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HILBERT

Matlab Implementation of Adaptive 2D BEM

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joint work with

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HILBERT?
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Weakly-Singular IE
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Introduction to HILBERT
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Conclusion & Outlook
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What is HILBERT ?

Hilbert Is a Lovely Boundary Element Research Tool

- Matlab library for h -adaptive Galerkin BEM
- currently, lowest-order elements for 2D Laplacian
 - \mathcal{P}^0 for fluxes
 - \mathcal{S}^1 for traces
- research code for FWF project P21732
 - overview on current state of the art
 - starting point for further investigations
 - free for academic use

HILBERT is a lovely boundary element research tool



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Introduction to HILBERT
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Conclusion & Outlook
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HILBERT?

HILBERT is a lovely boundary element research tool



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Introduction to HILBERT
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Conclusion & Outlook
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Features of HILBERT

- C implementation of integral operators via MEX interface
 - $V(\mathcal{P}_\Gamma^0 \times \mathcal{P}_\Gamma^0)$
 - $K(\mathcal{S}_\Gamma^1 \times \mathcal{P}_\Gamma^0)$
 - $W(\mathcal{S}_\Gamma^1 \times \mathcal{S}_\Gamma^1)$
 - $N(\mathcal{P}_\Omega^1 \times \mathcal{P}_\Gamma^0)$
- remaining codes in Matlab fully vectorized
 - different error estimators
 - different marking strategies
 - local mesh-refinement
- demo files and adaptive algorithms for
 - weakly-singular integral equation
 - hypersingular integral equation
 - symmetric integral formulation of mixed BVP
 - with/without volume force

HILBERT is a lovely boundary element research tool



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Current Restrictions of HILBERT

- $\Gamma = \partial\Omega$ piecewise affine boundary
- only lowest-order elements
- only canonical bases
 - characteristic functions for \mathcal{P}^0
 - hat functions for \mathcal{S}^1
- no matrix compression, i.e., dense matrices
- direct solution of linear systems by Matlab backslash operator



Weakly-Singular Integral Equation



Outline

- 1 HILBERT?
- 2 Weakly-Singular Integral Equation
- 3 Introduction to HILBERT
- 4 Conclusion & Outlook



Model Problem

Dirichlet Problem

$$\begin{aligned} -\Delta u &= 0 && \text{in } \Omega \\ u &= g && \text{on } \Gamma \end{aligned}$$

Weakly-Singular Integral Equation

$$V\phi = (K + 1/2)g$$

- $\mathcal{H} := H^{-1/2}(\Gamma)$ energy space
- $\|\cdot\|$ equivalent energy norm on \mathcal{H} induced by V



Variational Formulation

Continuous Formulation

Seek $\phi \in \mathcal{H} := H^{-1/2}(\Gamma)$ s.t.

$$\langle V\phi, \psi \rangle = \langle (K + 1/2)g, \psi \rangle \quad \text{for all } \psi \in \mathcal{H}$$

- \mathcal{T}_ℓ partition of Γ

Galerkin Formulation

Seek $\Phi_\ell^* \in X_\ell := \mathcal{P}^0(\mathcal{T}_\ell)$ s.t.

$$\langle V\Phi_\ell^*, \Psi_\ell \rangle = \langle (K + 1/2)g, \Psi_\ell \rangle \quad \text{for all } \Psi_\ell \in X_\ell$$

- **drawback:** RHS hard to compute in general



Perturbed Galerkin Formulation 2/2

Matrix Formulation of Perturbed Galerkin Formulation

Compute coefficient vector \mathbf{x} of Φ_ℓ by solving

$$\mathbf{V}\mathbf{x} = \left(\mathbf{K} + \frac{1}{2} \mathbf{M} \right) \mathbf{g}$$

where \mathbf{g} is nodal vector of g

- mass-type matrix \mathbf{M} analytically computable and sparse
- integral operator matrices \mathbf{V} , \mathbf{K} analytically computable



Perturbed Galerkin Formulation 1/2

- assume additional regularity $g \in H^1(\Gamma)$
- $G_\ell \in \mathcal{S}^1(\mathcal{T}_\ell)$ nodal interpolant of g

Perturbed Galerkin Formulation

Seek $\Phi_\ell \in X_\ell$ s.t.

$$\langle V\Phi_\ell, \Psi_\ell \rangle = \langle (K + 1/2)G_\ell, \Psi_\ell \rangle \quad \text{for all } \Psi_\ell \in X_\ell$$



Control Data Approximation

Reminder: Continuous Problem

$$V\phi = (K + 1/2)g$$

Reminder: Perturbed Continuous Problem

$$V\phi_\ell = (K + 1/2)G_\ell$$

Lemma (Aurada, Goldenits, P. '09)

$$\|\phi - \phi_\ell\| \lesssim \|g - G_\ell\|_{H^{1/2}(\Gamma)} \lesssim \|h_\ell^{1/2}(g - G_\ell)'\|_{L^2(\Gamma)} =: \text{osc}_\ell$$



