

Adaptive boundary element methods in industrial applications

Günther Of, Olaf Steinbach, Wolfgang Wendland

Outline

- Mixed boundary value problems of potential theory
- Symmetric **Galerkin** boundary integral equation formulation
- Realization of the boundary integral operators by the **Fast Multipole Method**
- Fast Multipole Method for **Gadaptive meshes**
- **Preconditioning** of the matrices of the hypersingular operator and the single layer potential
- Example: spray painting
- **Evaluation** of representation formula **on demand**
- Experimental setup and errors
- **Error estimator**

Mixed boundary value problem

Laplace equation:

$$\begin{aligned}
 -\Delta u(x) &= 0 && \text{for } x \in \Omega \subset \mathbb{R}^3, \\
 u(x) &= g_D(x) && \text{for } x \in \Gamma_D, \\
 t(x) := (T_x u)(x) = (\partial_n u)(x) &= g_N(x) && \text{for } x \in \Gamma_N.
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Representation formula:

$$u(x) = \int_{\Gamma} [U^*(x, y)]^{\top} t(y) ds_y - \int_{\Gamma} [T_y U^*(x, y)]^{\top} u(y) ds_y \quad \text{for } x \in \Omega.$$

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Calderon projector for the Cauchy data $u(x)$ and $t(x)$:

$$\begin{pmatrix} u \\ t \end{pmatrix} = \begin{pmatrix} \frac{1}{2}I - K & V \\ D & \frac{1}{2}I + K' \end{pmatrix} \begin{pmatrix} u \\ t \end{pmatrix} \quad \text{on } \Gamma$$

Boundary integral operators:

$$(Vt)(x) = \int_{\Gamma} [U^*(x, y)]^{\top} t(y) ds_y, \quad (Du)(x) = -T_x \int_{\Gamma} [T_y U^*(x, y)]^{\top} u(y) ds_y,$$

$$(Ku)(x) = \int_{\Gamma} [T_y U^*(x, y)]^{\top} u(y) ds_y, \quad (K't)(x) = \int_{\Gamma} [T_x U^*(x, y)]^{\top} t(y) ds_y.$$

Symmetric boundary integral formulation

Symmetric boundary integral formulation (Sirtori '79, Costabel '87):

$$\begin{aligned}
 (V\tilde{t})(x) - (K\tilde{u})(x) &= \left(\frac{1}{2}I + K\right)\tilde{g}_D(x) - (V\tilde{g}_N)(x) \quad \text{for } x \in \Gamma_D, \\
 (K'\tilde{t})(x) + (D\tilde{u})(x) &= \left(\frac{1}{2}I - K'\right)\tilde{g}_N(x) - (D\tilde{g}_D)(x) \quad \text{for } x \in \Gamma_N.
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Galerkin discretization with piecewise constant (φ_l) and piecewise linear (ψ_i) trial and test functions leads to a **system of linear equations**:

$$\begin{pmatrix} V_h & -K_h \\ K'_h & D_h \end{pmatrix} \begin{pmatrix} \tilde{t}_h \\ \tilde{u}_h \end{pmatrix} = \begin{pmatrix} \underline{f}_N \\ \underline{f}_D \end{pmatrix}.$$

Single Galerkin blocks for $k, l = 1, \dots, m$ and $i, j = 1, \dots, \tilde{m}$

$$\begin{aligned}
 V_h[l, k] &= \langle V\varphi_k, \varphi_l \rangle_{\Gamma_D}, & K_h[l, i] &= \langle K\psi_i, \varphi_l \rangle_{\Gamma_D}, \\
 K'_h[j, k] &= \langle K'\varphi_k, \psi_j \rangle_{\Gamma_N}, & D_h[j, i] &= \langle D\psi_i, \psi_j \rangle_{\Gamma_N}.
 \end{aligned}$$

Potential and Fast Multipole Method

Double layer potential:

$$(Ku)(x) = \frac{1}{4\pi} \int_{\Gamma} \frac{\partial}{\partial n_y} \frac{1}{|x-y|} u(y) ds_y \quad \text{for } x \in \Gamma$$

Single layer potential:

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In the farfield **numerical quadrature**:

$$\frac{1}{4\pi} \sum_{k=1}^N \int_{\tau_k} t(y) \frac{1}{|x-y|} ds_y \approx \frac{1}{4\pi} \sum_{k=1}^N \sum_{s=1}^{N_g} \underbrace{\Delta_k \omega_{k,s} t(y_{k,s})}_{=q_{k,s}} \frac{1}{|x-y_{k,s}|}.$$

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Fast Multipole Method: [Rohklin 1984; Greengard, Rohklin 1987; ...]

Fast evaluation of potentials in many particle systems:

$$\Phi(x_j) = \sum_{i=1}^N \frac{q_i}{|x_j - y_i|} \quad \text{for } j = 1, \dots, M.$$

Idea of the Fast Multipole Method

- Starting point: evaluation of sums in many points x_j

$$\Phi(x_j) = \sum_{i=1}^N q_i k(x_j, y_i) \quad j = 1, \dots, M.$$

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- Effect on the sums:

$$\Phi(x_j) = f(x_j) \sum_{i=1}^N q_i g(y_i) \quad j = 1, \dots, M.$$

reduced effort: $\mathcal{O}(N + M)$.

Realization of the Fast Multipole Method

- Actual realization in the **Fast Multipole Method**:

$$k(x, y) = \frac{1}{|x - y|} \approx \sum_{n=0}^p \sum_{m=-n}^n |x|^n Y_n^{-m}(\hat{x}) \frac{Y_n^m(\hat{y})}{|y|^{n+1}}$$

- with **spherical harmonics** for $m \geq 0$

$$Y_n^{\pm m}(\hat{x}) = \sqrt{\frac{(n-m)!}{(n+m)!}} (-1)^m \frac{d^m}{d\hat{x}_3^m} P_n(\hat{x}_3) (\hat{x}_1 \pm i\hat{x}_2)^m.$$

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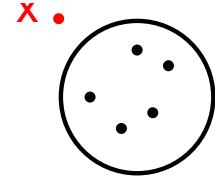
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- Effect on the sums:

$$\Phi(x_j) \approx \sum_{i \in NF(j)}^{1, N} q_i k(x_j, y_i) + \sum_{n=0}^p \sum_{m=-n}^n |x_j|^n Y_n^{-m}(\hat{x}_j) \underbrace{\sum_{i \in FF(j)}^{1, N} q_i \frac{Y_n^m(\hat{y}_i)}{|y_i|^{n+1}}}_{=L_n^m}.$$

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- **Multipole expansion** for $|x| > |y_j|$

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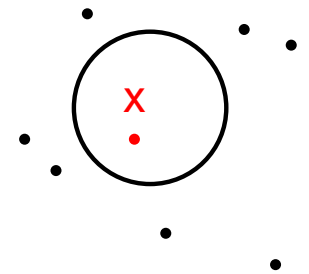
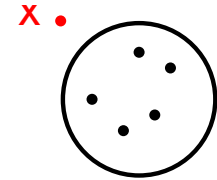
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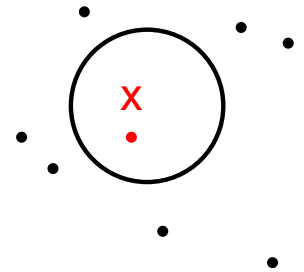
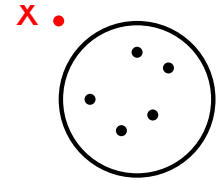
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 \Rightarrow **different farfield sums.**



