not enough

Large p required

Waveguide example: $p \approx 3$

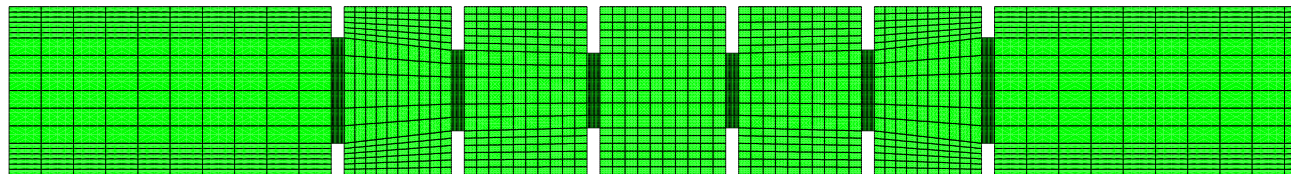
A two grid solver

Convergence of iterative solver

Insensitive to p -enrichment ($1 \leq p \leq 4$)

Coarse grid sufficiently fine

Waveguide example: $\lambda/h \approx 9$



Limitations of the hp -strategy for wave propagation problems:

We need large p and small h .

9. ELECTROMAGNETIC APPLICATIONS

Griding Techniques for the Waveguide Problem

Does convergence (or not) of the two grid solver depends upon h and/or p ? How?

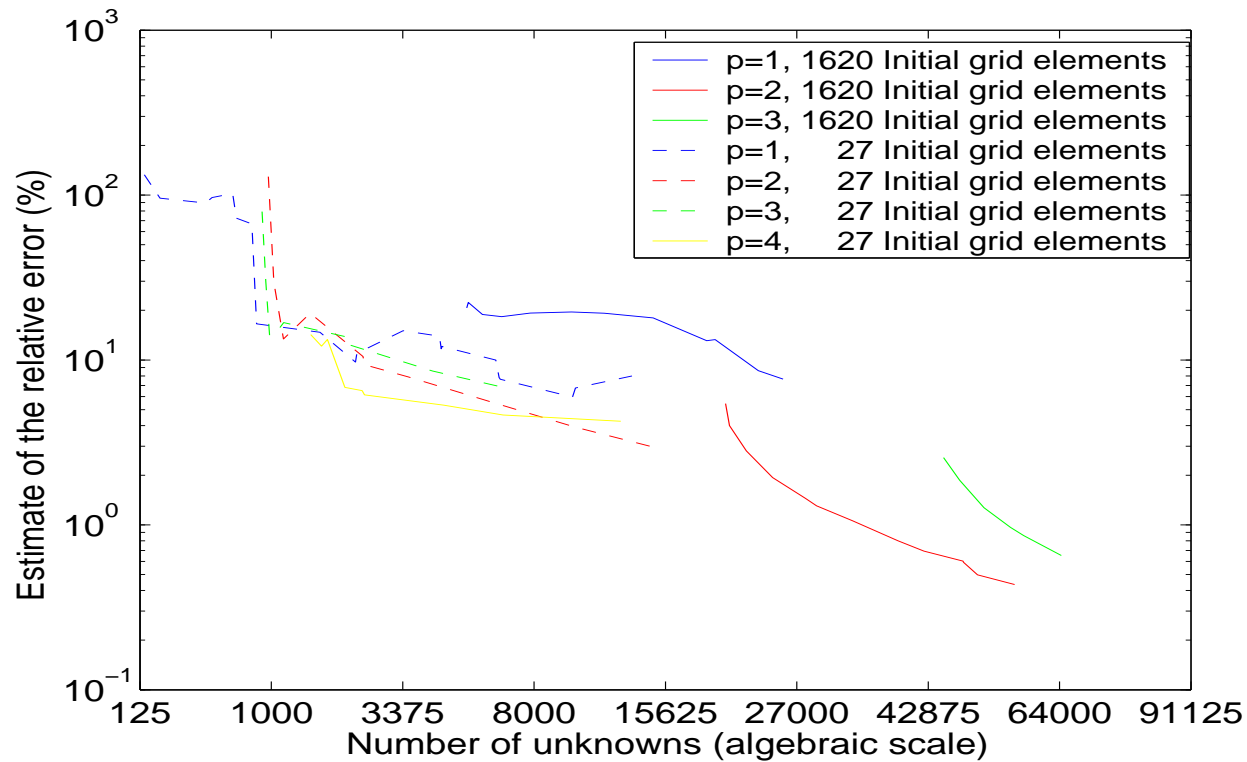
Convergence of two grid solver	$p = 1$	$p = 2$	$p = 3$	$p = 4$
Nr. of elements per $\lambda = 7, 13$	YES	YES	YES	YES
Nr. of elements per $\lambda = 7, 11$	NO	NO	NO	YES
Nr. of elements per $\lambda = 6, 13$	NO	NO	NO	NO

Convergence (or not) of the two grid solver is (almost) insensitive to p -enrichment.