Control Galerkin Error

- Φ_ℓ Galerkin solution w.r.t. \mathcal{T}_ℓ
- $\widehat{\Phi}_\ell$ Galerkin solution w.r.t. $\widehat{\mathcal{T}}_\ell$
- L^2 -projection Π_ℓ onto $X_\ell = \mathcal{P}^0(\mathcal{T}_\ell)$

Lemma (Ferraz-Leite, P. '08)

There holds $\eta_\ell \leq \tilde{\eta}_\ell \lesssim \tilde{\mu}_\ell \leq \mu_\ell \lesssim \eta_\ell$ for error estimators

$$\begin{aligned} \eta_\ell &= \|\widehat{\Phi}_\ell - \Phi_\ell\| & \tilde{\eta}_\ell &= \|\widehat{\Phi}_\ell - \Pi_\ell \widehat{\Phi}_\ell\| \\ \mu_\ell &= \|h_\ell^{1/2}(\widehat{\Phi}_\ell - \Phi_\ell)\|_{L^2(\Gamma)} & \tilde{\mu}_\ell &= \|h_\ell^{1/2}(\widehat{\Phi}_\ell - \Pi_\ell \widehat{\Phi}_\ell)\|_{L^2(\Gamma)} \end{aligned}$$



Local Refinement Indicators

$$\begin{aligned} \rho_\ell(T)^2 &:= \|h_\ell^{1/2}(\widehat{\Phi}_\ell - \Pi_\ell \widehat{\Phi}_\ell)\|_{L^2(T)}^2 + \|h_\ell^{1/2}(g - G_\ell)'\|_{L^2(T)}^2 \\ \|\phi - \Phi_\ell\|^2 &\lesssim (\eta_\ell + \text{osc}_\ell)^2 \lesssim \tilde{\mu}_\ell^2 + \text{osc}_\ell^2 = \sum_{T \in \mathcal{T}_\ell} \rho_\ell(T)^2 \end{aligned}$$

- everything works also with μ_ℓ instead of $\tilde{\mu}_\ell$



Control Galerkin and Data Approximation Error

Saturation Assumption for Non-Perturbed Problem

- Φ_ℓ^* Galerkin solution w.r.t. \mathcal{T}_ℓ and non-perturbed RHS
- $\widehat{\Phi}_\ell^*$ Galerkin solution w.r.t. $\widehat{\mathcal{T}}_\ell$ and non-perturbed RHS
- $\|\phi - \widehat{\Phi}_\ell^*\| \leq q \|\phi - \Phi_\ell^*\|$

Corollary (Aurada, Goldenits, P. '09)

- $\eta_\ell \lesssim \|\phi - \Phi_\ell\| + \text{osc}_\ell$
- saturation assumption $\implies \|\phi - \Phi_\ell\| \lesssim \eta_\ell + \text{osc}_\ell$



Adaptive Mesh-Refinement

- initial mesh \mathcal{T}_0
- adaptivity parameter $0 < \theta < 1$

Adaptive Algorithm

- 1 compute discrete solution $\widehat{\Phi}_\ell \in \widehat{X}_\ell = \mathcal{P}^0(\widehat{\mathcal{T}}_\ell)$
- 2 for all $T \in \mathcal{T}_\ell$, compute $\rho_\ell(T)$
- 3 find (minimal) set $\mathcal{M}_\ell \subseteq \mathcal{T}_\ell$ s.t.

$$\theta \sum_{T \in \mathcal{T}_\ell} \rho_\ell(T)^2 \leq \sum_{T \in \mathcal{M}_\ell} \rho_\ell(T)^2$$

- 4 refine (at least) marked elements $T \in \mathcal{M}_\ell$ to obtain $\mathcal{T}_{\ell+1}$
 - ensure $\kappa(\mathcal{T}_{\ell+1}) \leq 2\kappa(\mathcal{T}_0)$
- 5 increase counter $\ell \mapsto \ell + 1$ and iterate



Convergence of Adaptive Scheme

Theorem (Aurada, Goldenits, P. '09)

- Use refinement procedure of [AGP '09]
- Then, **concept of estimator reduction** applies, whence

$$\lim_{\ell \rightarrow \infty} (\tilde{\mu}_\ell^2 + \text{osc}_\ell^2) = 0$$

- Saturation assumption for non-perturbed problem implies

$$\lim_{\ell \rightarrow \infty} \|\phi - \hat{\Phi}_\ell\| = 0 = \lim_{\ell \rightarrow \infty} \|\phi - \Phi_\ell\|$$



