# DATA OSCILLATION AND CONVERGENCE OF ADAPTIVE FEM

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## The last week's seminar

#### Assumptions

- The initial mesh is sufficiently refined to resolve data within a tolerance  $\mu \epsilon \ll \epsilon$  (mesh fineness).
- The sum of the local error indicators of elements marked for refinement amounts to a fixed portion of the global error estimator (marking strategy).

## **Framework I**

- $\Omega \subset \mathbb{R}^d, d \in \{2, 3\}$  polygonal/polyhedral bounded domain
   $f \in L^2(\Omega)$

■ *A* is piecewise constant positive definite symmetric

- **Continuous:** Seek  $u \in H_0^1(\Omega) : (u, v)_{A,\Omega} = (f, v)_{0,\Omega}$   $v \in H_0^1(\Omega)$
- **T**<sub>*H*</sub> conforming regular triangulation of  $\Omega$  with piecewise constant mesh-size *H*, i.e.,  $H|_T = \text{diam}T$  for  $T \in \mathcal{T}_H$ .

 $\blacksquare$   $V^H$  space of continuous piecewise linear functions over  $\mathcal{T}_H$ 

 $\blacksquare V_0^H \subset V^H$  with vanishing boundary

**Discrete:** Seek  $u_H \in V_0^H : (u_H, \phi)_{A,\Omega} = (f, v)_{0,\Omega} \qquad \phi \in V_0^H$ 

## **Framework II**

- f<sub>H</sub> piecewise constant function over  $\mathcal{T}_H$  that is equal to mean value  $f_T$  of F on element  $T \in \mathcal{T}_H$ .
- $\blacksquare$   $S_H$  the set of inner sides of  $T_H$
- For  $S \in S_H$  define  $\Omega_S$  as union of the two elements in  $\mathcal{T}_H$  sharing S
- $\blacksquare$   $H_S$  denotes the diameter of S

## Assumptions

All partitions  $\mathcal{T}_H$  match the discontinuities of A, i.e., the jumps of A are located on  $\mathcal{S}_H$ .

## The DIFFERENCE to the last week's seminar I

Introduce <u>data oscillation</u>,

$$\operatorname{osc}(f, \mathcal{T}_H) := \left(\sum_{T \in \mathcal{T}_H} \|H(f - f_T)\|_{0,T}^2\right)^{1/2}$$

- $osc(f T_H)$  measures intrinsic information missing in the averaging process associated with finite elements, which fail to detect fine structures of f.
- The definition of osc(.) is <u>unrelated to quadrature</u> and quantifies data oscillation with the least amount of information per element, namely one degree of freedom associated with  $f_T$ .

## The DIFFERENCE to the last week's seminar II

Last week mesh fineness

$$\left(\sum_{T\in\mathcal{T}_H} \|Hf\|_H^2\right)^{1/2} \le \mu\epsilon.$$

This week <u>oscillations</u>

$$\left(\sum_{T\in\mathcal{T}_H} \|H(f-f_T)\|_H^2\right)^{1/2} \le \mu\epsilon.$$

# The MAIN result

#### Theorem

Let  $(u_k)_k$  be a sequence of FE solution produced by Algorithm C. Then there exist positive constants  $C_0$  and  $\beta < 1$ , depending only on f and the initial grid, such that

$$|u - u_k||_{A,\Omega} \le C_0 \beta^k,$$

with  $||u||^2_{A,\Omega} := (u, u)_{A,\Omega}$ .

## Comparison to PREVIOUS SEMINAR

Theorem does NOT imply that the error decays in every single step: It may be constant for a number of steps due to unresolved data oscillations

## **RESIDUAL-TYPE a posteriori error estimator**

#### Local error indicators

$$\eta_S^2 := \|H_S^{1/2} J_S\|_S^2 + \|Hf\|_{\Omega_S}^2$$

with  $J_S := [A \nabla u_H]_S \cdot \nu$ .

