

Can **functional-type** a posteriori error estimates unite adaptivity and error control in **BEM**?

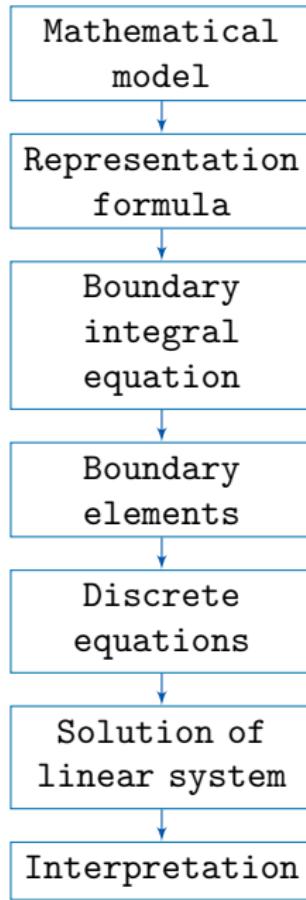
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Mathematical model

- ▶ $\Omega \subset \mathbb{R}^d$, $d = 2, 3$, polygonal boundary $\Gamma := \partial\Omega$,
- ▶ **homogeneous** Poisson problem

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

Representation formula

Let $G(z) = -\frac{1}{2\pi} \log|z|$ ($d = 2$) the **fundamental solution** of the Laplacian. Ansatz (make a choice!):

- ▶ $u(x) = \int_{\Gamma} G(x-y) \phi(y) dy =: [\tilde{V}\phi](x)$ (indir.)
- ▶ $u(x) = [\tilde{V}\phi](x) - \underbrace{\int_{\Gamma} \partial_{n(y)} G(x-y) g(y) dy}_{=: [\tilde{K}g](x)}$ (dir.)
- ▶ ...

For any $\phi \in H^{-1/2}(\Gamma)$ it holds $\Delta u = 0$.

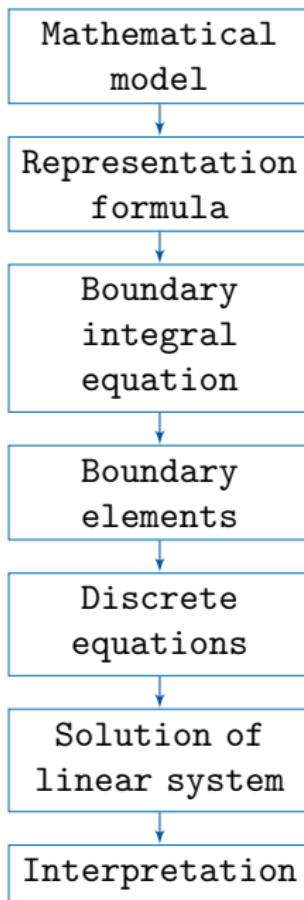
- ▶ Taking the trace of u (indir.) leads to boundary integral equation

$$V\phi := \gamma_0 \circ \tilde{V}\phi = g \quad \text{in } H^{1/2}(\Gamma).$$

- ▶ variational formulation: Find $\phi \in H^{-1/2}(\Gamma)$ s.t.

$$\langle\langle \phi, \psi \rangle\rangle := \langle \psi, V\phi \rangle_{\Gamma} = \langle \psi, g \rangle_{\Gamma}$$

for all $\psi \in H^{-1/2}(\Gamma)$.



Boundary elements

Let $\mathcal{E}_h = \{E_1, \dots, E_N\}$ be a regular triangulation of Γ .

- ▶ $\mathcal{P}^0(\mathcal{E}_h) := \{\psi \in L^\infty(\Gamma) : \psi|_E \text{ is const. } \forall E \in \mathcal{E}_h\} \subset H^{-1/2}(\Gamma)$
- ▶ $\mathcal{S}^1(\mathcal{E}_h) := \{\zeta \in C(\Gamma) : \zeta|_E \text{ is affine } \forall E \in \mathcal{E}_h\} \subset H^{1/2}(\Gamma)$

Discrete equations

Find $\phi_h \in \mathcal{P}^0(\mathcal{E}_h)$ s.t.

$$\langle \psi_h, V\phi_h \rangle_\Gamma = \langle \psi_h, g_h \rangle_\Gamma \quad \text{for all } \psi_h \in \mathcal{P}^0(\mathcal{E}_h).$$

L^2 -projection of g

Find $g_h \in \mathcal{S}^1(\mathcal{E}_h) \subset H^{1/2}(\Gamma)$ s.t.

$$\langle g_h, \zeta_k \rangle_\Gamma = \langle g, \zeta_k \rangle_\Gamma \quad \text{for all } \zeta_k \in \mathcal{S}^1(\mathcal{E}_h).$$

- ▶ oscillation error

$$\|g - g_h\|_{H^{1/2}(\Gamma)} \leq C \|h^{1/2}(g - g_h)'\|_{L^2(\Gamma)}$$

Solution of linear system

Interpretation

By means of chosen ansatz (representation formula)

$$u_h = \tilde{V}\phi_h$$

What to remember...

- ▶ compute boundary density $\phi_h \approx \phi \in H^{-1/2}(\Gamma)$
 - ▶ in a **direct ansatz**, we obtain the full Cauchy data $u|_\Gamma = g$ and $\partial_n u|_\Gamma \approx \phi_h$
 - ▶ in an **indirect ansatz**, the computed density has no physical relevance
- ▶ $\Delta u_h = 0$, in particular $\Delta(u - u_h) = 0$, in $L^2(\Omega)$.
- ▶ boundary datum g is discretized, e.g. by L^2 -projection
- ▶ lowest order discretization: $\mathcal{P}^0(\mathcal{E}_h) \subset H^{-1/2}(\Gamma)$,
 $\mathcal{S}^1(\mathcal{E}_h) \subset H^{1/2}(\Gamma)$
- ▶ energy error in $H^{-1/2}(\Gamma)$:

$$\|\phi - \phi_h\| := \sqrt{\langle \phi - \phi_h, V(\phi - \phi_h) \rangle_\Gamma} \simeq \|\phi - \phi_h\|_{H^{-1/2}(\Gamma)}$$

Motivation

Q: Can functional-type a posteriori error estimates unite **adaptivity** and **error control** in BEM?

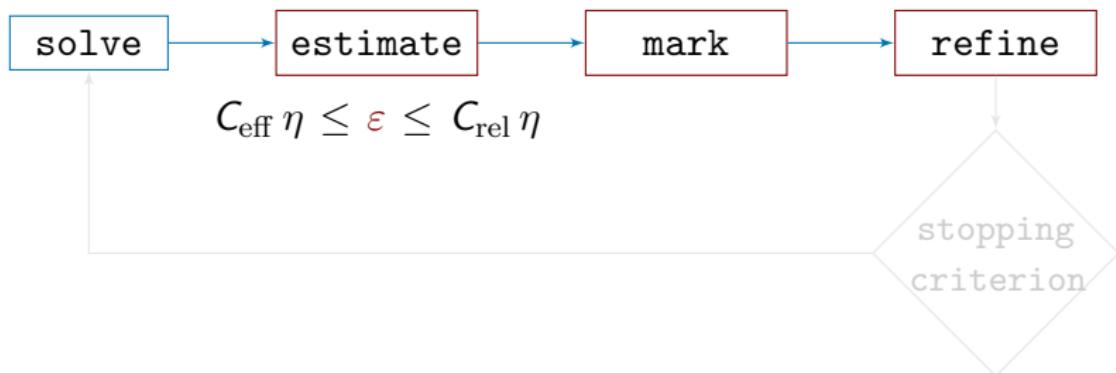
- ▶ **adaptivity?** $\rightsquigarrow \varepsilon = \|\phi - \phi_h\| \simeq \|\phi - \phi_h\|_{H^{-1/2}(\Gamma)}$
- ▶ **error control?** $\rightsquigarrow \varepsilon = \|\nabla(u - u_h)\|_{L^2(\Omega)}, u_h = \tilde{V}\phi_h$
- ▶ **unity?** (Which ε might enable **both**?)

Motivation

Q: Can functional-type a posteriori error estimates unite **adaptivity** and **error control** in BEM?

► **adaptivity?** $\rightsquigarrow \varepsilon = |\phi - \phi_h| \simeq \|\phi - \phi_h\|_{H^{-1/2}(\Gamma)}$

The standard adaptive algorithm:

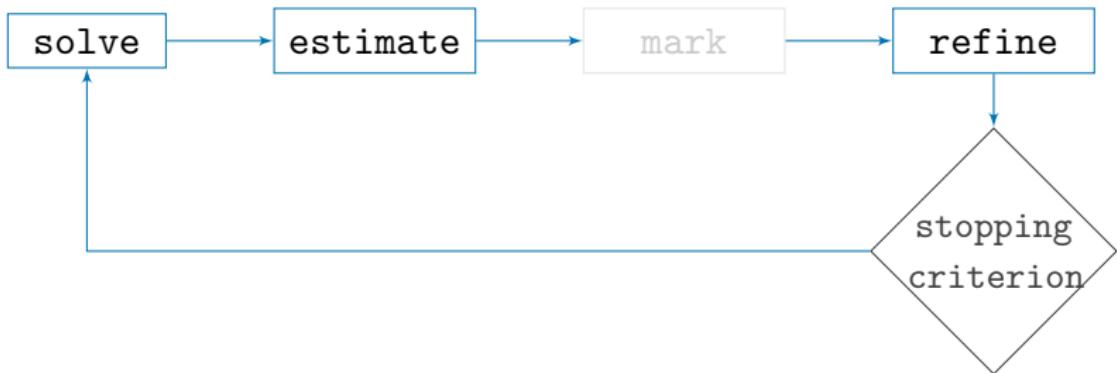


Motivation

Q: Can functional-type a posteriori error estimates unite **adaptivity** and **error control** in BEM?

► **error control?** $\rightsquigarrow \varepsilon = \|\nabla(u - u_h)\|_{L^2(\Omega)}, \quad u_h = \tilde{V}\phi_h$

The error-controlling algorithm (based on uniform refinement):

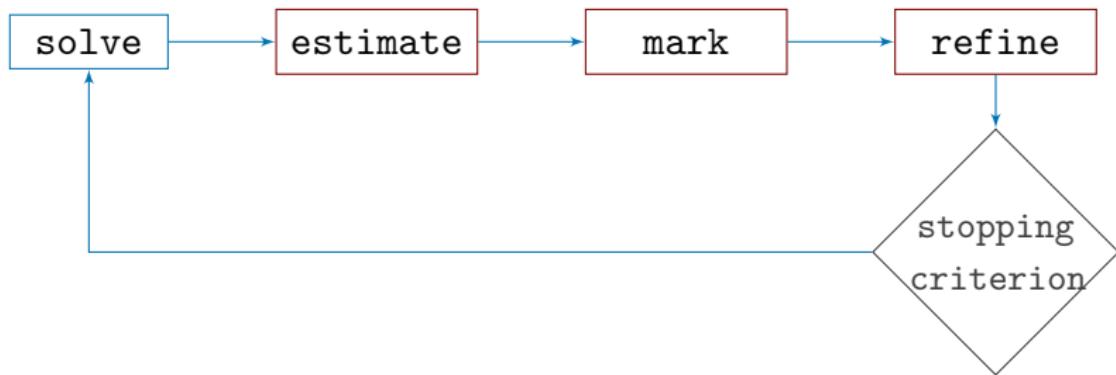


Motivation

Q: Can functional-type a posteriori error estimates unite **adaptivity** and **error control** in BEM?

► **unity?** (Which ε might enable **both**?)

The practical (ideal) adaptive algorithm:



What to remember...

- ▶ In BEM, intuitive error functionals for **adaptivity** and **error control** do generally not coincide
 - ▶ $\varepsilon = \|\phi - \phi_h\| \simeq \|\phi - \phi_h\|_{H^{-1/2}(\Gamma)}$
 - ▶ $\varepsilon = \|\nabla(u - u_h)\|_{L^2(\Omega)}$
- ▶ $\|\phi - \phi_h\| \simeq \|\phi - \phi_h\|_{H^{-1/2}(\Gamma)}$ is non-local, i.e. cannot be written as sum of local contributions
 - ▶ localization of $\|\cdot\|$ will involve incomputable constants
- ▶ $\|\nabla(u - u_h)\|_{L^2(\Omega)}$ is local, but (commonly demanded) properties like Galerkin orthogonality are not available on Ω

Key Message of this talk.

- ▶ The "**natural**" error functional is not the only option when aiming at adaptivity.
- ▶ **Functional-type** a posteriori error estimates are rather suitable for BEM. (does not require a priori knowledge on u_h , which we do not have, e.g. Galerkin orth.)

Error identity

For any approximation $v \in H^1(\Omega)$ with $\nabla v \in H(\text{div} = 0, \Omega)$, it holds that

$$\max_{\substack{\tau \in H(\text{div}, \Omega) \\ \text{div } \tau = 0}} \underline{\mathfrak{M}}(\tau) = \|\nabla(u - v)\|_{L^2(\Omega)}^2 = \min_{\substack{w \in H^1(\Omega) \\ w|_\Gamma = g - v|_\Gamma}} \overline{\mathfrak{M}}(w), \quad (1)$$

where

$$\underline{\mathfrak{M}}(\tau) := \left[2 \int_\Gamma (g - v|_\Gamma) \tau \cdot \mathbf{n} - \|\tau\|_{L^2(\Omega)}^2 \right] \quad (2)$$

and

$$\overline{\mathfrak{M}}(w) := \|\nabla w\|_{L^2(\Omega)}^2. \quad (3)$$

The unique maximizer is $\tau = \nabla(u - v)$, the unique minimizer is $w = u - v$.

Proof.

The proof is split into two steps.

Step 1 (Upper bound). Let $w \in H^1(\Omega)$ with $w|_{\Gamma} = u|_{\Gamma} = g$. Since $\nabla(u - v) \in H(\text{div} = 0, \Omega)$, integration by parts shows that

$$\begin{aligned}\|\nabla(u - v)\|_{L^2(\Omega)}^2 &= \langle \nabla(u - w), \nabla(u - v) \rangle_{\Omega} + \langle \nabla(w - v), \nabla(u - v) \rangle_{\Omega} \\ &= \langle \nabla(w - v), \nabla(u - v) \rangle_{\Omega}.\end{aligned}$$

With the Cauchy-Schwarz inequality, we are led to

$$\|\nabla(u - v)\|_{L^2(\Omega)} \leq \|\nabla(w - v)\|_{L^2(\Omega)}.$$

By substitution, this proves that

$$\|\nabla(u - v)\|_{L^2(\Omega)} \leq \inf_{\substack{w \in H^1(\Omega) \\ w|_{\Gamma} = g - v|_{\Gamma}}} \|\nabla w\|_{L^2(\Omega)},$$

and the infimum is clearly attained for $w = u - v$.



Proof.

Step 2 (Lower bound). In any Hilbert space H , it holds that

$$\|a\|_H^2 = \max_{b \in H} [2\langle a, b \rangle_H - \|b\|_H^2] \quad \text{for all } a \in H,$$

where the maximum is attained for $b = a$. With $\nabla(u - v) \in H(\operatorname{div} = 0\Omega) =: H$, integration by parts shows that

$$\begin{aligned}\|\nabla(u - v)\|_{L^2(\Omega)}^2 &= \|\nabla(u - v)\|_{H(\operatorname{div}, \Omega)}^2 \\ &= \max_{\substack{\tau \in H(\operatorname{div}, \Omega) \\ \operatorname{div} \tau = 0}} \left[2 \int_{\Omega} \nabla(u - v) \cdot \tau - \|\tau\|_{L^2(\Omega)}^2 \right] \\ &= \max_{\substack{\tau \in H(\operatorname{div}, \Omega) \\ \operatorname{div} \tau = 0}} \left[2 \int_{\Gamma} (g - v|_{\Gamma}) \tau \cdot \mathbf{n} - \|\tau\|_{L^2(\Omega)}^2 \right].\end{aligned}$$

In particular, the maximum is attained for $\tau = \nabla(u - v)$. This concludes the proof. □

Boundary residual

Inside the error identity ($\|\cdot\| := \|\cdot\|_{L^2(\Omega)}$)

$$\max_{\substack{\tau \in H(\text{div}, \Omega) \\ \text{div } \tau = 0}} \left[2 \int_{\Gamma} \tau \cdot n (g - v|_{\Gamma}) - \|\tau\|^2 \right] = \|\nabla(u - v)\|^2 = \min_{\substack{w \in H^1(\Omega) \\ w|_{\Gamma} = g - v|_{\Gamma}}} \|\nabla w\|^2,$$

the boundary residual is essential and contains all relevant information.

- ▶ evaluations of a BEM approximation v might be expensive at the boundary!
- ▶ numerical approximations of $w_h \approx w$ w.r.t. to the majorant will never satisfy the **boundary condition** exactly, i.e.

$$\underline{\mathfrak{M}}(\tau_h) \leq \max_{\substack{\tau \in H(\text{div}, \Omega) \\ \text{div } \tau = 0}} \underline{\mathfrak{M}}(\tau) = \|\nabla(u - v)\| = \min_{\substack{w \in H^1(\Omega) \\ w|_{\Gamma} = g - v|_{\Gamma}}} \overline{\mathfrak{M}}(w) \leq \overline{\mathfrak{M}}(w_h)$$

will, in general, **not** hold true.

BEM example

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Error identity

Majorant

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Majorant

Practical Majorant

Let $J_h : H^{1/2}(\Gamma) \rightarrow H^{1/2}(\Gamma)$. For any $v \in H^1(\Omega)$ with $\nabla v \in H(\operatorname{div} = 0, \Omega)$, it then follows that

$$\|\nabla(u - v)\|_{L^2(\Omega)} \leq \min_{\substack{\widehat{w} \in H^1(\Omega) \\ \widehat{w}|_\Gamma = J_h(g - v)|_\Gamma}} \|\nabla \widehat{w}\|_{L^2(\Omega)} + 2 \|(1 - J_h)(g - v)|_\Gamma\|_{H^{1/2}(\Gamma)}. \quad (4)$$

Proof.

Idea: Let $w = u - v$ the exact solution to

$$\min_{\substack{w \in H^1(\Omega) \\ w|_\Gamma = g - v|_\Gamma}} \overline{\mathfrak{M}}(w)$$

and \widehat{w} the solution to

$$\min_{\substack{w \in H^1(\Omega) \\ w|_\Gamma = J_h(g - v)|_\Gamma}} \overline{\mathfrak{M}}(w).$$

Then,

$$\begin{aligned} \|\nabla(u - v)\|_{L^2(\Omega)} &= \|\nabla w\|_{L^2(\Omega)} \leq \|\nabla \widehat{w}\|_{L^2(\Omega)} + \underbrace{\|\nabla(w - \widehat{w})\|_{L^2(\Omega)}}_{\leq \|(1 - J_h)(g - v)|_\Gamma\|_{H^{1/2}(\Gamma)}}. \end{aligned}$$

Corollary

Let $J_h : H^{1/2}(\Gamma) \rightarrow H^{1/2}(\Gamma)$. Let $v \in H^1(\Omega)$ with $\nabla v \in H(\text{div} = 0, \Omega)$. Let $S \subseteq \Omega$ be a Lipschitz domain with $\Gamma \subseteq \partial S$ and $\bar{w} \in H^1(S)$ the solution of the inhomogeneous Dirichlet problem

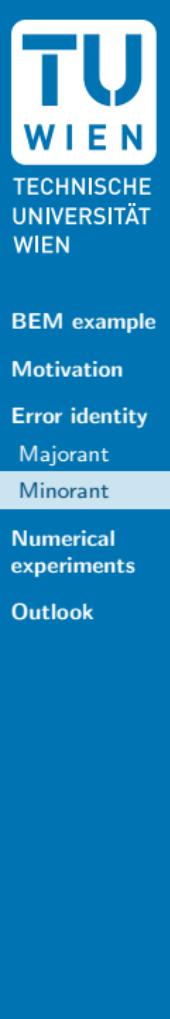
$$\langle \nabla \bar{w}, \nabla \varphi \rangle_S = 0 \text{ for all } \varphi \in \mathring{H}^1(S) \text{ subject to } \bar{w}|_{\partial S} = \begin{cases} J_h(g - v|_\Gamma) & \text{on } \Gamma \subseteq \partial S, \\ 0 & \text{on } \partial S \setminus \Gamma. \end{cases} \quad (5)$$

Then, $\|\nabla(u - v)\|_{L^2(\Omega)} \leq \|\nabla \bar{w}\|_{L^2(S)} + 2 \| (1 - J_h)(g - v|_\Gamma) \|_{H^{1/2}(\Gamma)}.$ (6)

Moreover, let \mathcal{T}_h be a conforming triangulation of S and suppose that $\text{range}(J_h) \subseteq \{v_h|_\Gamma : v_h \in \mathcal{S}^1(\mathcal{T}_h)\}$. Consider the FEM approximation of \bar{w} , i.e., $\bar{w}_h \in \mathcal{S}^1(\mathcal{T}_h)$ satisfies that

$$\langle \nabla \bar{w}_h, \nabla \varphi_h \rangle_S = 0 \text{ for all } \varphi_h \in \mathcal{S}_0^1(\mathcal{T}_h) \text{ with } \bar{w}_h|_{\partial S} = \begin{cases} J_h(g - v|_\Gamma) & \text{on } \Gamma \subseteq \partial S, \\ 0 & \text{on } \partial S \setminus \Gamma. \end{cases} \quad (7)$$

Then, (6) also holds with \bar{w} being replaced by \bar{w}_h .



Minorant

Proposition

For any approximation $v \in H^1(\Omega)$ with $\nabla v \in H(\operatorname{div} = 0, \Omega)$, let $(\tau, p) \in H(\operatorname{div}, \Omega) \times L^2(\Omega)$ be the solution to the mixed problem

$$\begin{aligned}\langle \tau, \sigma \rangle_\Omega + \langle \operatorname{div} \sigma, p \rangle_\Omega &= \langle \sigma \cdot \mathbf{n}, g - v|_\Gamma \rangle_\Gamma \\ \langle \operatorname{div} \tau, q \rangle_\Omega &= 0\end{aligned}$$

for all $(\sigma, q) \in H(\operatorname{div}, \Omega) \times L^2(\Omega)$.

Then, it holds that

$$2 \langle \tau \cdot \mathbf{n}, g - v|_\Gamma \rangle_\Gamma - \|\tau\|_{L^2(\Omega)}^2 = \|\nabla(u - v)\|_{L^2(\Omega)}^2.$$

↪ Implementation available from C. Bahriawati, C. Carstensen in [6].

Proposition (yet implemented version) only $d = 2$

For topologically trivial domains $\Omega \subset \mathbb{R}^2$, we have

$H(\operatorname{div} = 0, \Omega) = \operatorname{rot}_z H(\operatorname{rot}_z, \Omega)$. Hence by $\tau = \operatorname{rot}_z \omega = \begin{pmatrix} -\partial_y \omega \\ \partial_x \omega \end{pmatrix}$, we are led to

$$\min_{\omega \in H^1(\Omega)} \left[2 \langle \operatorname{rot}_z \omega \cdot \mathbf{n}, g - v|_\Gamma \rangle_\Gamma - \|\nabla \omega\|_{L^2(\Omega)}^2 \right] = \|\nabla(u - u_h)\|_{L^2(\Omega)}^2.$$

Proposition

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$$\begin{aligned}\langle \tau, \sigma \rangle_{\Omega} + \langle \operatorname{div} \sigma, p \rangle_{\Omega} &= \langle \sigma \cdot \mathbf{n}, g - v|_{\Gamma} \rangle_{\Gamma} \\ \langle \operatorname{div} \tau, q \rangle_{\Omega} &= 0\end{aligned}$$

for all $(\sigma, q) \in H(\operatorname{div}, \Omega) \times L^2(\Omega)$.

Then, it holds that

$$\underline{\mathfrak{M}}(\tau_0) \leq 2 \langle \tau \cdot \mathbf{n}, g - v|_{\Gamma} \rangle_{\Gamma} - \|\tau\|_{L^2(\Omega)}^2 = \|\nabla(u - v)\|_{L^2(\Omega)}^2.$$

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$$\min_{\omega \in H^1(\Omega)} \left[2 \langle \operatorname{rot}_z \omega \cdot \mathbf{n}, g - v|_{\Gamma} \rangle_{\Gamma} - \|\nabla \omega\|_{L^2(\Omega)}^2 \right] = \|\nabla(u - u_h)\|_{L^2(\Omega)}^2.$$

Algorithm

Set $\ell = 1$.

1. Extract BEM-mesh \mathcal{E}_ℓ from given FEM-mesh \mathcal{T}_ℓ
2. Compute L^2 -projection $g_h \in \mathcal{S}^1(\mathcal{E}_\ell)$ of $g \in H^1(\Gamma)$
3. Solve boundary integral equation in $\mathcal{P}^0(\mathcal{E}_\ell)$
4. Solve Majorant / Minorant problems via P1-FEM on second order patch Σ , i.e. with $\mathcal{S}^1(\mathcal{T}_\ell^\Sigma) = \mathcal{S}^1(\mathcal{T}_\ell)|_\Sigma$
5. compute error estimator on \mathcal{T}_ℓ^Σ :

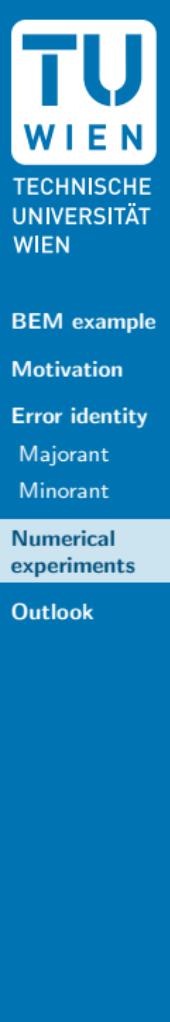
$$\eta_\ell(T) = \|\nabla \bar{w}_h\|_{L^2(T)}$$

for $T \in \mathcal{T}_\ell^\Sigma$, and 0 in $\mathcal{T}_\ell \setminus \mathcal{T}_\ell^\Sigma$

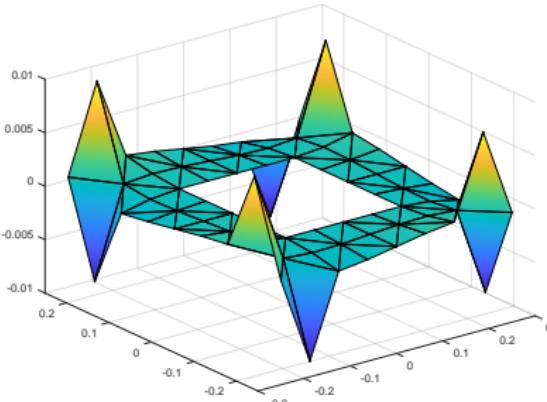
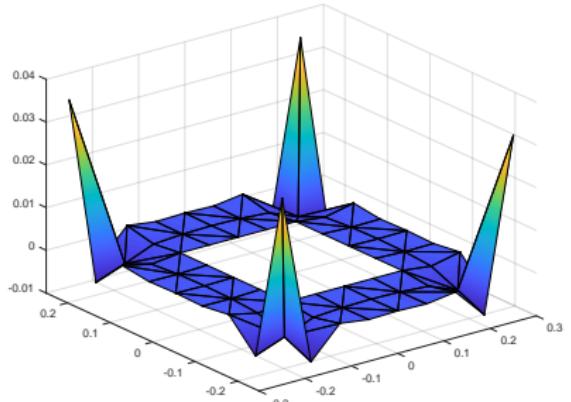
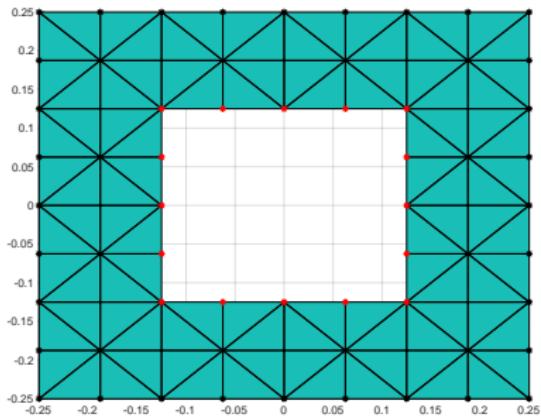
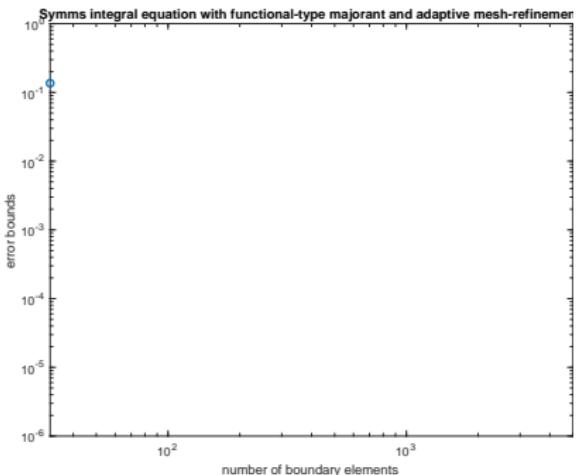
6. Determine set $\mathcal{M}_\ell \subset \mathcal{T}_\ell$ of minimal cardinality such that

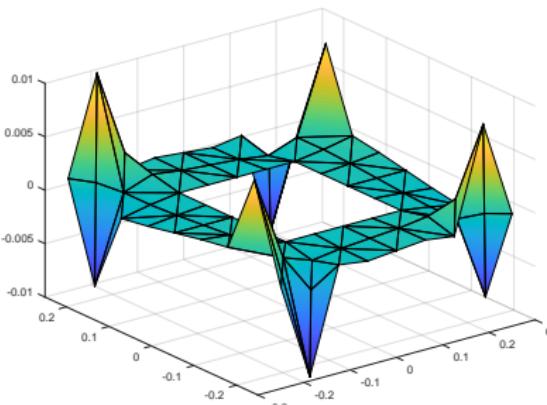
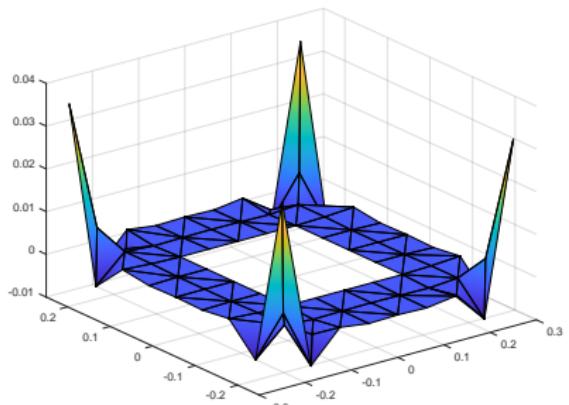
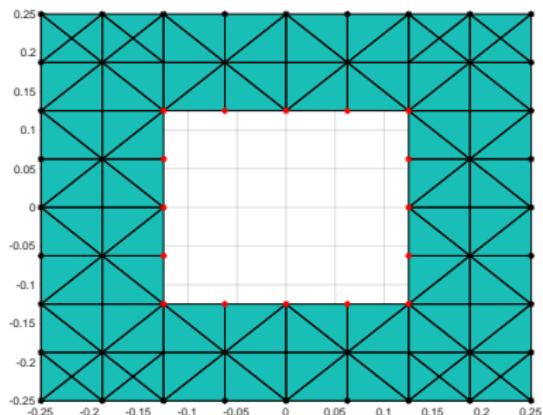
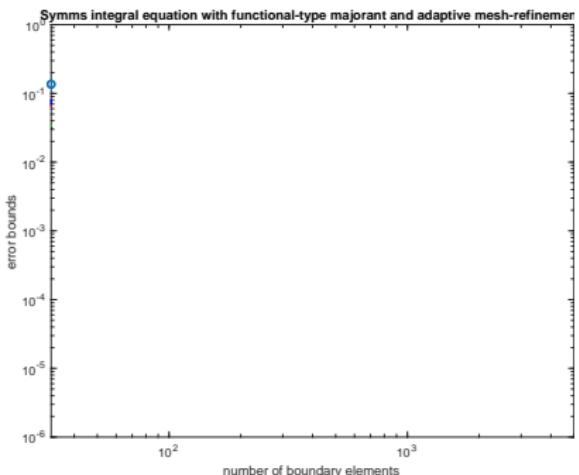
$$\theta \sum_{T \in \mathcal{T}_\ell} \eta_\ell^2(T) \leq \sum_{T \in \mathcal{M}_\ell} \eta_\ell^2(T).$$

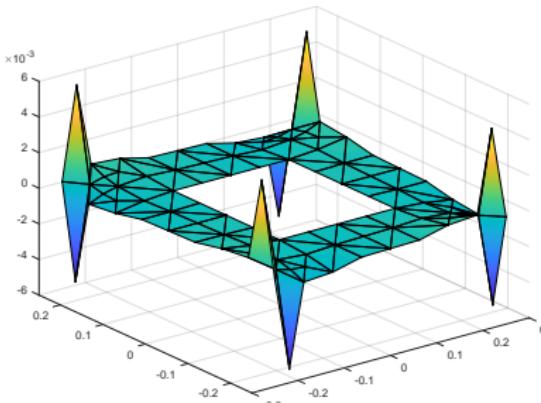
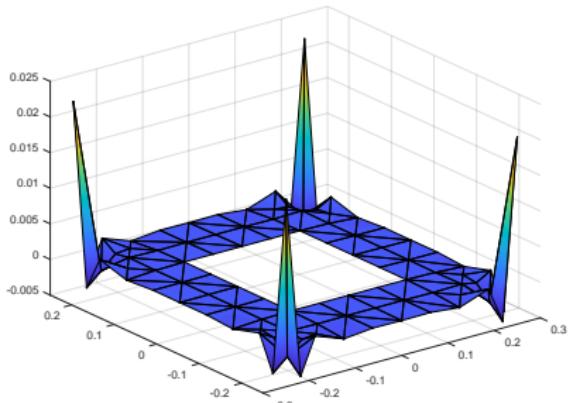
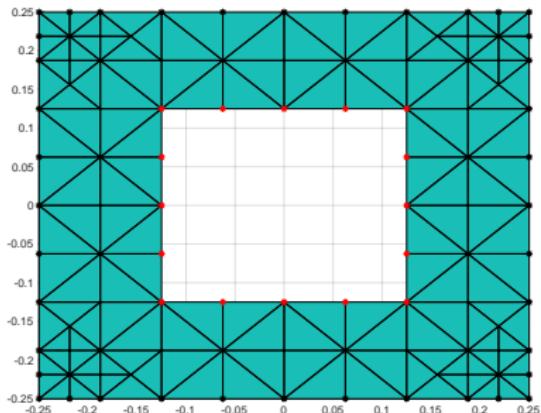
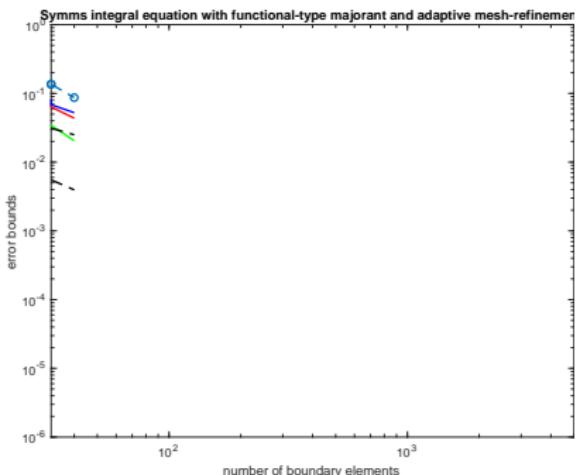
7. Refine marked elements \mathcal{M}_ℓ by newest vertex bisection to obtain $\mathcal{T}_{\ell+1}$

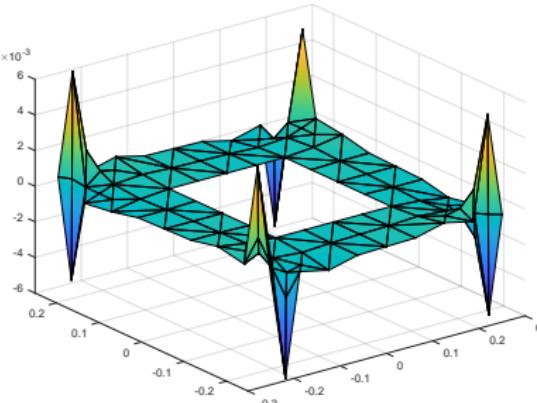
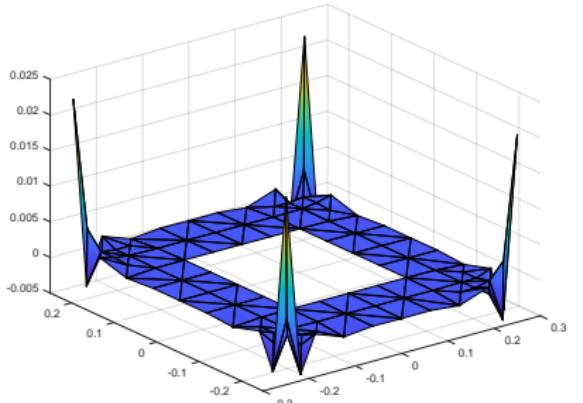
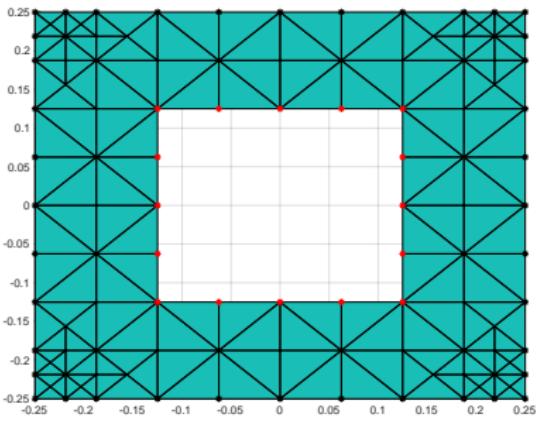
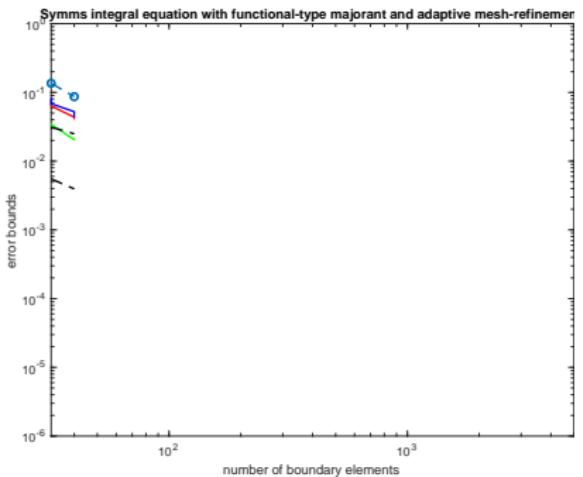


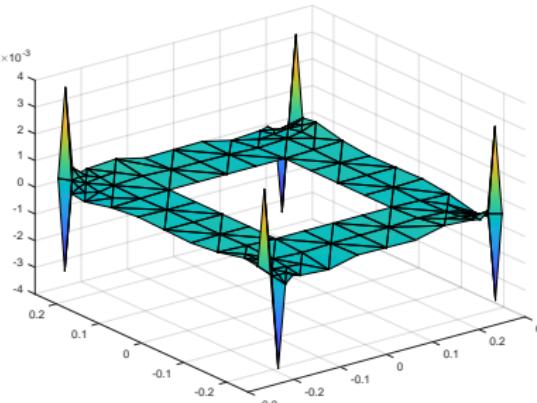
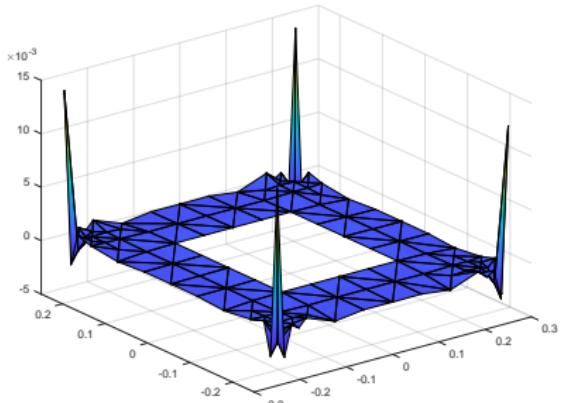
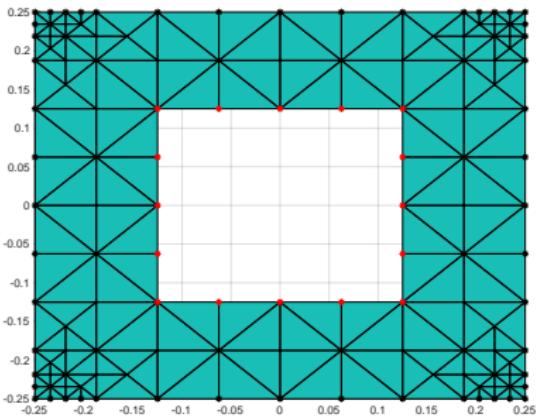
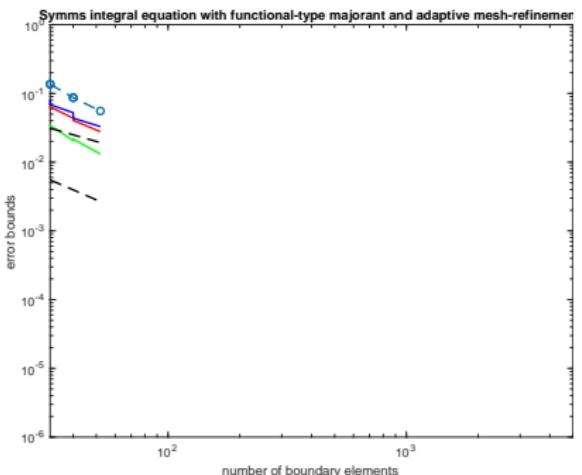
Unit Square