Data Structure

- $\mathcal{T}_\ell = \{T_1, \dots, T_N\}$ partition
- $\mathcal{N}_\ell = \{z_1, \dots, z_N\}$ nodes

coordinates

- $(N \times 2)$ double-array
- $z_j = (x, y) \in \mathbb{R}^2$ corresponds to `coordinates(j, :) = [x, y]`

elements

- $(N \times 2)$ int-array
- $T_i = \text{conv}\{z_j, z_k\}$ corresponds to `elements(i, :) = [j, k]`



Introduction to HILBERT



Build Galerkin Data

buildV

- assembles $\mathbf{V} = V(\mathcal{P}^0 \times \mathcal{P}^0)$
- use of symmetry
- outer integration over element with $\text{length}(T_j) \leq \text{length}(T_k)$
- outer integration by Gaussian quadrature for far-field entries

buildK

- assembles $\mathbf{K} = K(\mathcal{S}^1 \times \mathcal{P}^0)$

buildSymmRHS

- assembles \mathbf{b} for perturbed RHS $(K + 1/2)G_\ell$



Local Mesh-Refinement

markElements

- returns index vector for marked elements $\mathcal{M}_\ell \subseteq \mathcal{T}_\ell$ from

$$\theta \sum_{T \in \mathcal{T}_\ell} \rho_\ell(T)^2 \leq \sum_{T \in \mathcal{M}_\ell} \rho_\ell(T)^2$$

- even multiple triangulations, e.g., FEM-BEM, mixed BVP

refineBoundaryMesh

- algorithm from [Aurada, Goldenits, P. '09]
- complexity $\mathcal{O}(N \log N)$ for local mesh-refinement
- complexity $\mathcal{O}(N)$ for uniform mesh-refinement

 $(h - h/2)$ -Error Estimators 1/2

computeEstSlpEta

- returns $\eta_\ell^2 = \|\hat{\Phi}_\ell - \Phi_\ell\|^2$.

computeEstSlpEtaTilde

- returns $\tilde{\eta}_\ell^2 = \|\hat{\Phi}_\ell - \Pi_\ell \hat{\Phi}_\ell\|^2$.



Data Oscillations

computeOscDirichlet

- returns $\mathbf{v} \in \mathbb{R}^N$ with $\text{osc}_\ell^2 \approx \tilde{\text{osc}}_\ell^2 = \sum_{j=1}^N \mathbf{v}_j$, where

$$\mathbf{v}_j \approx \text{length}(T_j) \|(g - G_\ell)'\|_{L^2(T_j)}^2$$

- quadrature located at nodes and midpoints
- exact for $g \in \mathcal{S}^2$
- $\text{osc}_\ell = \mathcal{O}(h^{3/2})$ and $|\text{osc}_\ell - \tilde{\text{osc}}_\ell| = \mathcal{O}(h^{5/2})$ for smooth g

 $(h - h/2)$ -Error Estimators 2/2

computeEstSlpMu

- returns $\mathbf{v} \in \mathbb{R}^N$ with $\mu_\ell^2 = \sum_{j=1}^N \mathbf{v}_j$, where

$$\mathbf{v}_j = \text{length}(T_j) \|\hat{\Phi}_\ell - \Phi_\ell\|_{L^2(T_j)}^2$$

computeEstSlpMuTilde

- returns $\tilde{\mathbf{v}} \in \mathbb{R}^N$ with $\tilde{\mu}_\ell^2 = \sum_{j=1}^N \tilde{\mathbf{v}}_j$, where

$$\tilde{\mathbf{v}}_j = \text{length}(T_j) \|\hat{\Phi}_\ell - \Pi_\ell \hat{\Phi}_\ell\|_{L^2(T_j)}^2$$