Global error estimator

$$\eta_H^2 := \sum_{S \in \mathcal{S}_H} \eta_S^2$$

#### Theorem

$$\begin{aligned} \|u - u_h\|_{A,\Omega}^2 &\leq C_1 \eta_H^2 \\ \|u - u_h\|_{A,\Omega_S}^2 &\geq C_2 \eta_S^2 - C_3 \|H(f - f_h)\|_{0,\Omega_s}^2 \end{aligned}$$

# Marking I

## Marking Strategy E

Given a parameter  $0 < \theta < 1$ 

1. Construct a subset  $\hat{\mathcal{S}}_H \subset \mathcal{S}_H$  such that

$$\left(\sum_{S\in\hat{\mathcal{S}}_H}\eta_S^2\right)^{1/2} \ge \theta\eta_H.$$

2. Let  $\hat{\mathcal{T}}_H$  be the set of elements with one side in  $\hat{\mathcal{S}}_H$  and mark all these elements.

# Marking II

#### Theorem (error reduction)

Let  $\mathcal{T}_h$  be a refinement of  $\mathcal{T}_H$  such that each element of  $\hat{\mathcal{T}}_H$ , as well as each side in  $\hat{\mathcal{S}}_H$ , contains a node of  $\mathcal{T}_h$  in its interior.

Then there exist constants  $\mu > 0$  and  $0 < \alpha < 1$ , depending only on the initial triangulation, such that for any  $\epsilon > 0$ 

$$osc(f, \mathcal{T}_H) \le \mu \epsilon \implies \|u - u_H\|_{A,\Omega} \le \epsilon \lor \|u - u_h\|_{A,\Omega} \le \alpha \|u - u_H\|_{A,\Omega}.$$

## Lemmata I

## Lemma (Error reduction = $||u_H - u_h||^2_{A,\Omega}$ )

Let  $\mathcal{T}_h$  be a local refinement of  $\mathcal{T}_H$  such that  $V^H \subset V^h$ . Then

$$||u - u_h||_{A,\Omega}^2 = ||u - u_H||_{A,\Omega}^2 - ||u_H - u_h||_{A,\Omega}^2.$$

#### Proof.

Galerkin orthogonality.

$$(u - u_h, v_h)_{A,\Omega} = 0, \forall v_h \in V^h \implies (u - u_h, \underbrace{u_h - u_H}_{=u - u_h + u_h - u_H})_{A,\Omega} = 0$$

## Lemmata II

# Lemma $(||u_H - u_h||^2_{A,\Omega} \ge ???||u - u_H||^2_{A,\Omega}$ proportional error decrease)

Let  $\mathcal{T}_h$  be a refinement of  $\mathcal{T}_H$  satisfying the assumption of the THEOREM. Then there exist constants  $C_4, C_5$  depending only on the initial triangulation such that

$$\eta_S^2 \le C_4 \|u_h - u_H\|_{A,\Omega_S}^2 + C_5 \|H(f - f_H)\|_{0,\Omega_S}^2 \qquad \forall S \in \hat{\mathcal{S}}_H.$$

#### Proof.

CONSTRUCTIVE: Integration by parts, Poincare inequality, triangle inequality.

## Lemmata III

## Corollary (GLOBAL lower bound for the error decrease)

Assumptions as in THEOREM. Then

$$\|u_h - u_H\|_{A,\Omega}^2 \ge \frac{\theta^2}{2C_4C_1} \|u - u_H\|_{A,\Omega}^2 - \frac{C_5}{C_4} osc(f, \mathcal{T}_H)^2.$$