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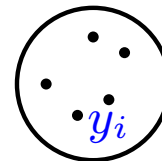
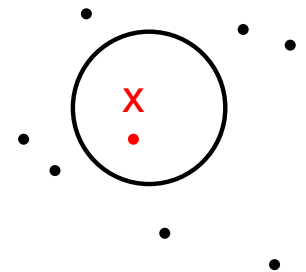
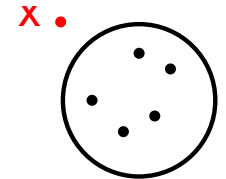
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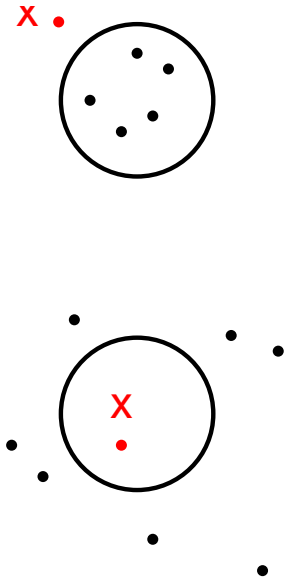
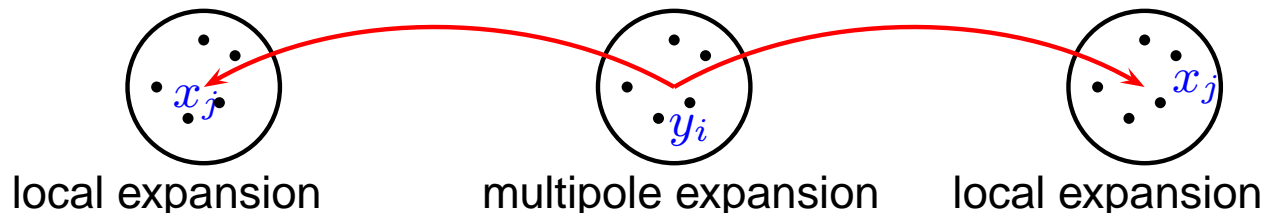
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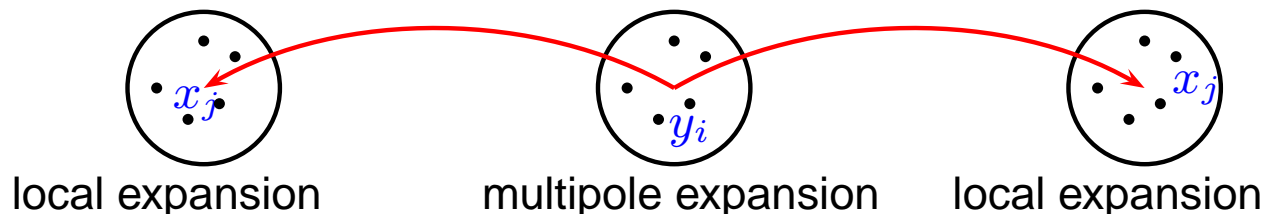
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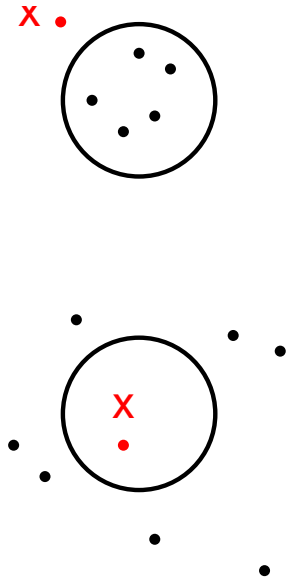
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- efficient computation by the use of a **hierarchical structure**

\Rightarrow complexity (time and memory) with error control: $\mathcal{O}(N \log^2 N)$ for $N = M$.



Adaptive version of the Fast multipole method

Adaptive versions already exist:

Cheng, Greengard, Rokhlin; Nabors, Kormsmeier, Leighton, White.

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Otherwise huge nearfield or huge tree (many expansions).

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Maintain the the **symmetry** of the matrices and **error control** in the Fast Multipole Method.

Lemma. *The Fast Multipole Method is **symmetric** for two points P_1 and P_2 with unit charge using a finite expansion degree, if the paths of the transformations of the expansions agree with each other except of the direction.*

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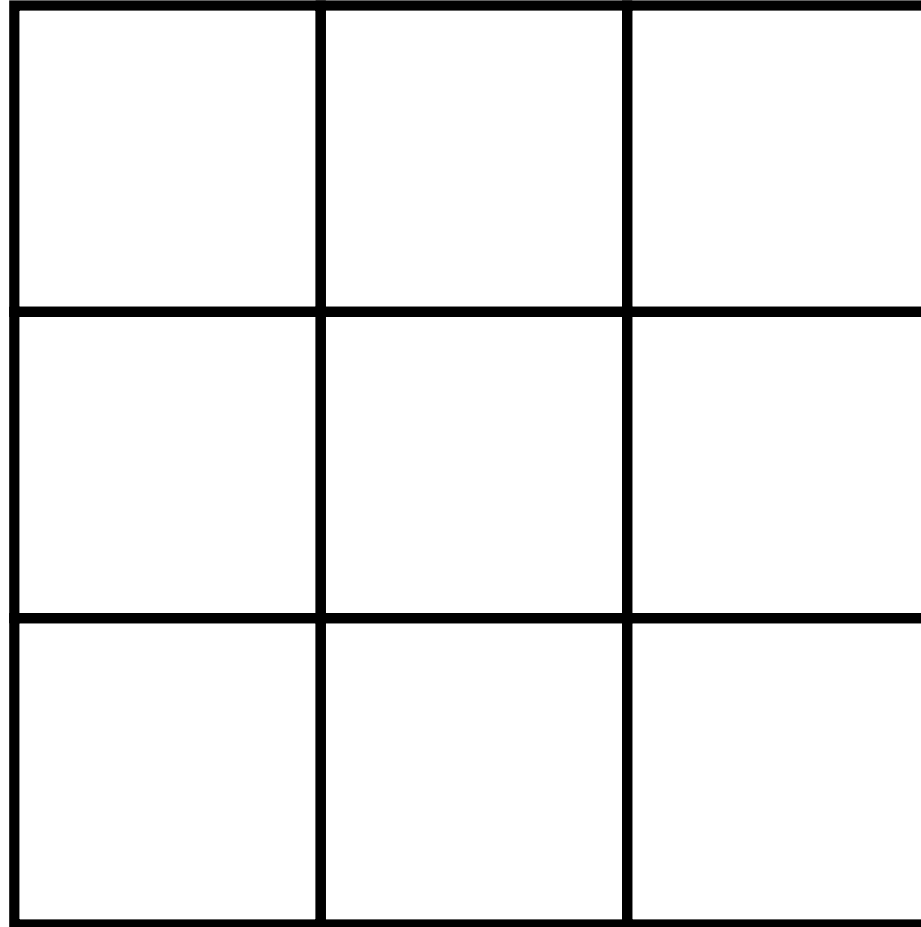
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⇒ Condition for the cluster tree: **symmetry of the nearfields.**