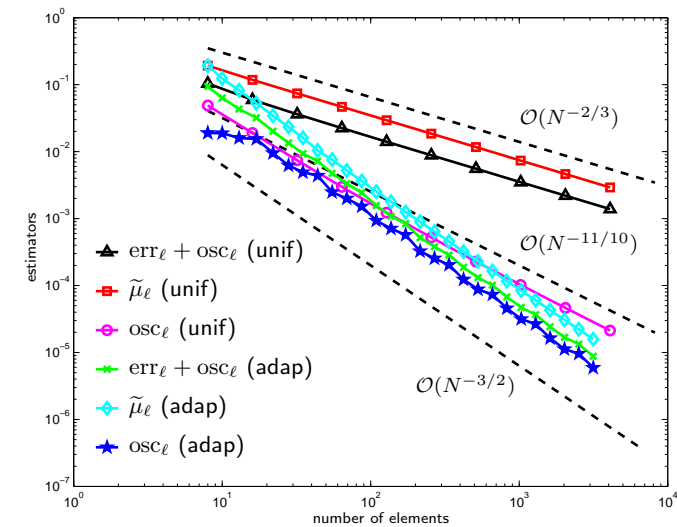
Adaptive Mesh-Refining Algorithm

```

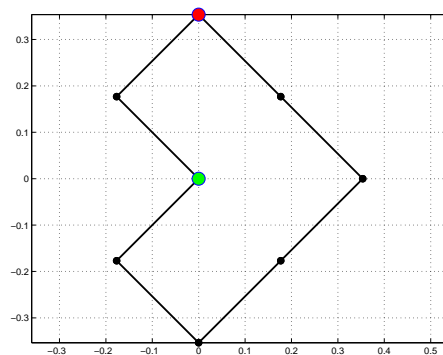
1 while size(elements,1) < nEmax
2
3     %*** build uniformly refined mesh
4     [coordinates_fine, elements_fine, father2son] ...
5         = refineBoundaryMesh(coordinates, elements);
6
7     %*** compute fine-mesh solution
8     V_fine = buildV(coordinates_fine, elements_fine);
9     b_fine = buildSymmRHS(coordinates_fine, elements_fine, @g);
10    x_fine = V_fine\b_fine;
11
12    %*** compute (h-h/2)-error estimator tilde-mu
13    mu_tilde = computeEstSlpMuTilde(coordinates, elements, father2son, ...
14        x_fine);
15
16    %*** compute data oscillations
17    osc_fine = computeOscDirichlet(coordinates_fine, elements_fine, @g);
18    osc = osc_fine(father2son(:,1)) + osc_fine(father2son(:,2));
19
20    %*** mark elements for refinement
21    marked = markElements(theta, mu_tilde + osc);
22
23    %*** generate new mesh
24    [coordinates, elements] = refineBoundaryMesh(coordinates, elements, marked);
25 end

```

Numerical Experiment 2/2



Numerical Experiment 1/2

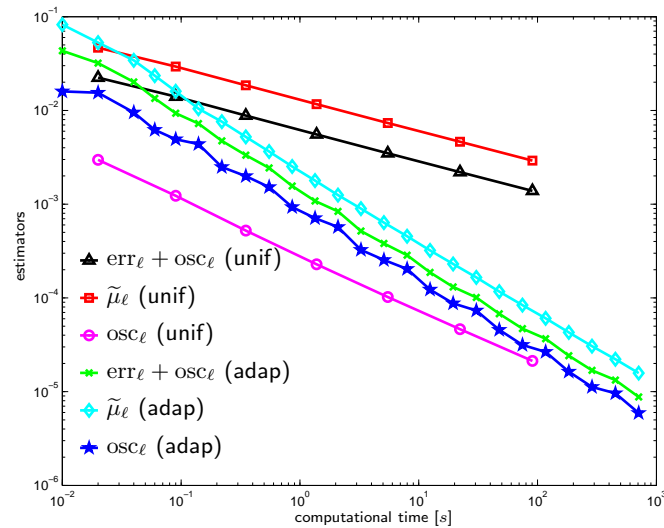


- known exact solution ϕ
- generic singularity of ϕ at reentrant corner (green)
- singularity of g at uppermost corner (red)

Adaptivity Pays 1/2

- adaptive pays w.r.t. number of elements
 - observe optimal convergence behaviour
- does adaptivity pay w.r.t. computational time?
 - $t_\ell^{(\text{unif})} := \text{unif}^{(\ell)}(\mathcal{T}_0) + \text{assemble} + \text{solve}$
 - $t_0^{(\text{adap})} := \text{assemble} + \text{solve} + \text{estimate} + \text{mark} + \text{refine}$
 - $t_{\ell+1}^{(\text{adap})} := t_\ell^{(\text{adap})} + \dots$

Adaptivity Pays 2/2



HILBERT — Release One

- release date: **Friday 13. November 2009**
- available at: <http://www.asc.tuwien.ac.at/abem>
- Release One contains lowest-order BEM for 2D Laplacian
 - Dirichlet problem
 - Neumann problem
 - mixed Dirichlet-Neumann problem
 - with/without volume forces
- Release One includes
 - $(h - h/2)$ -error estimators for all problems
 - local mesh-refinement for boundary/volume meshes
 - adaptive algorithms & demo files
 - several visualization tools
 - full documentation of all Matlab codes (about 60 pages)

Conclusion & Outlook

Short-Term Extensions

- further error estimators, e.g.,
 - weighted-residual error estimators
 - two-level error estimators
 - Faermann residual error estimator
 - Steinbach-Schulz error estimator
- adaptive FEM-BEM coupling
- Lamé equation

Thank You for Listening

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