## Lemmata IV

#### Proof.

By previous LEMMA and MARKING STRATEGY E we have

$$\begin{aligned} \theta^2 \eta_H^2 &\leq \sum_{S \in \hat{\mathcal{S}}_H} \eta_S^2 \\ &\leq C_4 \sum_{S \in \hat{\mathcal{S}}_H} \|u_h - u_H\|_{A,\Omega_S}^2 + C_5 \sum_{S \in \hat{\mathcal{S}}_H} \|H(f - f_H)\|_{0,\Omega_s}^2 \\ &\leq 2C_4 \|u_h - u_H\|_{A,\Omega}^2 + 2C_5 \|H(f - f_H)\|_{0,\Omega_s}^2. \\ &\implies \|u_h - u_H\|_{A,\Omega}^2 \geq \frac{\theta^2}{2C_4} \eta_H^2 - \frac{C_5}{C_4} \|H(f - f_H)\|_{0,\Omega_s}^2. \end{aligned}$$

Insert error-estimator-LEMMA.

## **Proof of THEOREM**

#### Proof.

$$\begin{aligned} \|u - u_h\|_{A,\Omega}^2 &= \|u - u_H\|_{A,\Omega}^2 - \|u_H - u_h\|_{A,\Omega}^2 \\ &\leq \|u - u_H\|_{A,\Omega}^2 \Big(1 - \frac{\theta^2}{2C_4C_1}\Big) + \frac{C_5}{C_4} \mathsf{osc}(f,\mathcal{T}_H)^2. \end{aligned}$$

Case  $||u - u_H||_{A,\Omega} > \epsilon$ . Hence

$$\|u - u_h\|_{A,\Omega}^2 \le \|u - u_H\|_{A,\Omega}^2 \underbrace{\left(1 - \frac{\theta^2}{2C_4C_1} + \frac{C_5}{C_4}\mu^2\right)}_{\mathbf{u} - \mathbf{u}_H}$$

<1 for  $\mu > 0$  sufficiently small

## **EXAMPLES:** Ingredients for CONVERGERNCE I

#### Interior node 1

Necessity of creating an interior node inside each refined triangle

$$A = \mathsf{Id}, f \equiv 1, \Omega = (0, 1)^2$$
  
" $u_H = (1/12)$ "  
" $u_h = 1/24(1, 1, 2, 1, 1)$ "

## **EXAMPLES: Ingredients for CONVERGERNCE II**

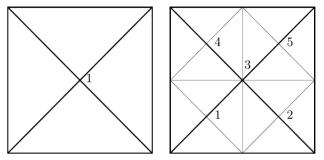


FIG. 3.1. Refinement by bisecting all triangles twice.

## **EXAMPLES: Ingredients for CONVERGERNCE III**

Interior node 2

Also happens "later" for  $osc(f, T_n) = 0$ 

f is orthogonal to the basis functions of  $\mathcal{T}_k, k = 0, 1, 2 \implies u_k \equiv 0, k = 0, 1, 2$ .

 $u_k = 0, k = 3, 4, ...$  on "squares" where *f* changes sign (symmetry of problem).  $u_3, u_4$  behave like in previous example, ie.  $u_3 = u_4$ 

## **EXAMPLES:** Ingredients for CONVERGERNCE IV

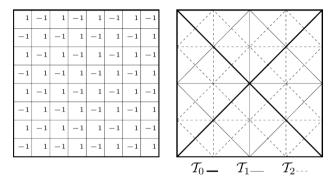


FIG. 3.2. Values of the function f of Example 3.6 for n = 3 (left), and grids  $T_k$  for k = 0, 1, 2 (right).

## **EXAMPLES:** Ingredients for CONVERGERNCE V

Data oscillation

 $osc(f, T_H)$  has to be small

See previous example with additional refinement (interior nodes)

## **EXAMPLES:** Ingredients for CONVERGERNCE VI

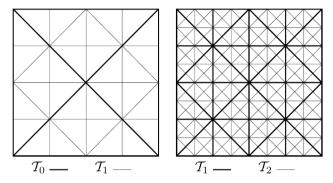


FIG. 3.3. Resulting grid  $\mathcal{T}_1$  (left) and  $\mathcal{T}_2$  (right) after performing three bisections on each element of  $\mathcal{T}_0$  and  $\mathcal{T}_1$ , respectively.