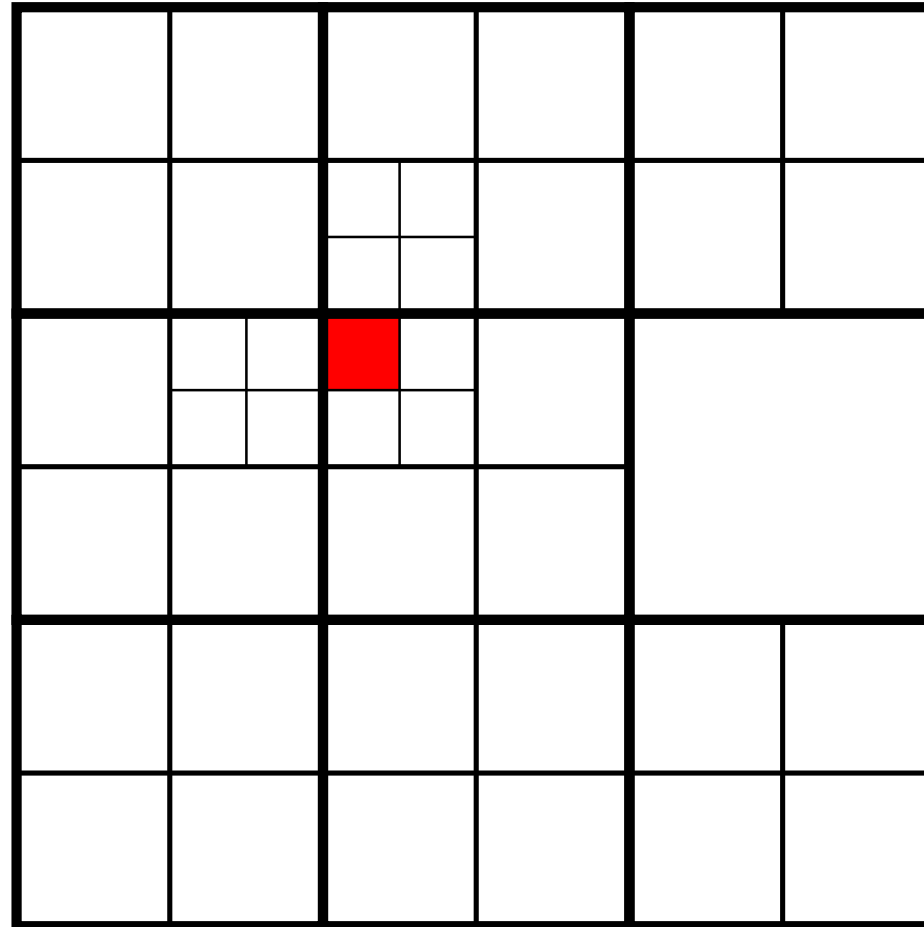
Construction of the adaptive cluster tree

Symmetry of the nearfields and distance control. Example:



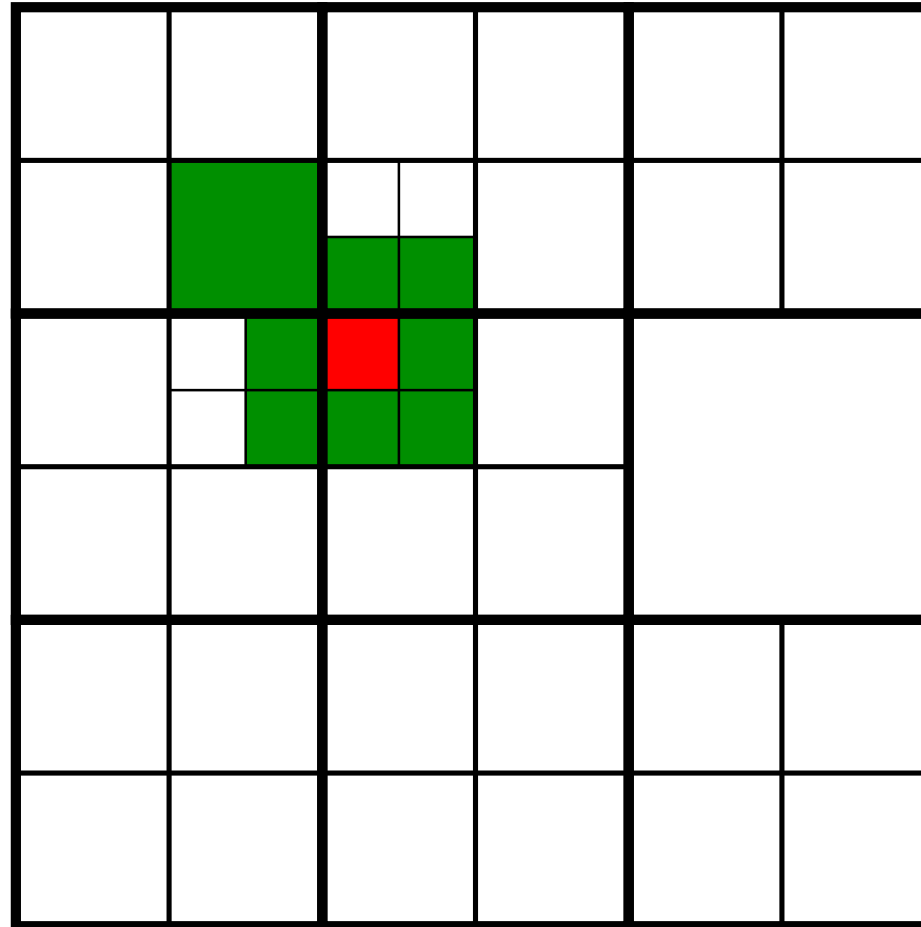
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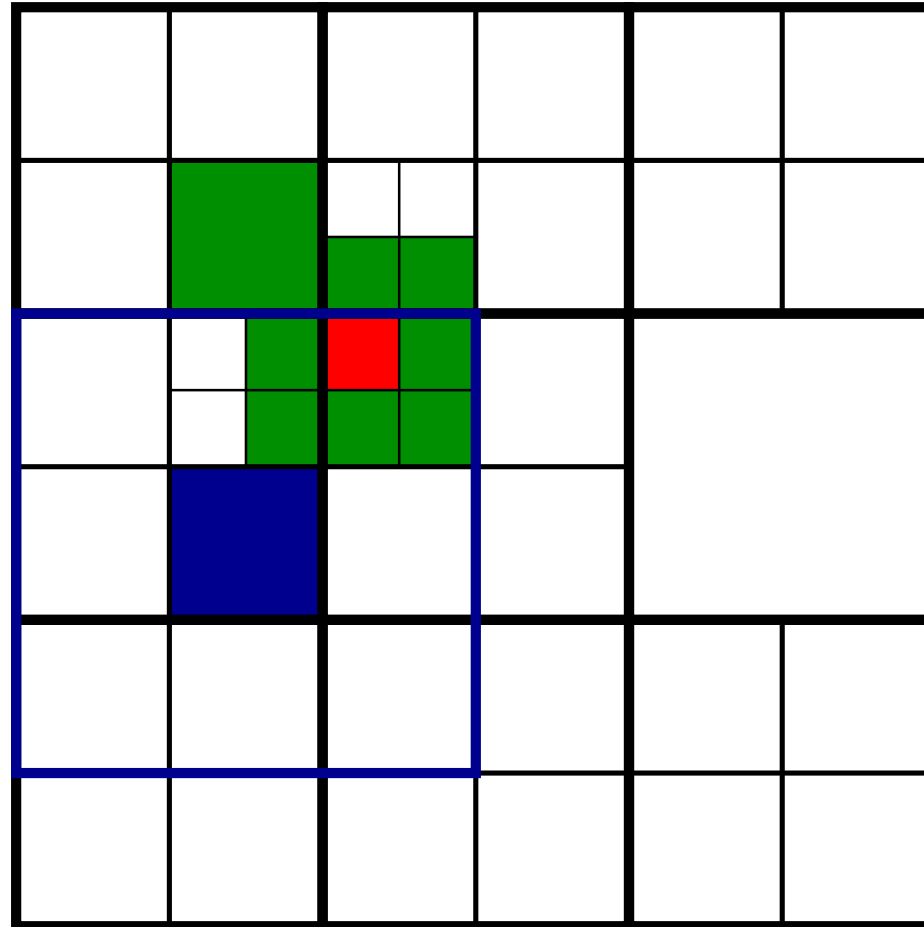
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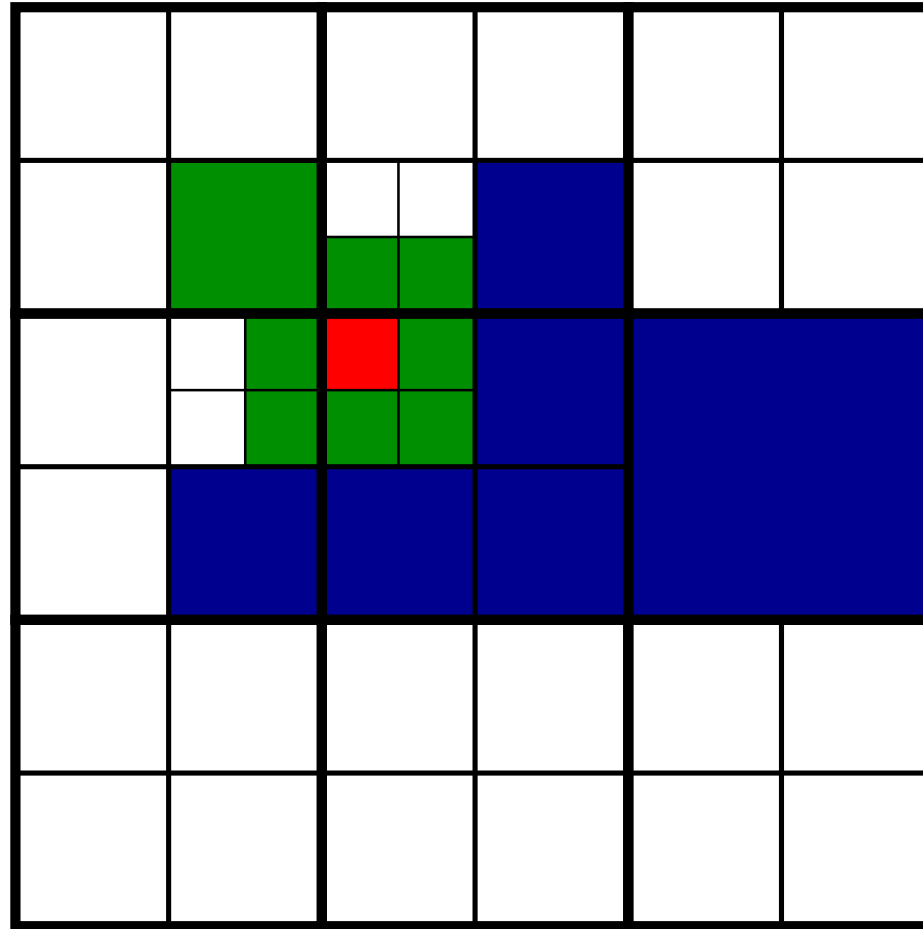
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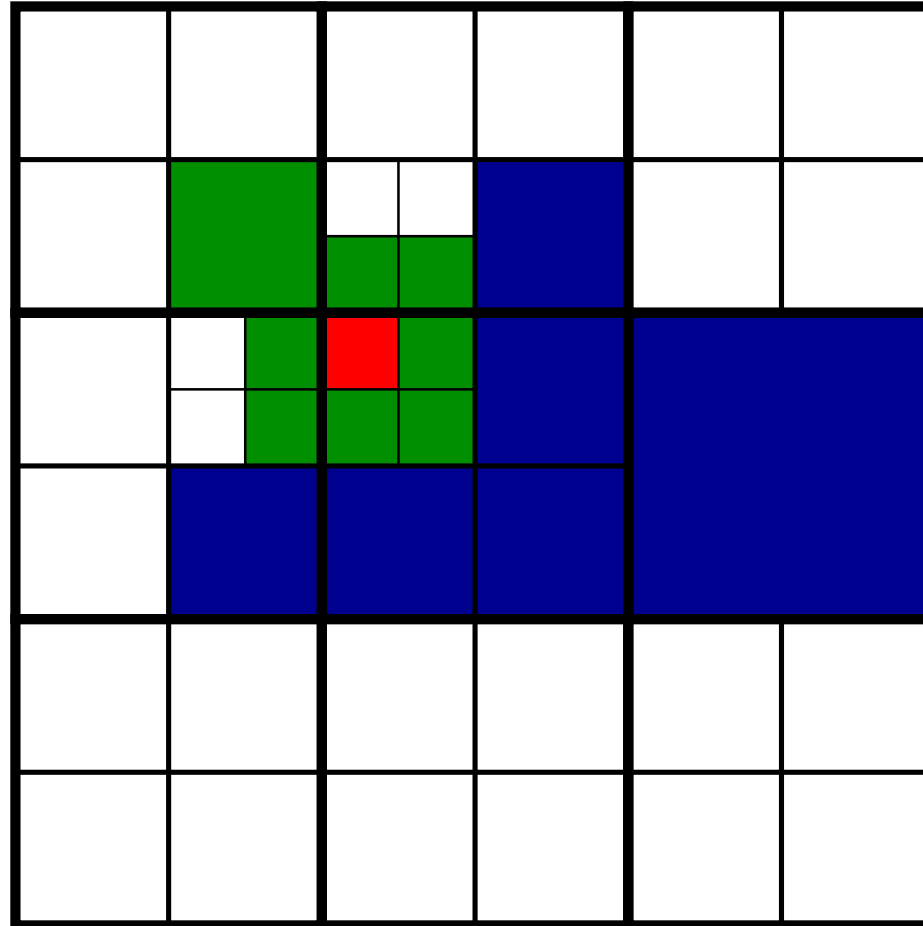
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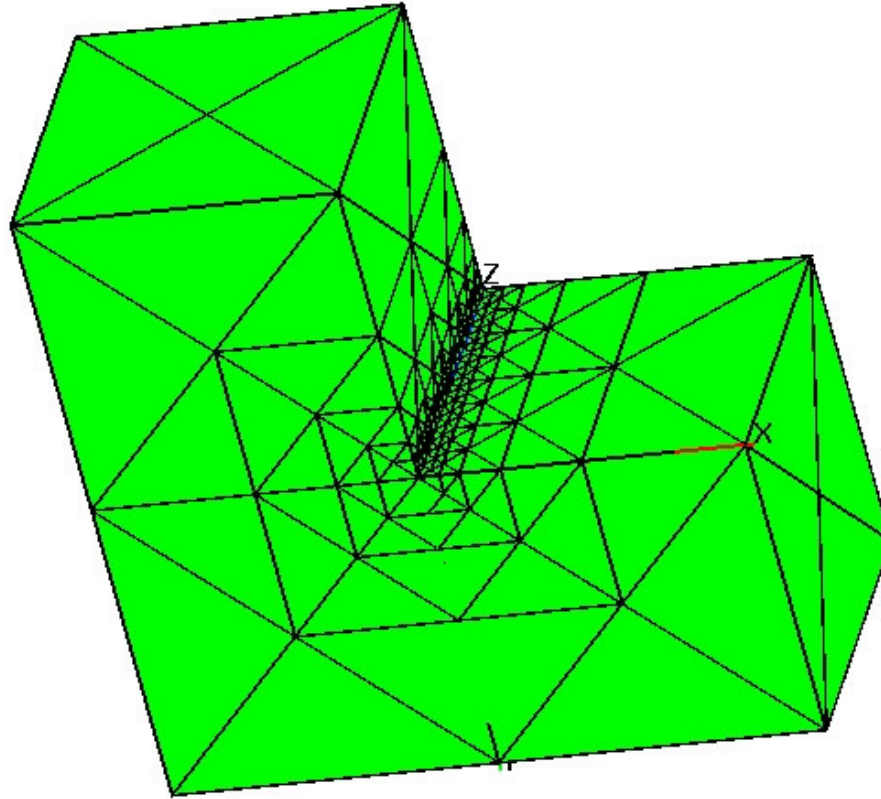
Symmetry of the nearfields and distance control. Example:



Postprocessing of the nearfields on son levels possible.