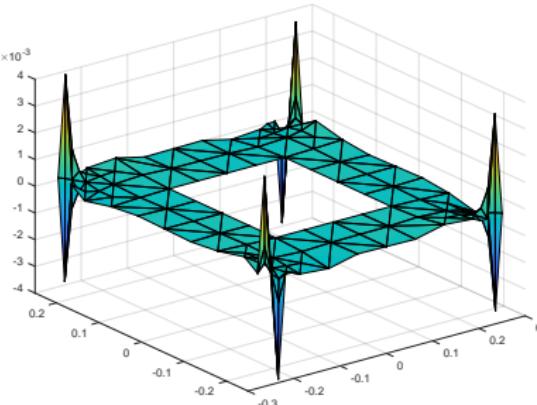
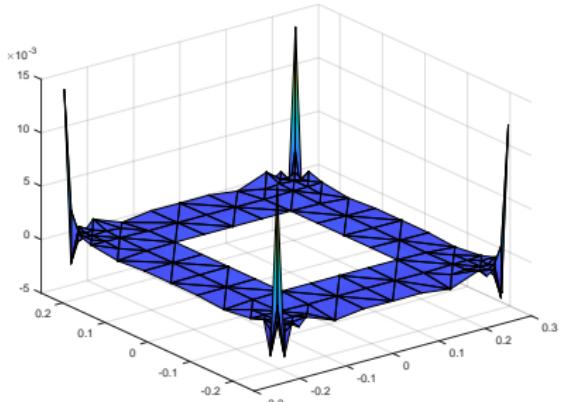
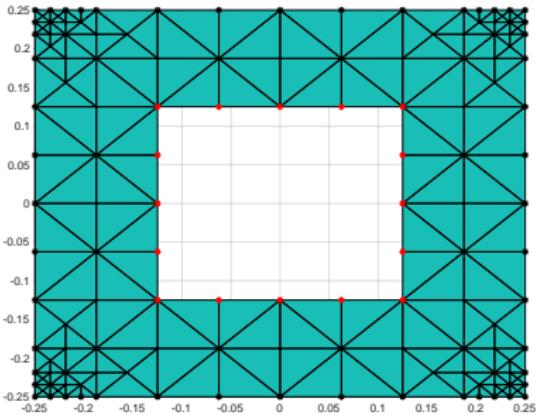
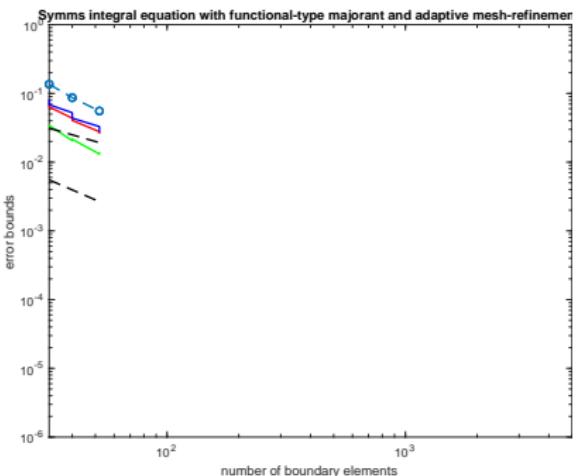


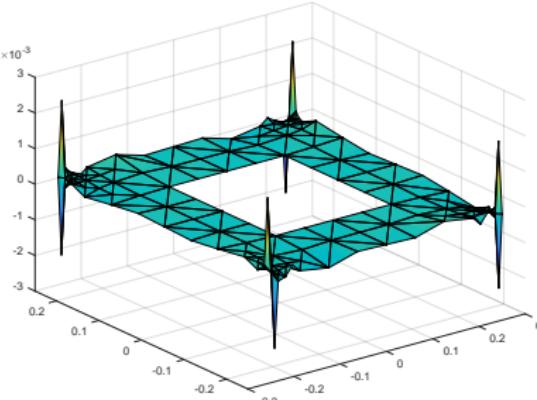
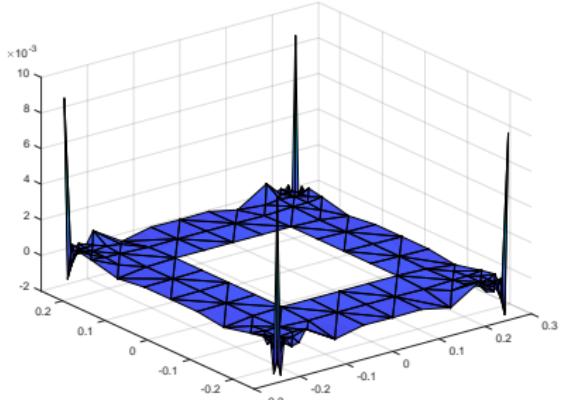
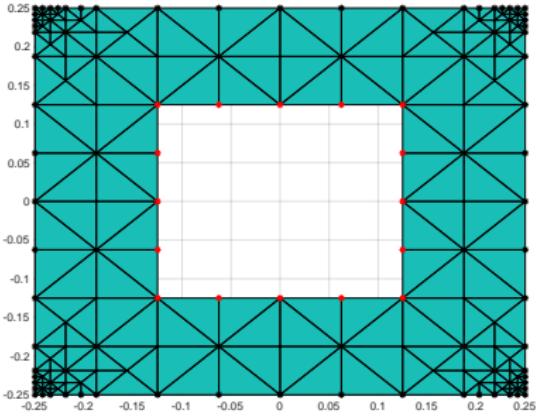
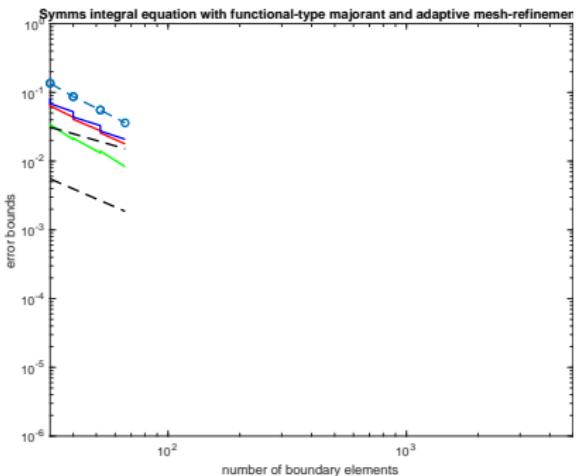


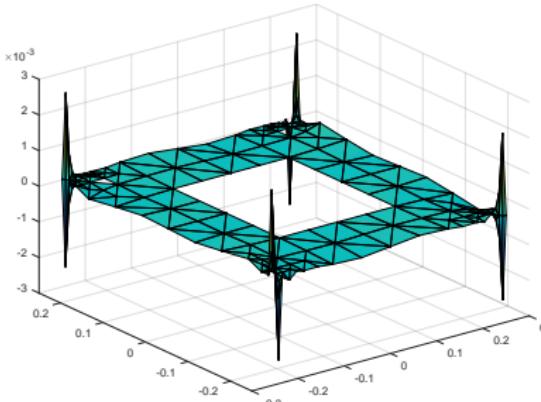
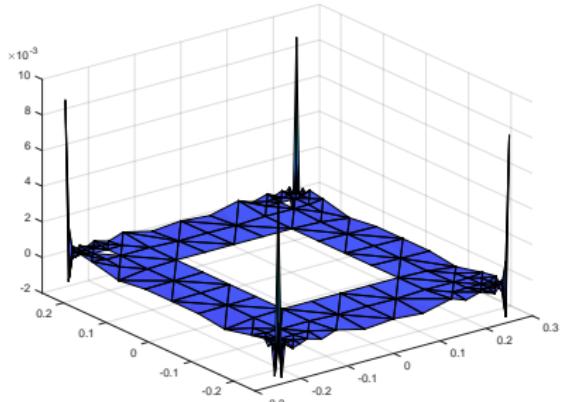
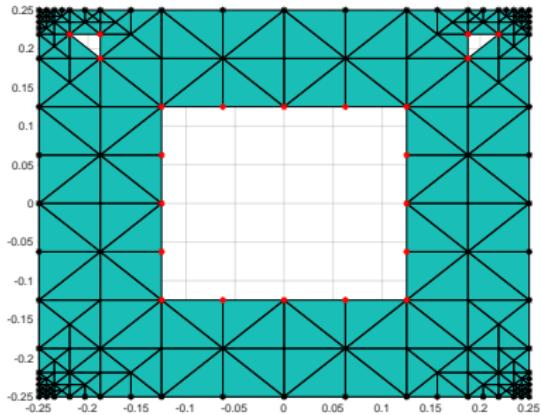
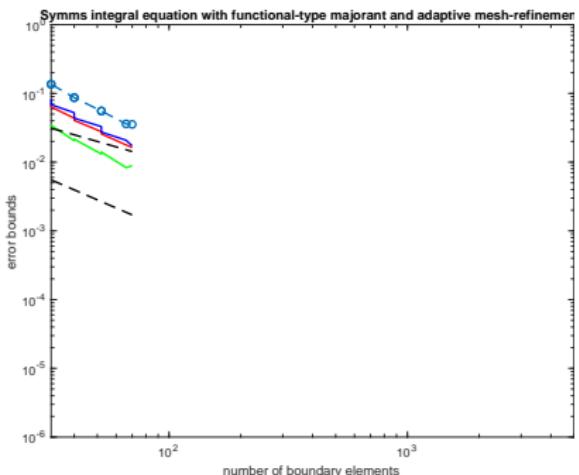


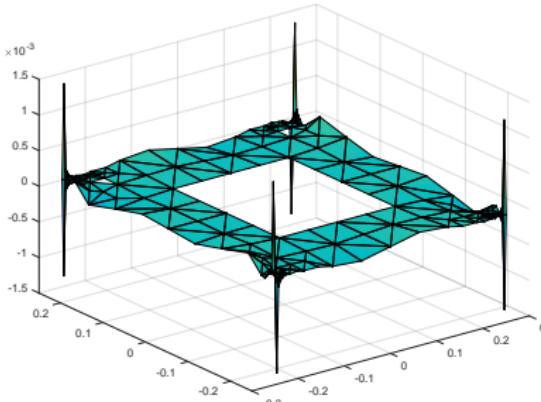
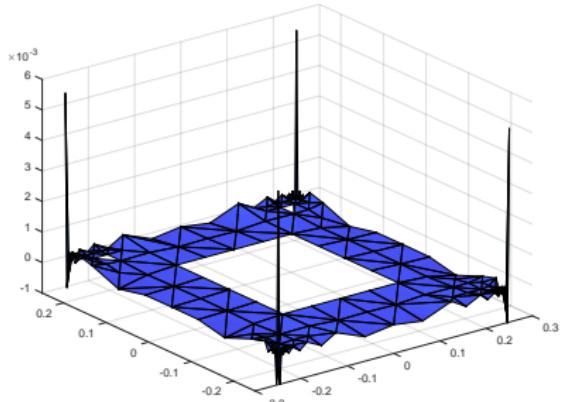
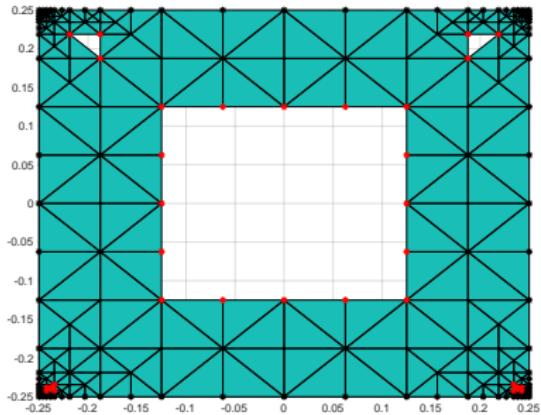
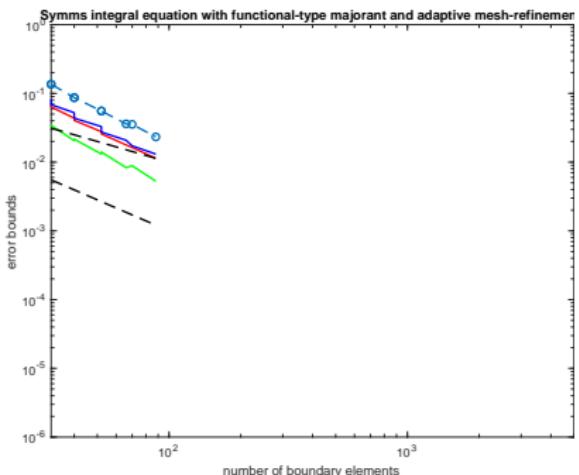


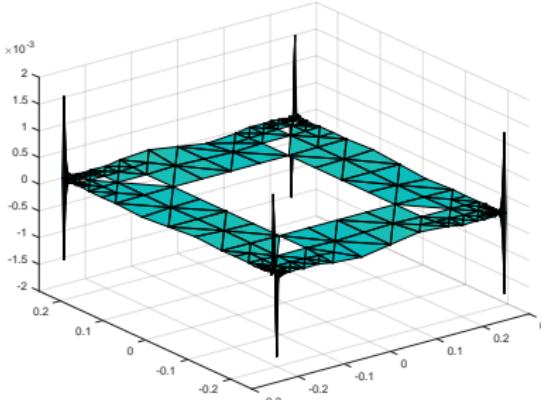
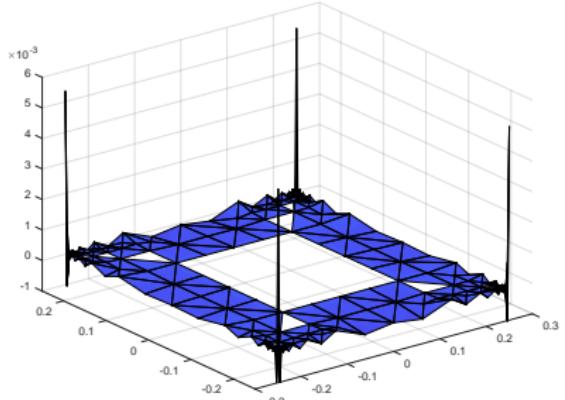
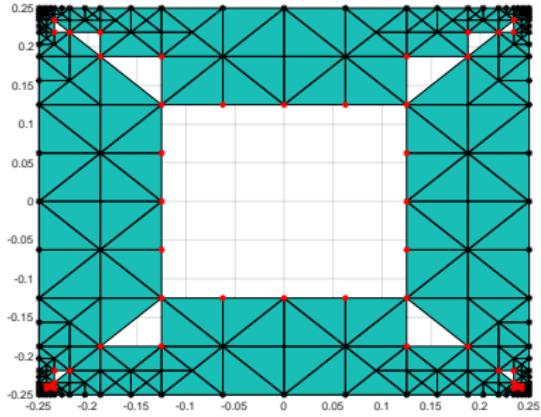
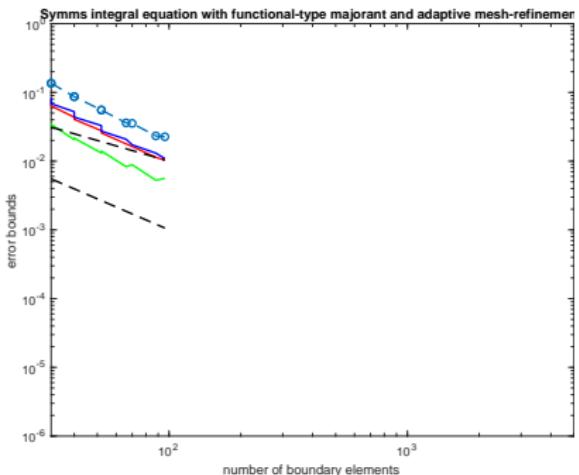


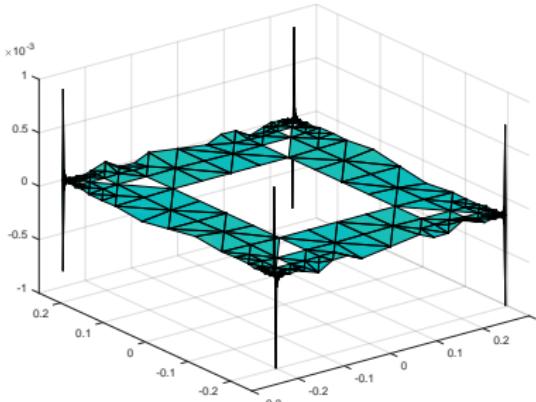
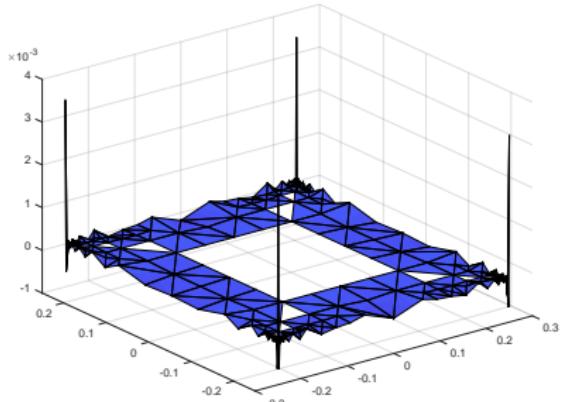
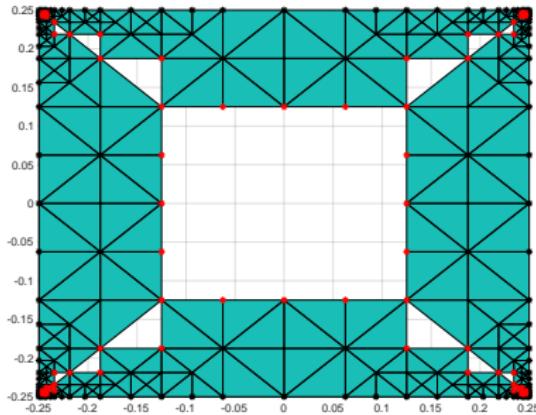
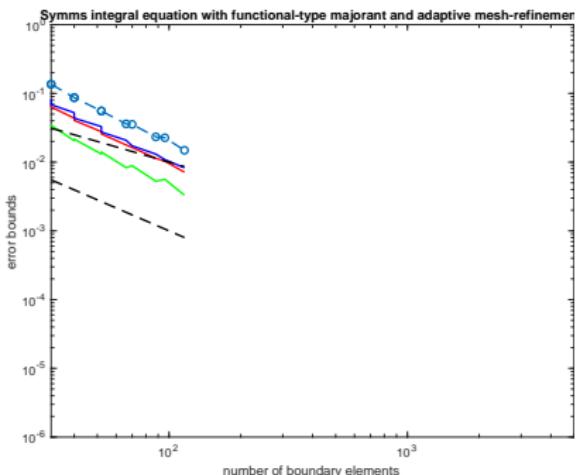


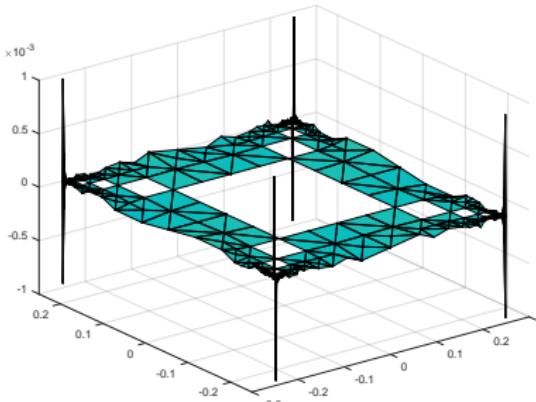
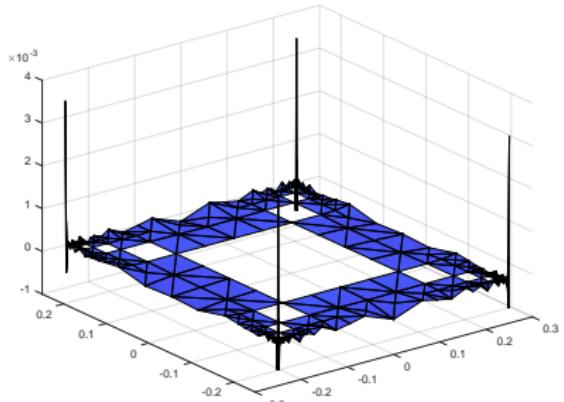
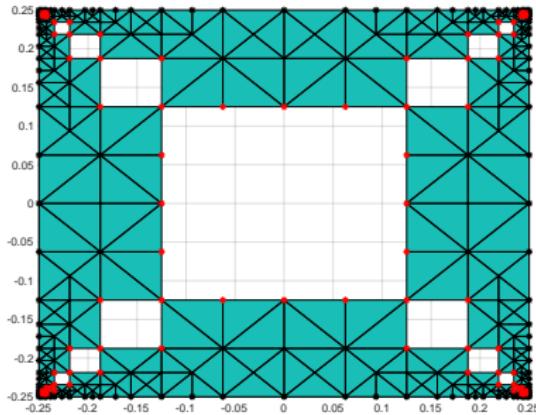
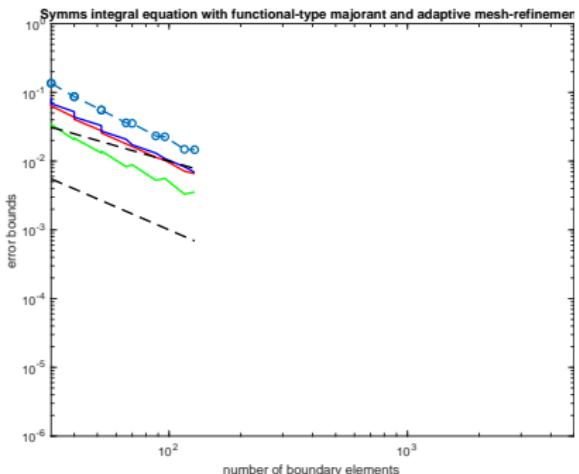


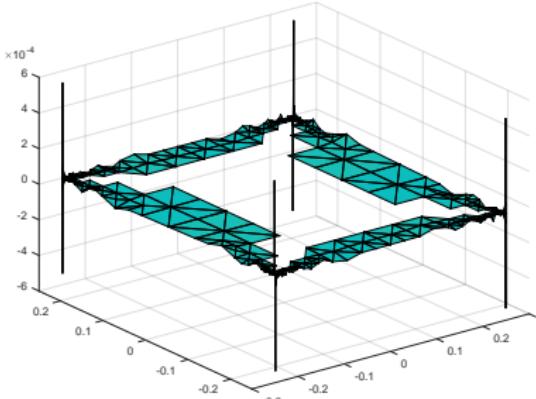
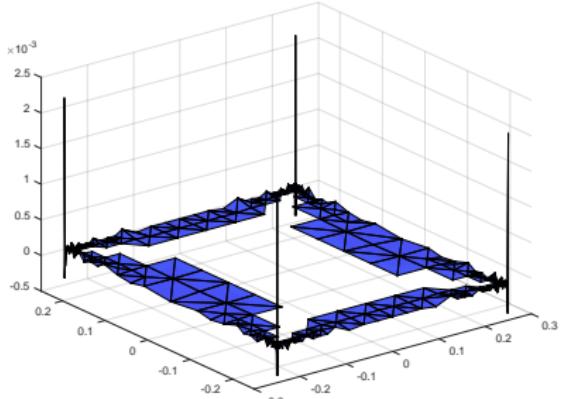
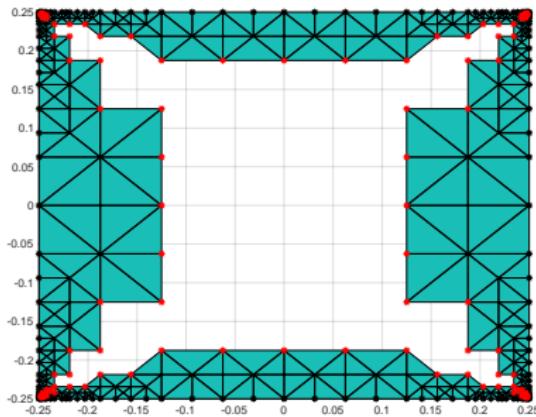
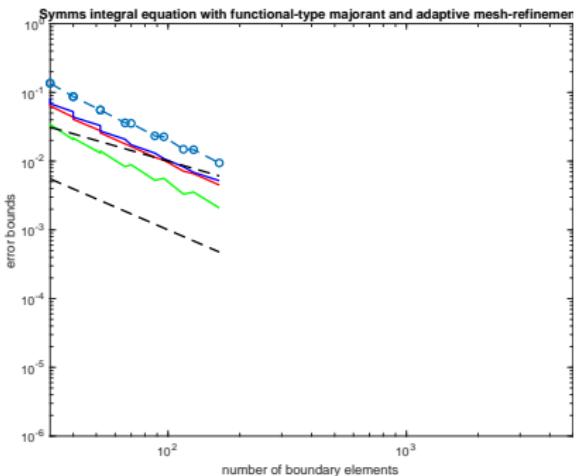


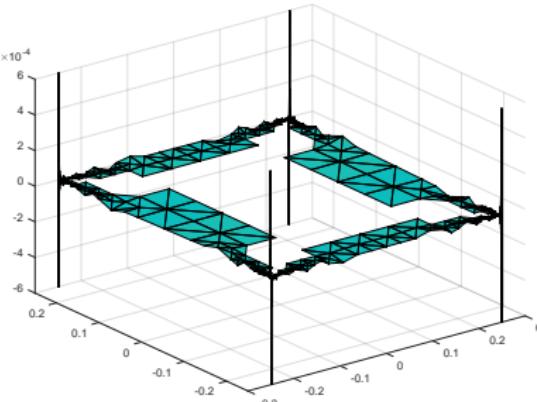
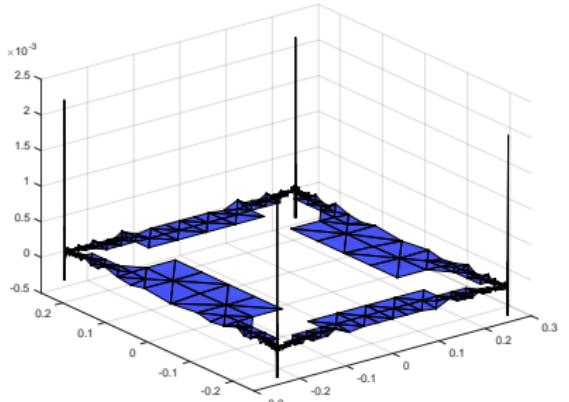
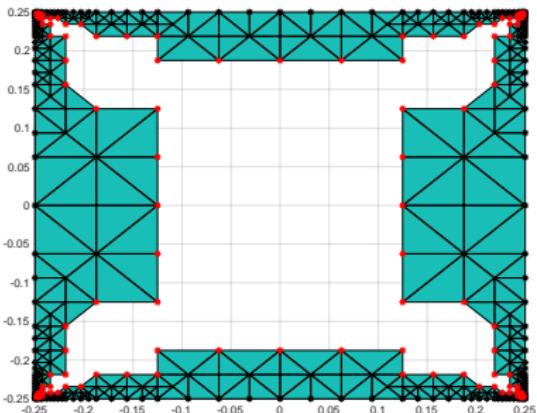
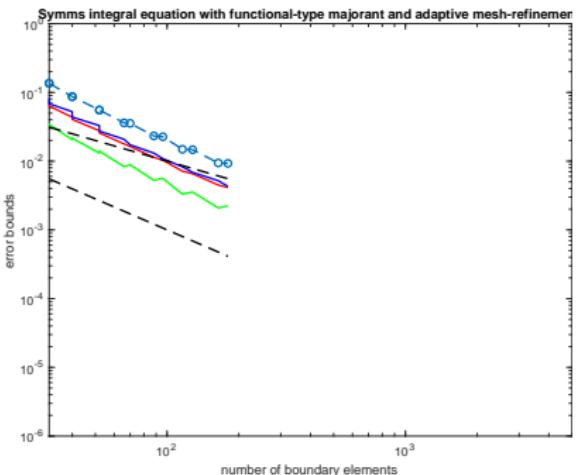


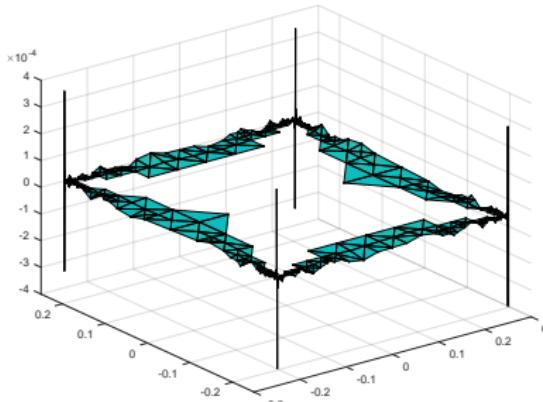
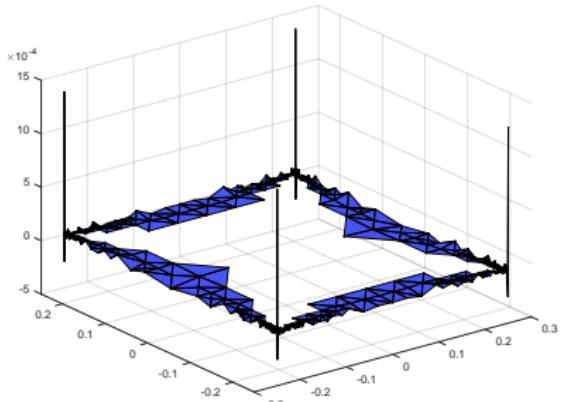
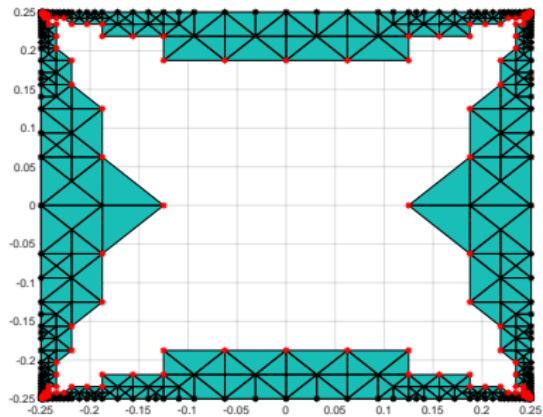
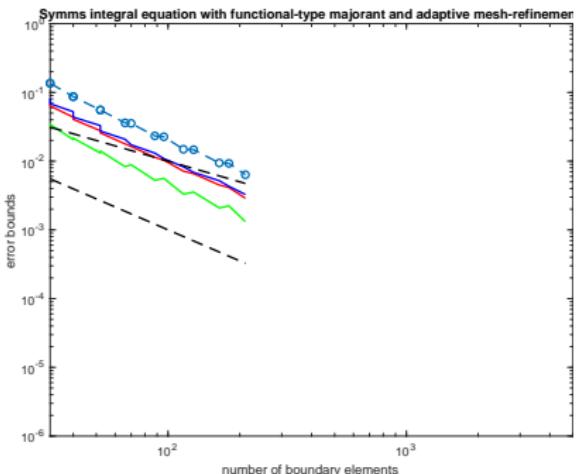