9. ELECTROMAGNETIC APPLICATIONS

Gridding Techniques for the Waveguide Problem

Convergence history for different initial grids

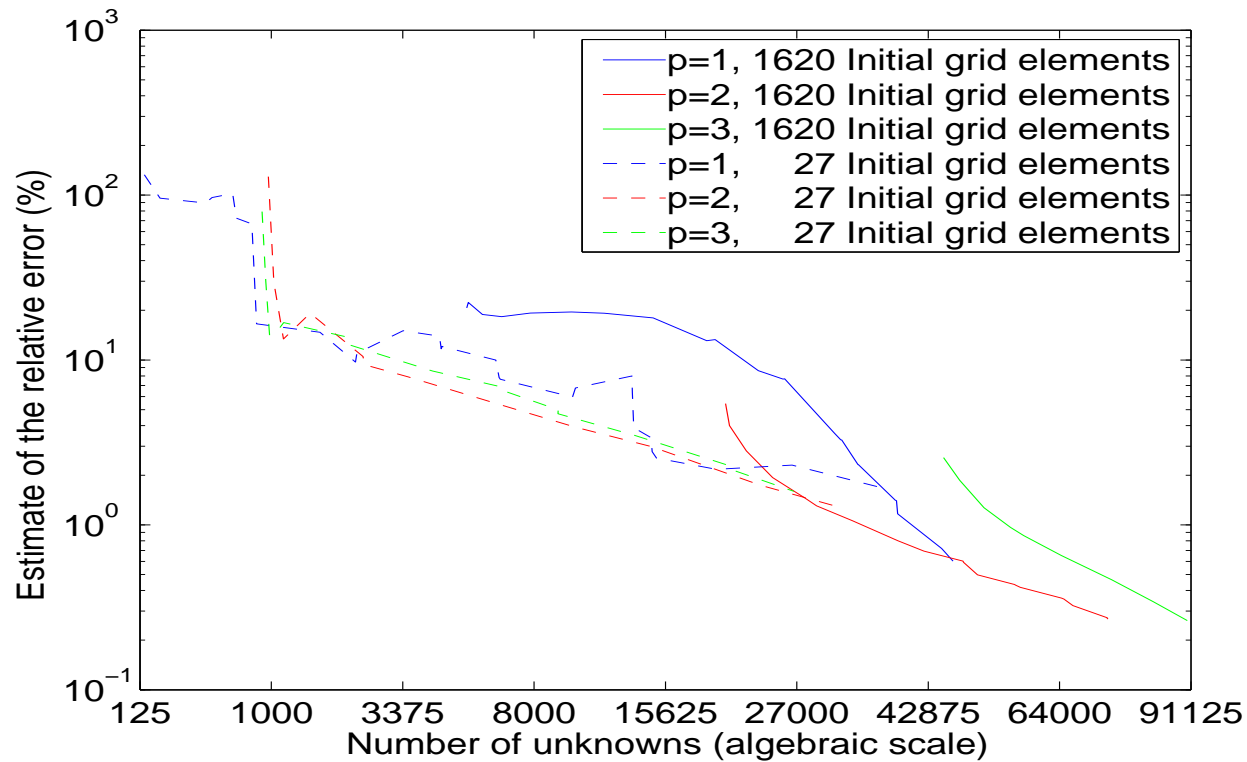


Conclusion : We need to control the dispersion error.

9. ELECTROMAGNETIC APPLICATIONS

Gridding Techniques for the Waveguide Problem

Convergence history for different initial grids



Conclusion : Do we need to control the dispersion error?

10. CONCLUSIONS AND FUTURE WORK

- **Exponential convergence** is achieved for real world problems by using a fully automatic hp -adaptive strategy.
- **Multigrid** for highly nonuniform hp -adaptive grids is an **efficient** iterative solver.
- It is possible to guide hp -adaptivity with partially converged solutions.
- There is a compromise between large p and small h on the design of the initial grid.
- **This numerical method can be applied to a variety of real world EM problems.**

10. CONCLUSIONS AND FUTURE WORK

Completed tasks

- Designed and implemented a 2D and 3D version of the two grid solver for elliptic problems.
- Studied numerically the 2D and 3D versions of the two grid solver.
- Designed, studied and implemented a two grid solver for 2D Maxwell's equations.
- Studied and designed an error estimator for a two grid solver for Maxwell's equations.
- Studied performance of different smoothers (in context of the two grid solver) for Maxwell's equations.
- Designed, studied, and implemented a flexible CG/GMRES method that is suitable to accelerate the two grid solver for Maxwell's equations.
- Developed a convergence theory for all algorithms mentioned above.
- Applied the *hp*-adaptive strategy combined with the two grid solver in order to solve a number of problems related to waveguide filters design, and modeling of LWD electromagnetic measuring devices.

Future Tasks

- Solve the 3D Fickera problem using *hp*-adaptivity and the two grid solver.
- Implement and study a two grid solver for 3D Maxwell's equations.
- Utilize this technology to solve a 3D model problem related to Radar Cross Section (RCS) analysis.
- Write and defend dissertation.

Completion date

- NOV 2003
- DEC 2003
- JAN 2004
- MAR 2004