Example of an adaptive grid

Institut für Angewandte Analysis und Numerische Simulation
 SFB 404 Mehrfeldprobleme in der Kontinuumsmechanik
 Universität Stuttgart



Example of an adaptive Fast Multipole Method

level	# elements	setup	solving	nearfield	$ u(x^*) - u_h(x^*) $
0	338	5	2	26.91 %	1.676e-3
		6	1	40.42 %	1,677e-3
1	1352	20	12	6.22 %	4.101e-4
		58	3	26.43 %	4.103e-4
2	5408	60	42	1.77 %	1.025e-4
		113	13	4.51 %	1.021e-4
3	21632	203	188	0.44 %	2.699e-5
		183	43	0.67 %	2.674e-5
4	86528	720	849	0.11 %	6.414e-6
		556	284	0.16 %	6.561e-6
5	346112			%	
		1909	1596	0.04 %	1.610e-6

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$$\frac{c_1^V c_1^D}{1 + c_R} \langle V^{-1}v, v \rangle_\Gamma \leq \langle \tilde{D}v, v \rangle_\Gamma \leq \frac{1}{4(1 - c_R)} \langle V^{-1}v, v \rangle_\Gamma.$$

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- single layer potential with piecewise linear basis functions
- spectral equivalent **preconditioning matrix** $C_{\tilde{D}}$ for \tilde{D}_h :

$$C_{\tilde{D}}[j, i] = \langle V^{-1}\psi_i, \psi_j \rangle_\Gamma \quad \text{for } i, j = 1, \dots, M.$$

Approximation of the preconditioning matrix $C_{\tilde{D}}$

$$\tilde{C}_{\tilde{D}} = M_h^T V_h^{-1} M_h, \quad \tilde{C}_{\tilde{D}}^{-1} = M_h^{-1} V_h M_h^{-T}$$

with

$$V_h[j, i] = \langle V\psi_i, \psi_j \rangle_\Gamma, \quad M_h[j, i] = \langle \psi_i, \psi_j \rangle_\Gamma.$$

Artificial Multilevel Preconditioner

[Steinbach 2003]

Sequence of nested boundary element spaces (globally quasiuniform)

$$Z_0 \subset Z_1 \subset \dots \subset Z_J = Z_h \subset H^{-1/2}(\Gamma)$$

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L_2 projection:

$$Q_j w \in Z_j : \langle Q_j w, v_j \rangle_\Gamma = \langle w, v_j \rangle_\Gamma \quad \text{for all } v_j \in Z_j$$

Multilevel operator: [Bramble, Pasciak, Xu 1990]

$$A^s w := \sum_{j=0}^J h_j^{-2s} (Q_j - Q_{j-1}) w$$

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Spectral equivalence inequalities [Oswald 1998]

$$c_1 \|w\|_{H^{-1/2}(\Gamma)}^2 \leq \langle A^{-1/2} w, w \rangle_\Gamma \leq c_2 J^2 \|w\|_{H^{-1/2}(\Gamma)}^2 \quad \text{for all } w \in Z_J.$$

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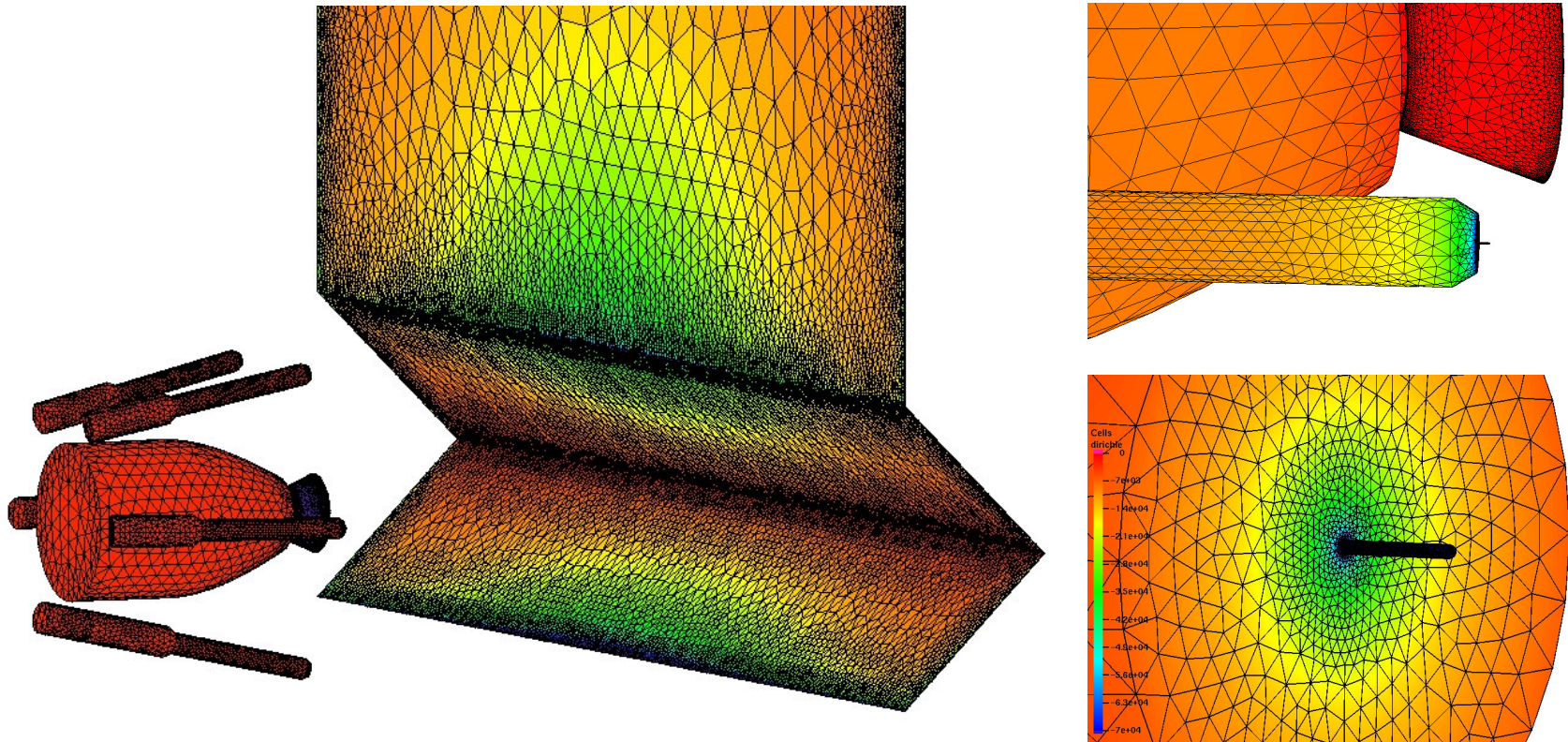
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The sequence of nested boundary element spaces is build from the **clustering of the boundary elements by a hierarchical structure** as used in the Fast BEM.

Extendable to **adaptive grids**.