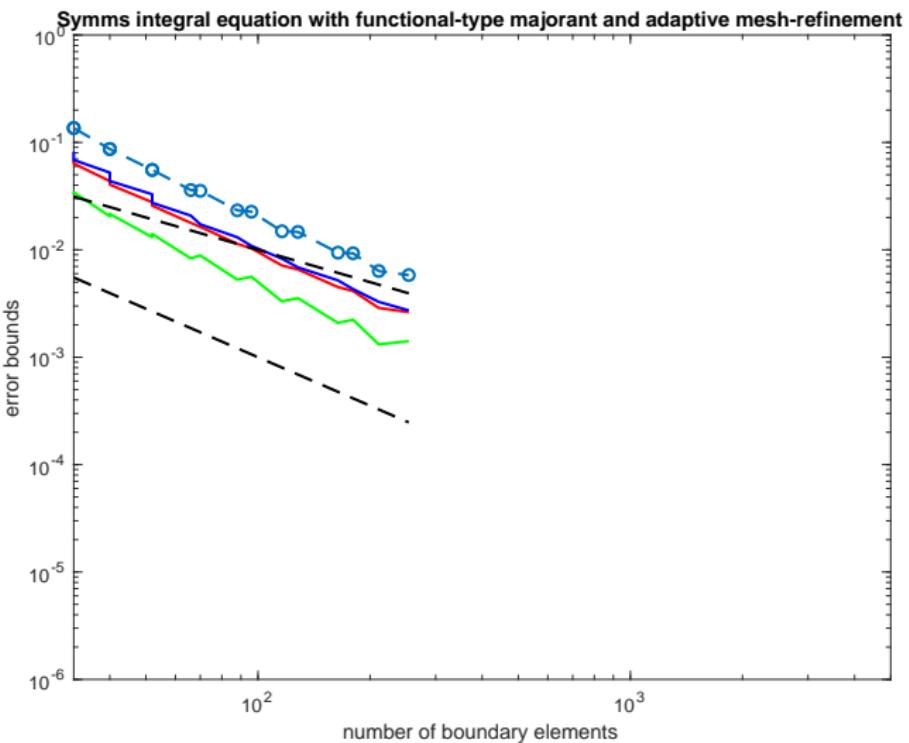


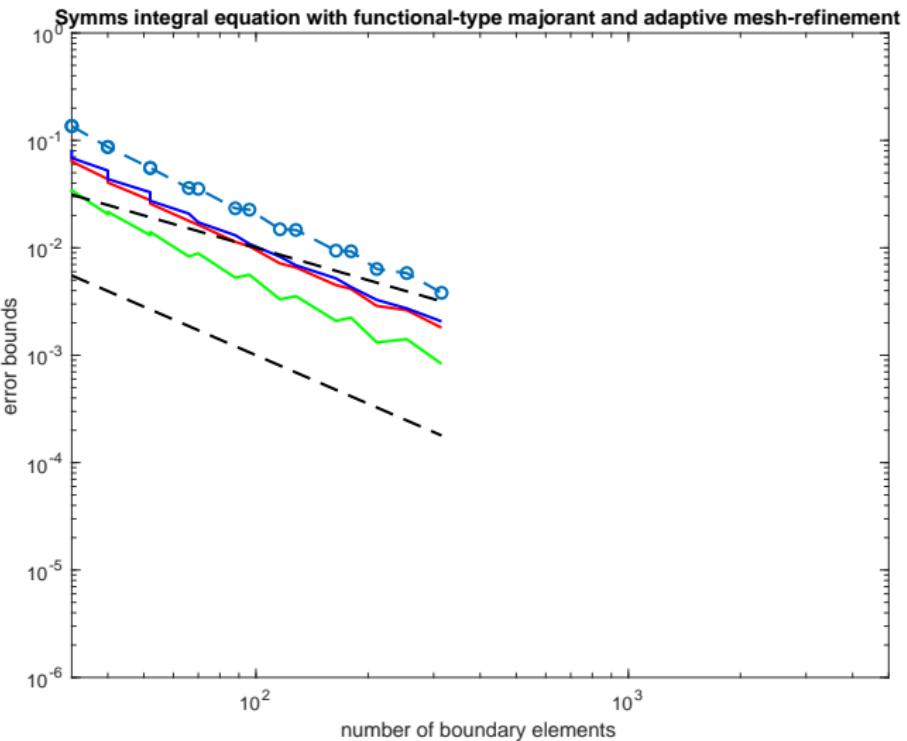


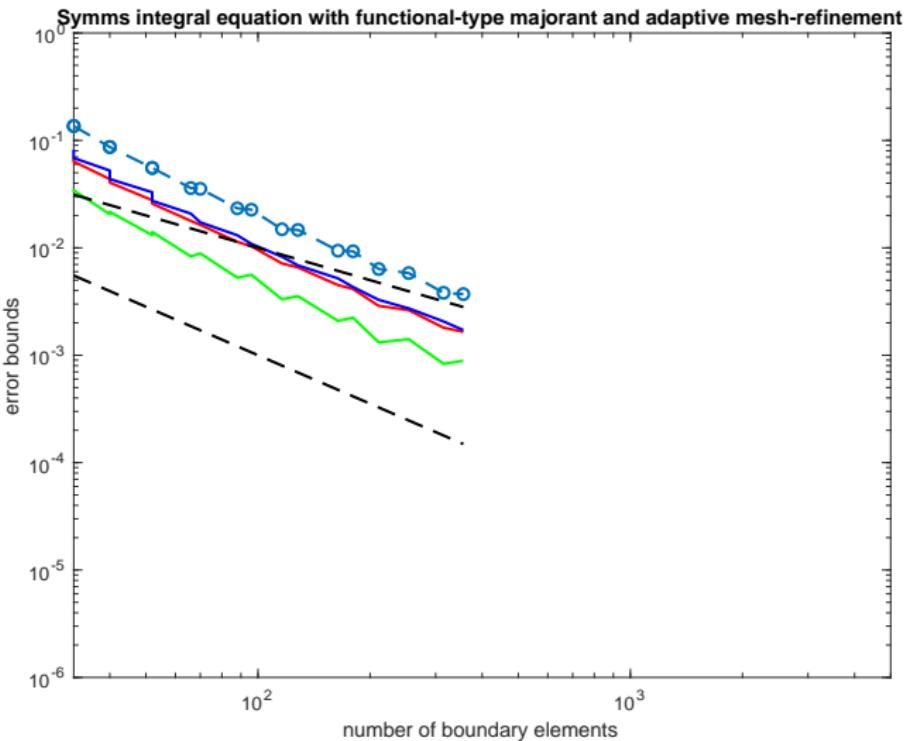


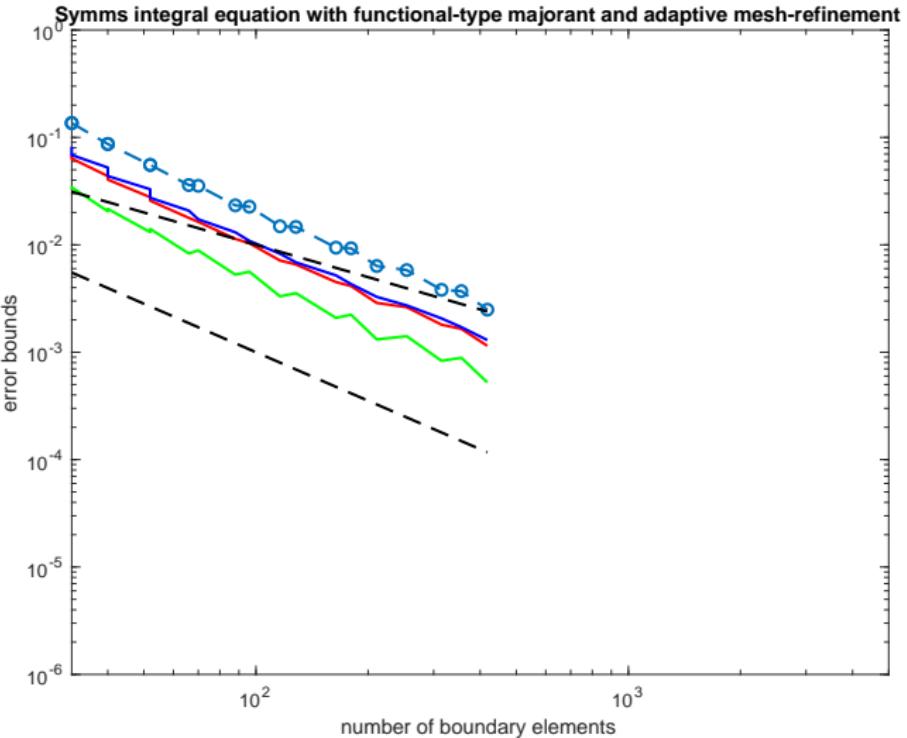


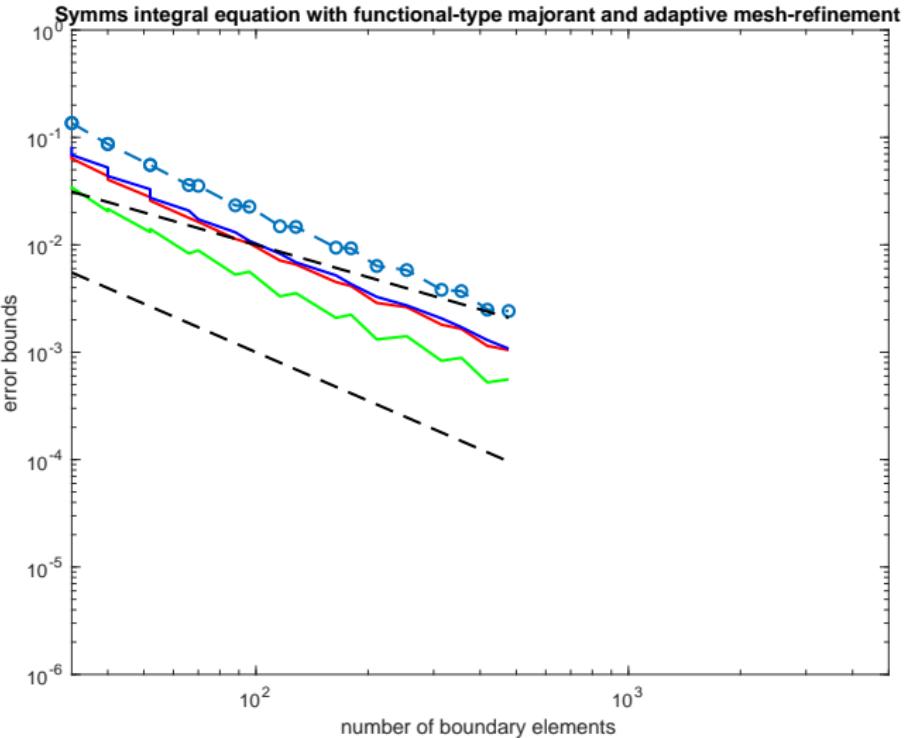


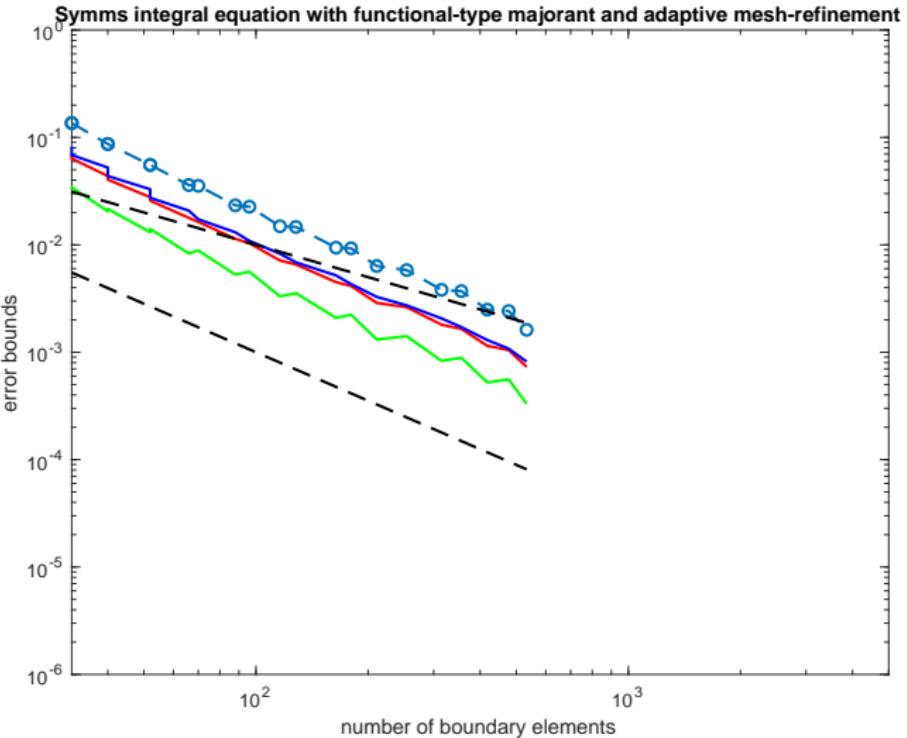


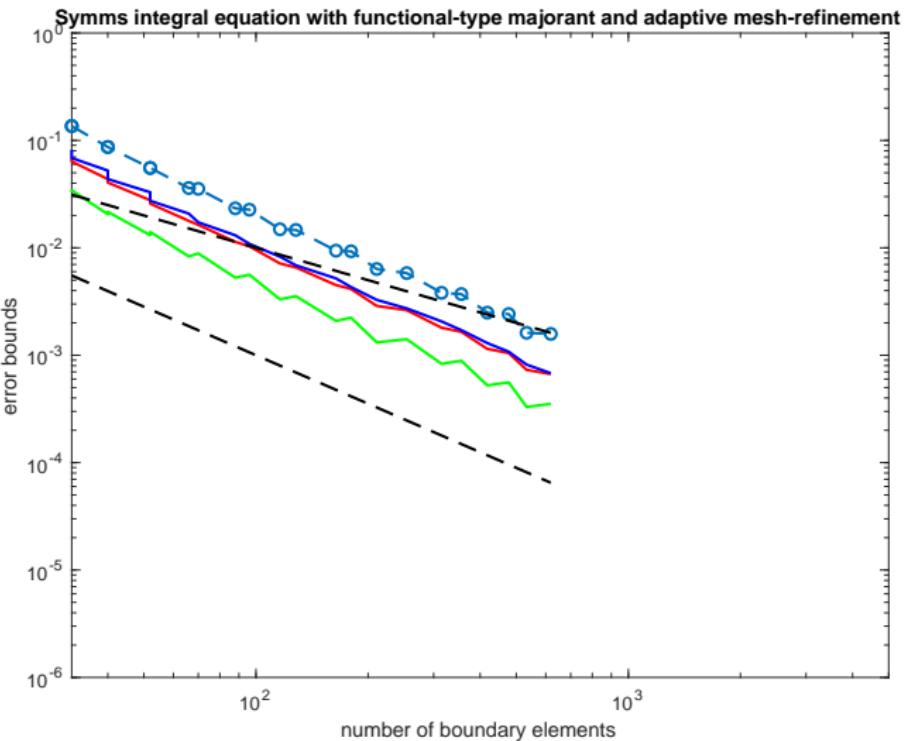


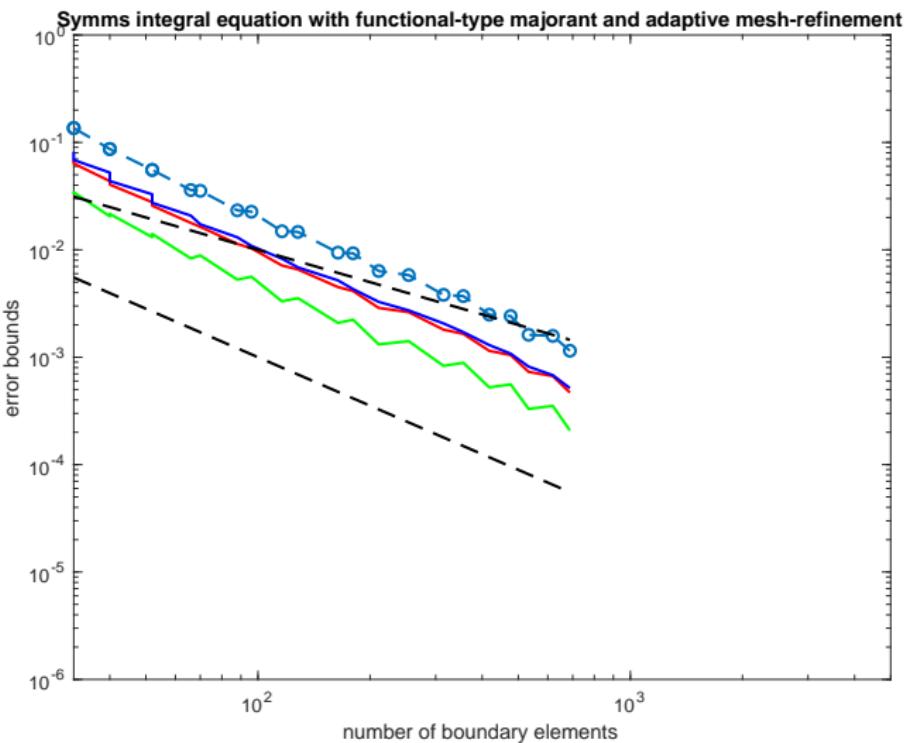


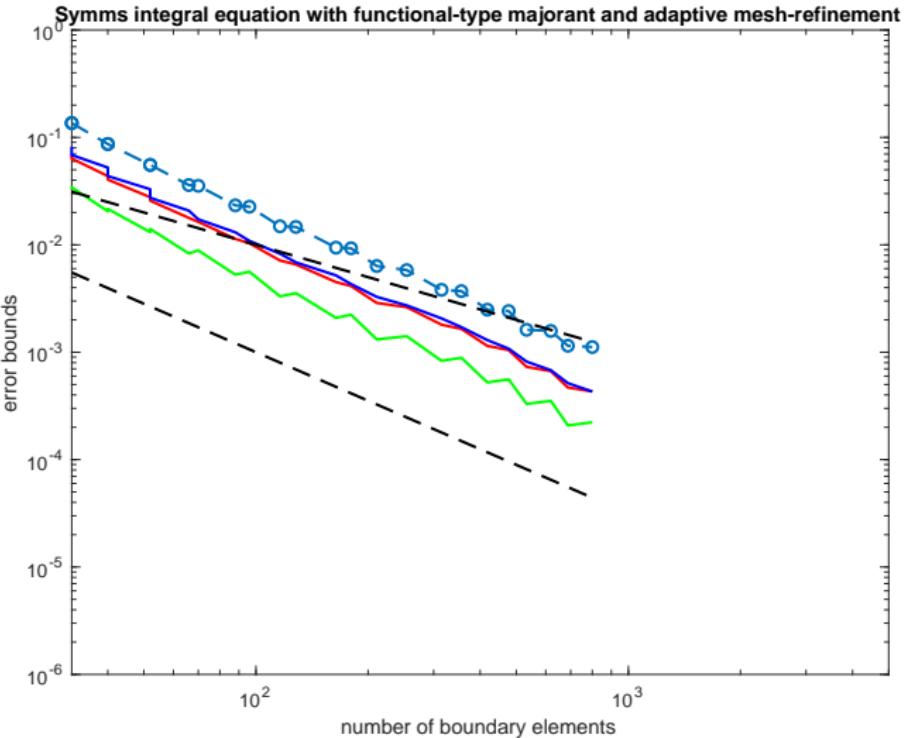


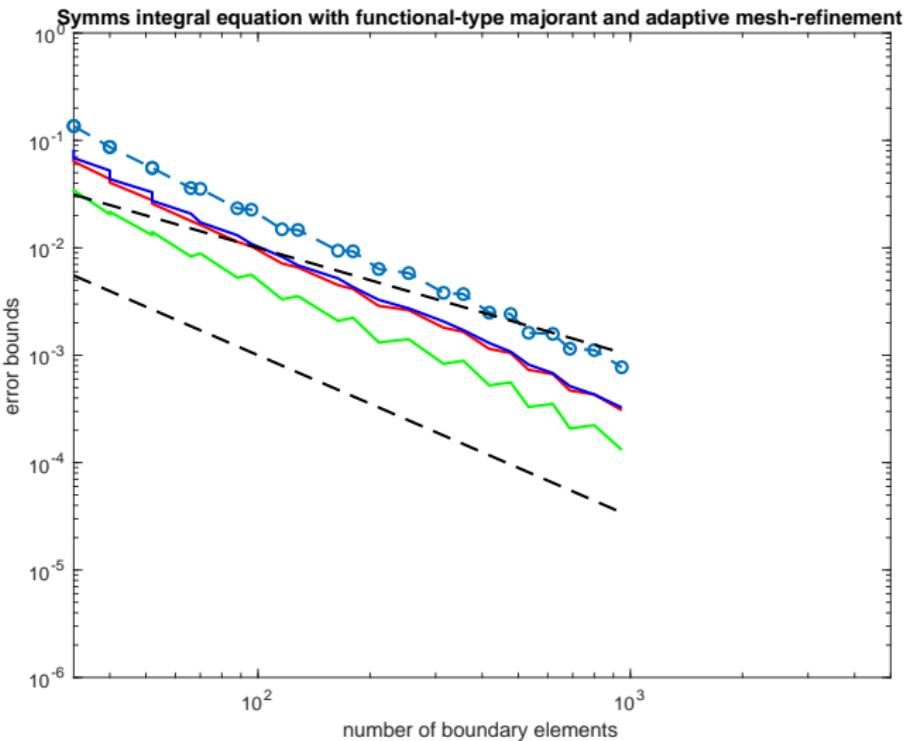


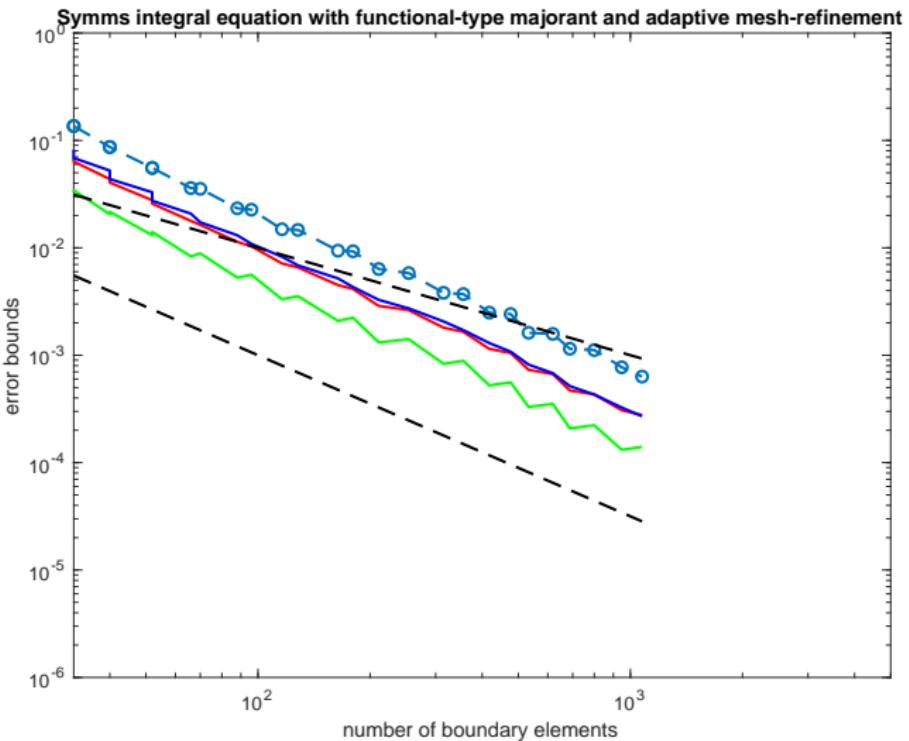


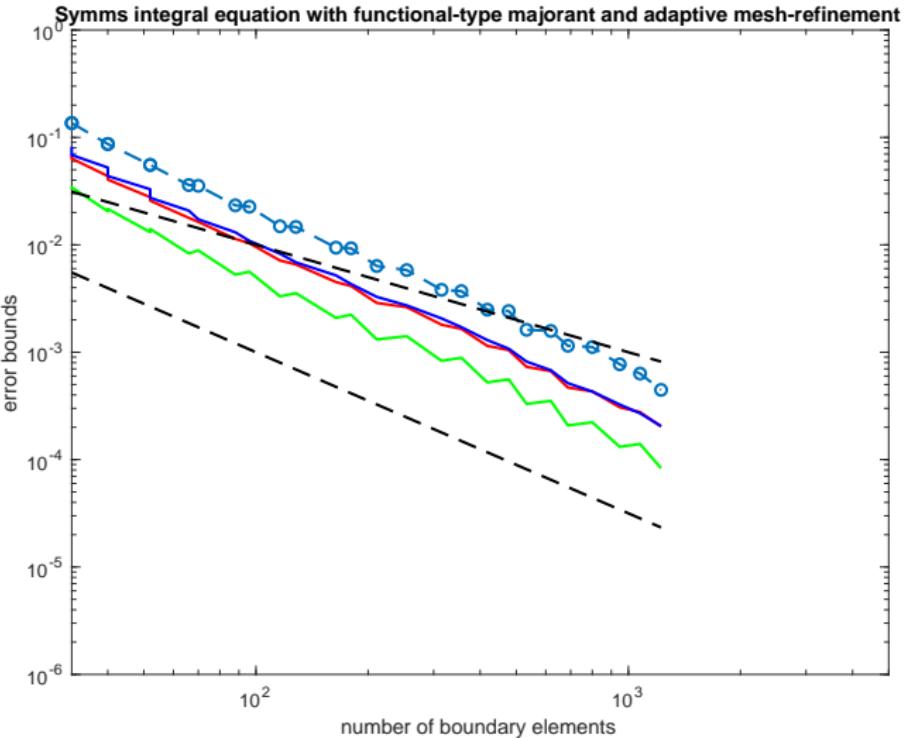


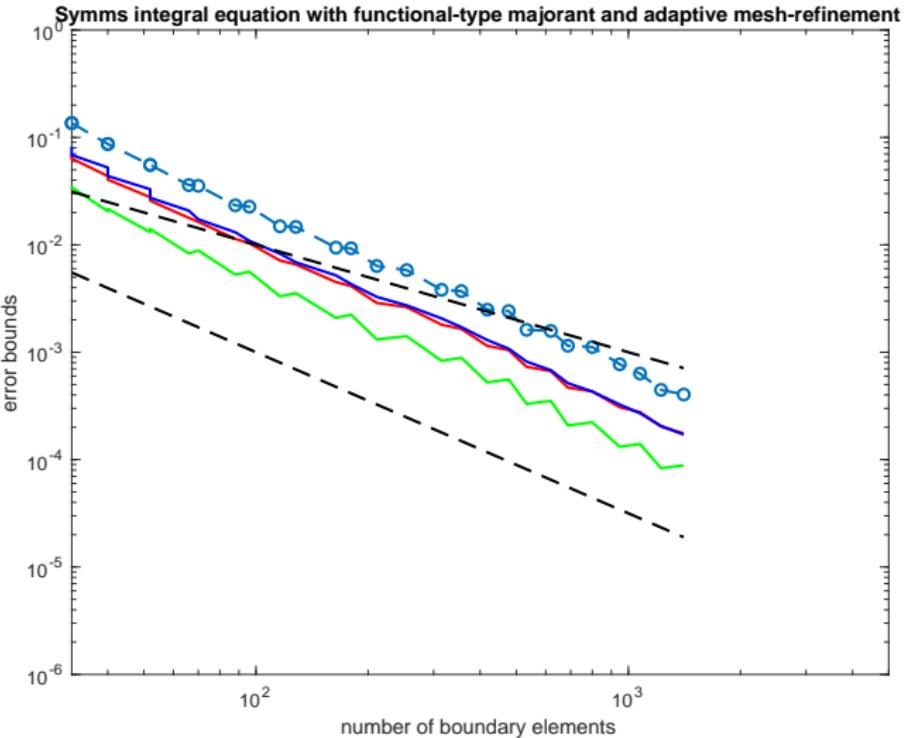


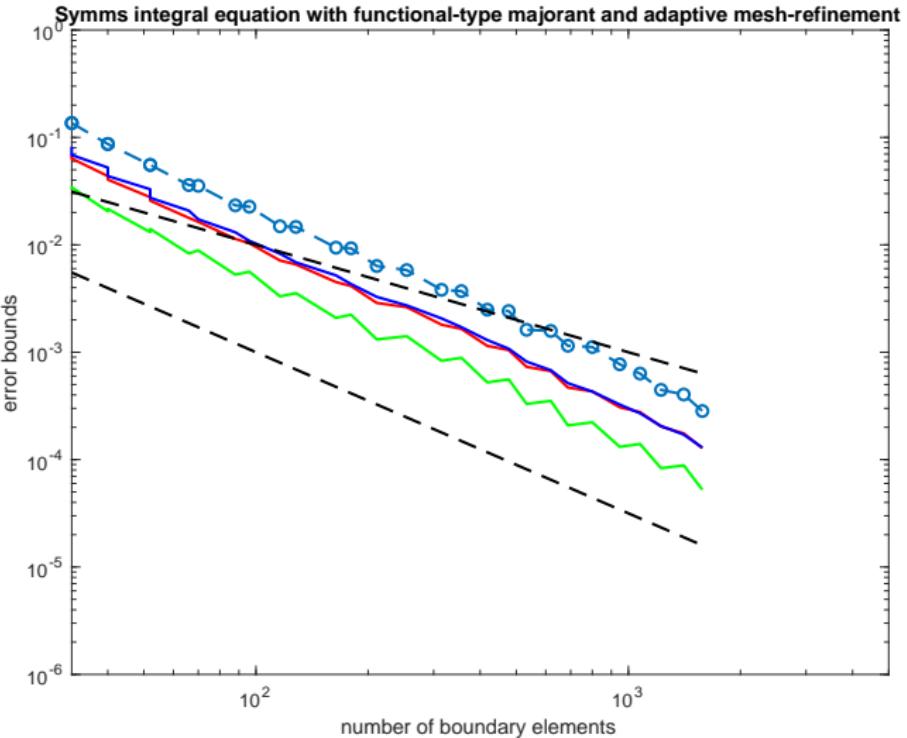


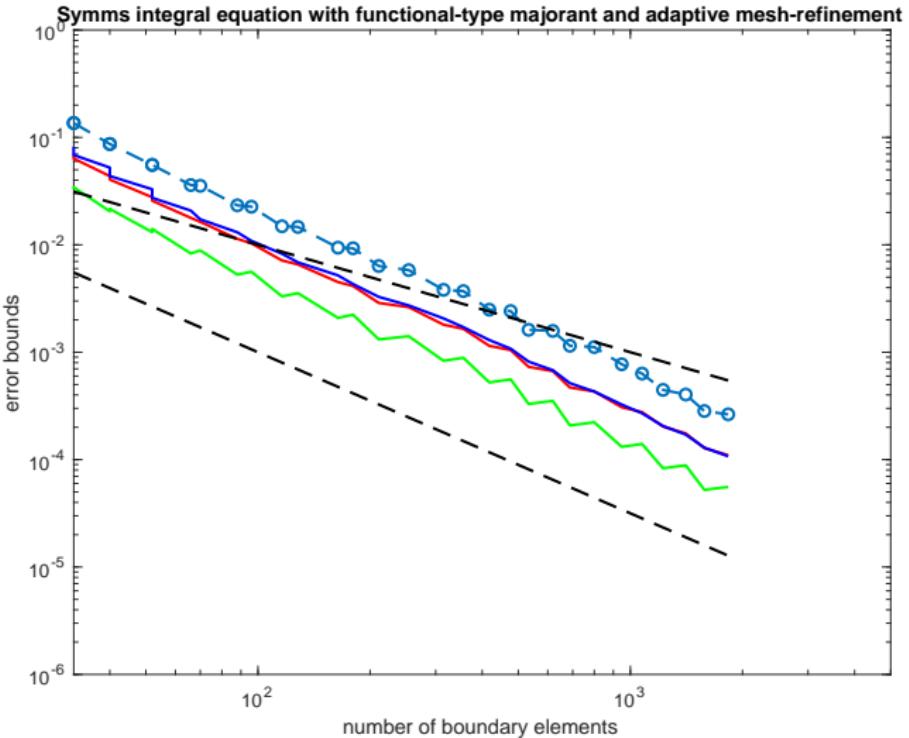


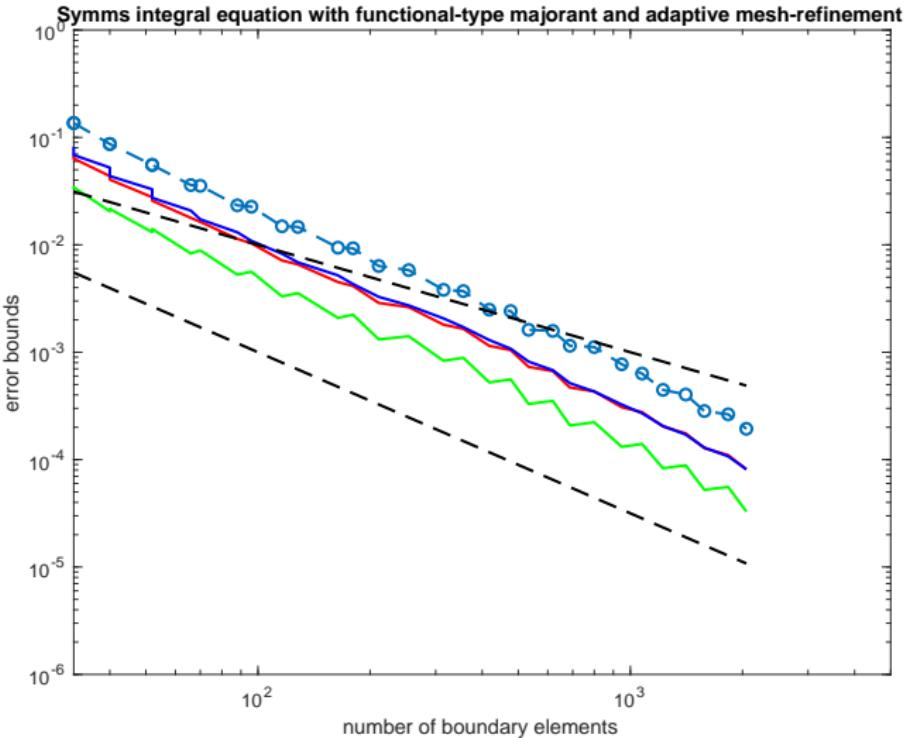


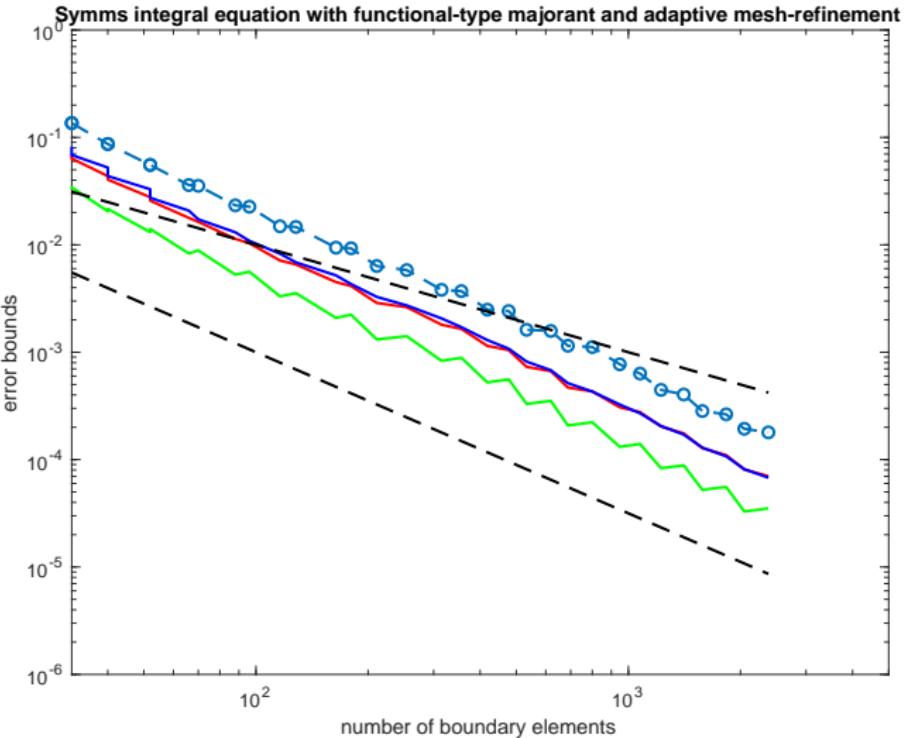


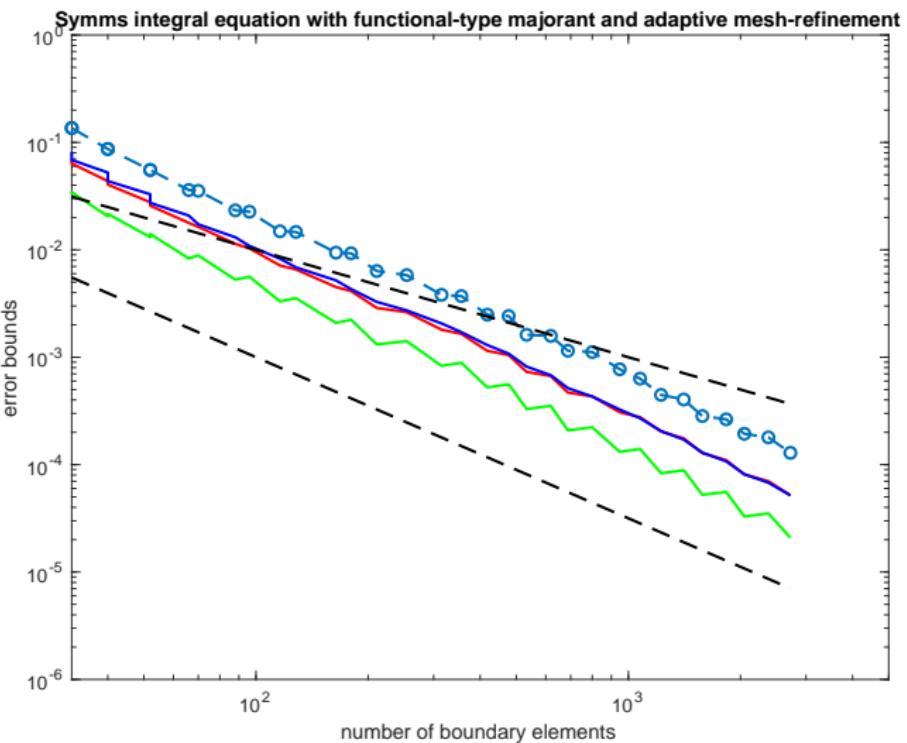


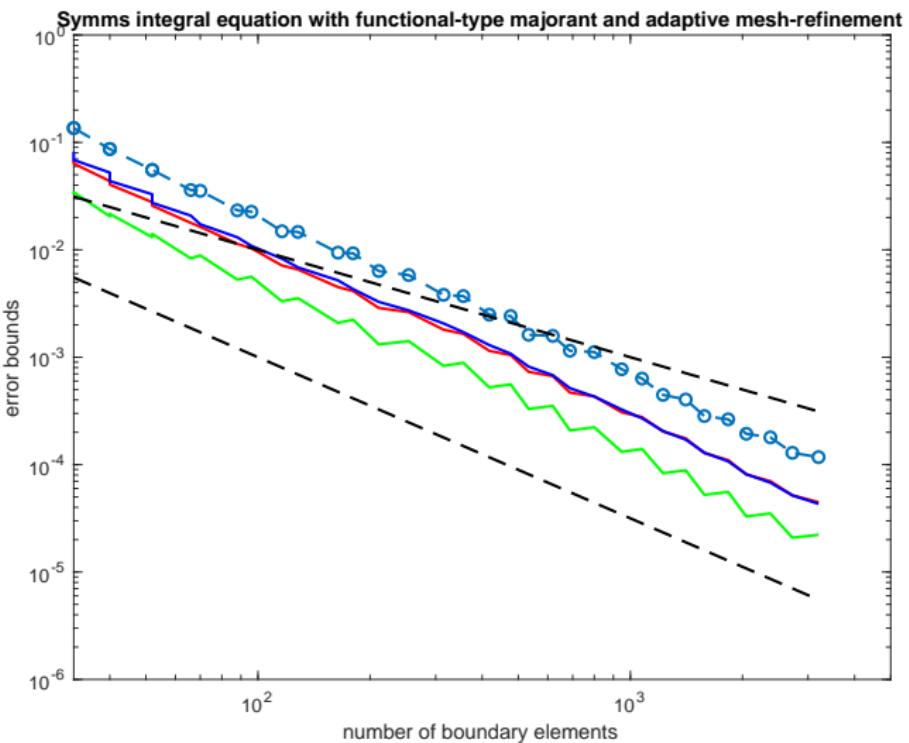


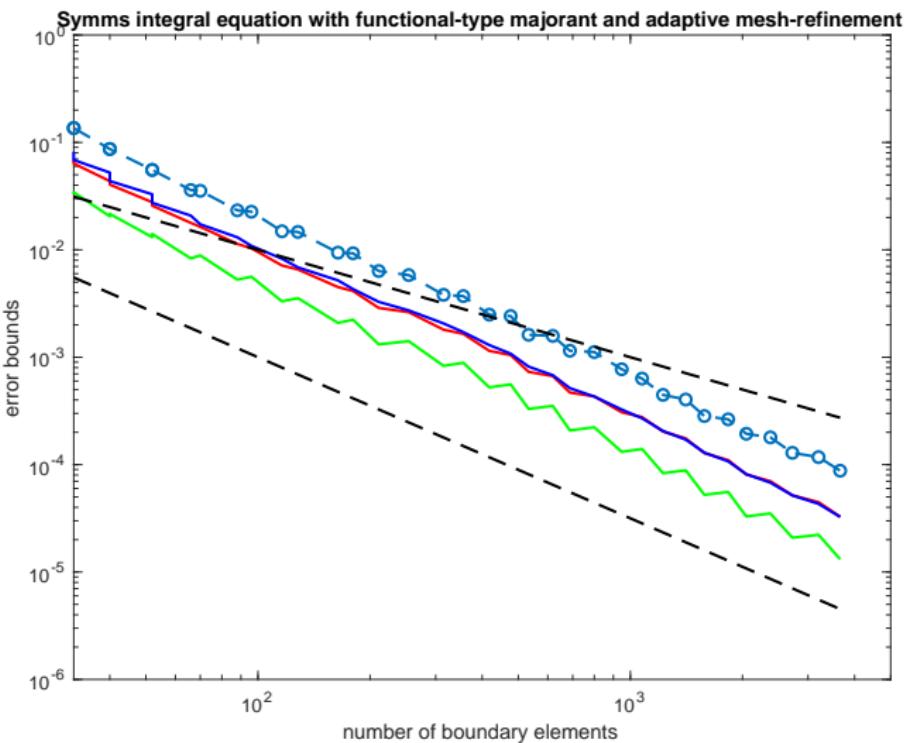


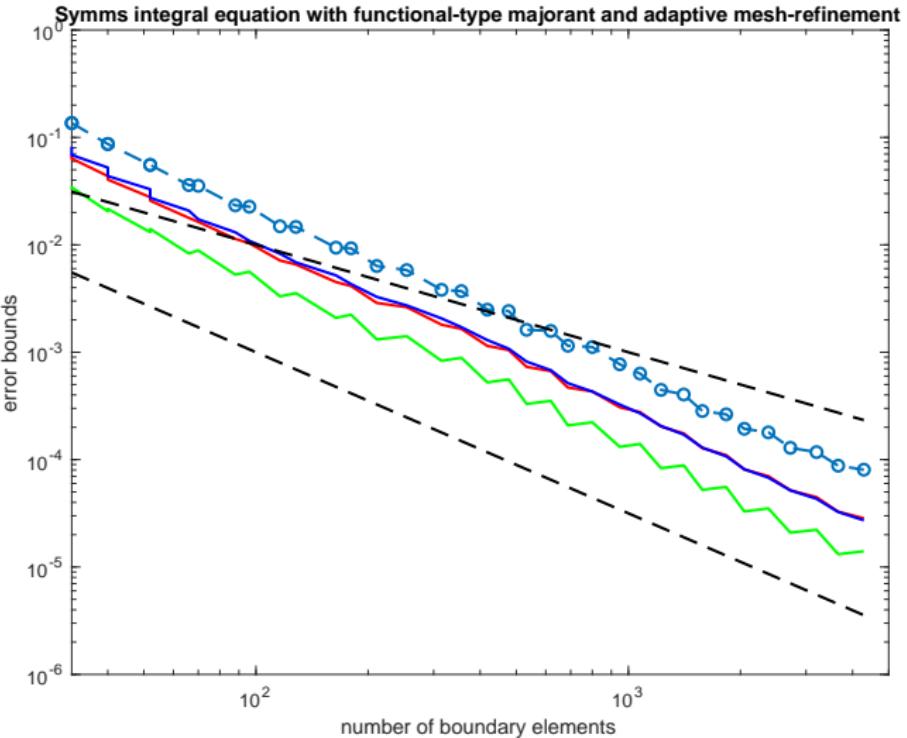


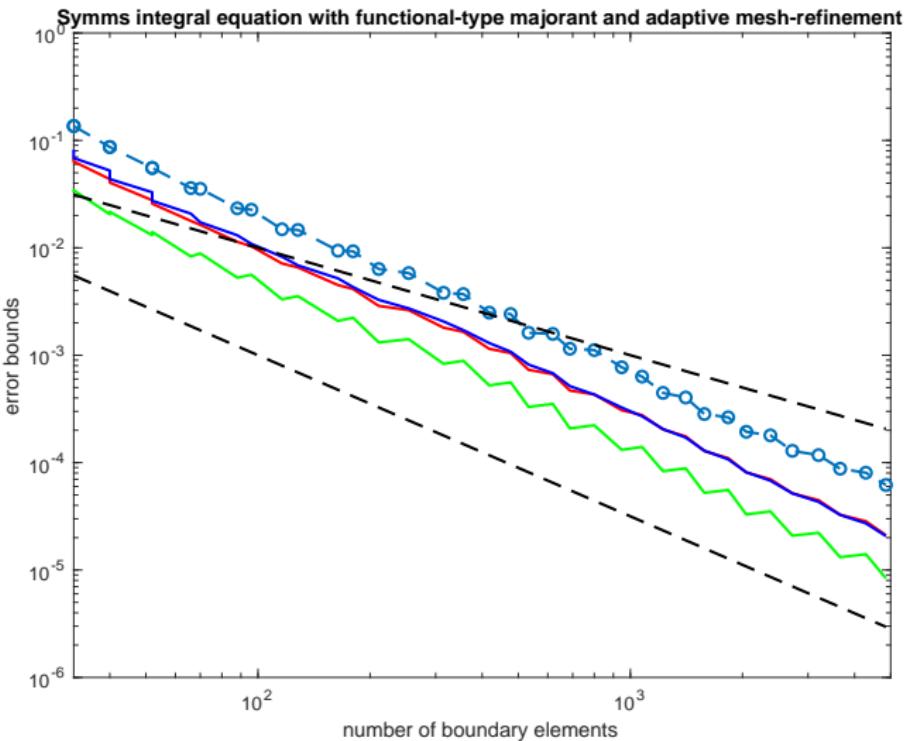












BEM example

Motivation

Error identity

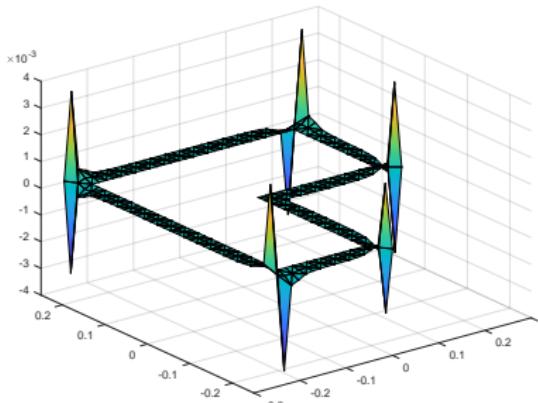
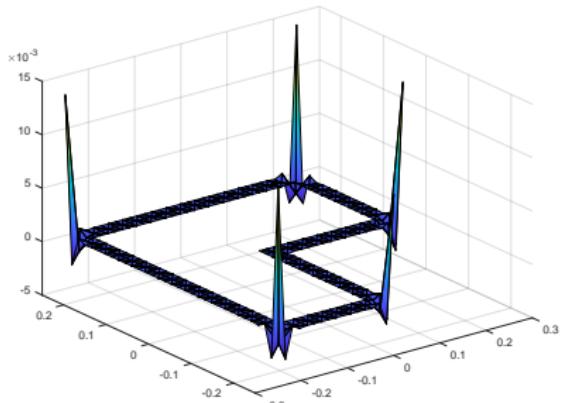
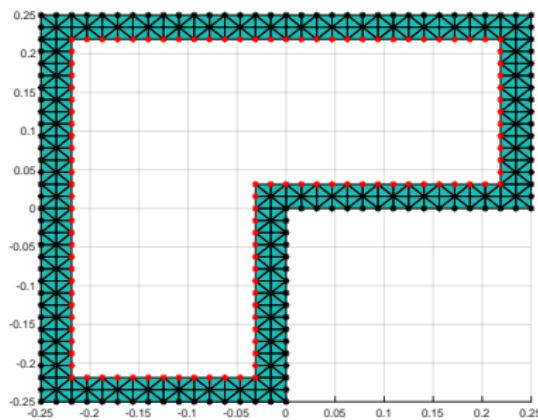
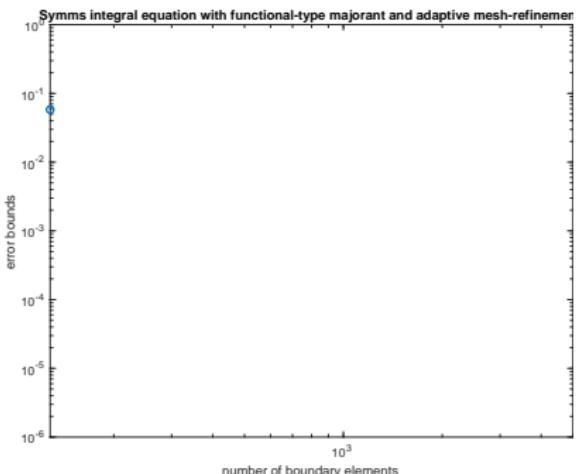
Majorant