Simulation of spray painting

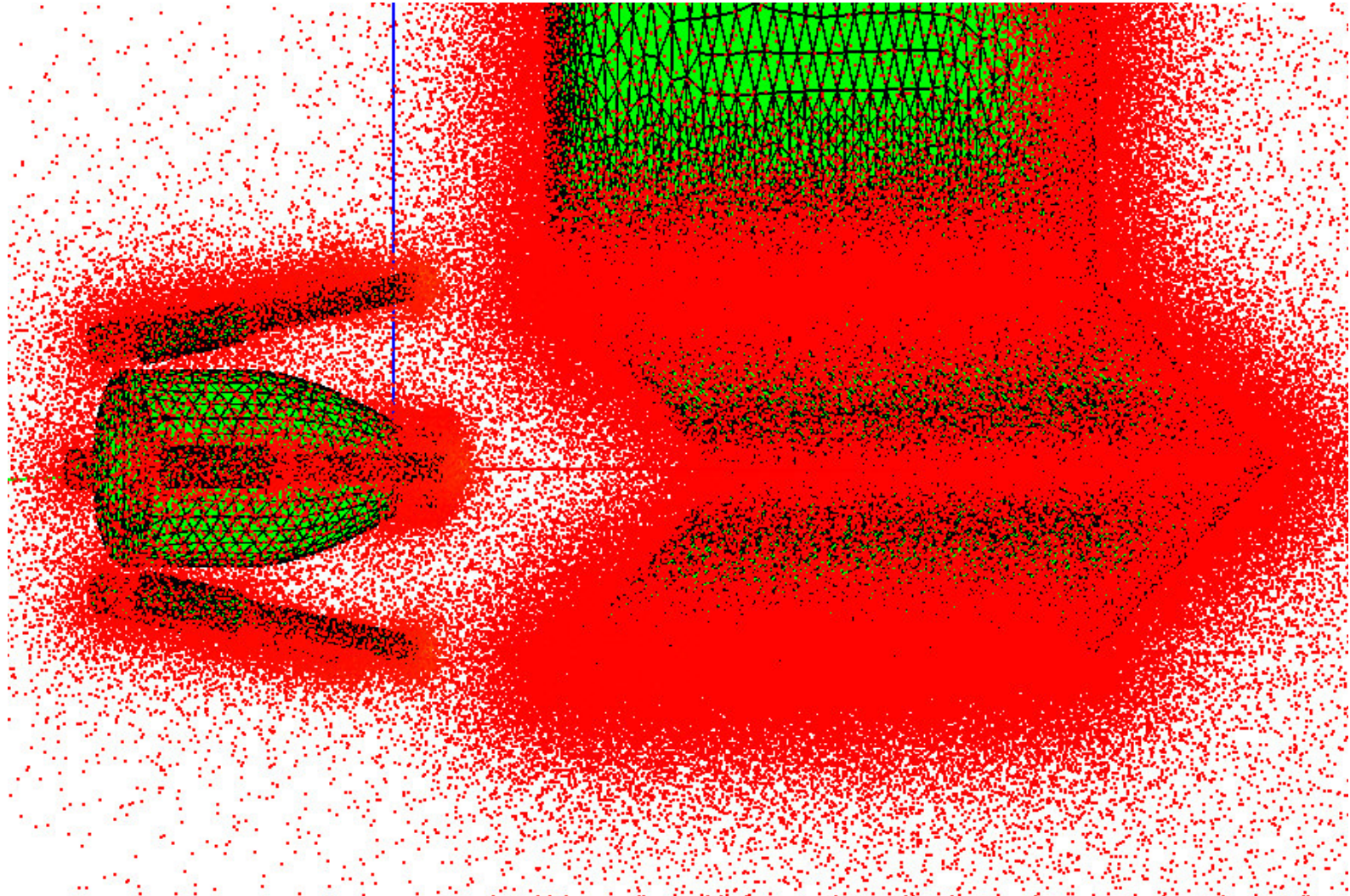
(with R. Sonnenschein, Daimler Chrysler, Dornier)



- 112146 boundary elements
- mesh ratio $\frac{h_{\max}}{h_{\min}} \approx 1454,5$
- 570930 evaluation points
- 150 iteration steps
- thickness of the wall: 0.8 mm
- size of the wall: about 1 m
- flux: $-2.3 \cdot 10^5 \dots 5.5 \cdot 10^8$

Field evaluation

in 570930 points or better interactive on demand. \implies Fast Multipole Method



Evaluation on demand

- Evaluation of the representation formula at many points one by one.
- exact evaluation instead of interpolation on a grid (FEM)
- no grid to construct (FEM)
- build cluster tree without information on the evaluation points
- build separate trees for the geometry and the evaluation.
- new admissibility condition: number of nearfield panels limited
- extra feature: prediction of the point of intersection of the boundary and a straight line (point and direction (electric field))

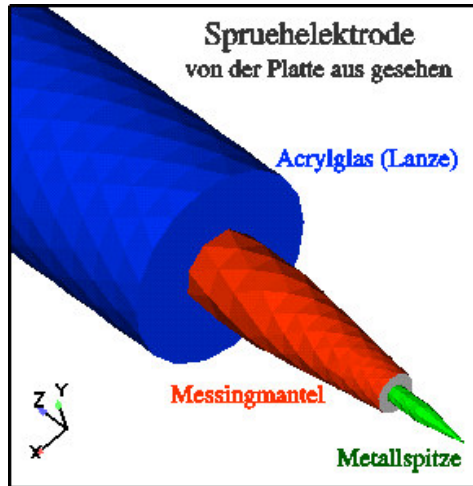
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- extra feature: prediction of the point of intersection of the boundary and a straight line (point and direction (electric field))
- example: 267069 evaluation points

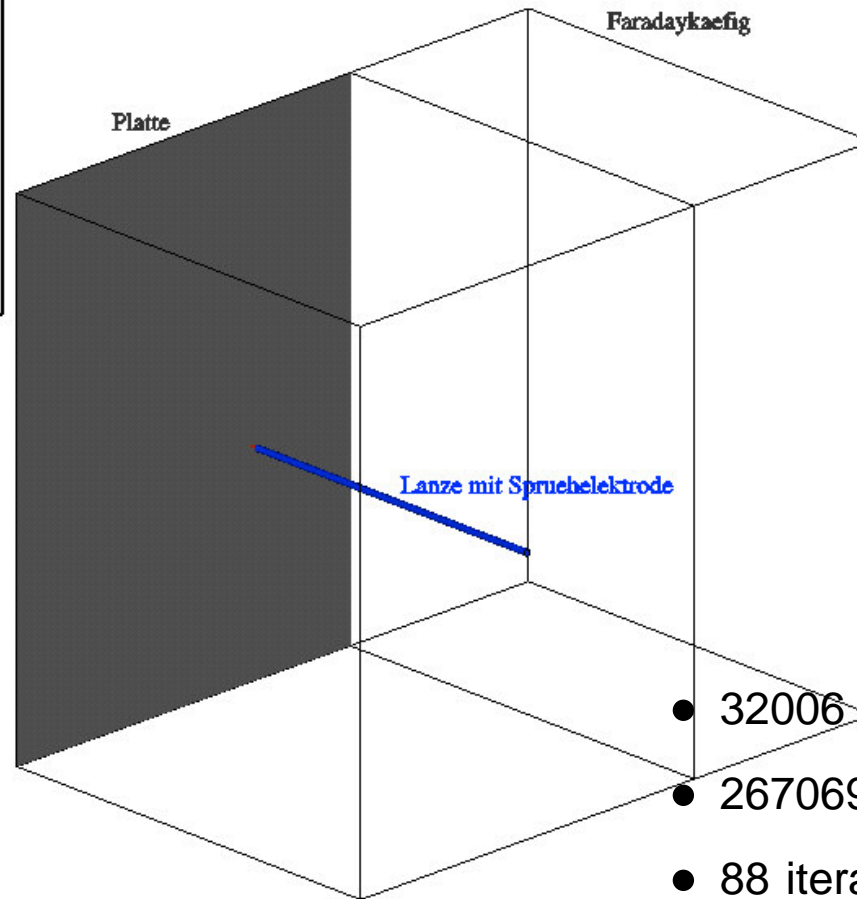
on demand	at once
254 sec	333 sec + 158 sec

Experimental setup

(with R. Sonnenschein, Daimler Chrysler, Dornier)

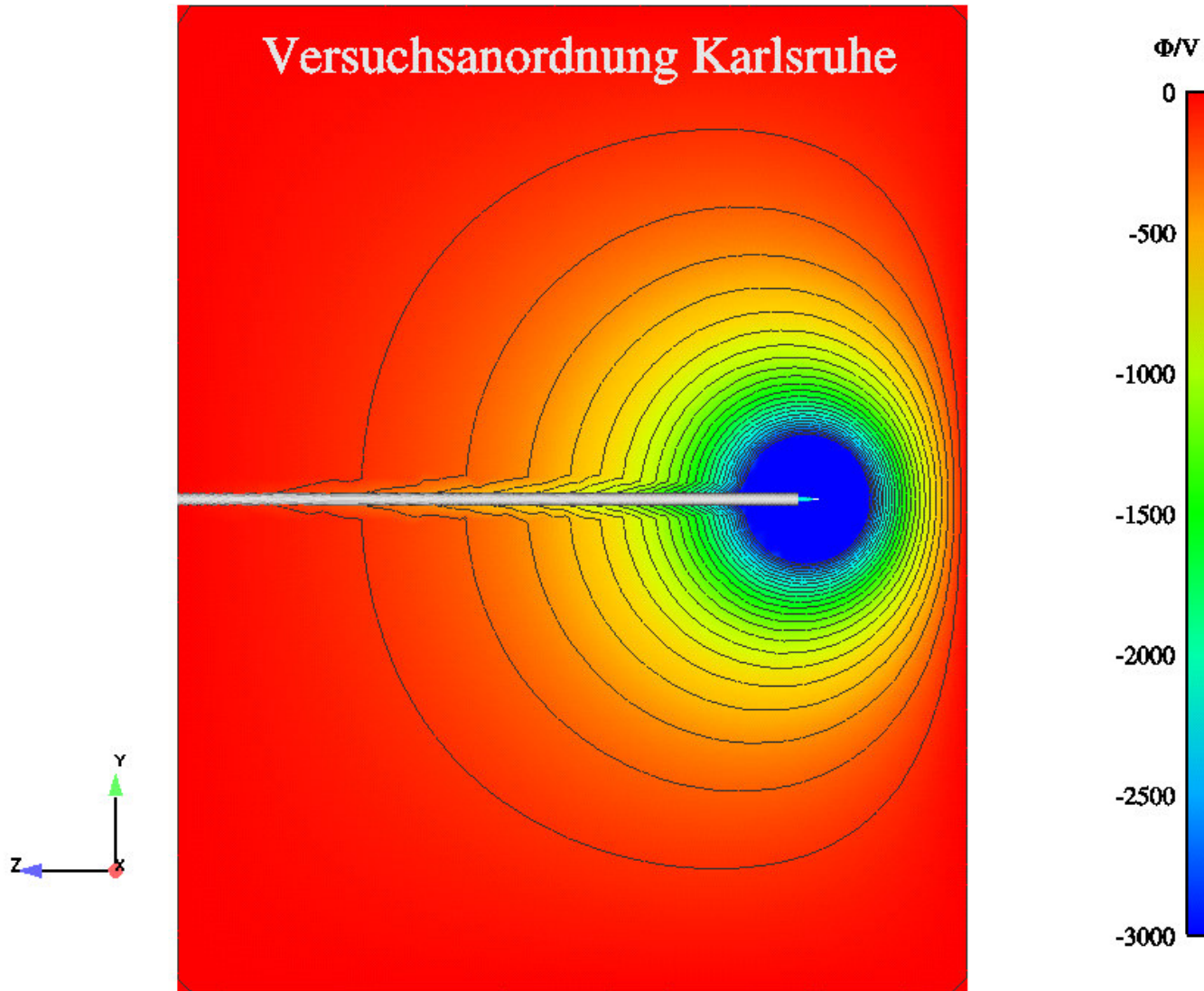


Gesamtansicht Messeinrichtung "Spitze-Platte"



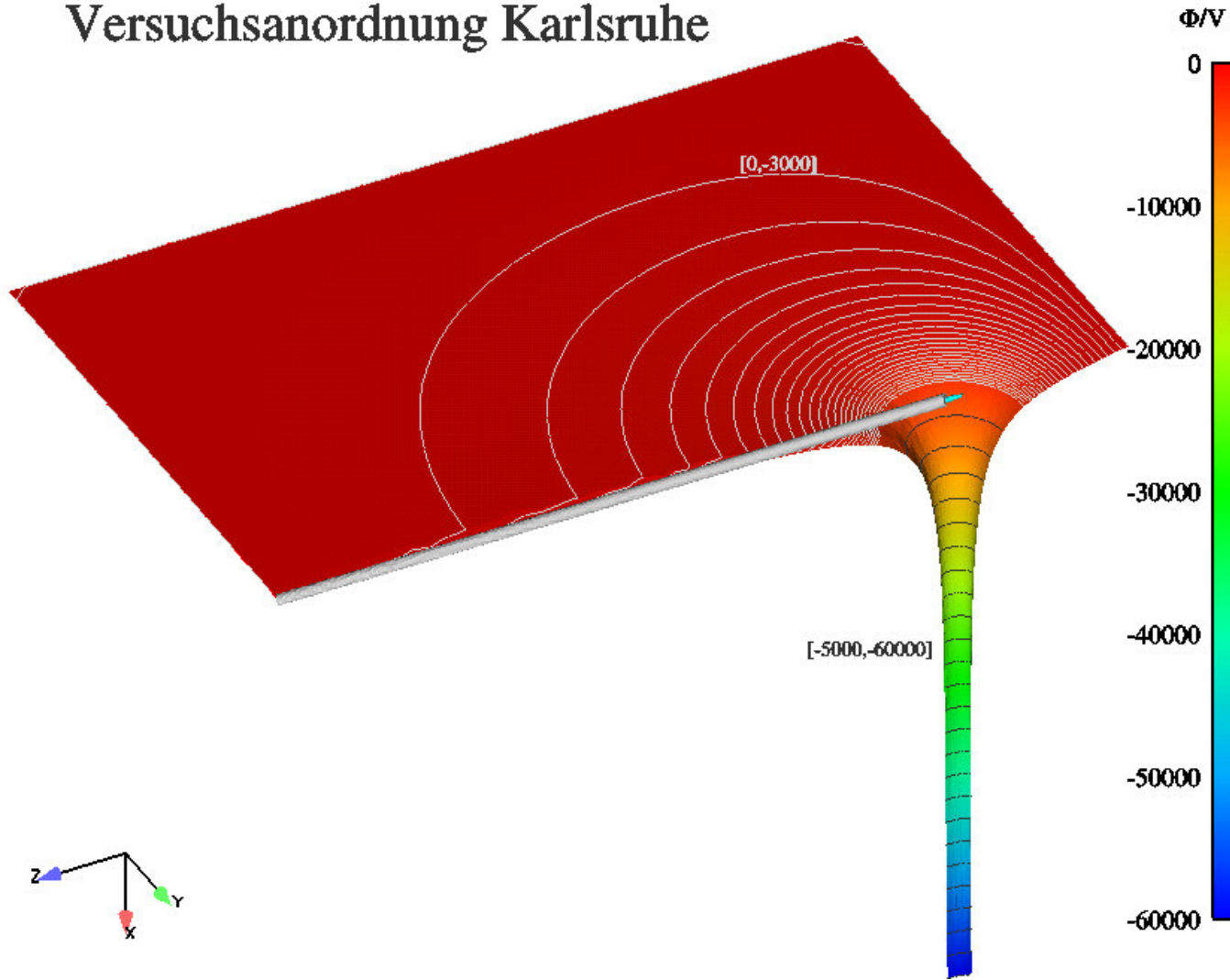
- 32006 boundary elements
- 267069 evaluation points
- 88 iteration steps (1 × re-
fined 92 iteration steps)

Isolines of the potential

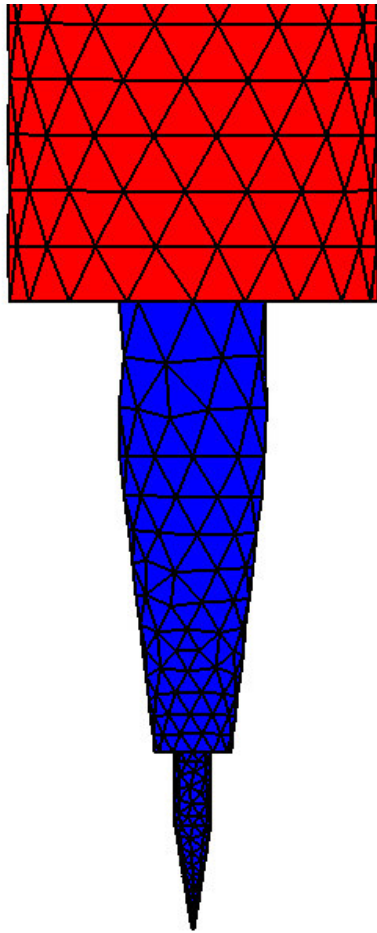


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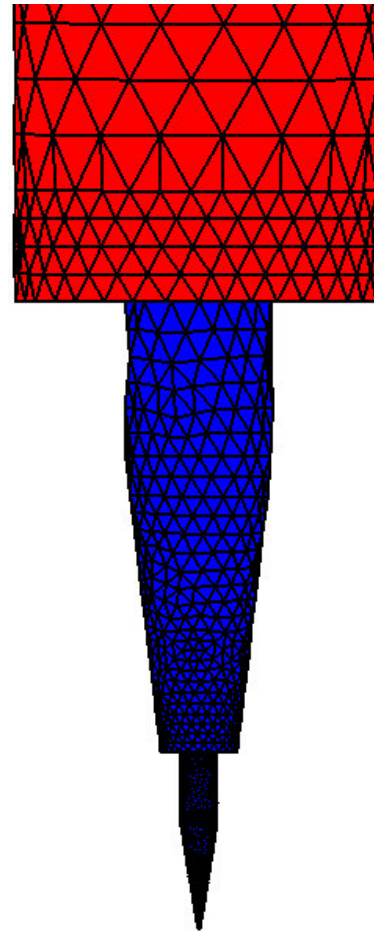
Versuchsanordnung Karlsruhe