Minorant

Numerical
experiments

Outlook

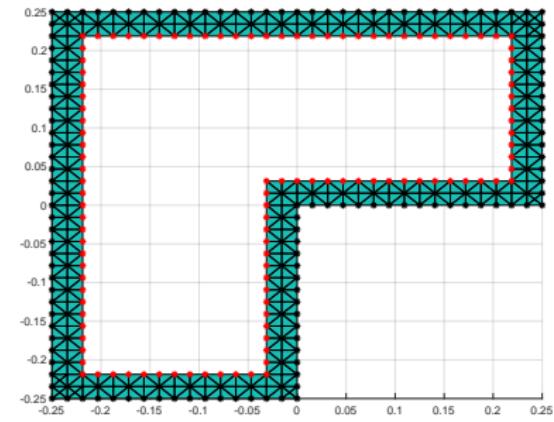
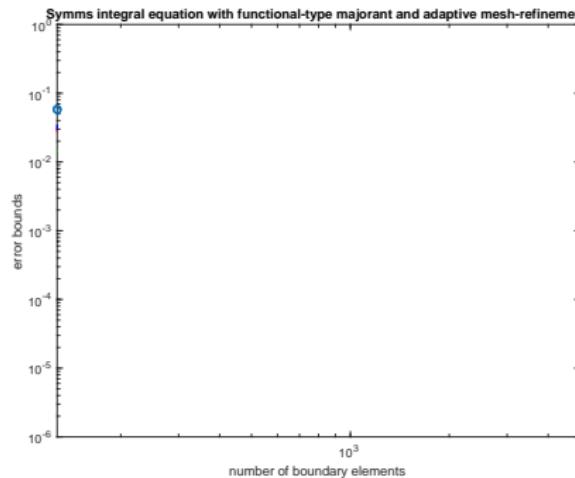
L-Shaped



BEM example**Motivation****Error identity**

Majorant

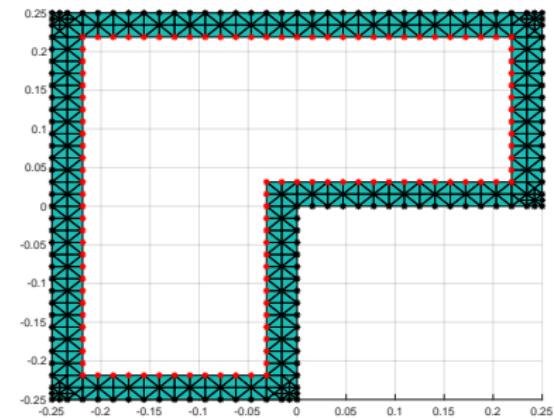
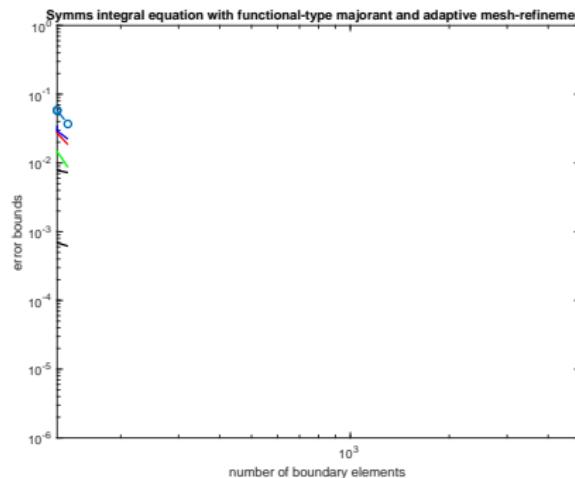
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

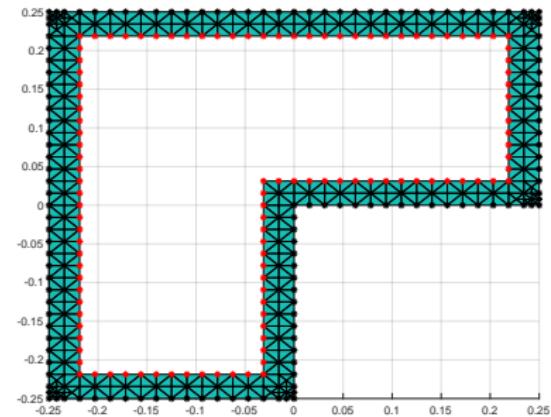
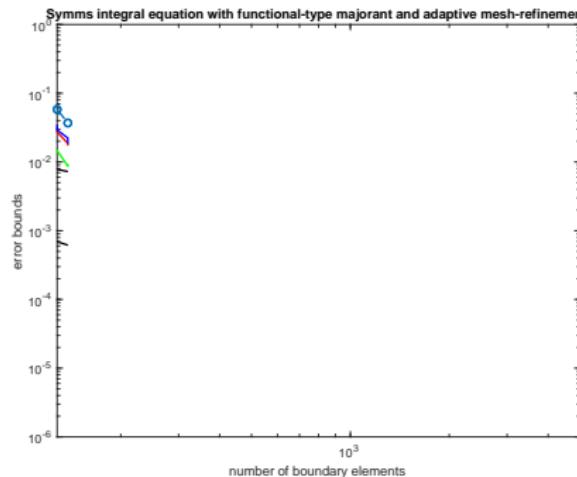
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

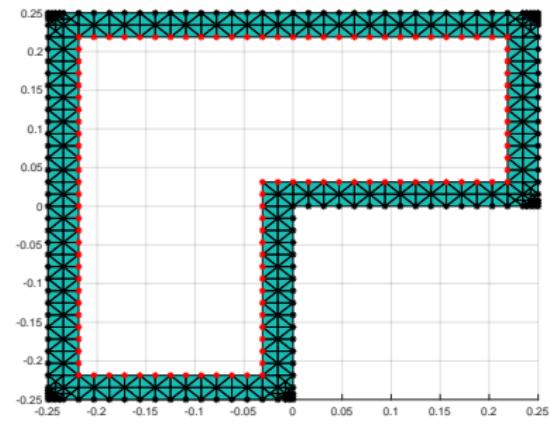
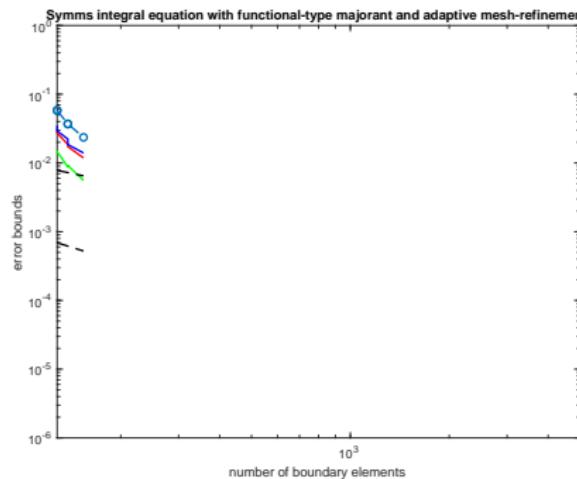
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

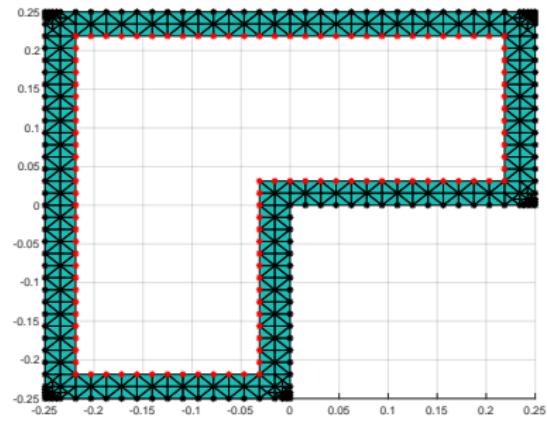
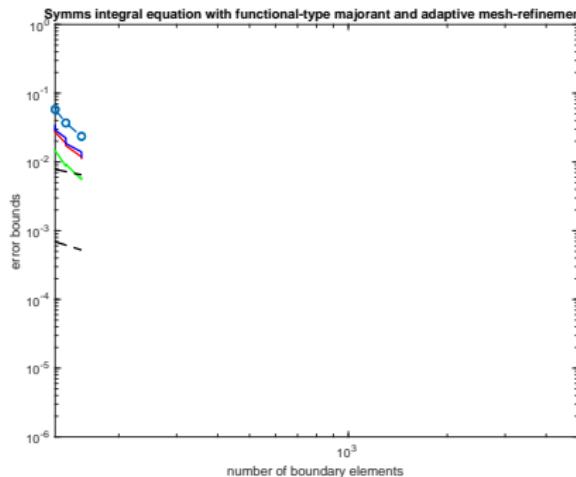
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

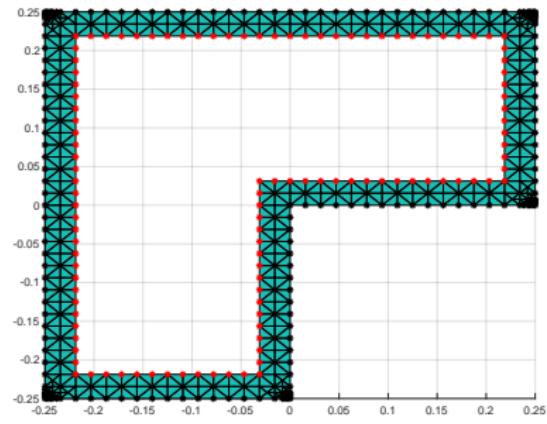
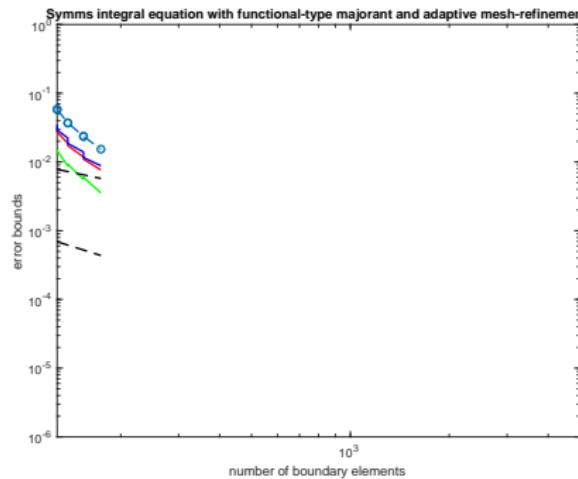
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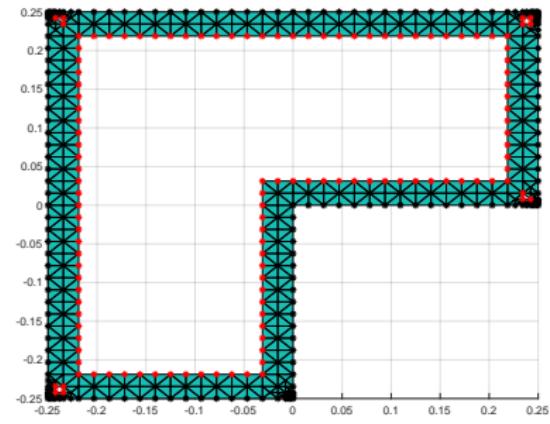
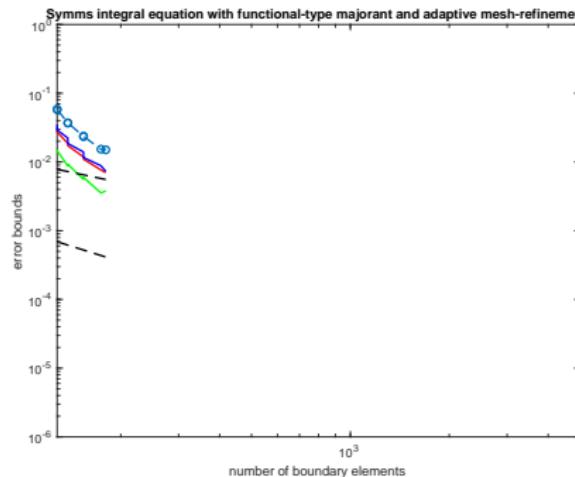
**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

Minorant

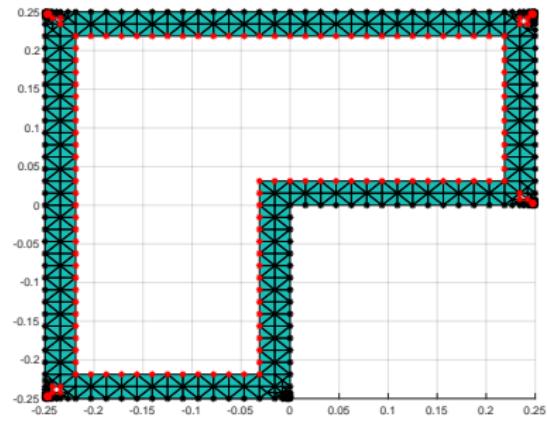
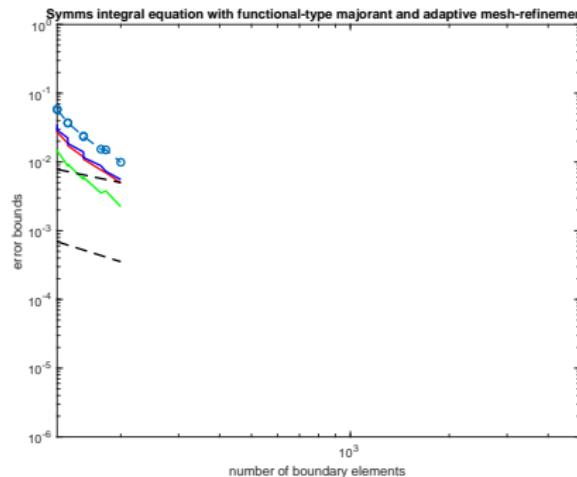
**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

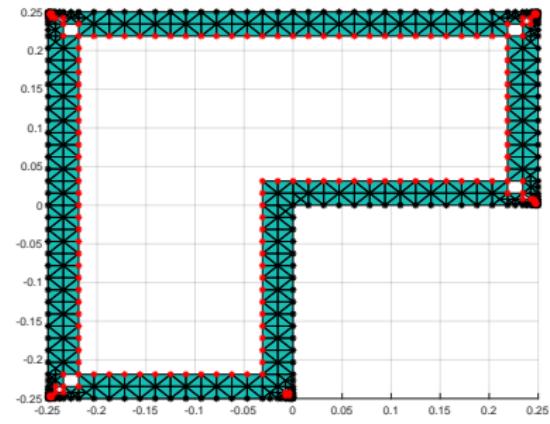
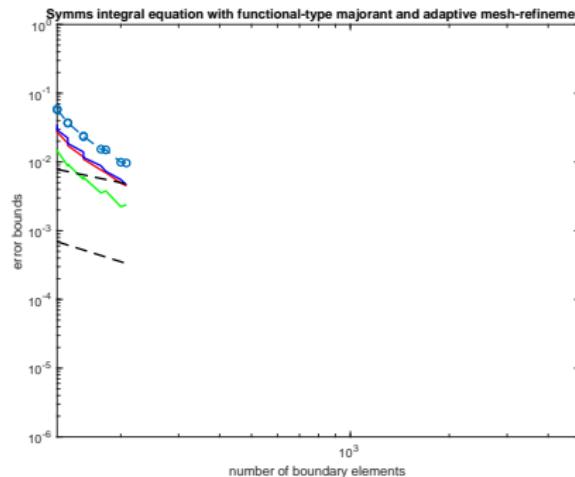
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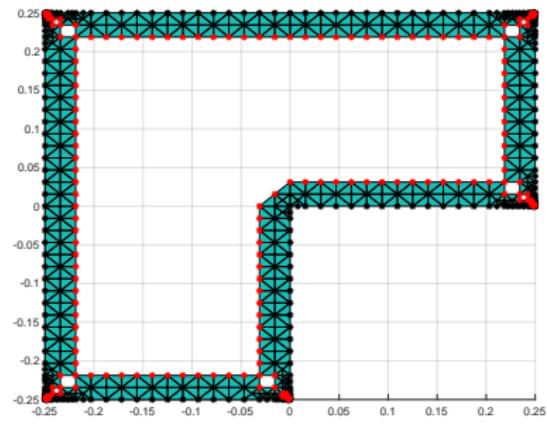
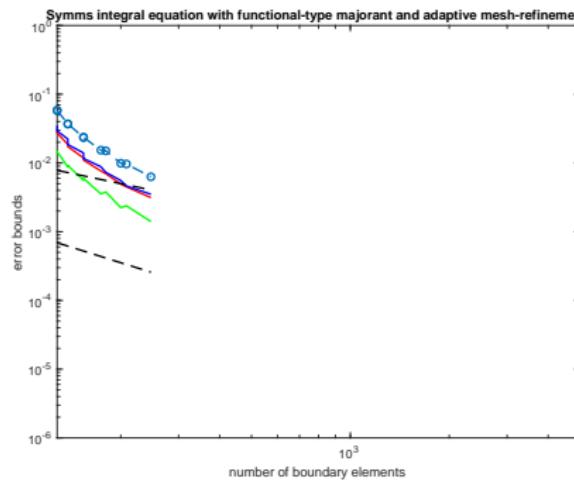
**Numerical
experiments****Outlook**

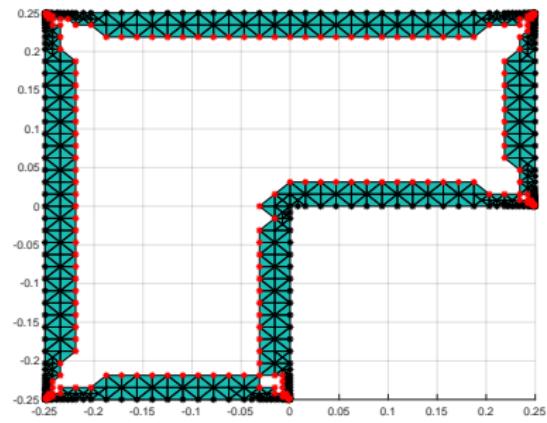
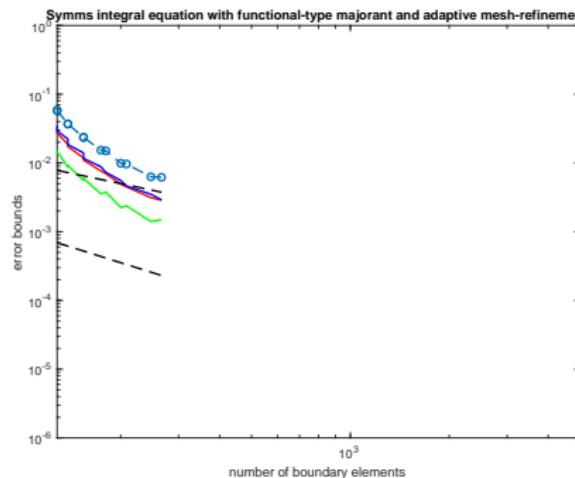
BEM example**Motivation****Error identity**

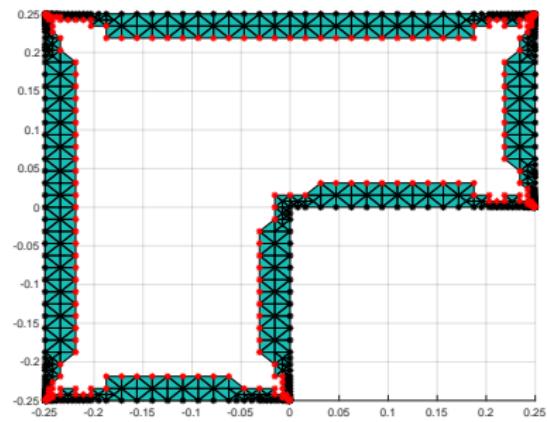
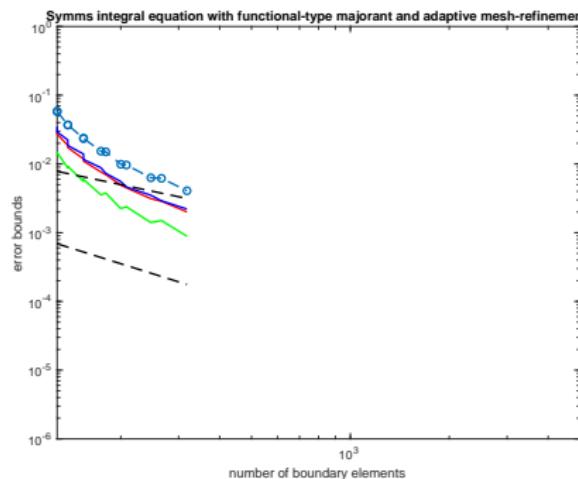
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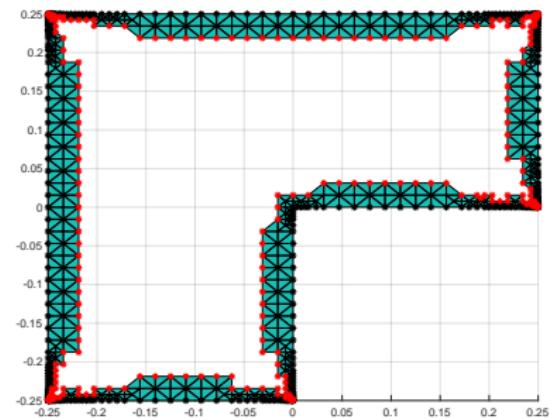
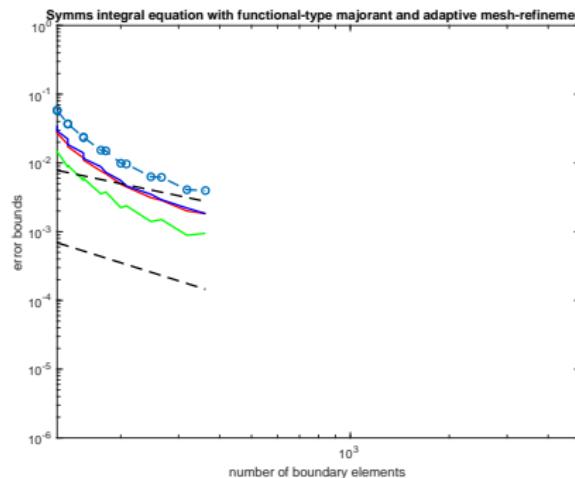
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**



BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

BEM example

Motivation

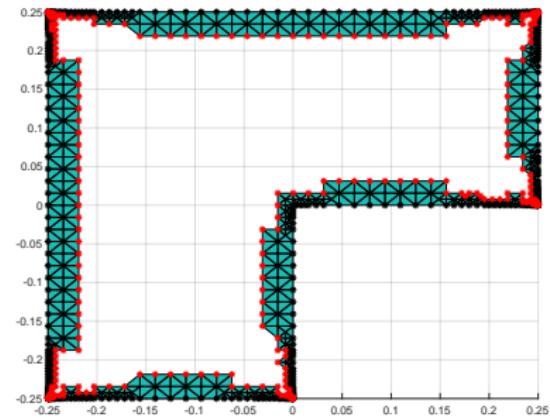
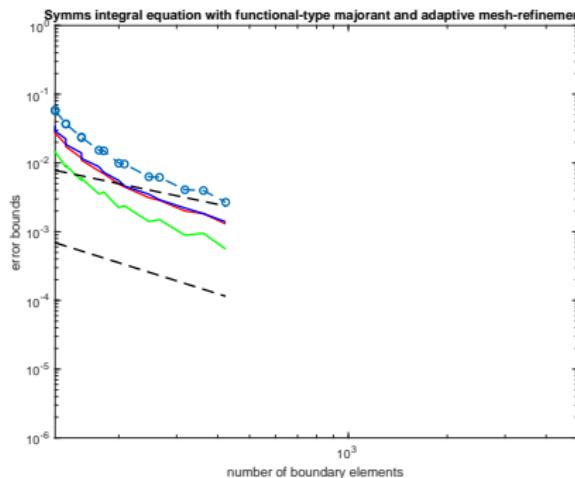
Error identity

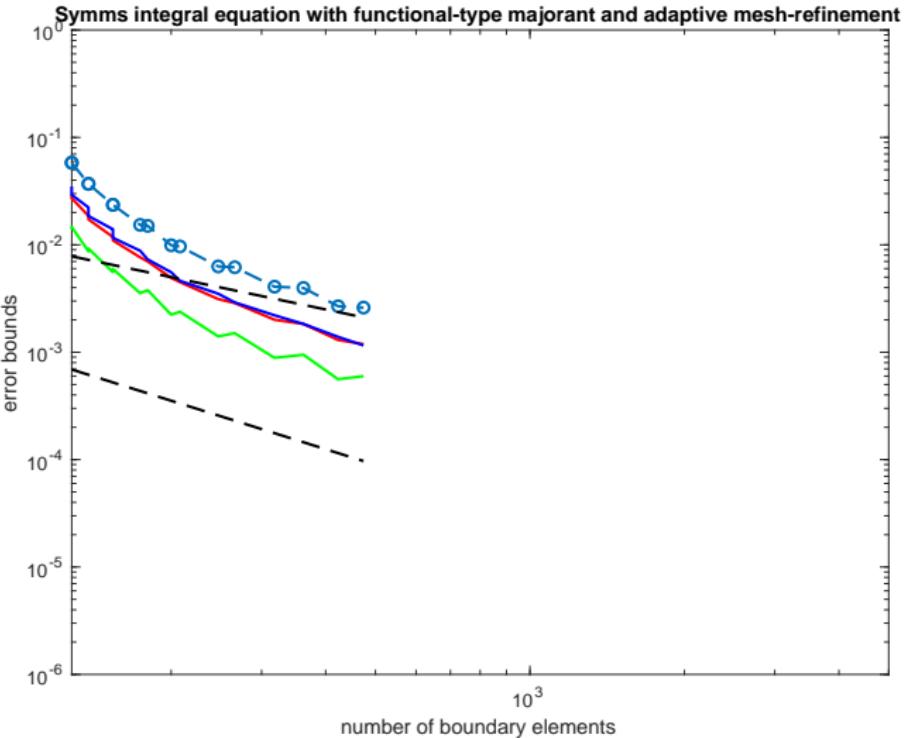
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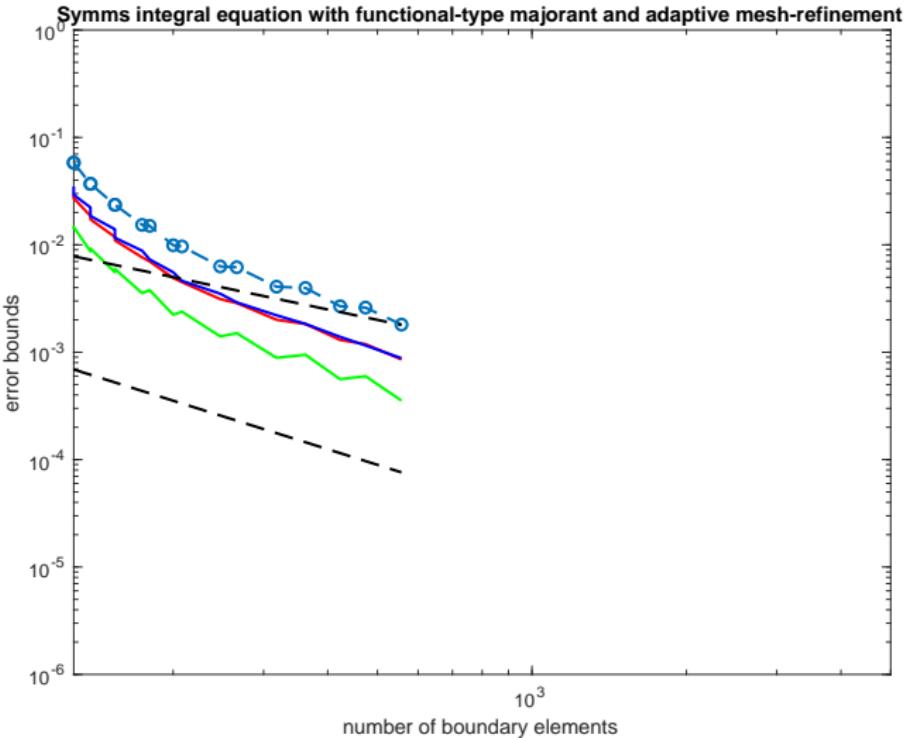
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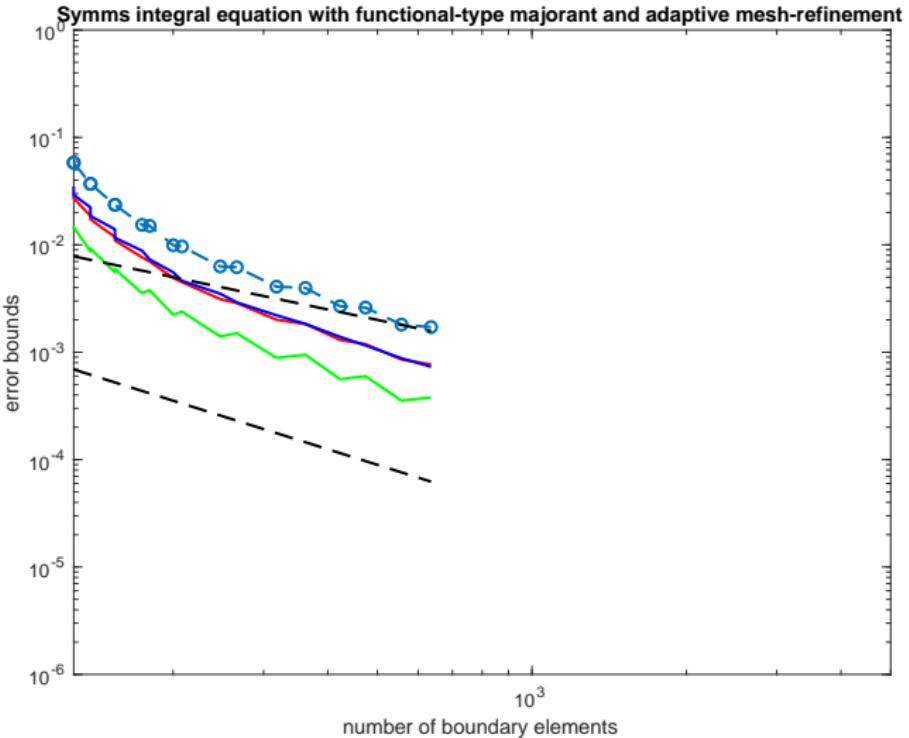
Numerical
experiments

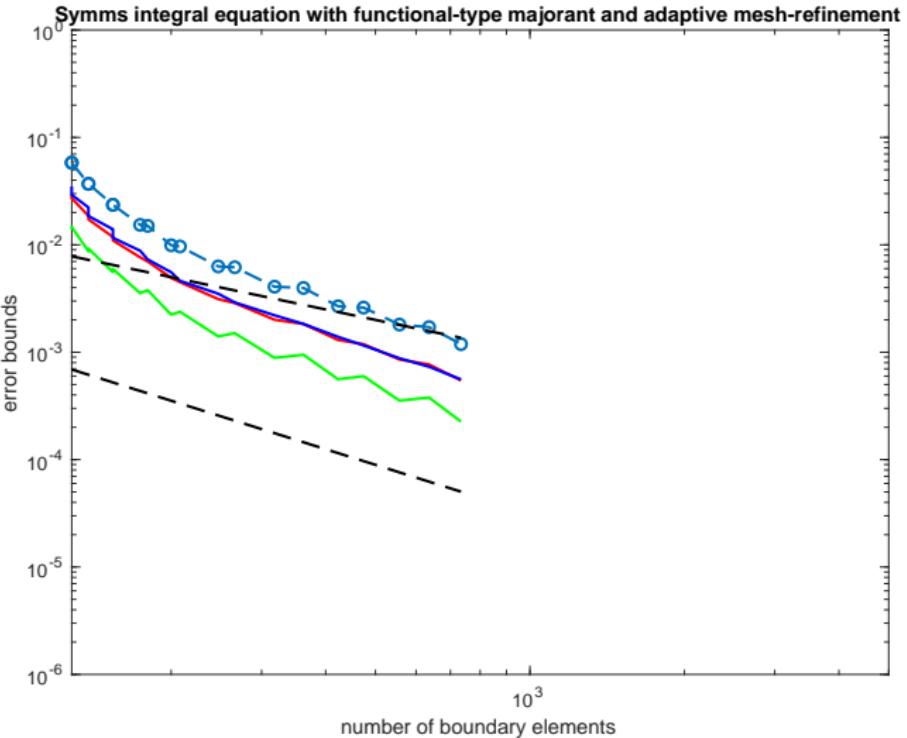
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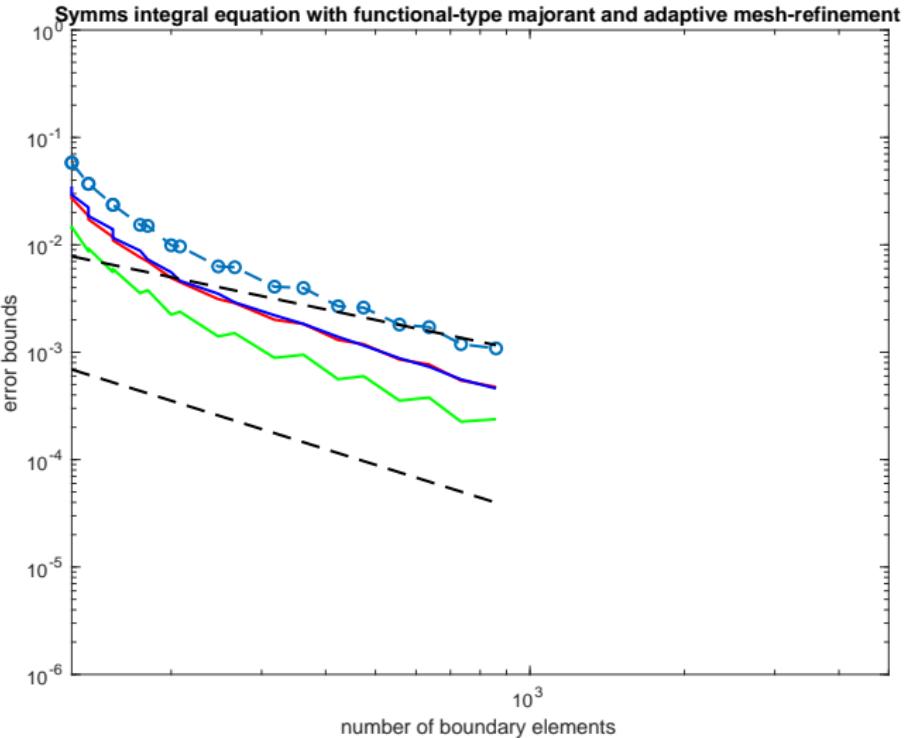


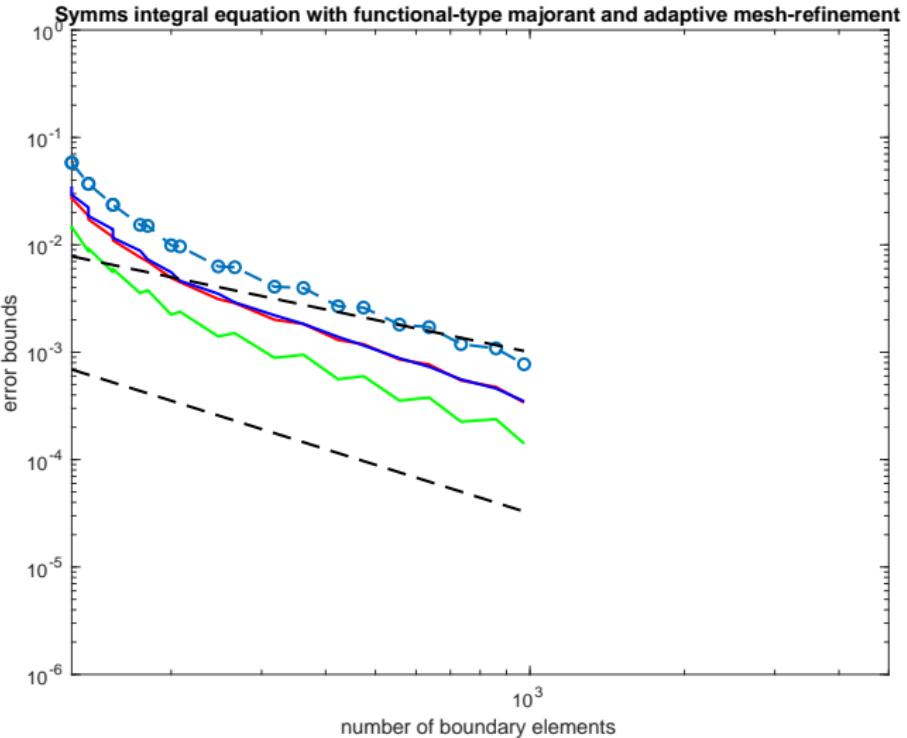


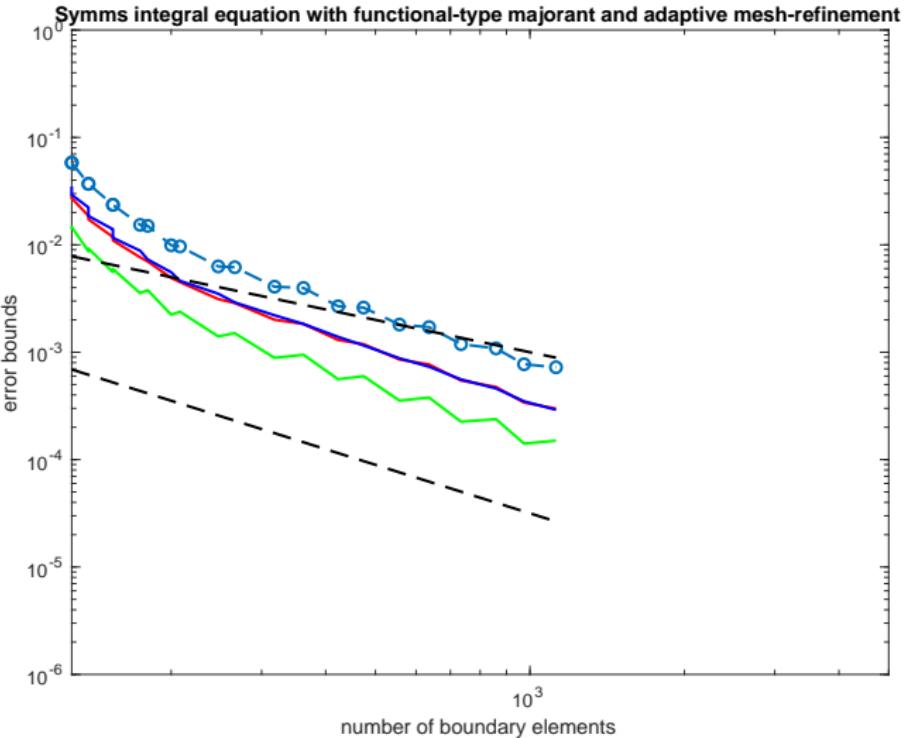


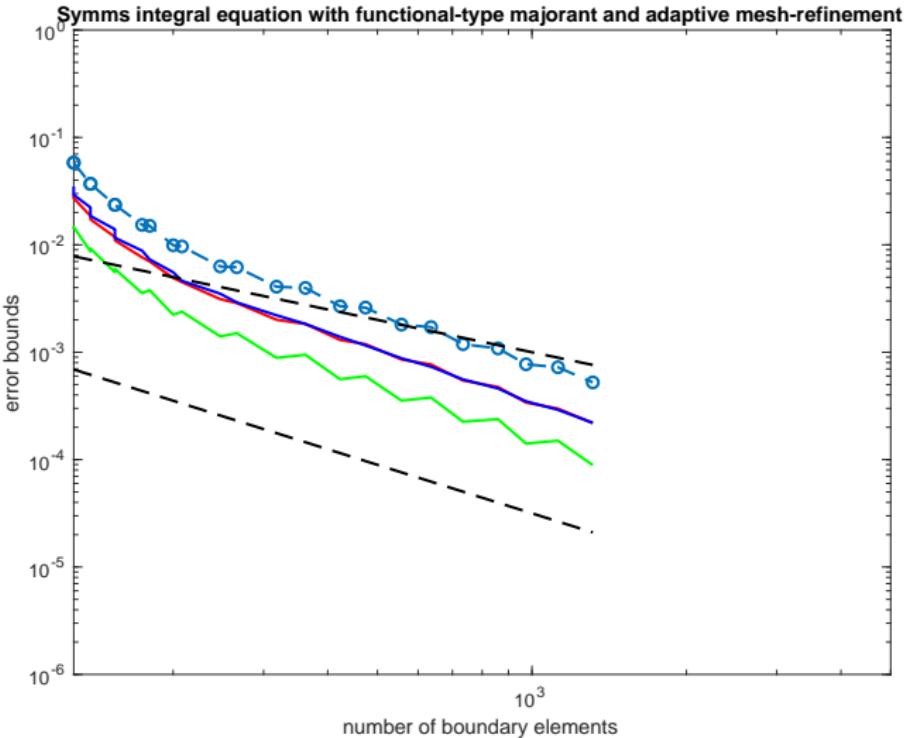


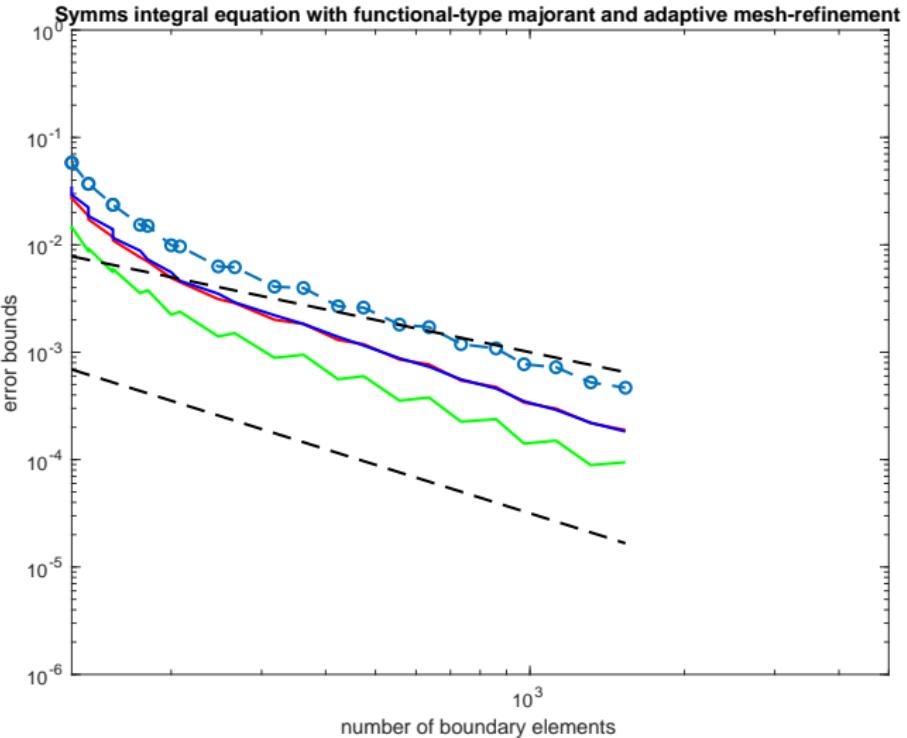


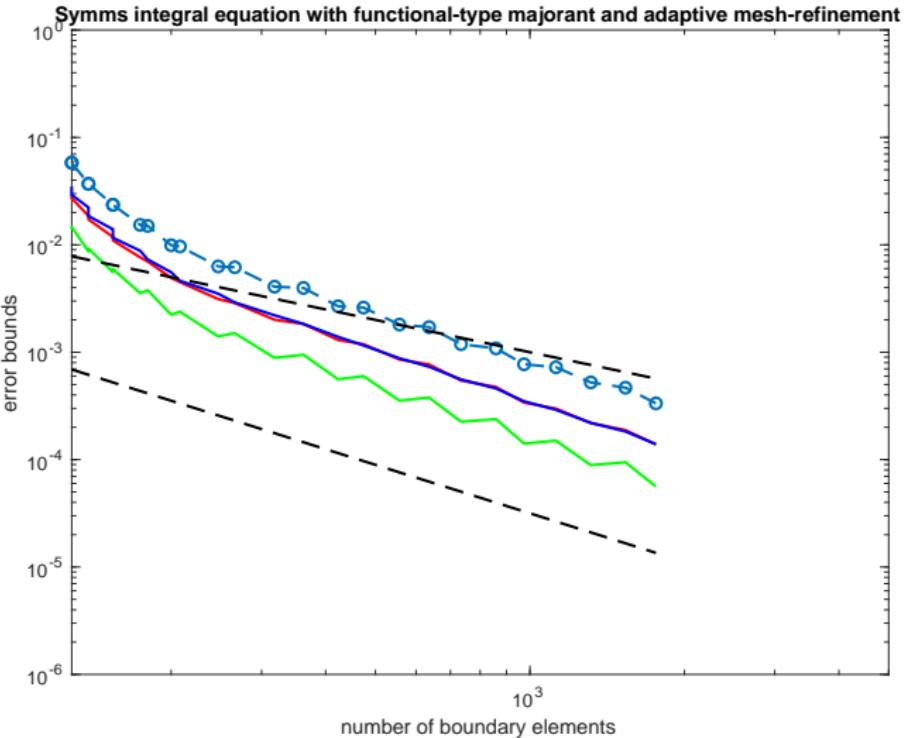


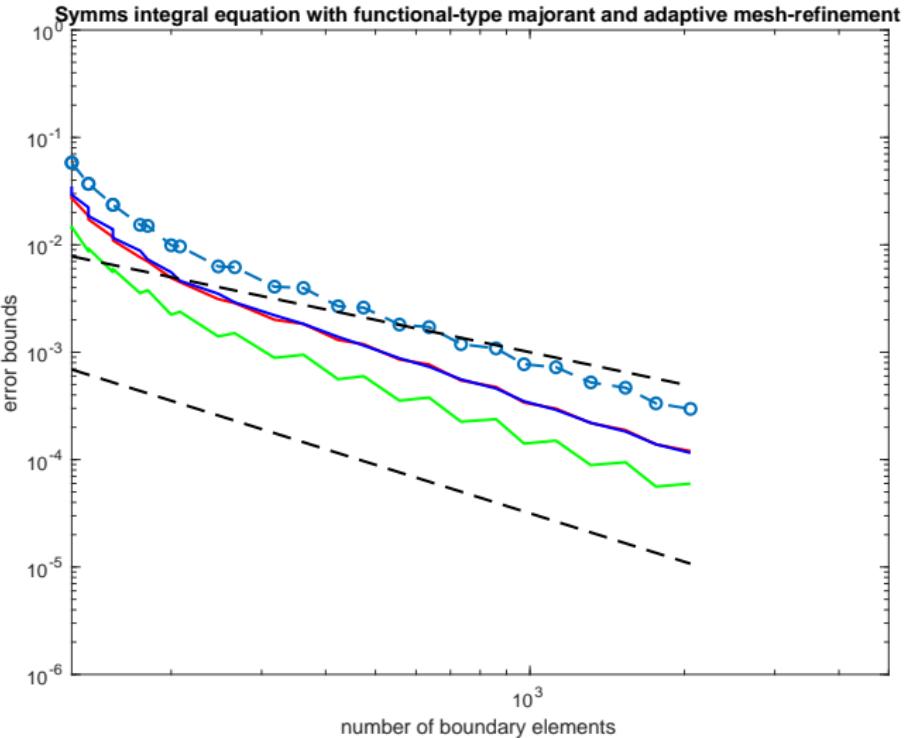


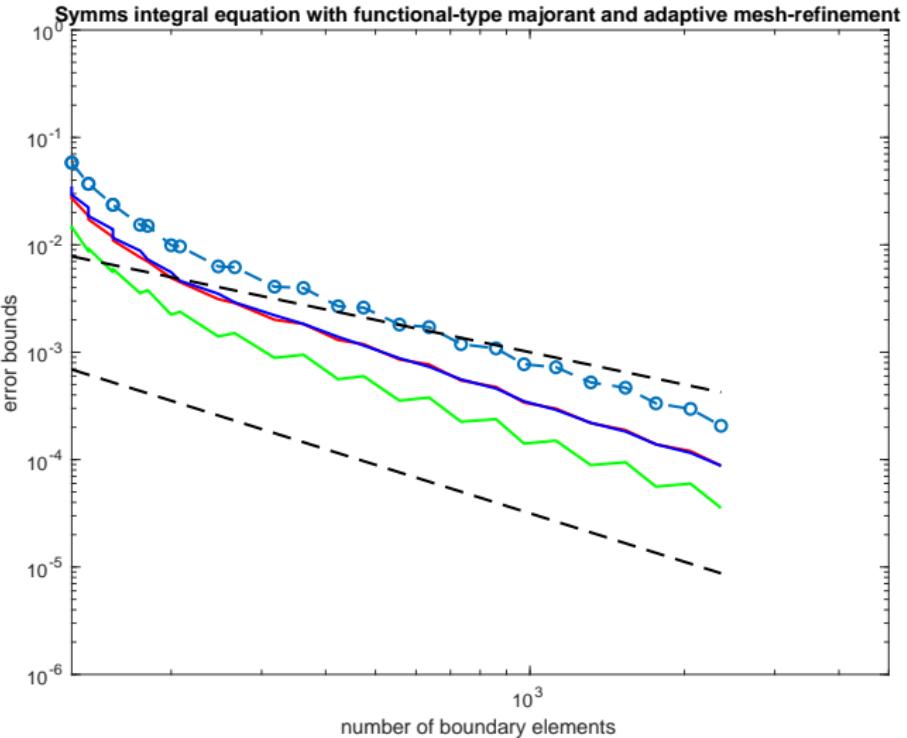


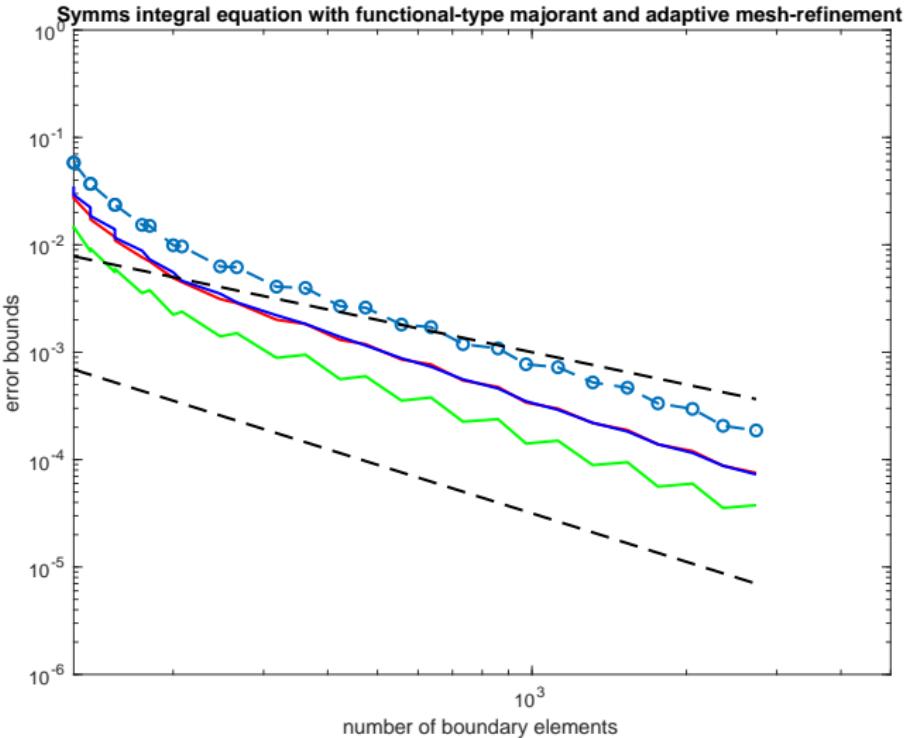


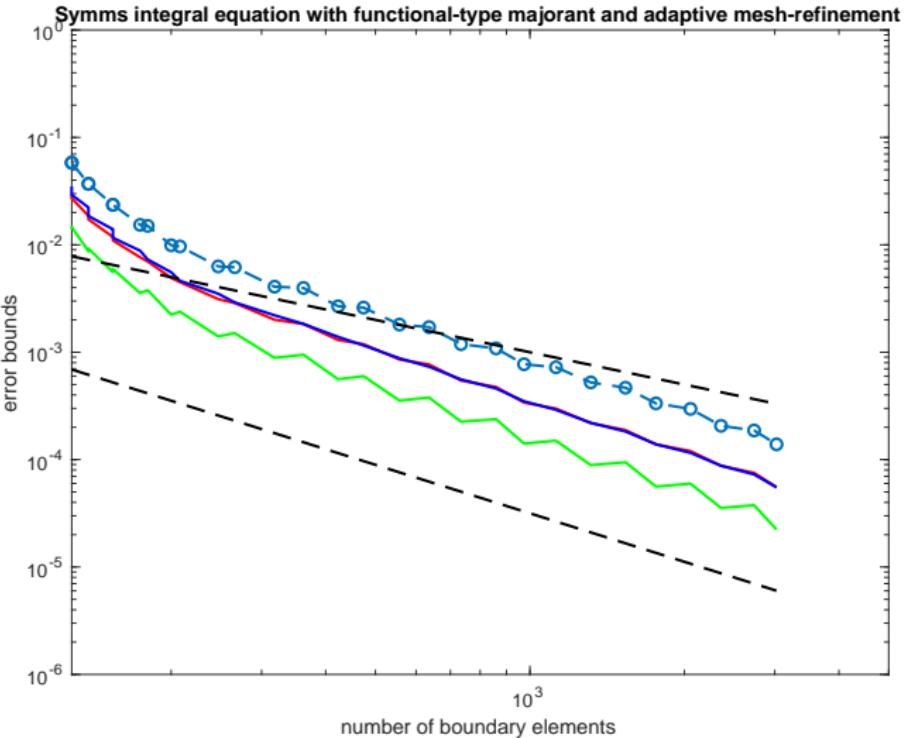


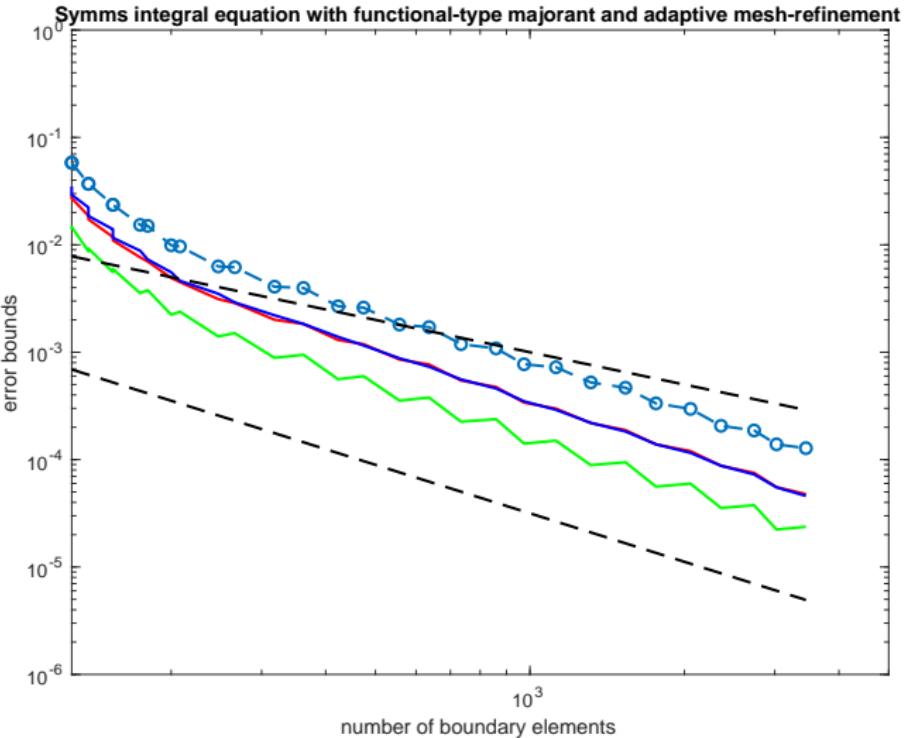


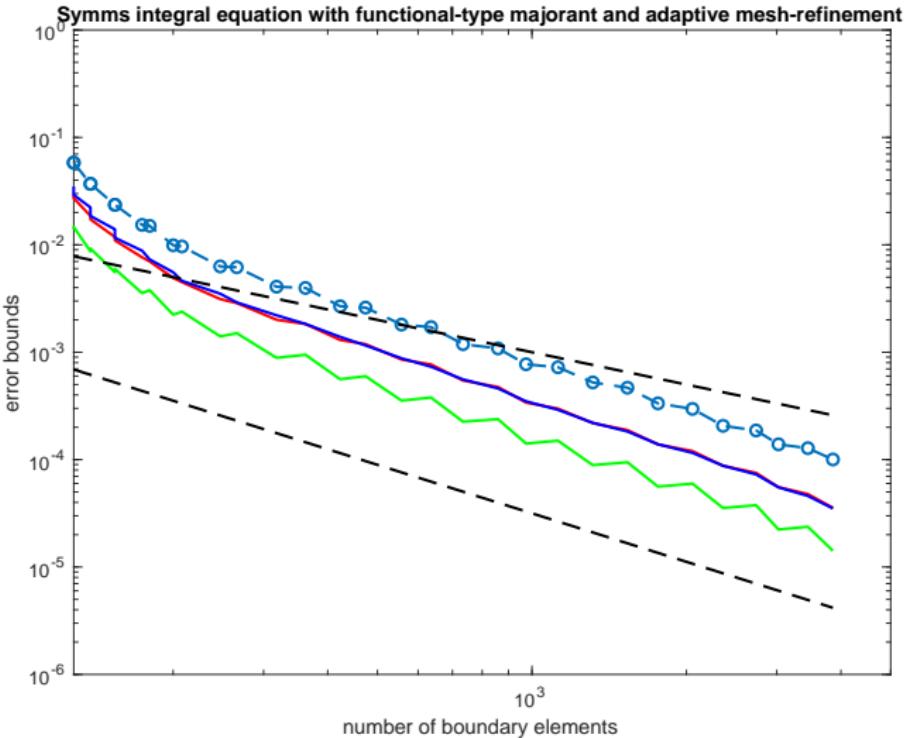


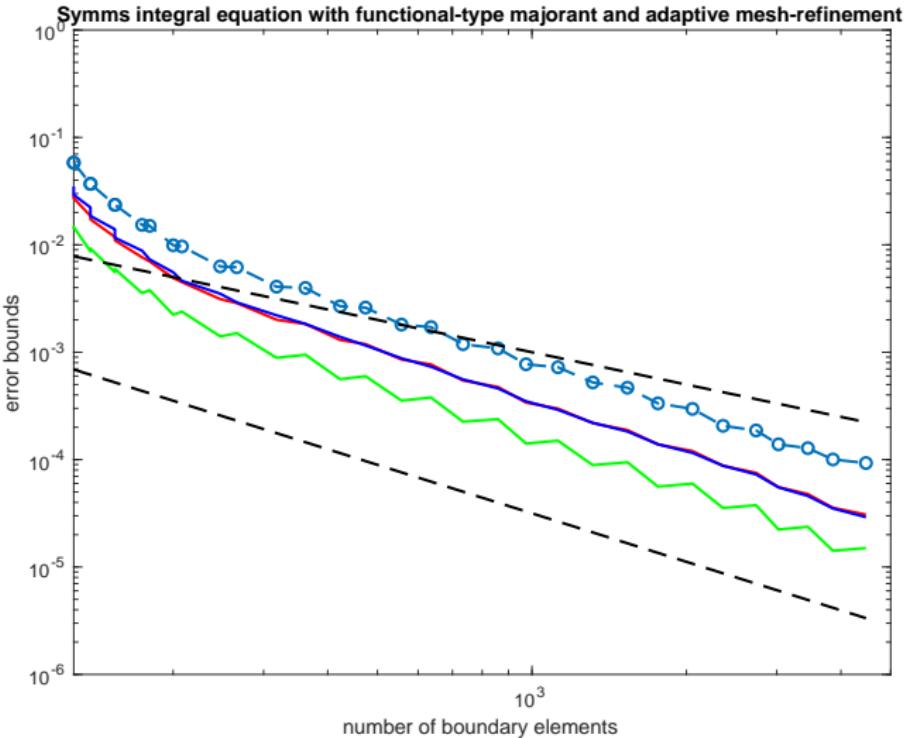


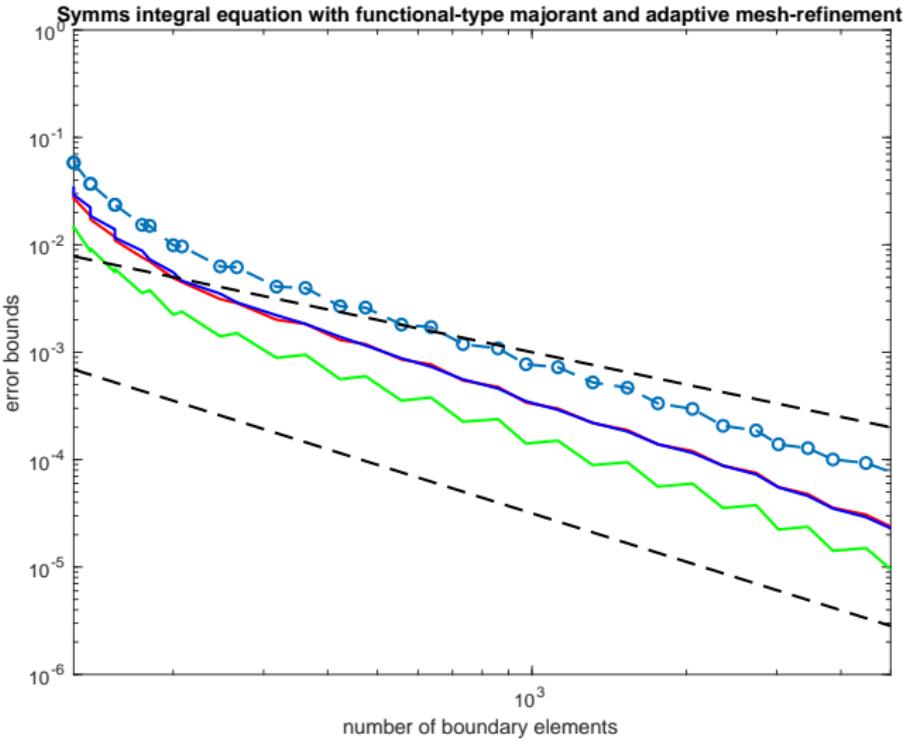












BEM example

Motivation

Error identity

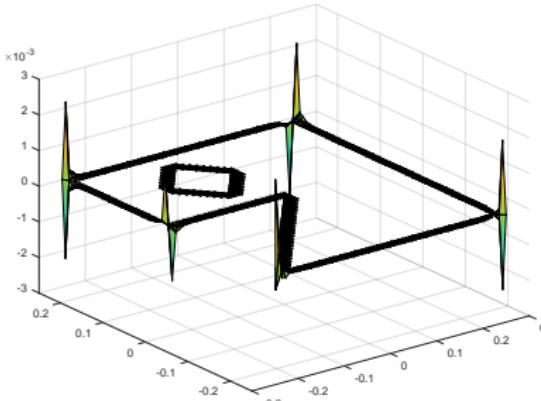
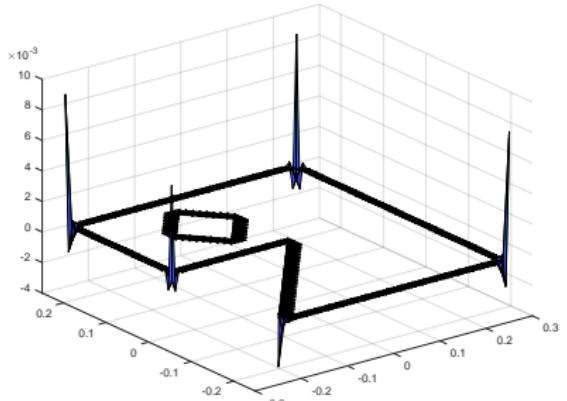
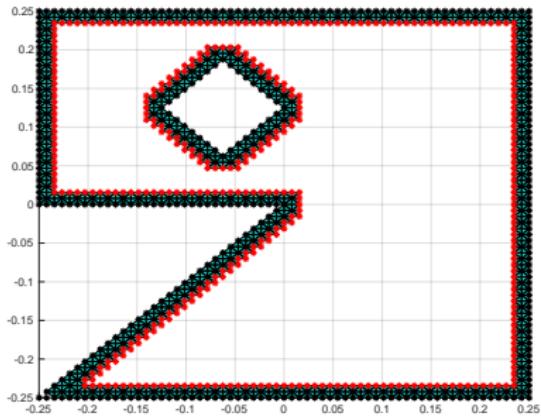
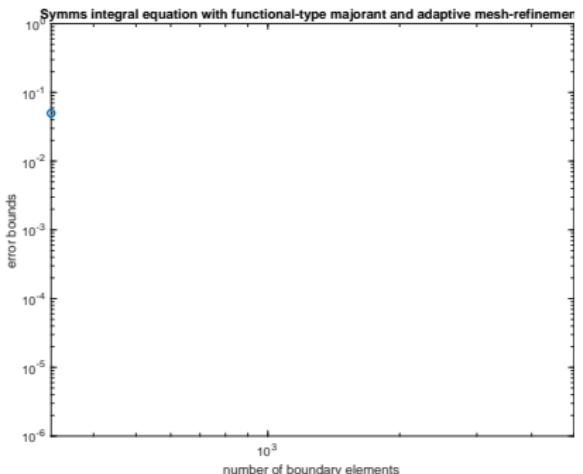
Majorant

Minorant

Numerical
experiments

Outlook

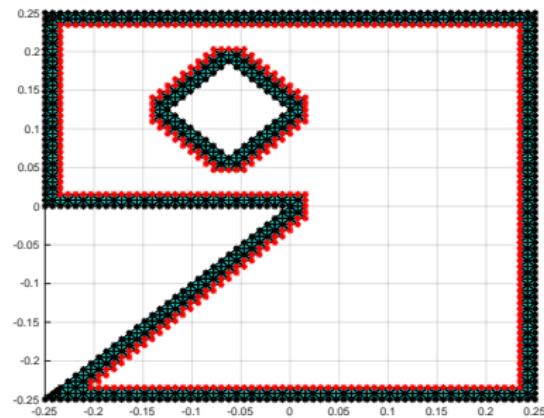
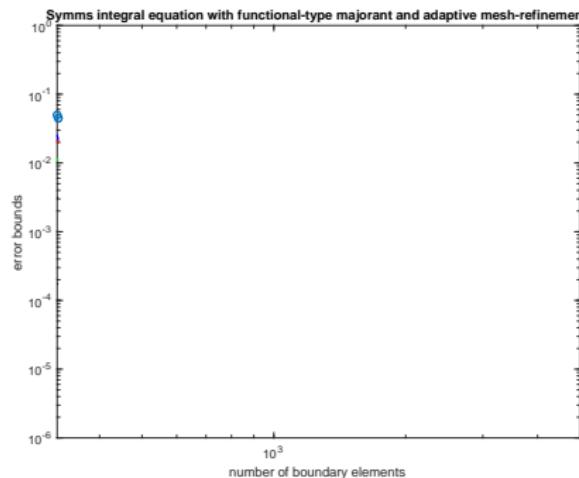
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BEM example**Motivation****Error identity**

Majorant

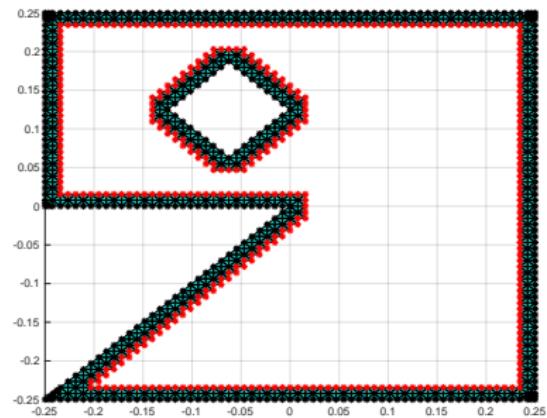
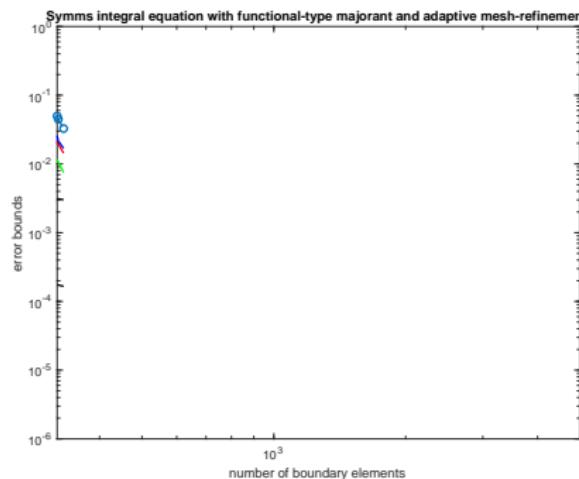
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

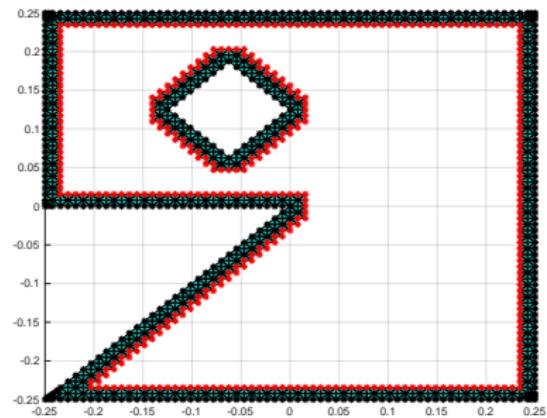
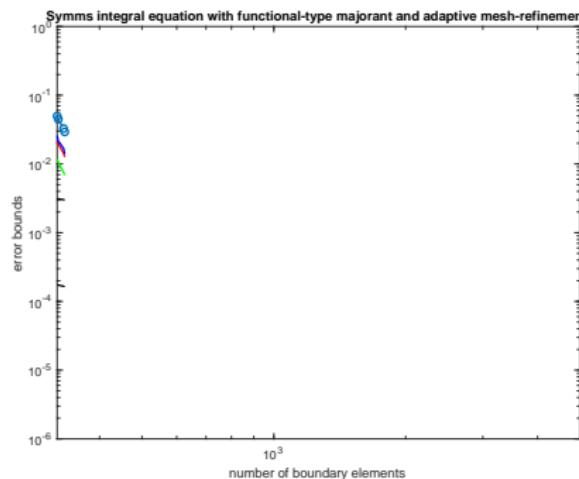
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

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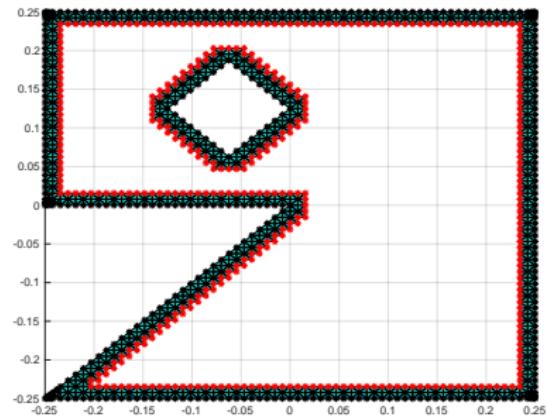
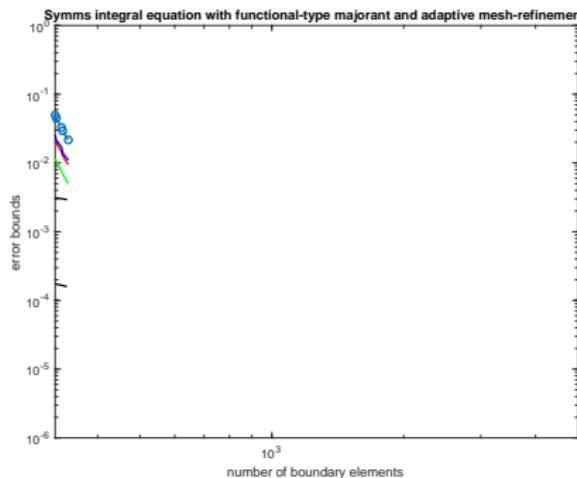
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

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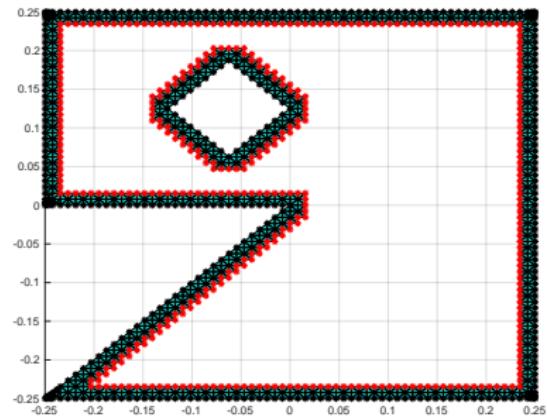
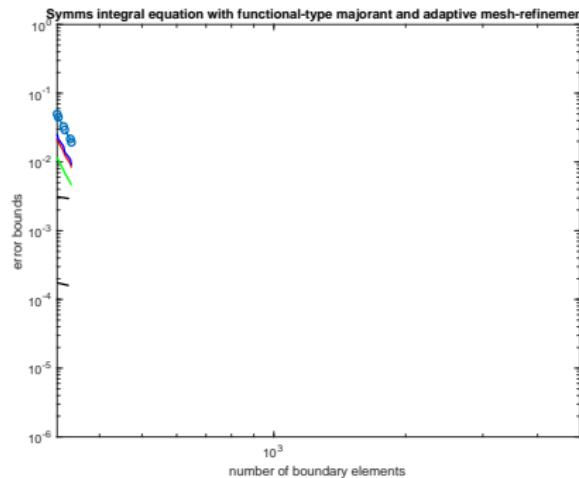
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

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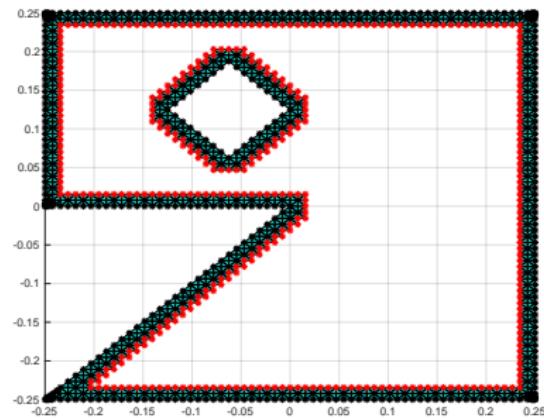
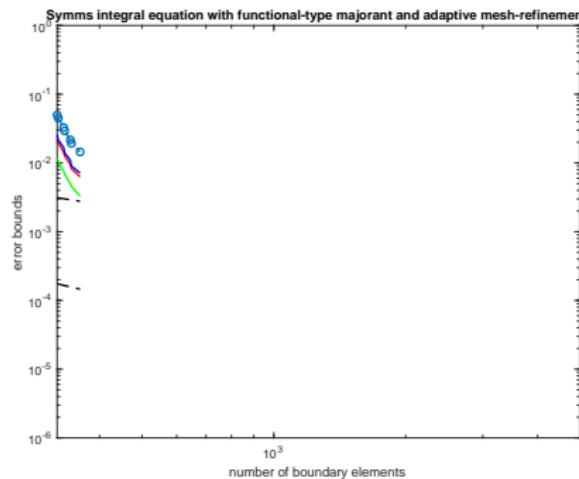
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

Majorant

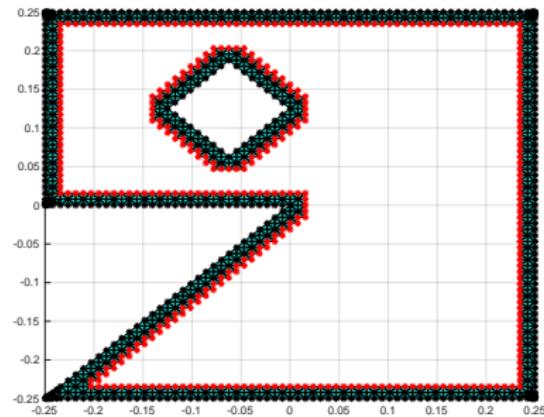
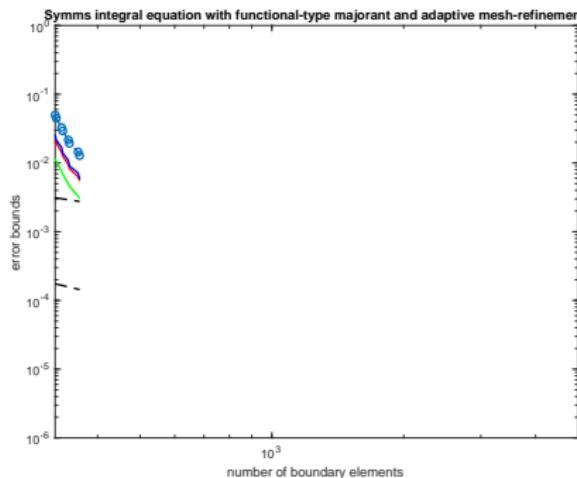
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**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity**

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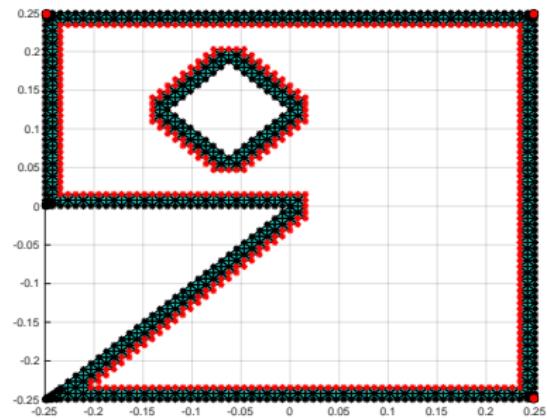
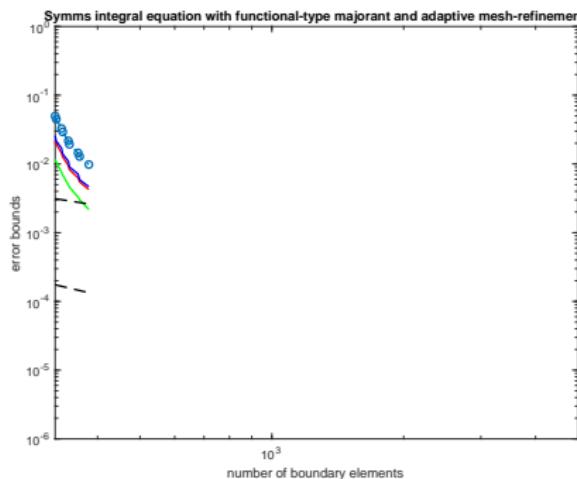
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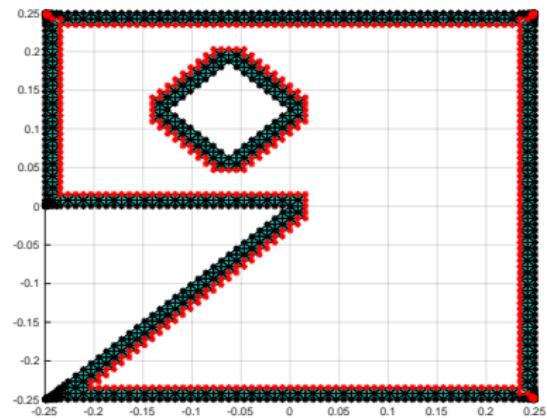
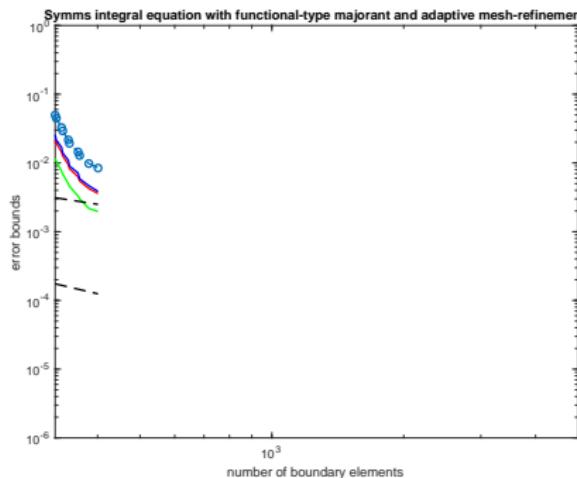
**Numerical
experiments****Outlook**