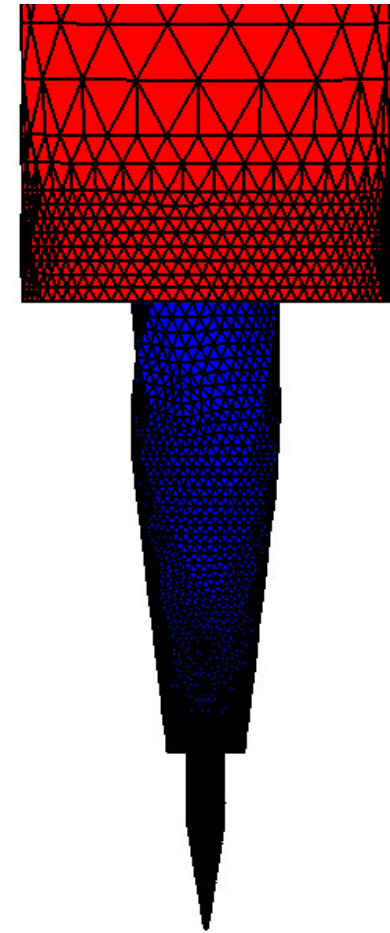
Adaptive meshes



32006 triangles



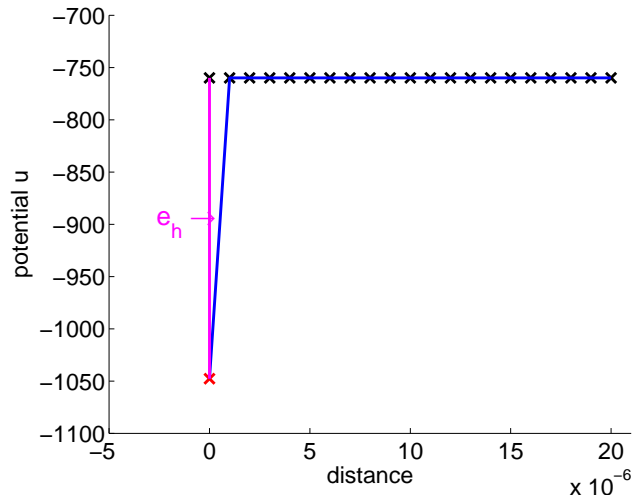
34424 triangles



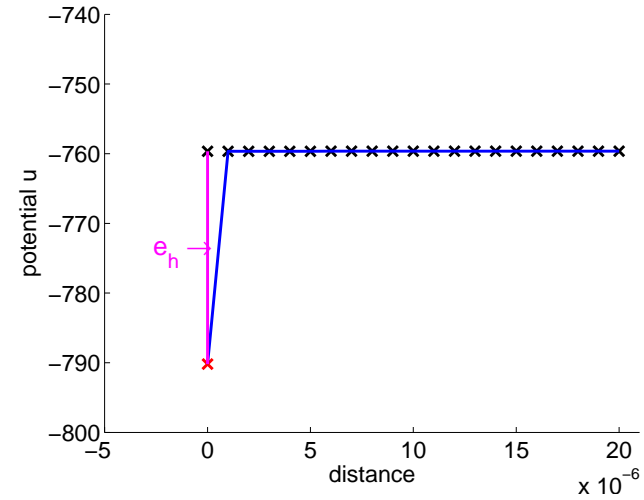
44168 triangles

Cauchy data and representation formula

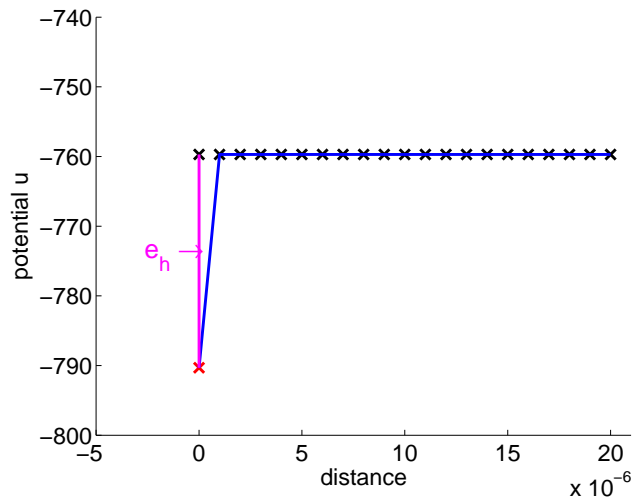
unrefined mesh



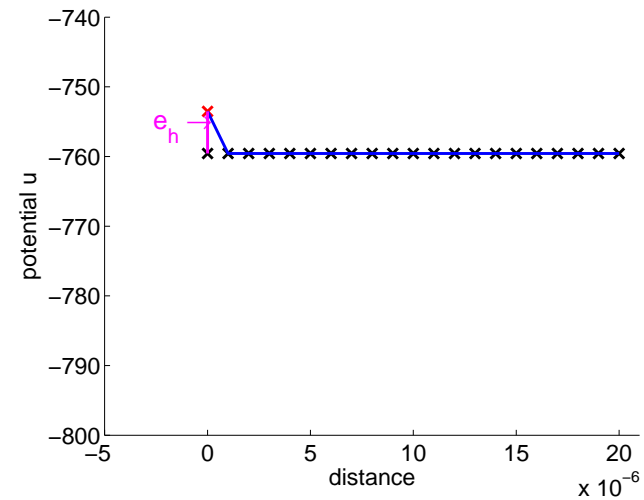
once adaptively refined mesh



uniform refined mesh



twice adaptively refined mesh



A posteriori error estimator

Neumann problem:

$$\begin{aligned}
 -\Delta u(x) &= 0 && \text{for } x \in \Omega \subset \mathbb{R}^3, \\
 t(x) := (T_x u)(x) = (\partial_n u)(x) &= g(x) && \text{for } x \in \Gamma.
 \end{aligned}$$

compatibility condition:

$$\int_{\Gamma} g(x) \cdot 1 \, ds_x = 0$$

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Define \tilde{u} for a discrete solution u_h :

$$\tilde{u}(x) = Vg(y) + ((1 - \sigma(x))I - K)u_h(x) \quad \text{for } x \in \Gamma.$$

Lemma (Schulz, Steinbach). *The **error** $u - u_h$ of the approximation u_h is a solution of the boundary integral equation*

$$(\sigma(x)I + K)(u - u_h)(x) = (\tilde{u} - u_h)(x) \quad \text{for } x \in \Gamma.$$

A simple **error estimator** e_h :

$$\frac{1}{1 + c_K} \|\tilde{u} - u_h\|_D \leq \|u - u_h\|_D \leq \frac{1}{1 - c_K} \|\tilde{u} - u_h\|_D$$

Current and future work

- **domain decomposition methods** for spray painting geometry
 ⇒ **parallel** solvers
- Boundary Element Tearing and Interconnecting methods (**BETI**)
- **automatic** generation of domain decompositions
- adaptive meshes based on the **error estimator**
- **industrial applications**
- ...