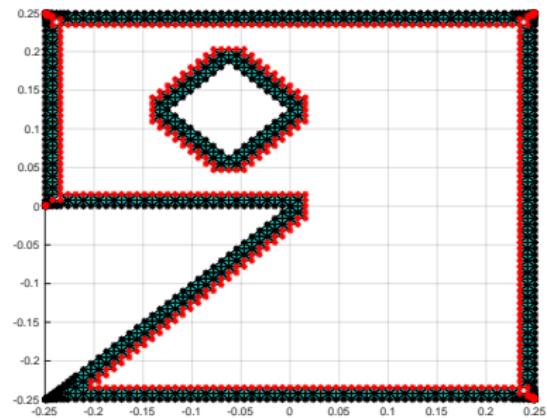
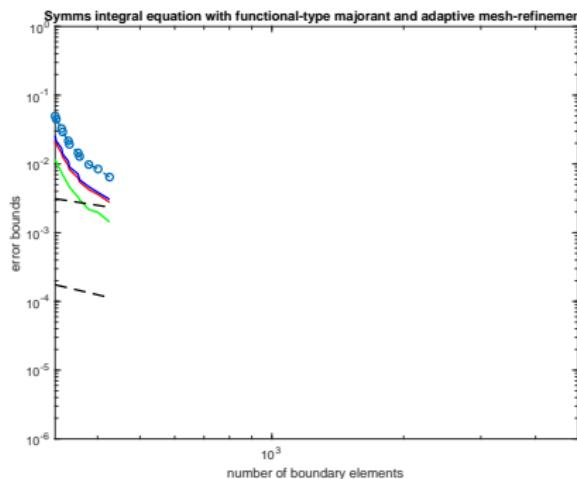
BEM example**Motivation****Error identity**

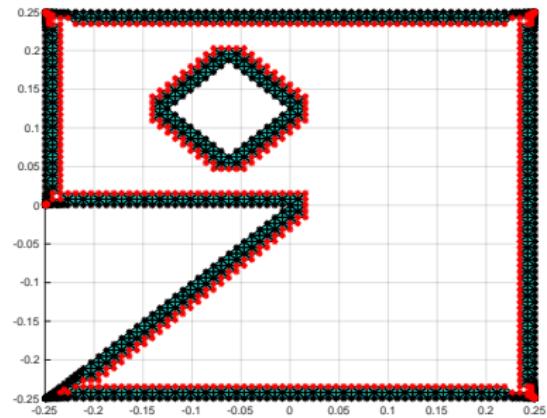
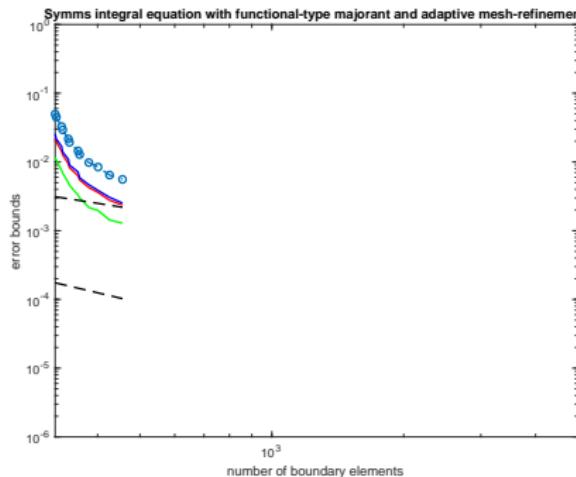
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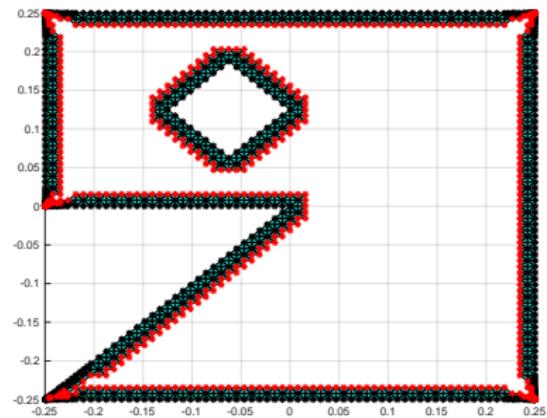
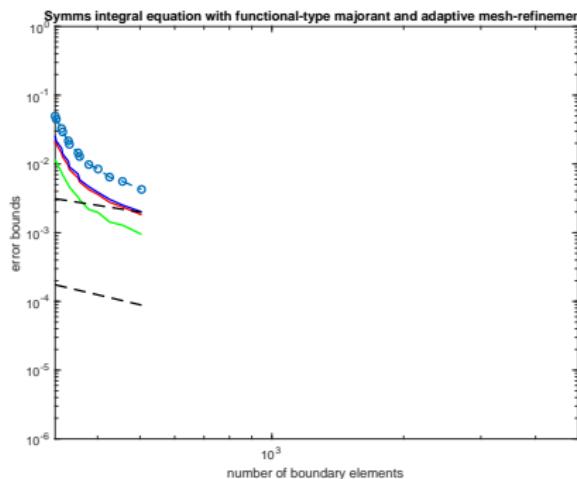
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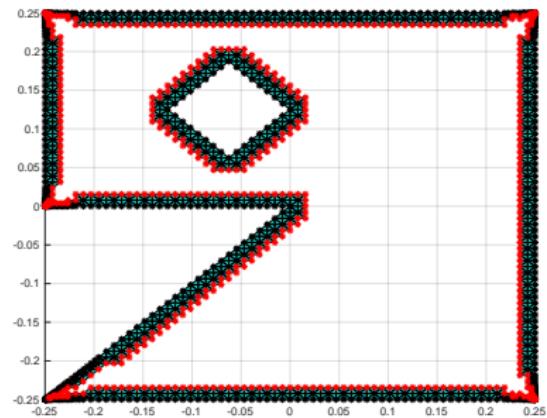
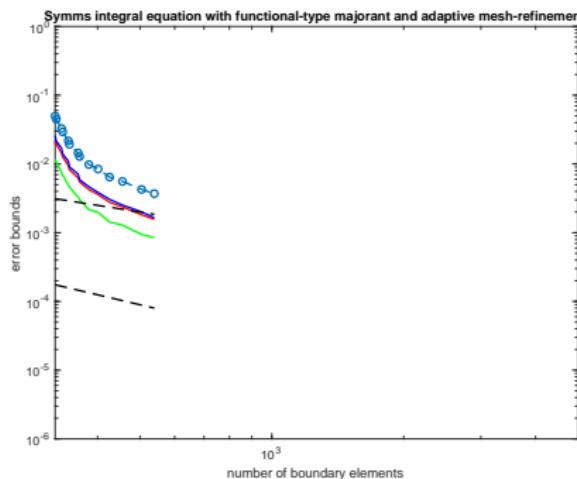
**Numerical
experiments****Outlook**

BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

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BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

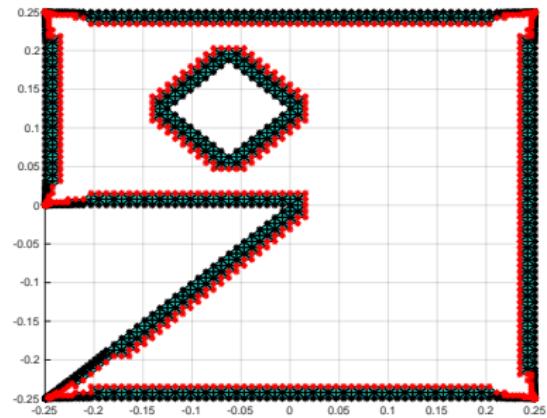
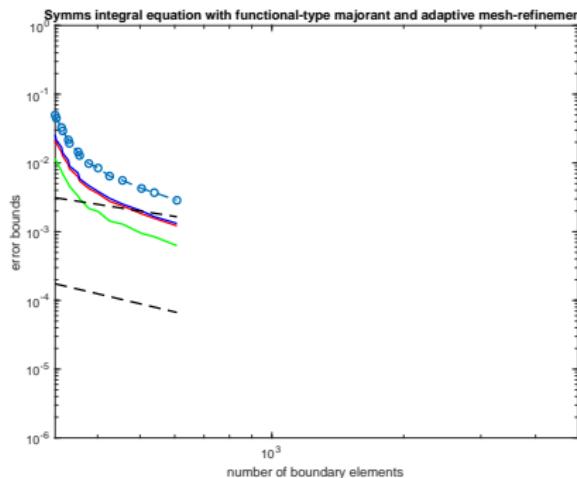
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experiments****Outlook**

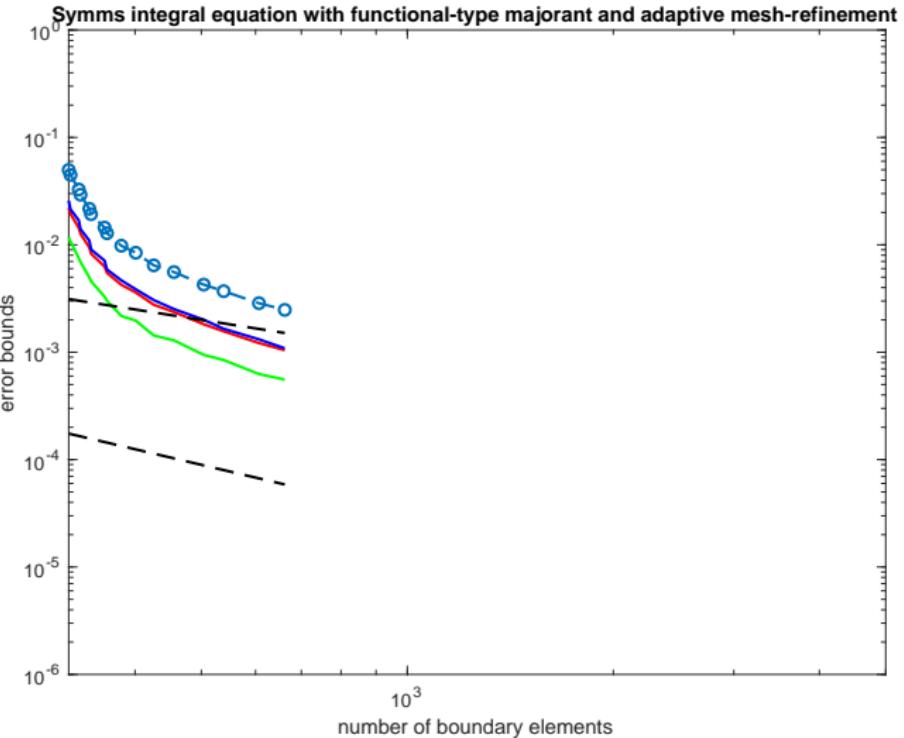
BEM example**Motivation****Error identity****Majorant****Minorant****Numerical
experiments****Outlook**

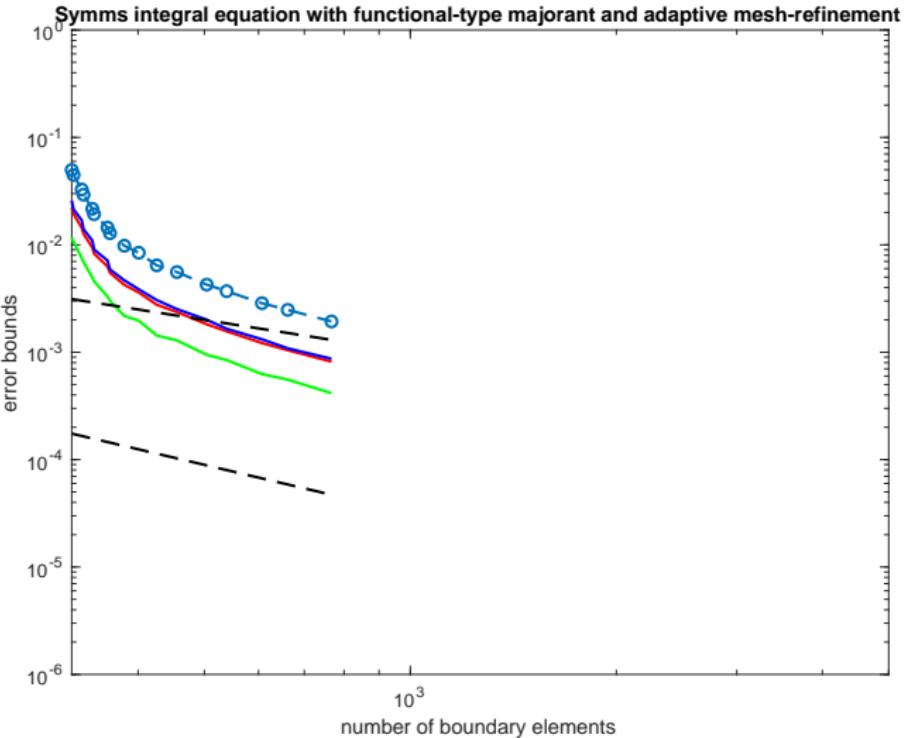
BEM example**Motivation****Error identity**

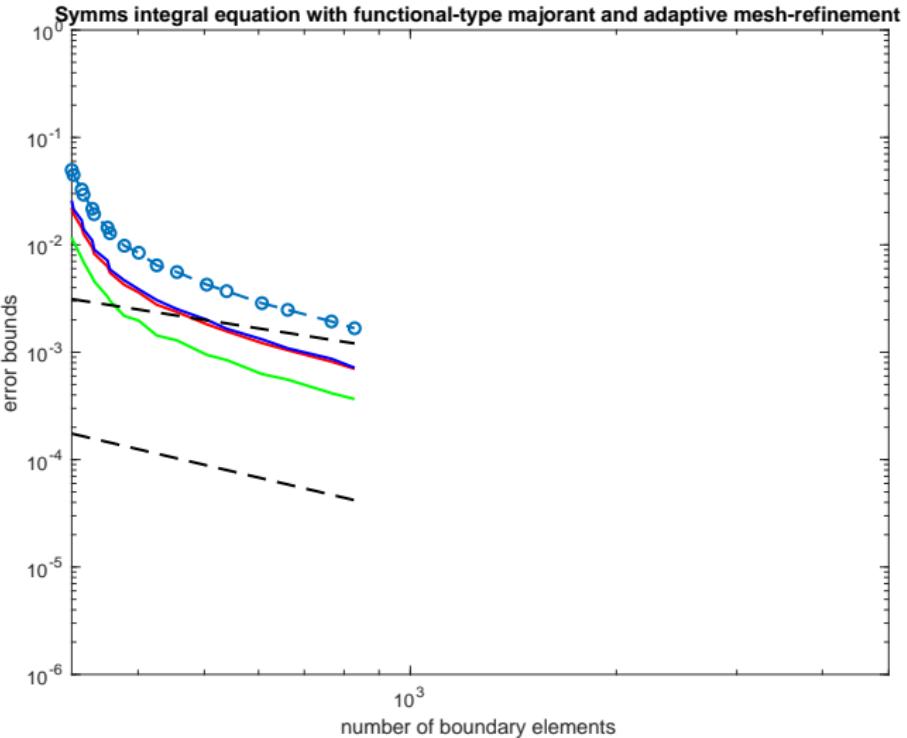
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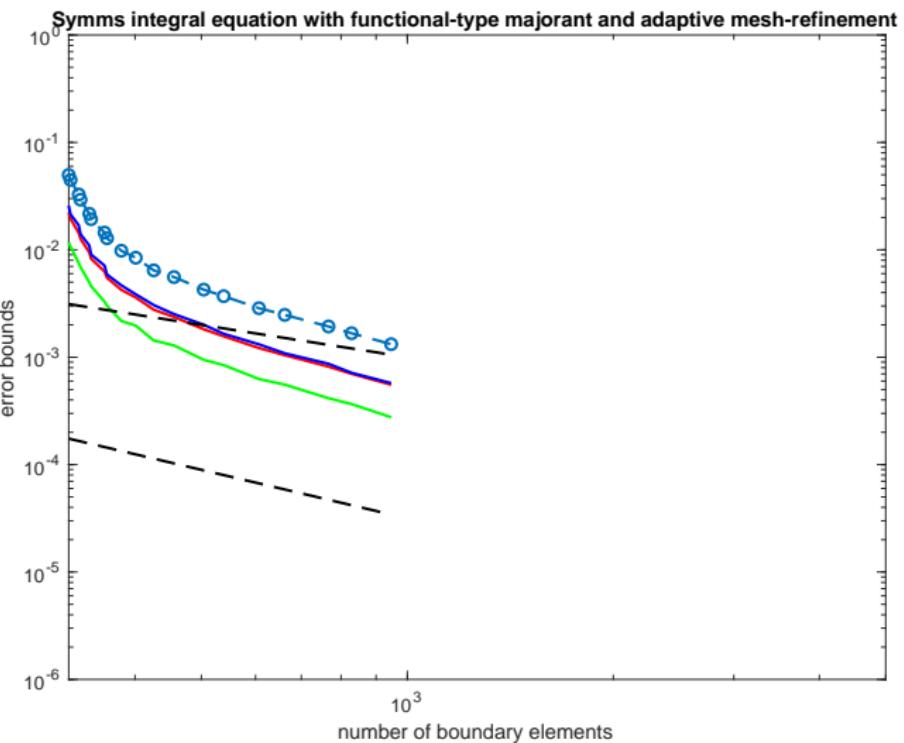
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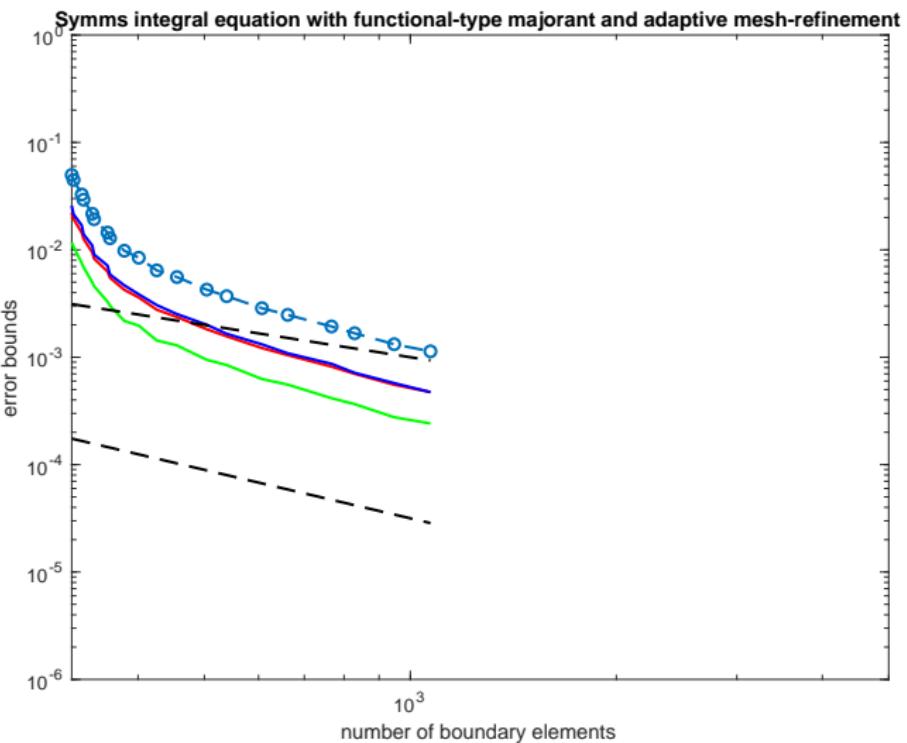
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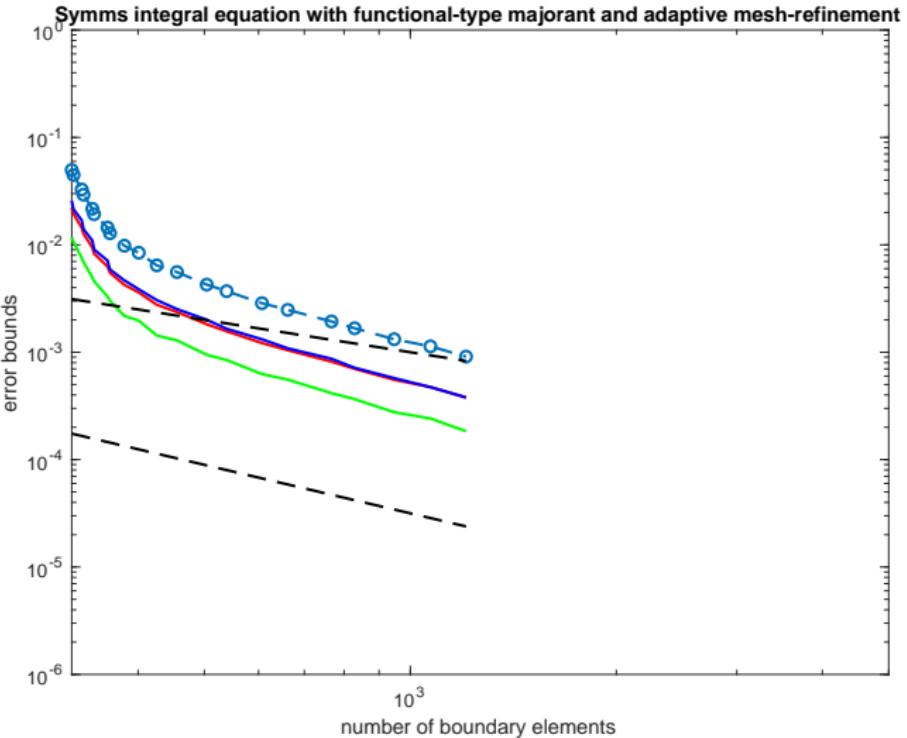


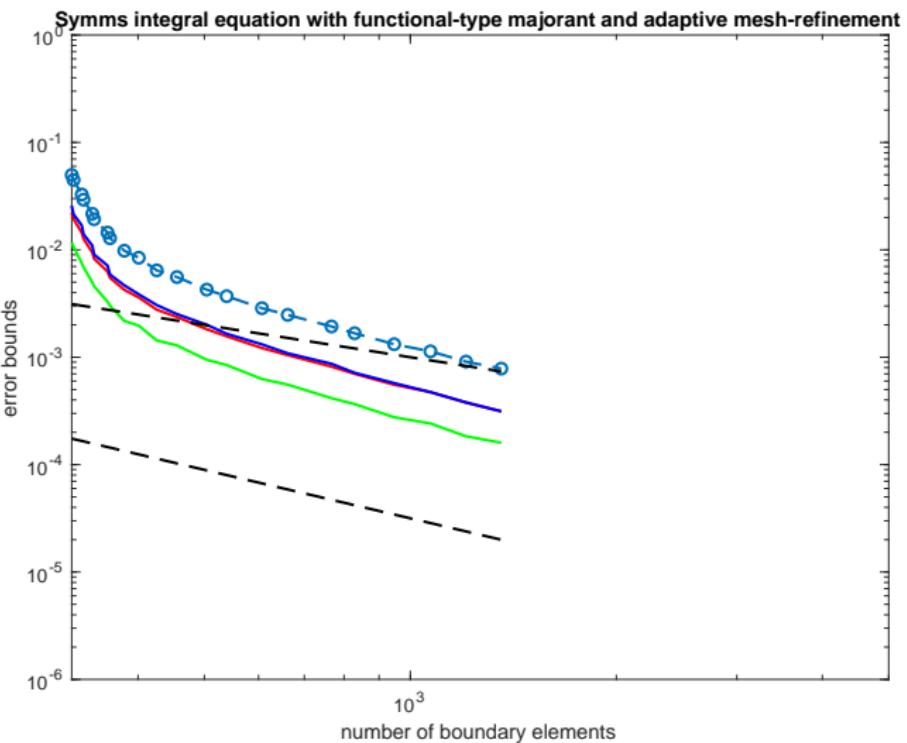


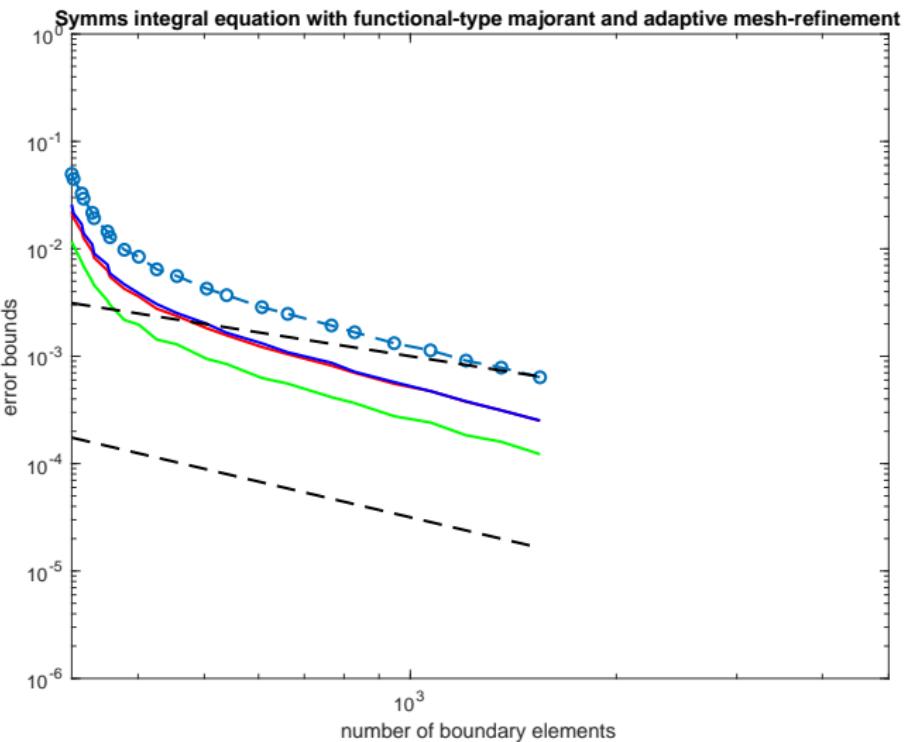


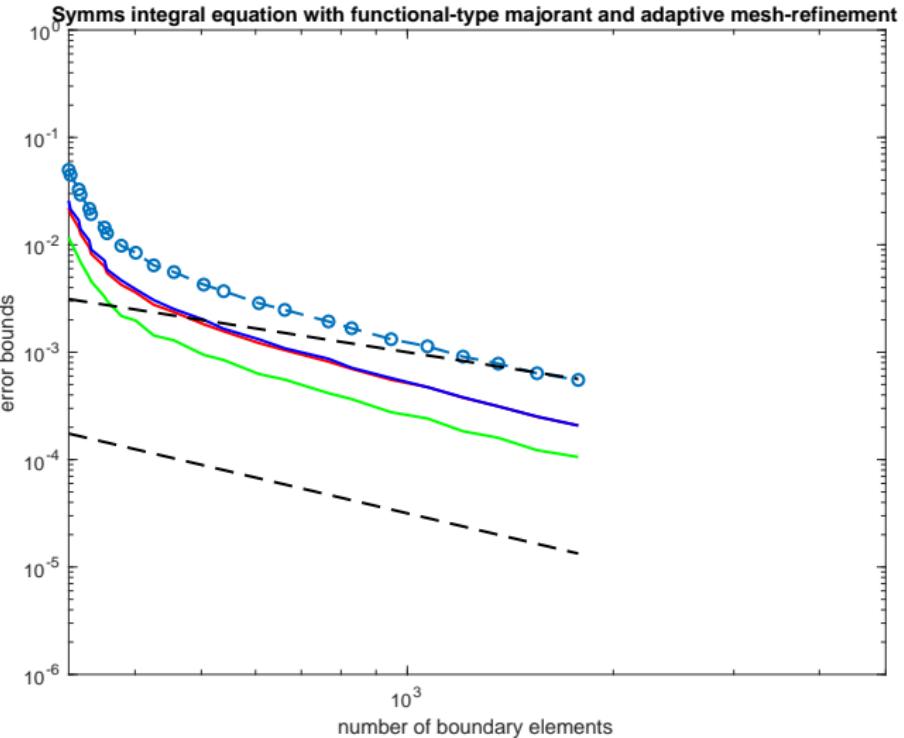


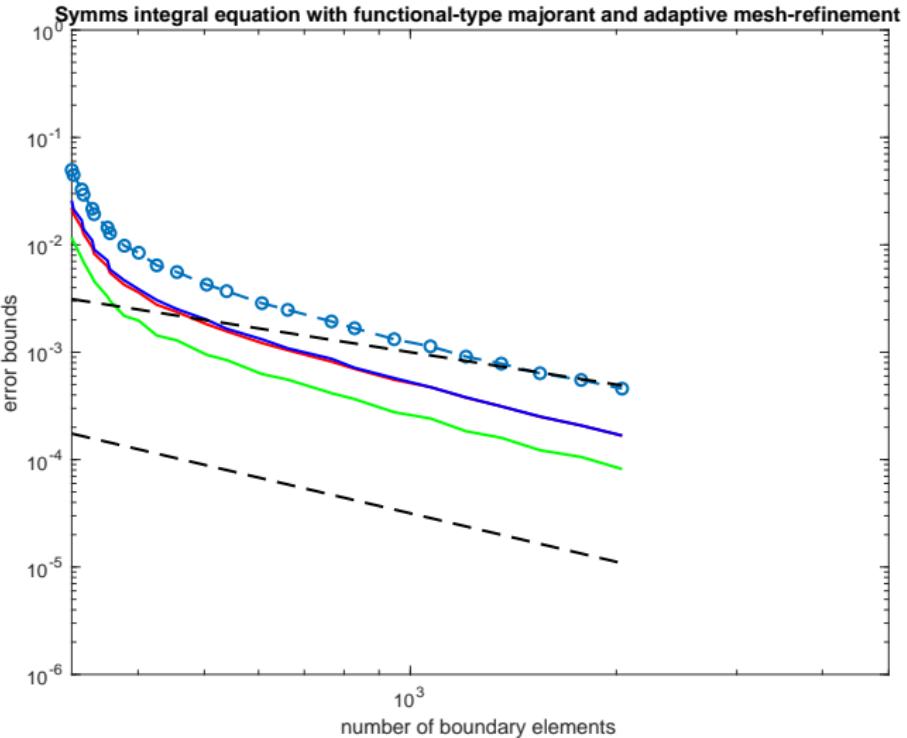


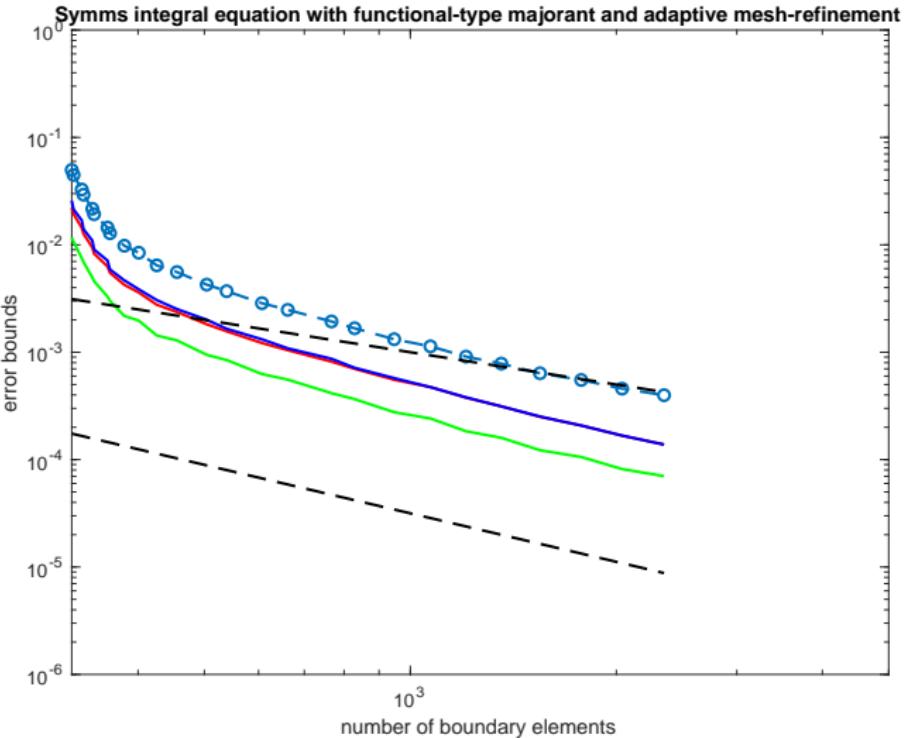


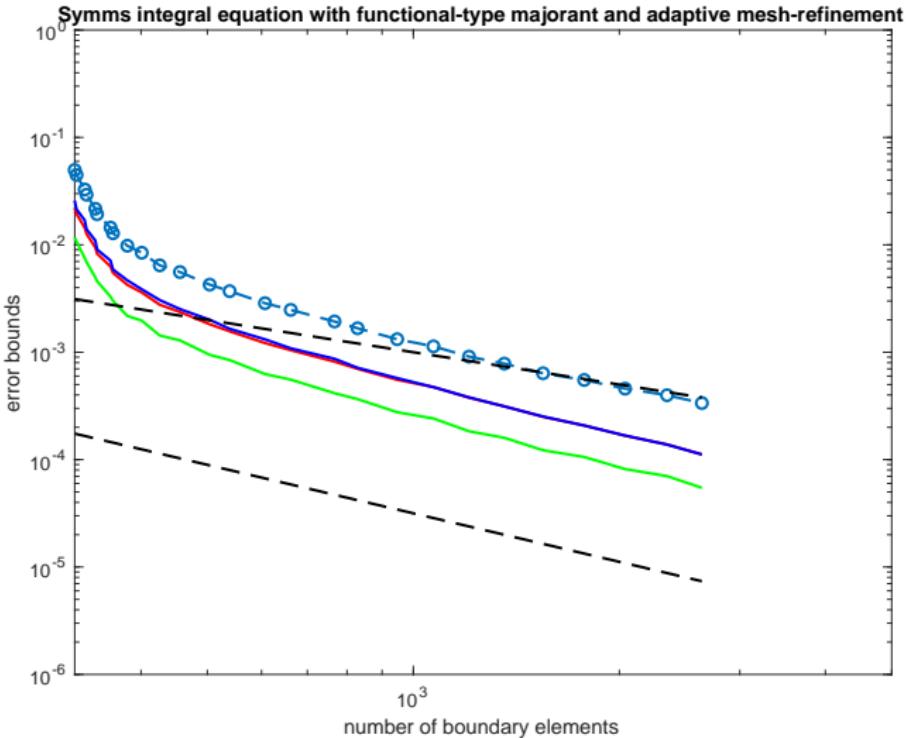


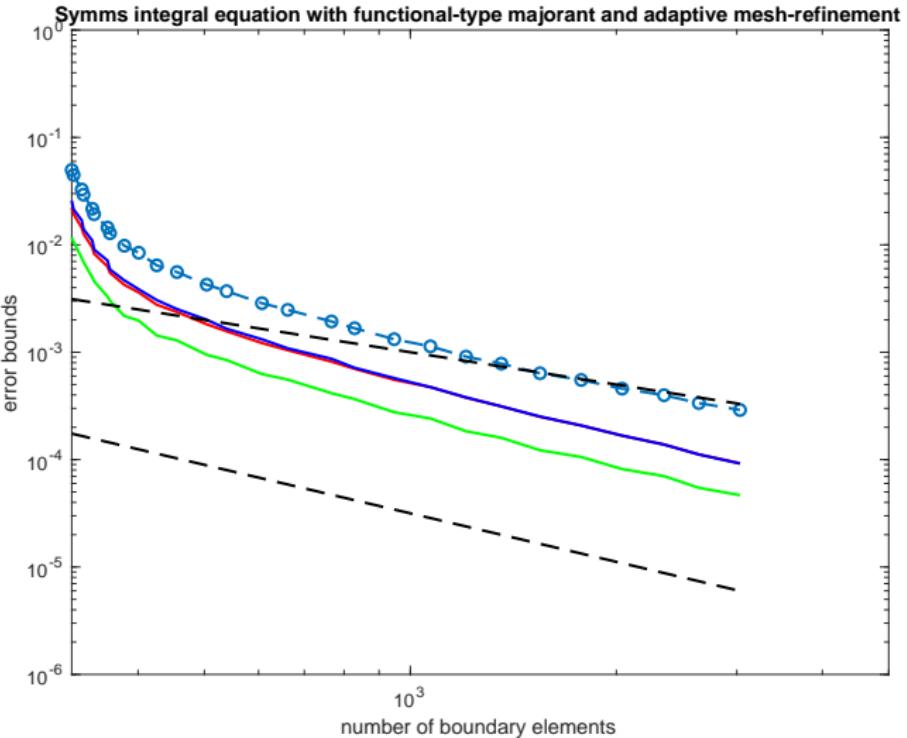


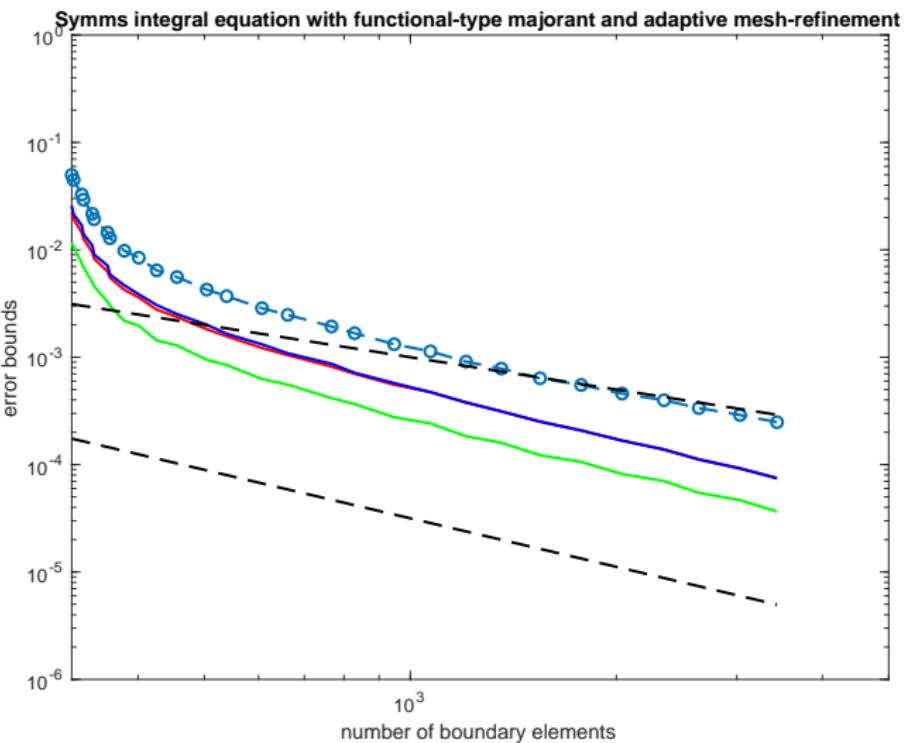


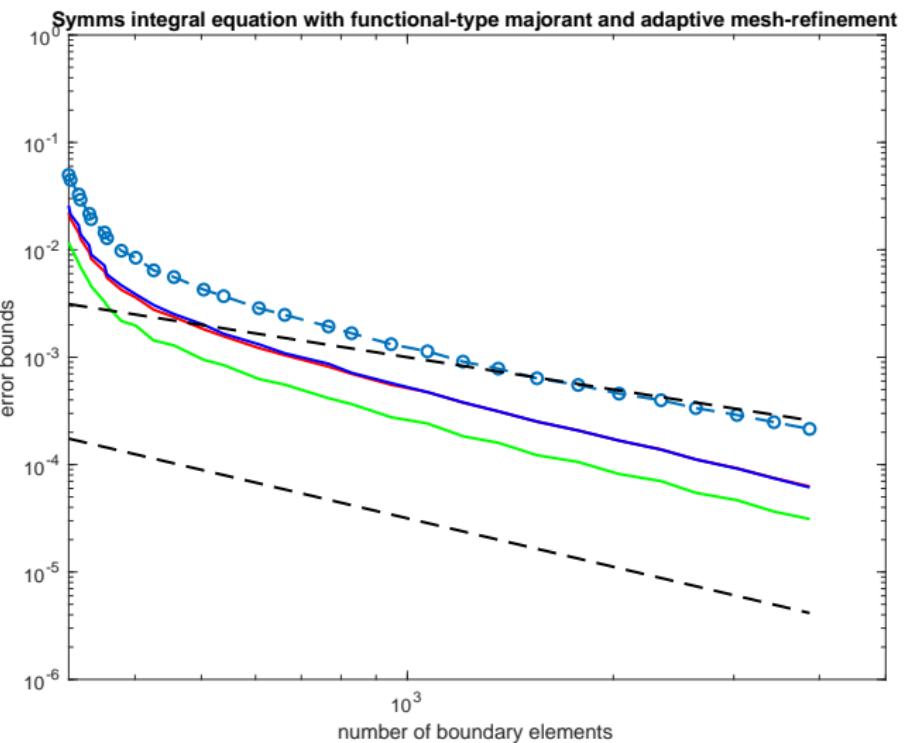


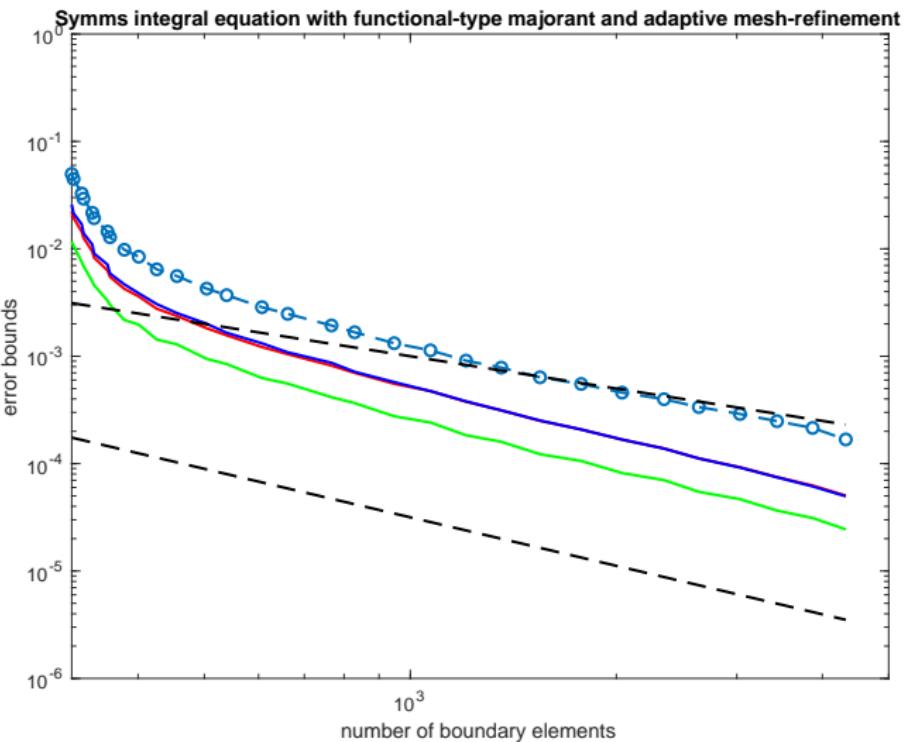


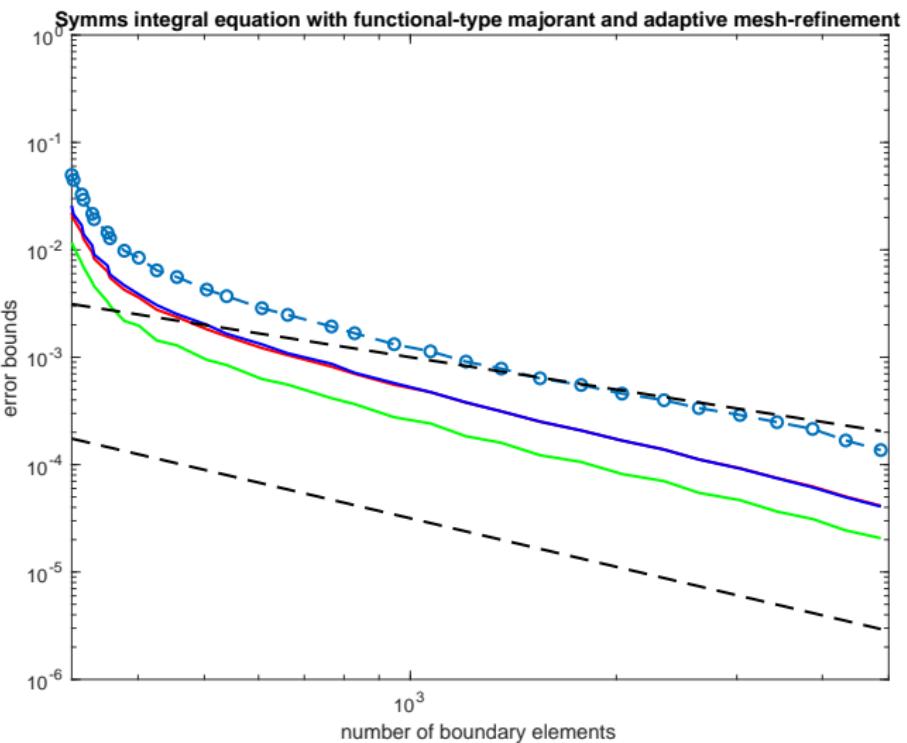


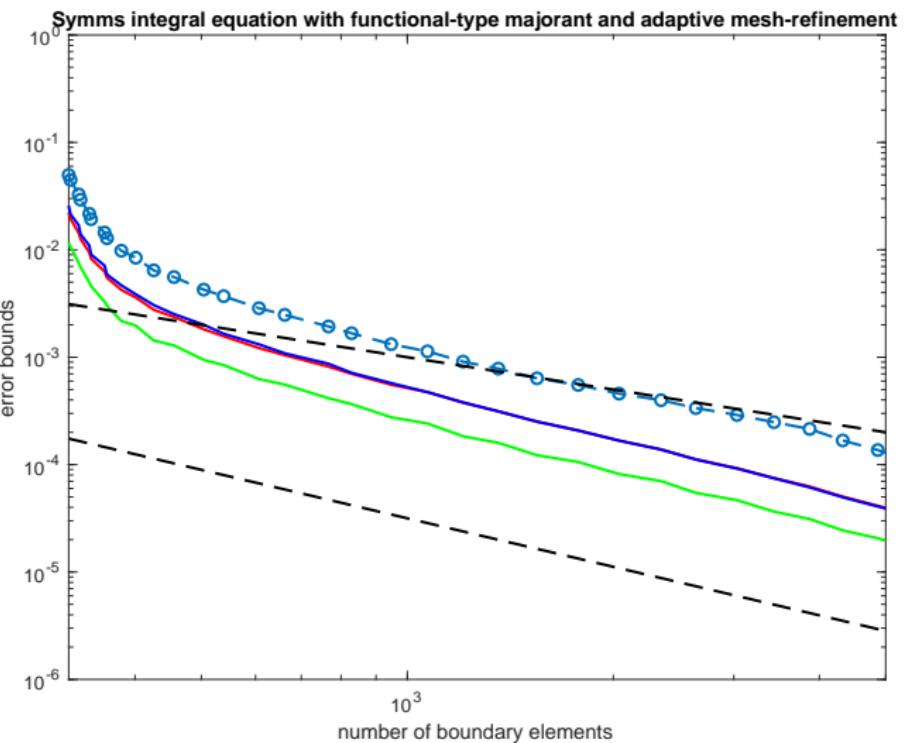












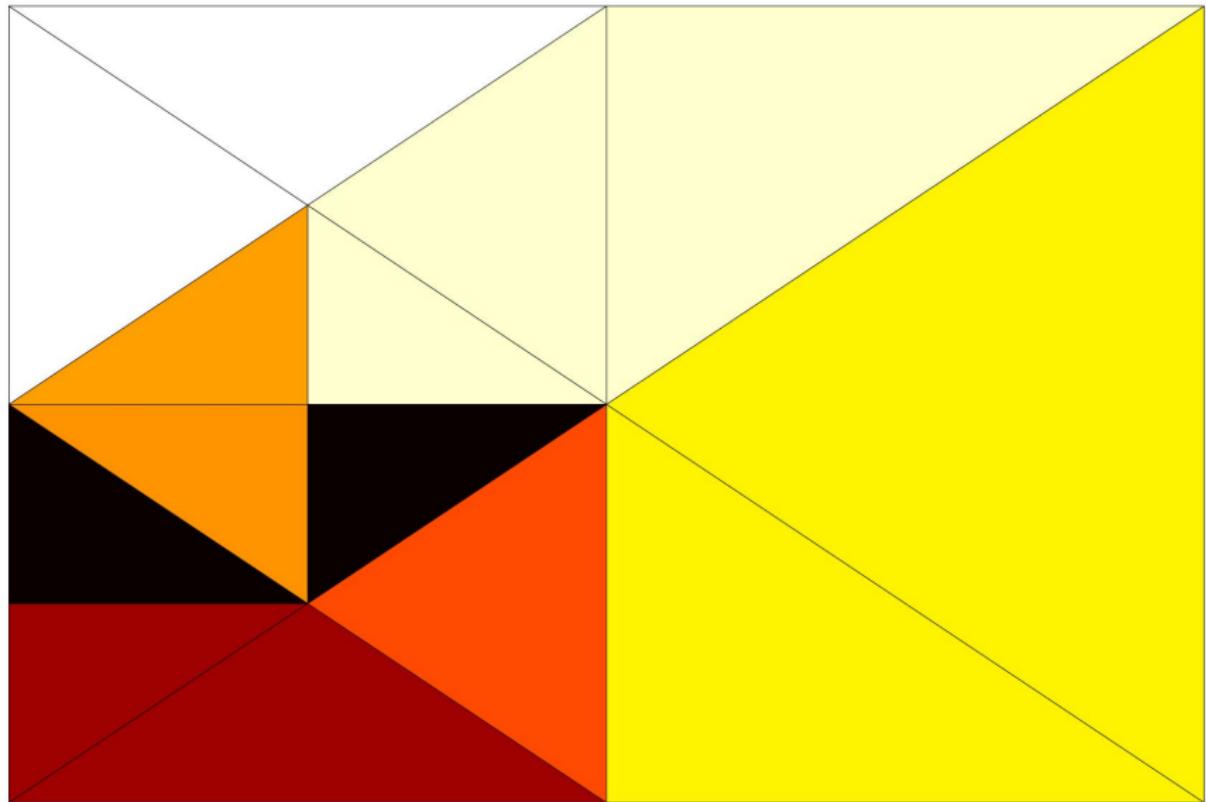
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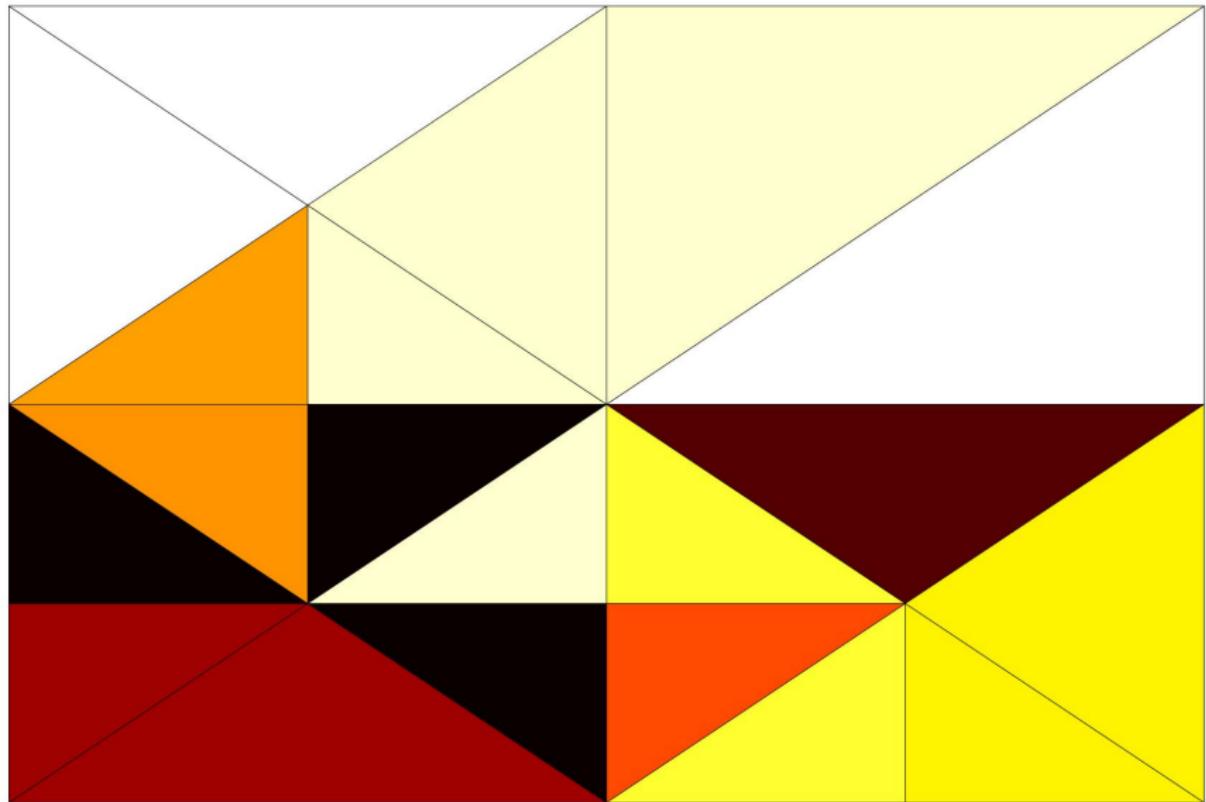
Future work

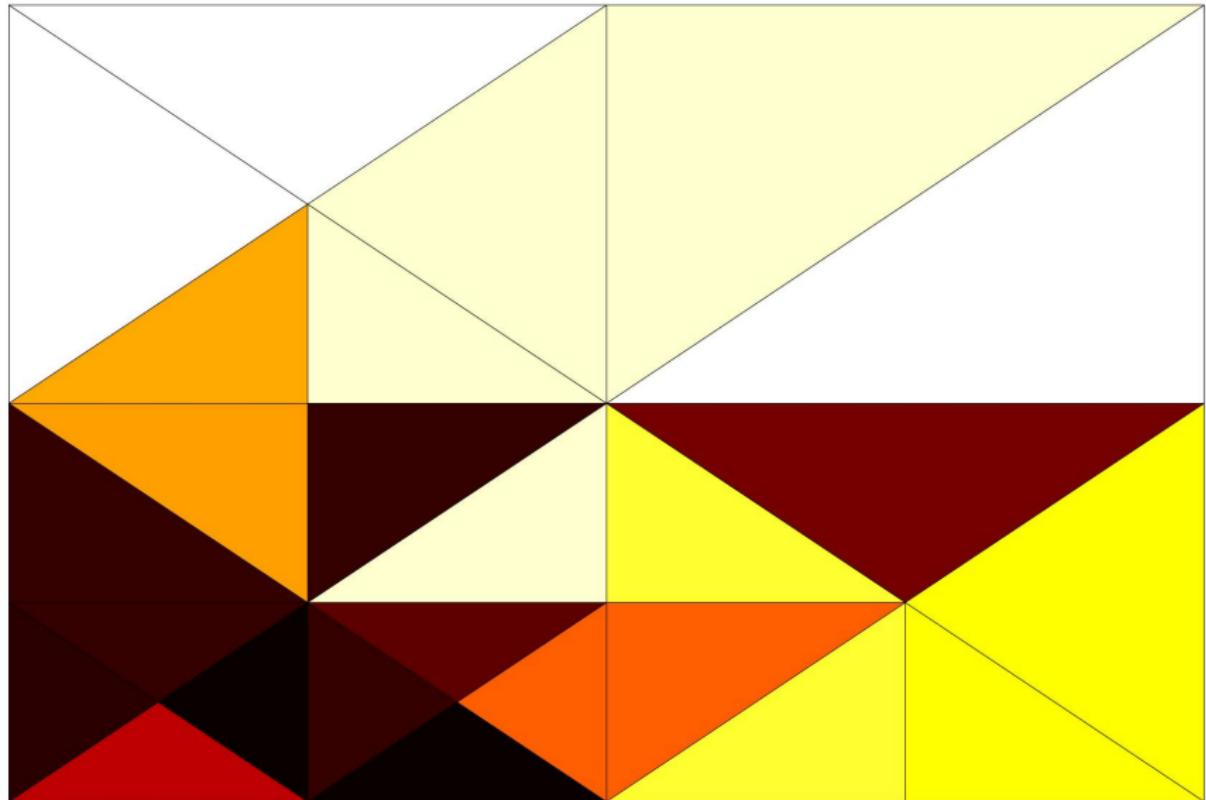
- ▶ Implementation of P_2 AFEM (boundary residual approximation, oscillation errors)
- ▶ Solving the same (maj/min) problems with BEM?
- ▶ Implementation of direct method.
- ▶ 3D implementation

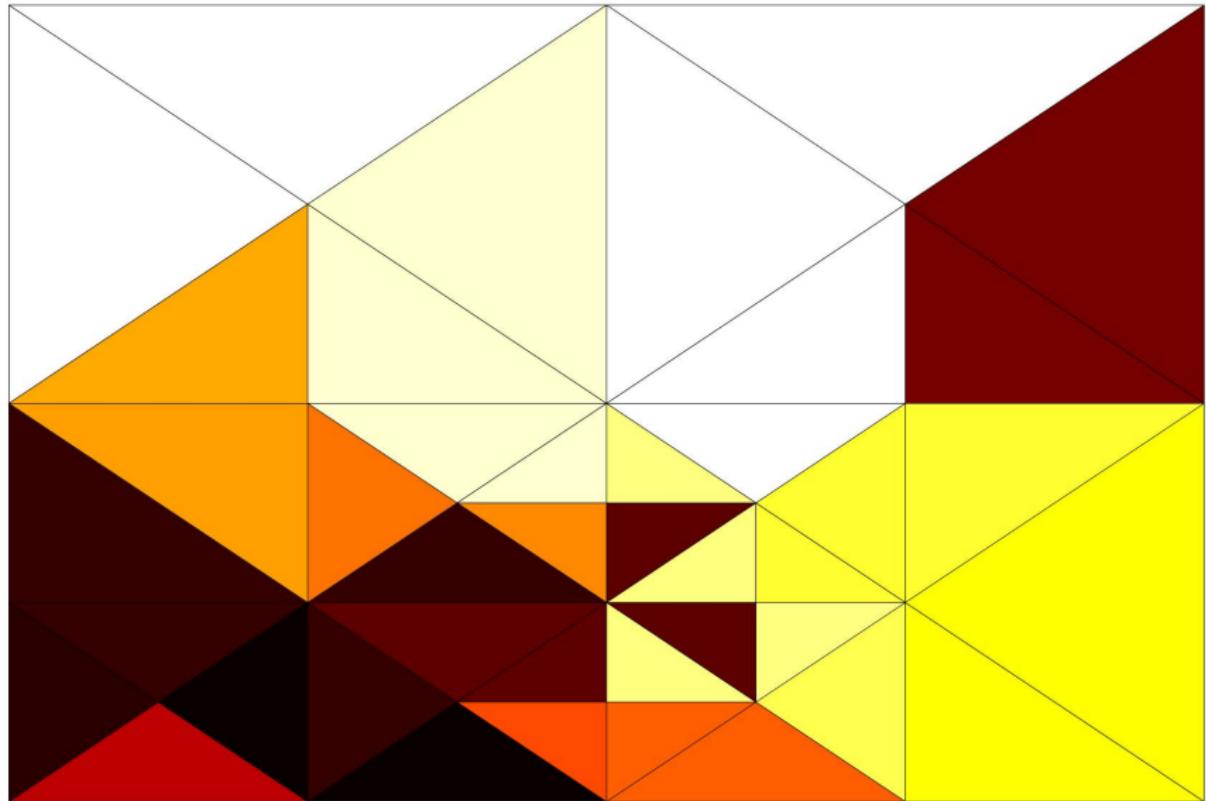
References

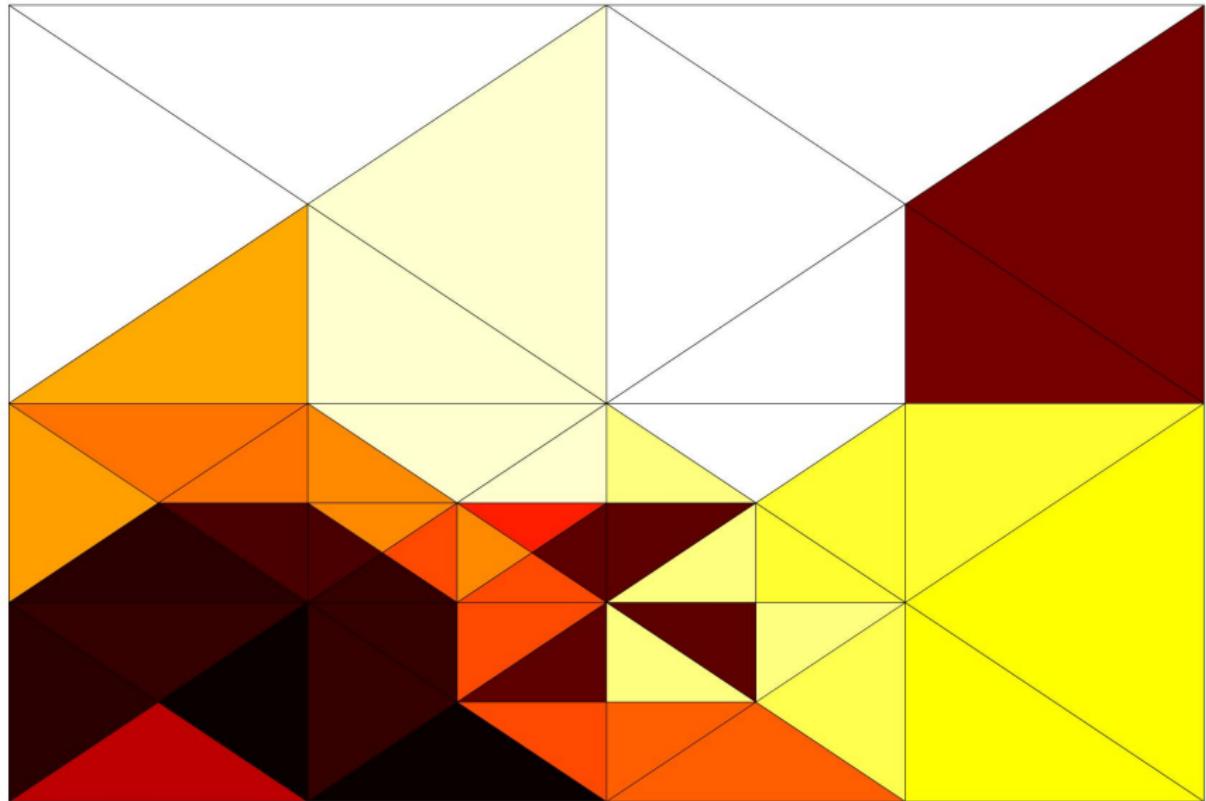
-  S. Repin. [A posteriori estimates for partial differential equations.](#)
Walter de Gruyter, Berlin, 2008.
-  S. Repin and S. Kurz. [Basic Introduction into the Boundary Element Method and Related Functional Error Estimates.](#) 27.10.2017.
-  D. Pauly, S. Repin, D. Praetorius and S. Kurz. [Private communication.](#)
-  M. Aurada, M. Ebner, M. Feischl, S. Ferraz-Leite, T. Führer, P. Goldenits, M. Karkulik, M. Mayr, D. Praetorius. [HILBERT - a Matlab implementation of adaptive BEM \(Release 3\).](#) May 2013.
-  S. Funken, D. Praetorius, P. Wissgott. [Efficient implementation of adaptive P1-FEM in Matlab.](#)
-  C. Bahriawati, C. Carstensen [Three MATLAB implementations of the lowest-order Raviart-Thomas MFEM with a posteriori error control.](#)

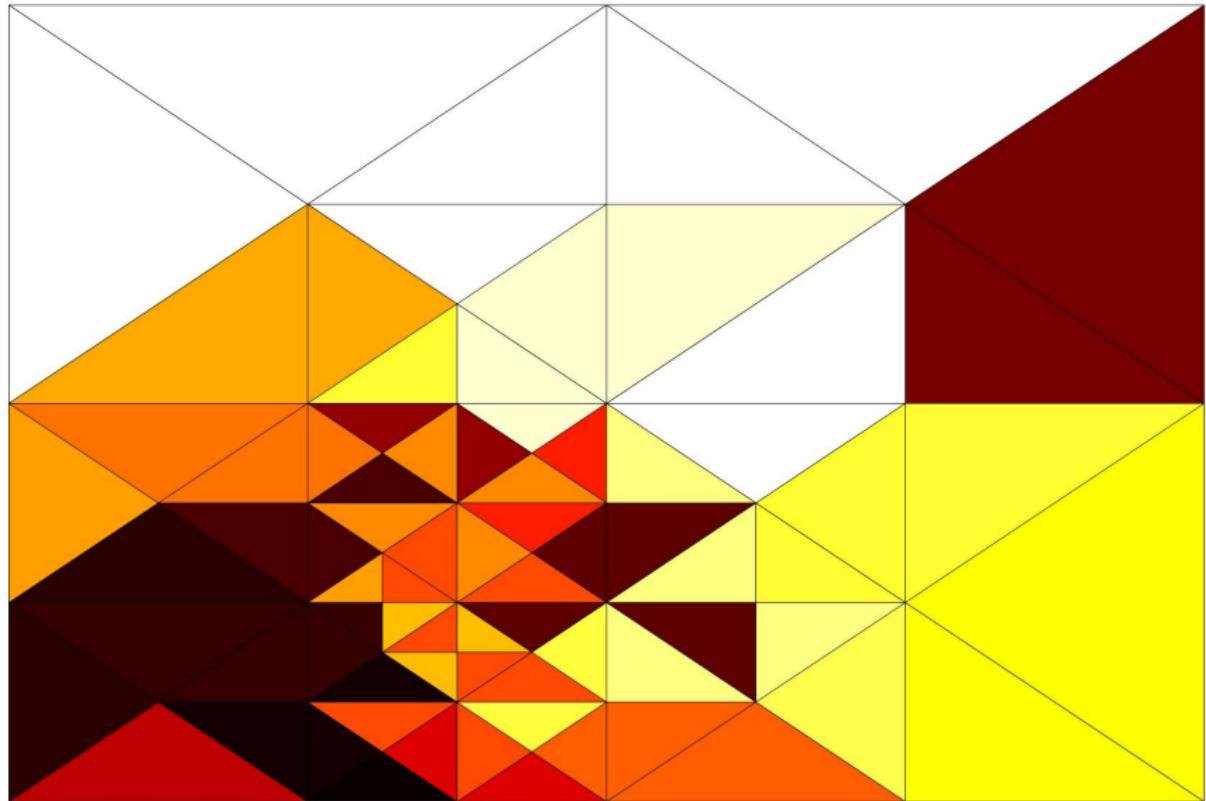


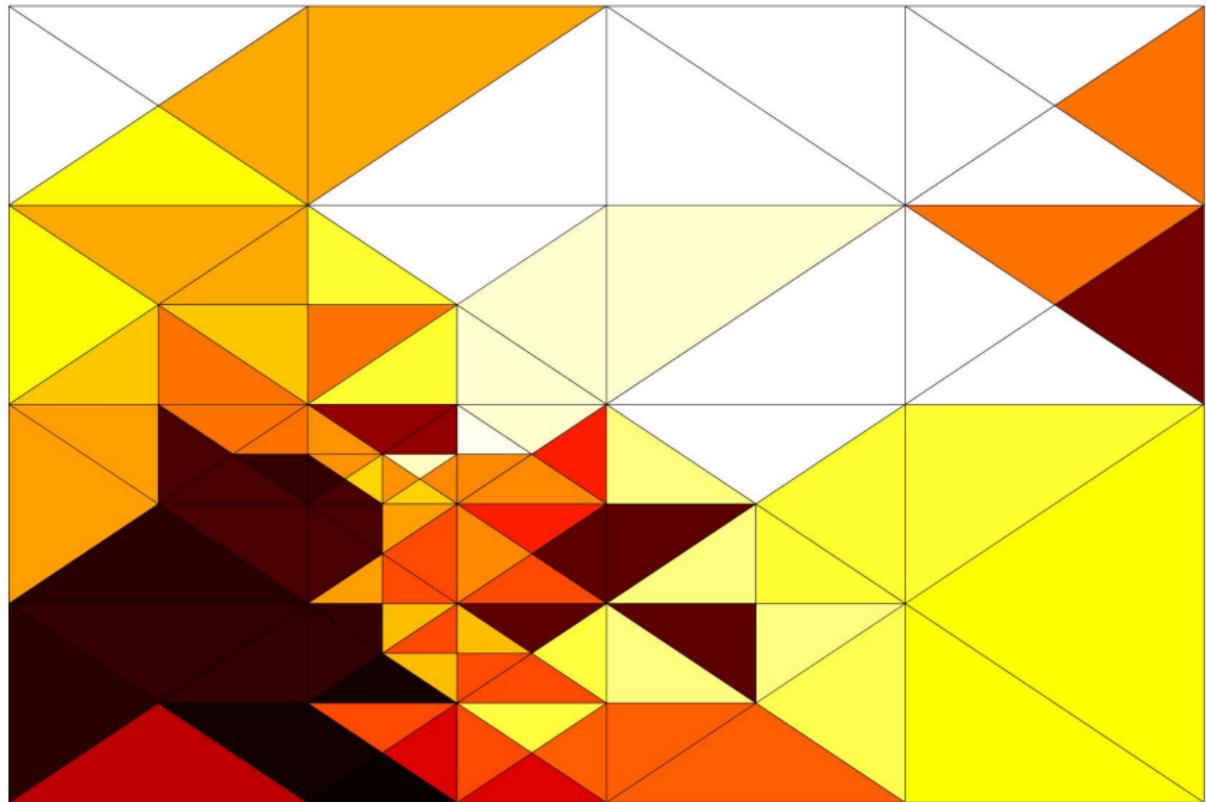


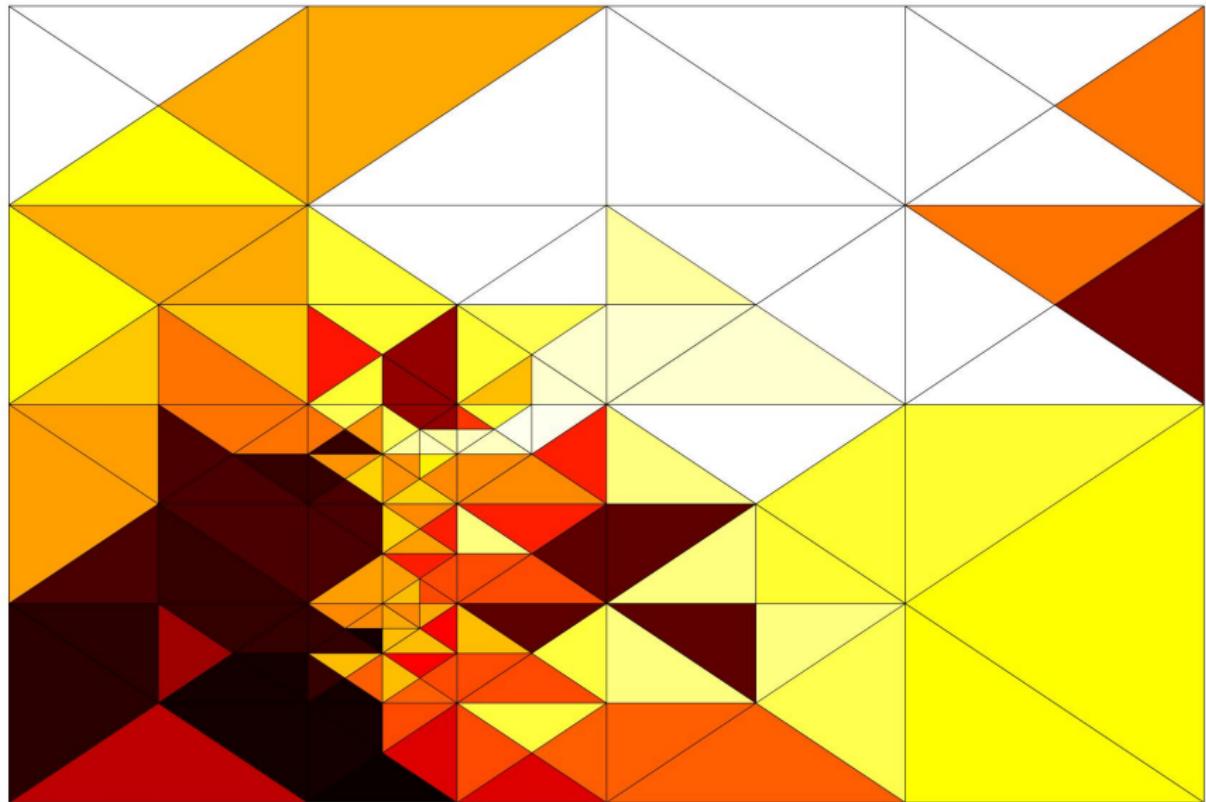


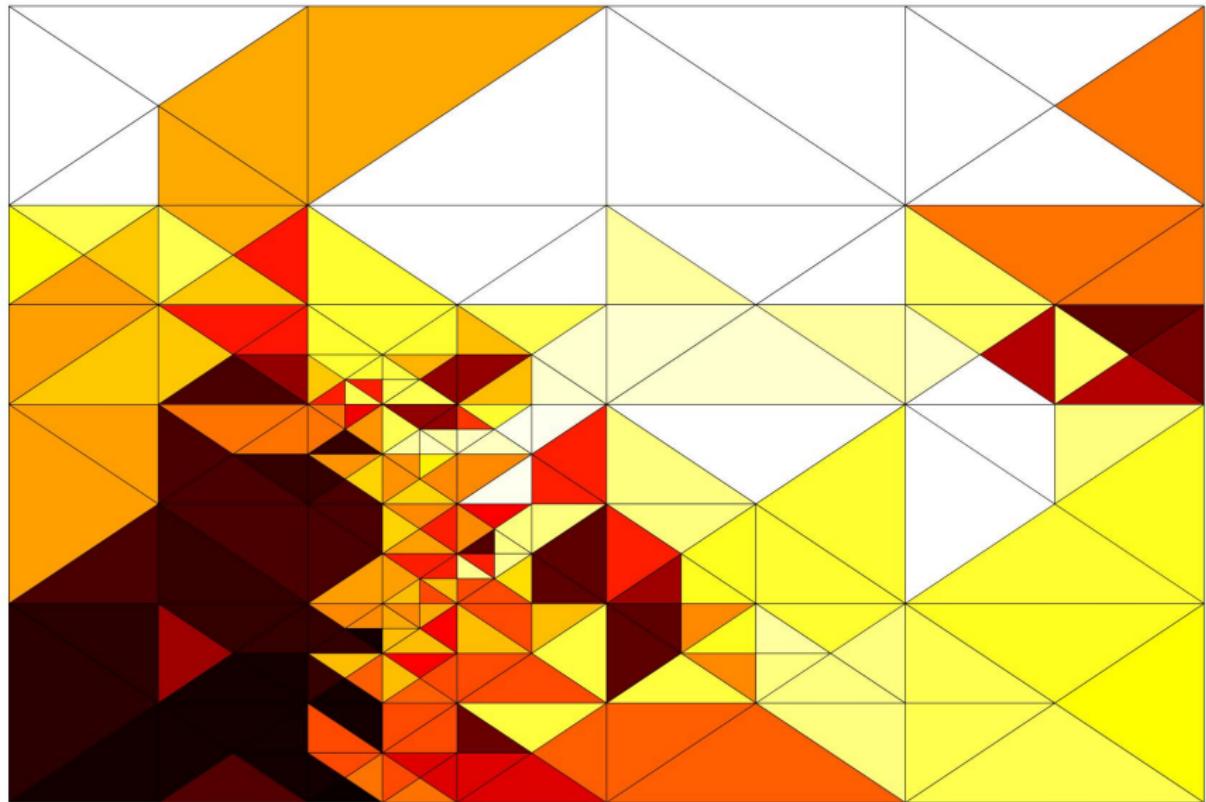


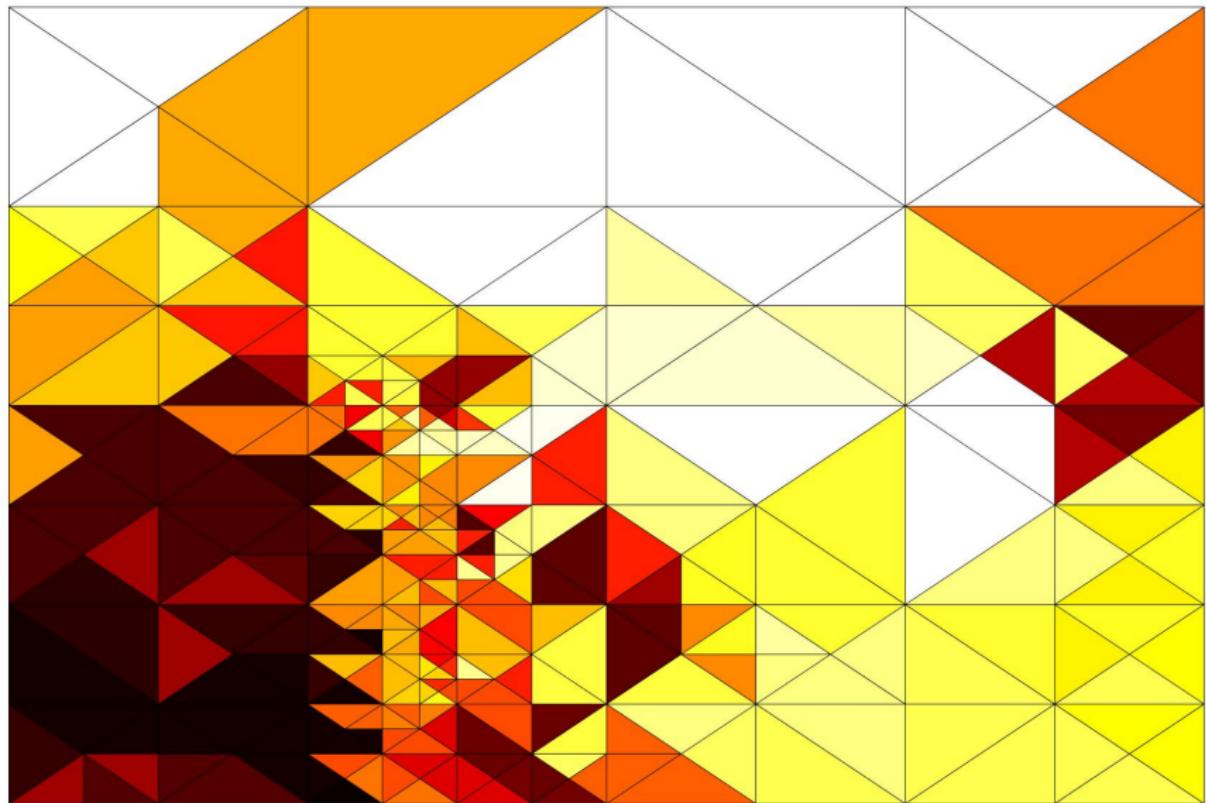


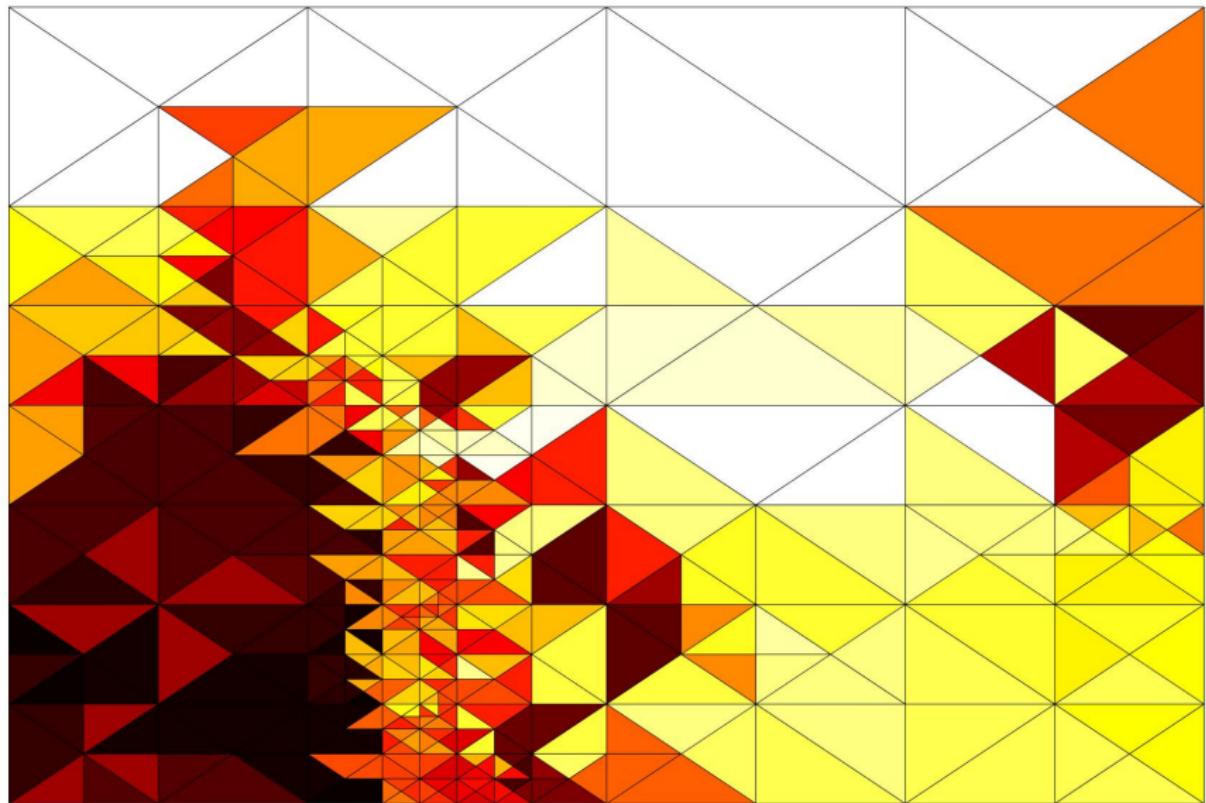


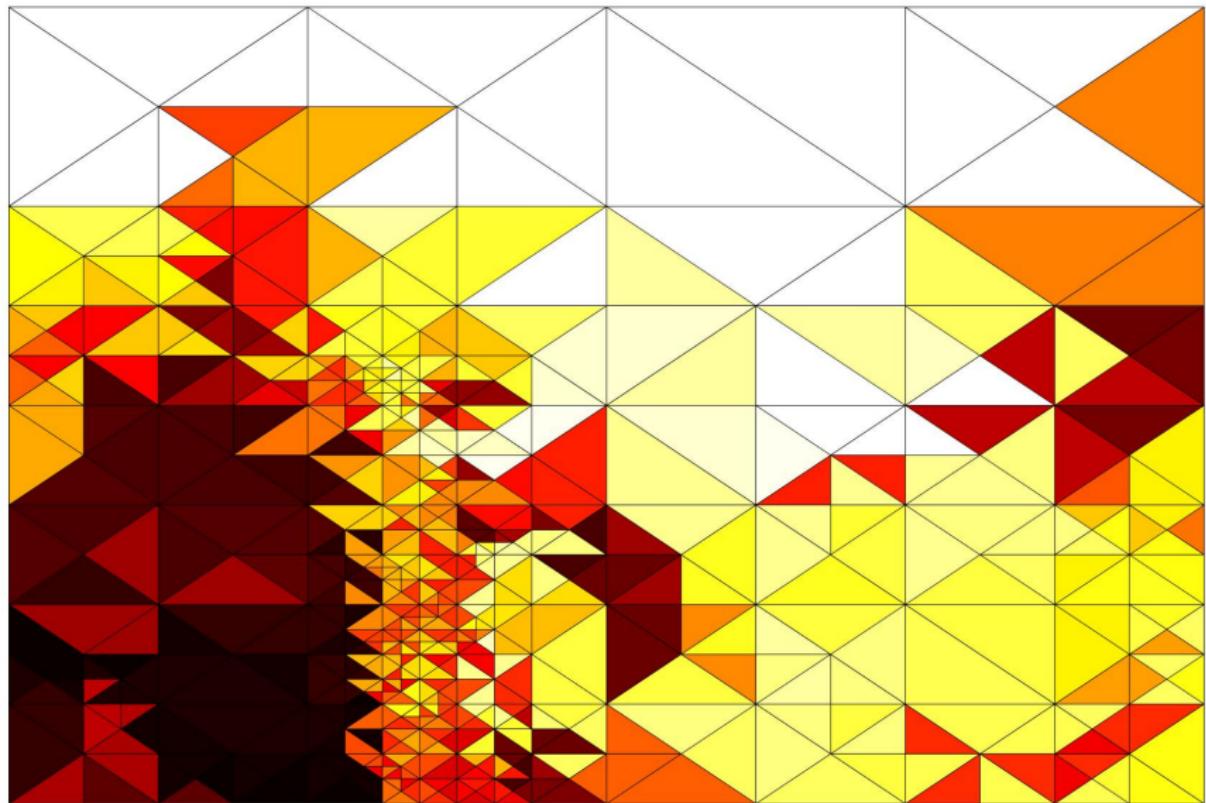


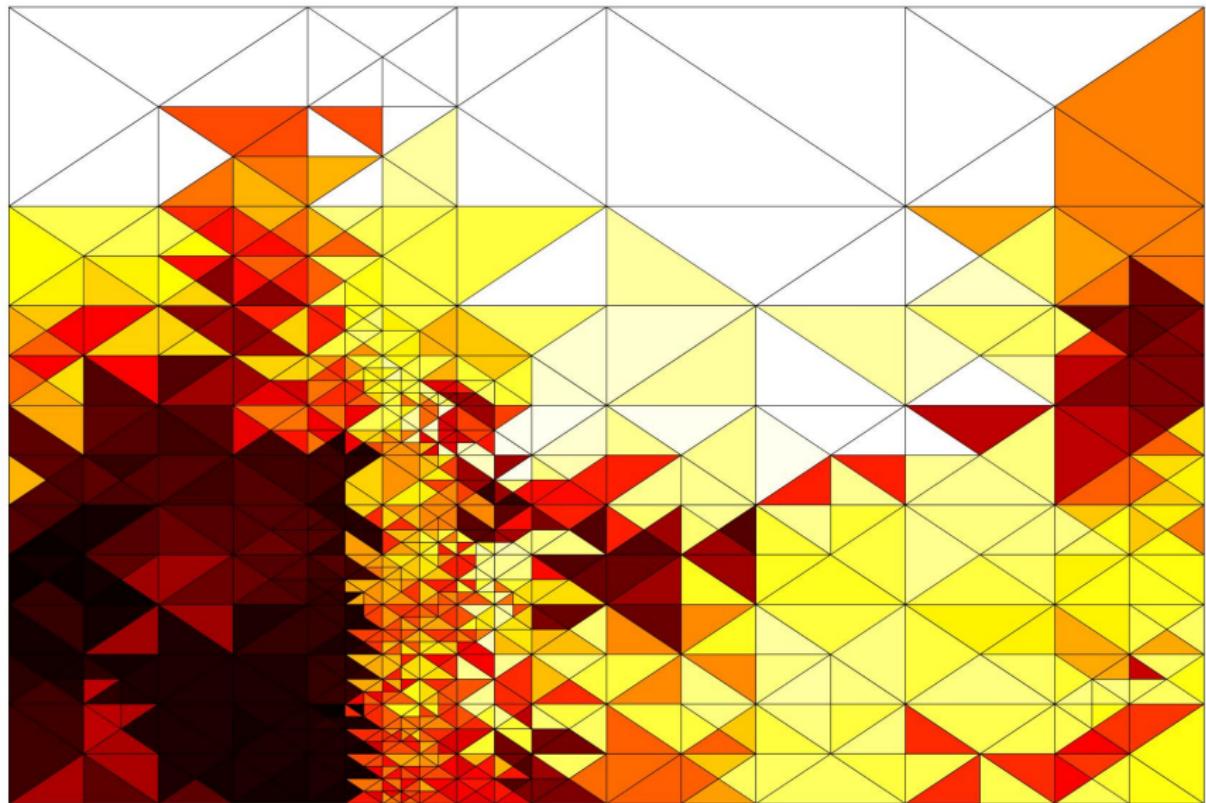


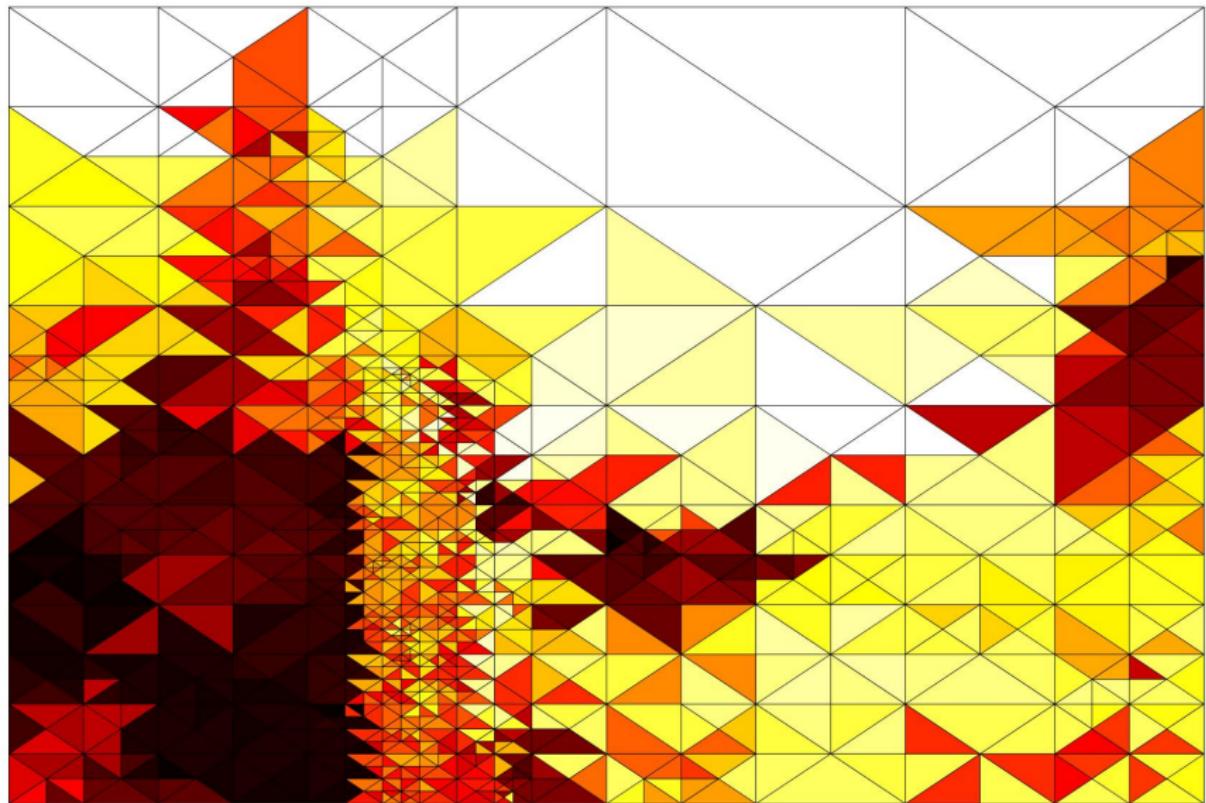


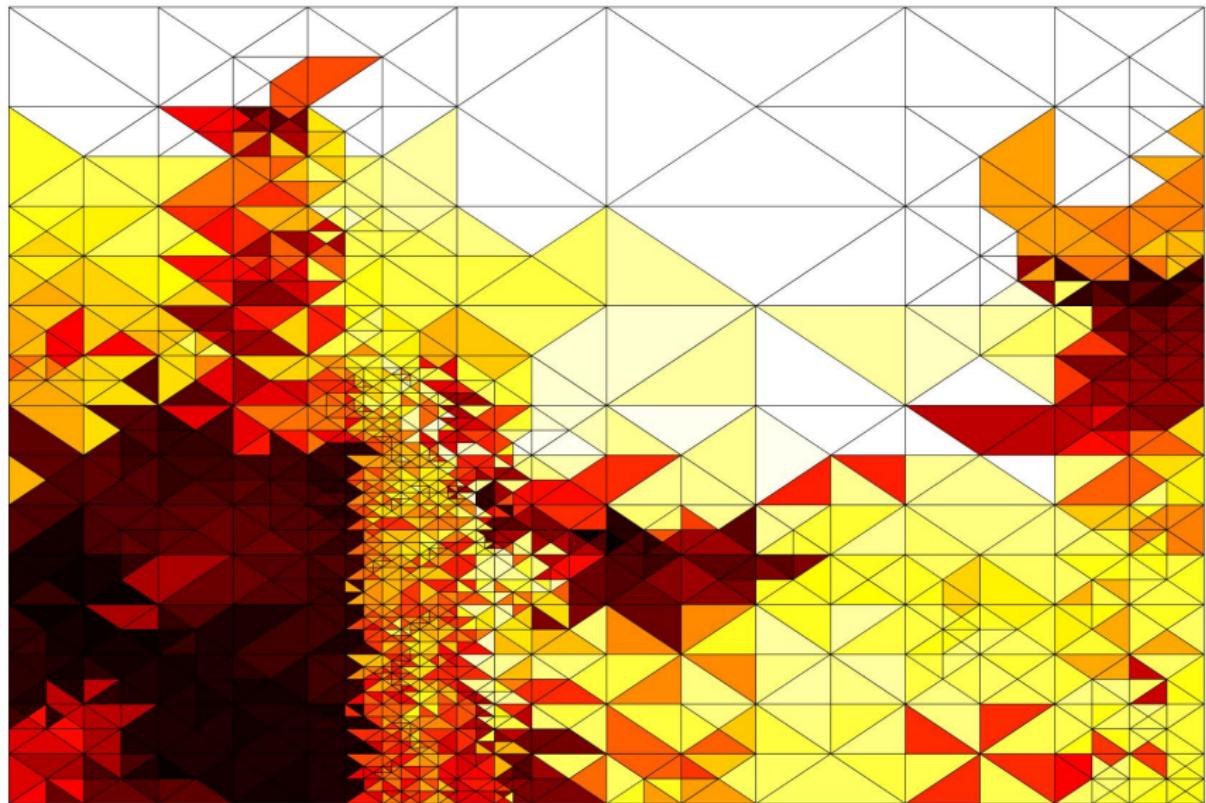


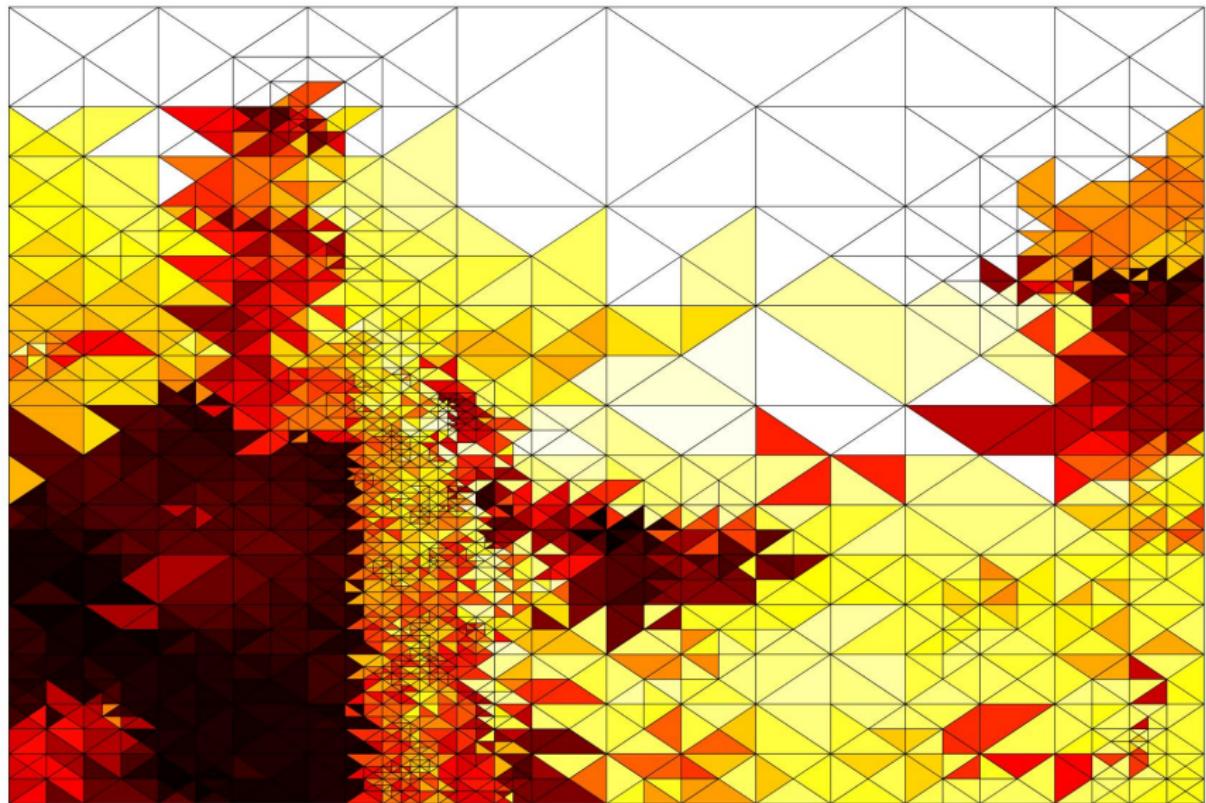


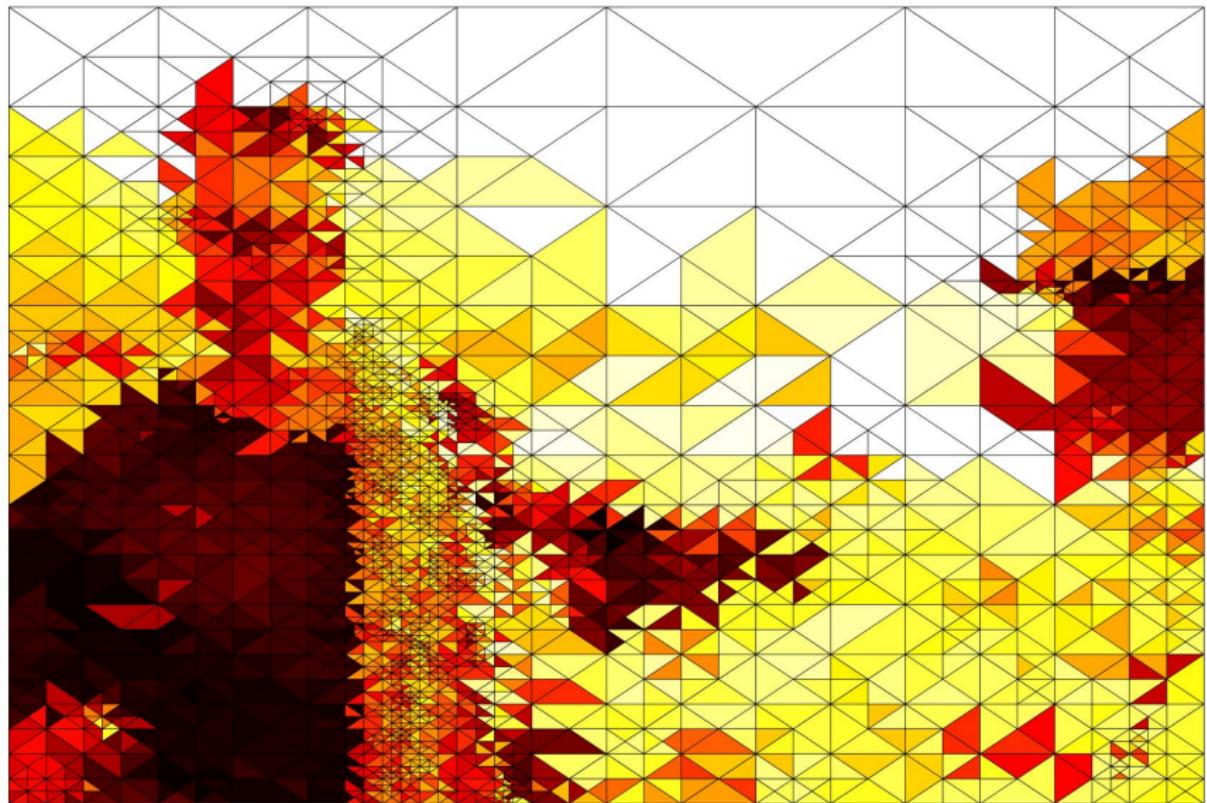


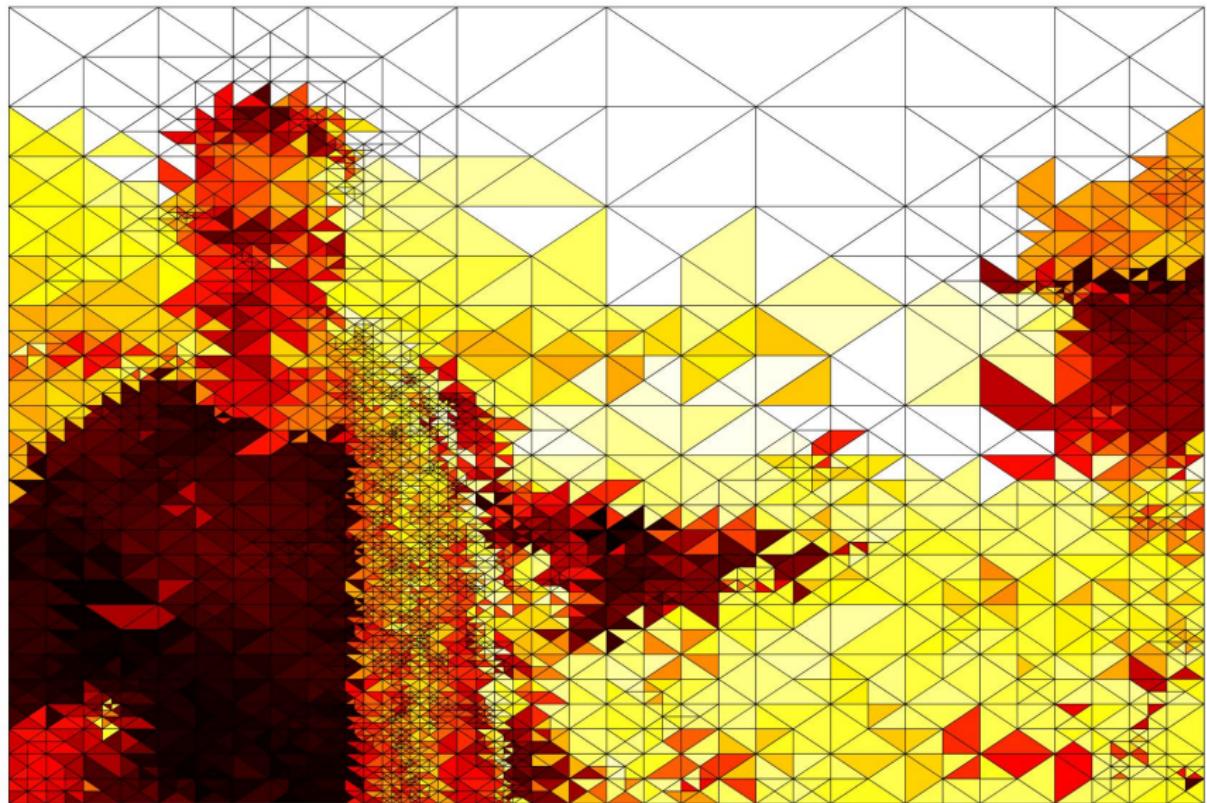


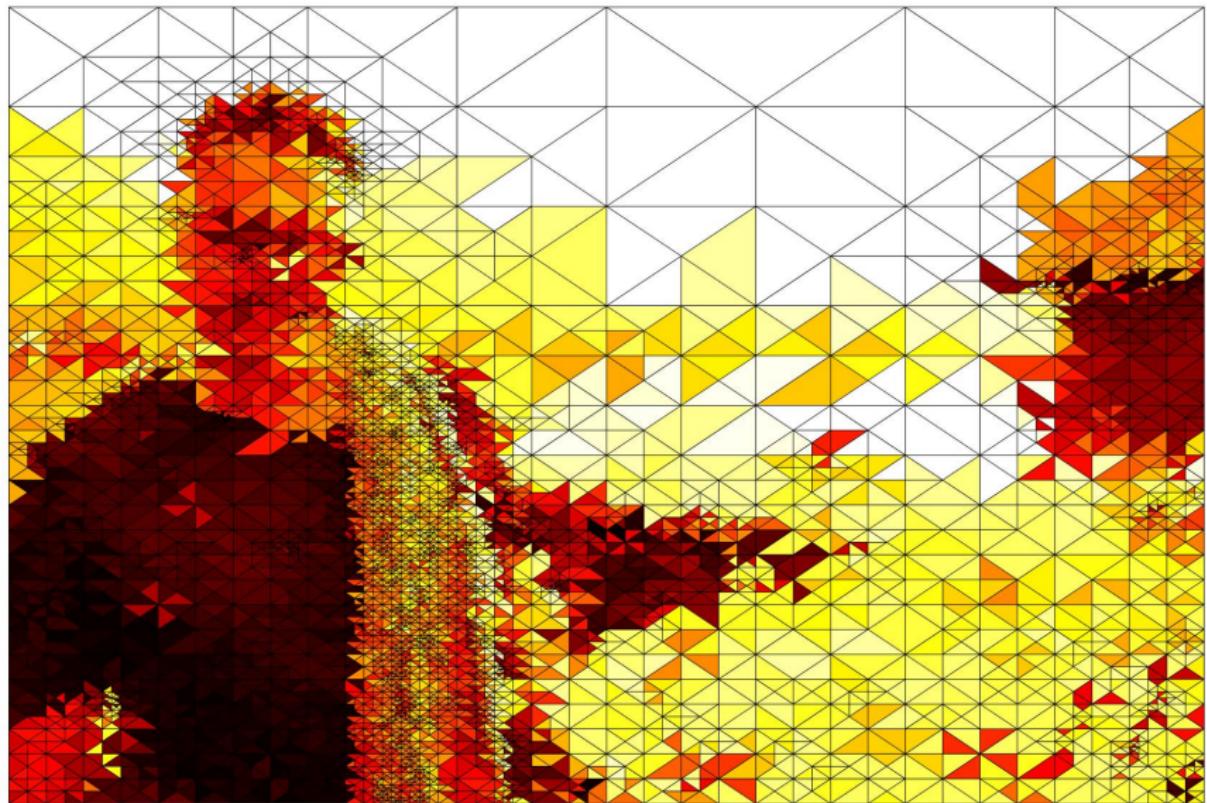


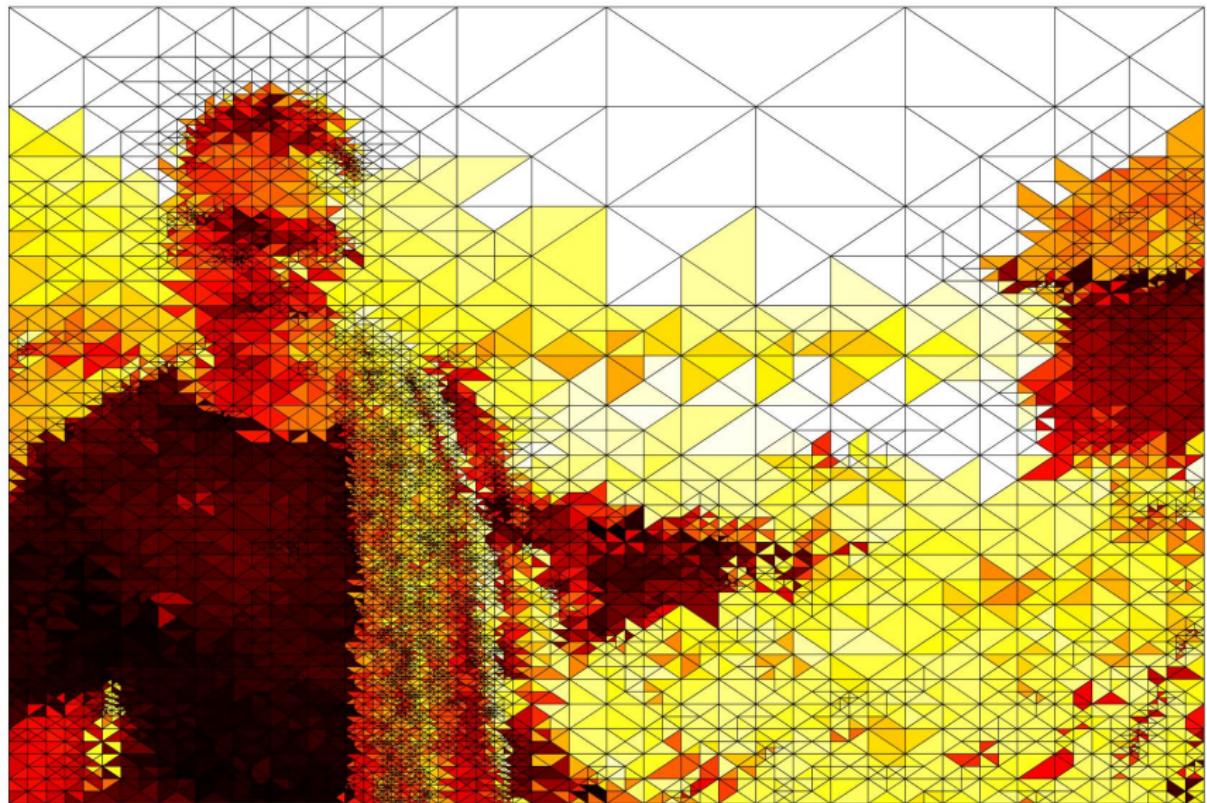


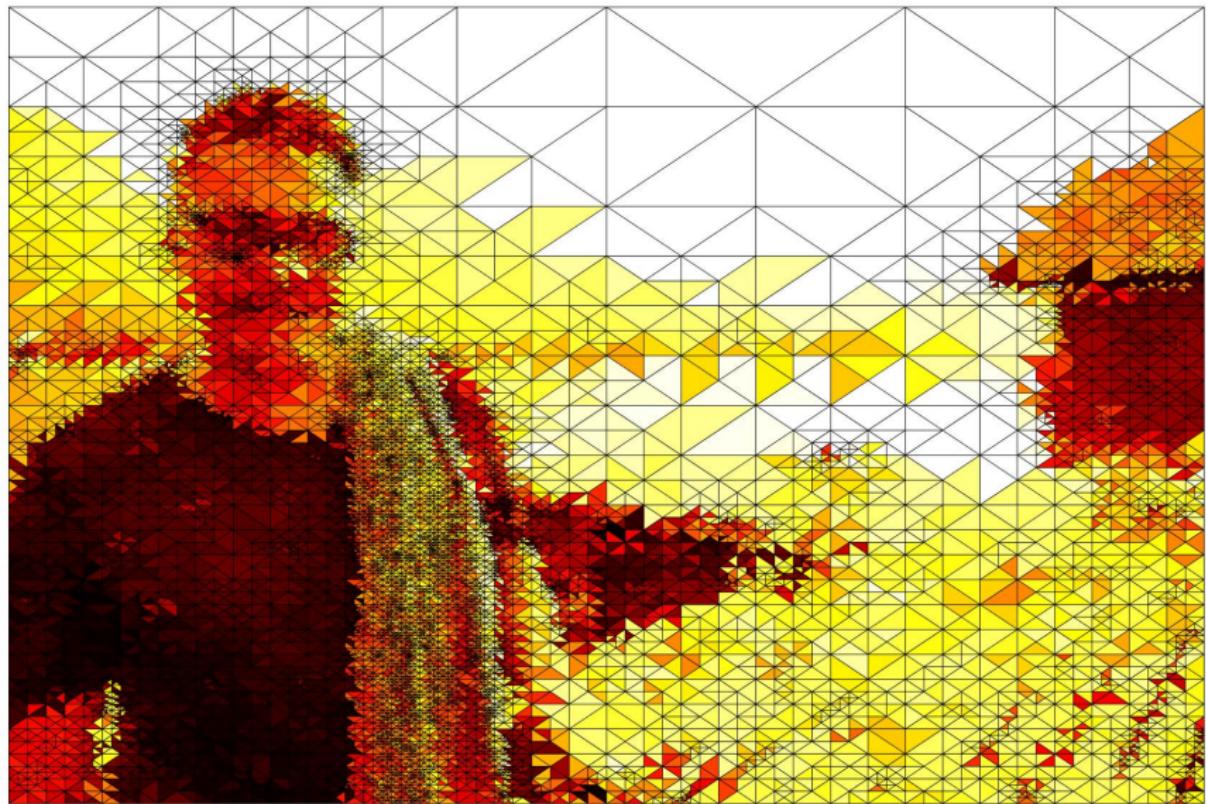


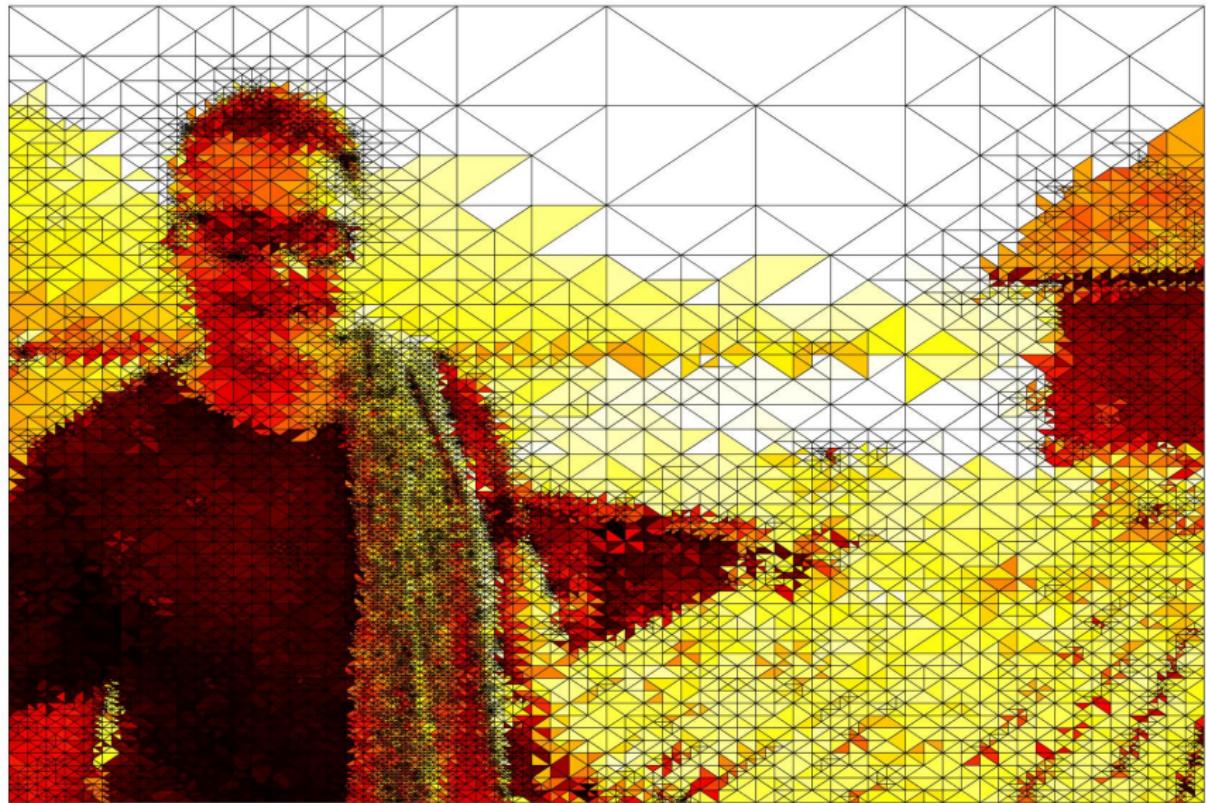


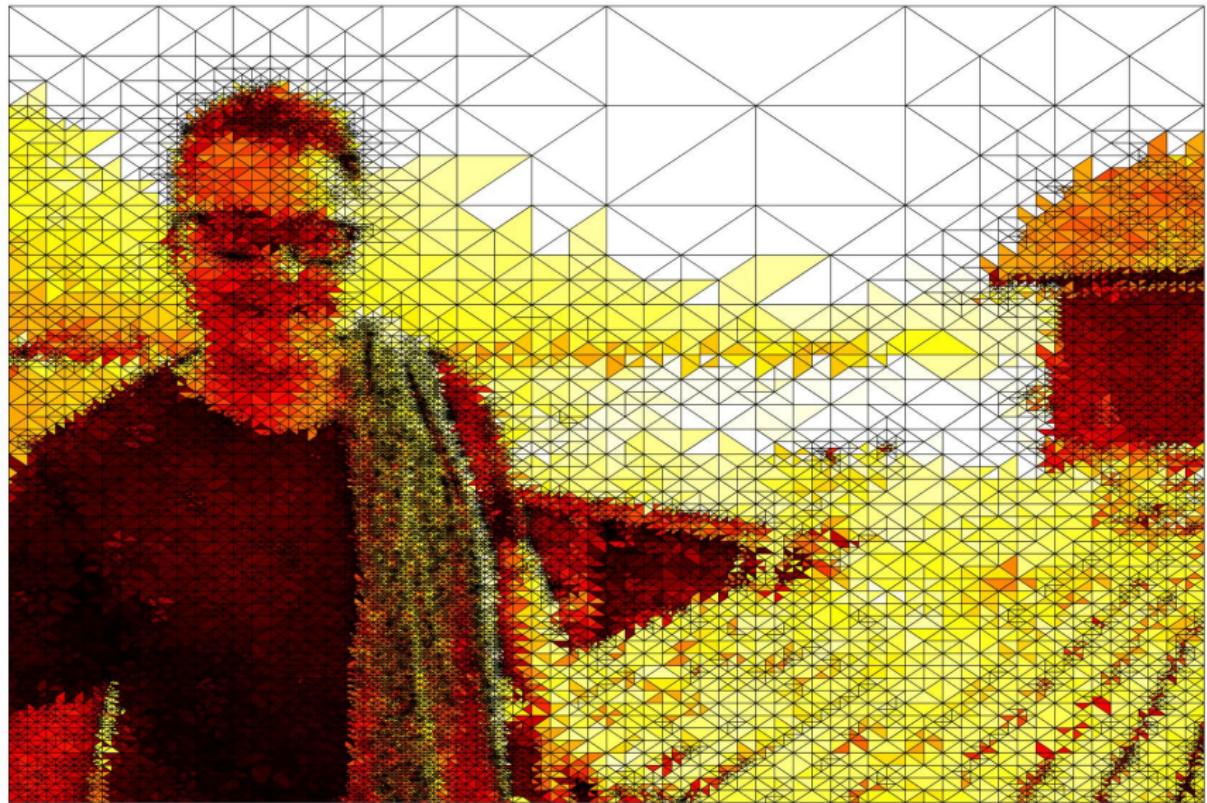




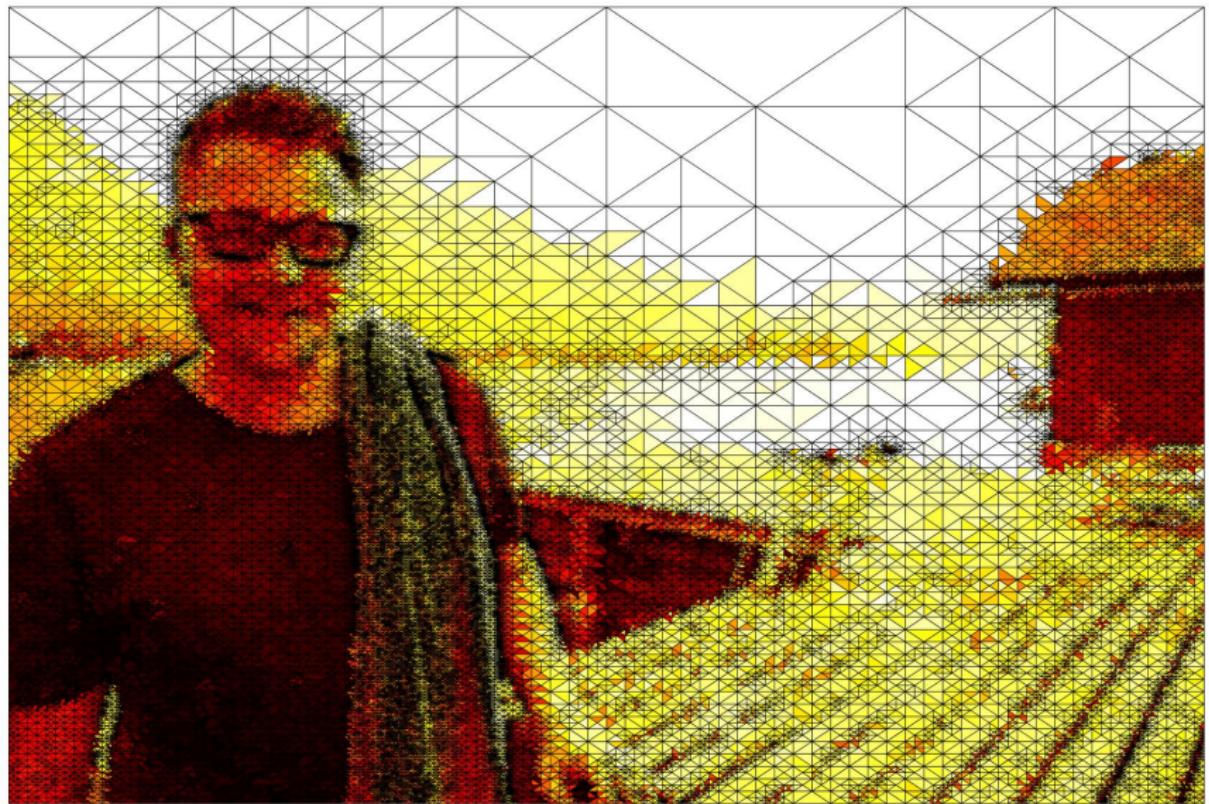
















Daniel Sebastian, Strobl am Wolfgangsee, July 1, 2019



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