# **EXAMPLES: Ingredients for CONVERGERNCE VII**

## CONCLUSION

- Interior nodes are necessary for error decrease.
- Interior nodes are not sufficient if mesh does not sufficiently resolve oscillation.
- We must readjust the mesh to resolve  $osc(f, T_H)$  according to a decreasing tolerance.

# **EXAMPLES: Ingredients for CONVERGERNCE VIII**

#### Lemma

Let  $0 < \gamma < 1$  reduction factor of element size in one refinement step. Let  $0 < \hat{\theta} < 1, \hat{\alpha} := (1 - (1 - \gamma^2)\hat{\theta}^2)^{1/2}$ . Let  $\hat{\mathcal{T}}_H \subset \mathcal{T}_H$  such that

 $osc(f, \hat{\mathcal{T}}_H) \geq \hat{\theta}osc(f, \mathcal{T}_H).$ 

Then if  $\mathcal{T}_h$  is obtained from  $\mathcal{T}_H$  by refining AT LEAST  $\hat{\mathcal{T}}_H$  one has

 $osc(f, \mathcal{T}_h) \leq \hat{\alpha}osc(f, \mathcal{T}_H).$ 

# **EXAMPLES: Ingredients for CONVERGERNCE IX**

#### Proof.

Per definition,  $f_T = |T|^{-1} \int_T f$  is  $L^2$ -projection of f onto piecewise constants on T. Let  $T \in \mathcal{T}_h, \hat{T} \in \hat{\mathcal{T}}_H, T \subset \hat{T}$ . Hence  $||f - f_T||_T \leq ||f - f_{\hat{T}}||_T$ . Per definition  $h_T \leq \gamma h_{\hat{T}}$ .

$$\begin{aligned} &\mathsf{osc}(f,\mathcal{T}_{h})^{2} = \sum_{T\in\mathcal{T}_{h}} h_{T}^{2} \|f - f_{T}\|_{0,T}^{2} \\ &\leq \gamma^{2} \sum_{\hat{T}\in\hat{\mathcal{T}}_{H}} h_{\hat{T}}^{2} \|f - f_{\hat{T}}\|_{0,\hat{T}}^{2} + \sum_{T\in\mathcal{T}_{H}\setminus\hat{\mathcal{T}}_{H}} h_{T}^{2} \|f - f_{T}\|_{0,T}^{2} \\ &= (\gamma^{2} - 1)\mathsf{osc}(f,\hat{\mathcal{T}}_{H})^{2} + \mathsf{osc}(f,\mathcal{T}_{H})^{2} \leq \hat{\alpha}^{2}\mathsf{osc}(f,\mathcal{T}_{H})^{2} \end{aligned}$$

# **EXAMPLES: Ingredients for CONVERGERNCE X**

#### Lemma

Let f be piecewise  $H^s, 0 < s \le 1$  over initial mesh. Redefine

$$osc(f, \mathcal{T}_h) := \Big(\sum_{T \in \mathcal{T}_h} h_T^{2+2s} \|D^s f\|_{0,T}^2 \Big)^{1/2}.$$

Let 
$$\hat{\alpha} := (1 - (1 - \gamma^{2+2s})\hat{\theta}^2)^{1/2}$$
. Then  $osc(f, \mathcal{T}_h) \leq \hat{\alpha}osc(f, \mathcal{T}_H)$ .

#### Proof.

Analogous to previous lemma.

# **EXAMPLES:** Ingredients for CONVERGERNCE XI

## Marking Strategy D

Given a parameter  $0 < \hat{\theta} < 1$  and the subset  $\hat{\mathcal{T}}_H \subset \mathcal{T}_H$  produced by Marking Strategy E:

**1**. Enlarge  $\hat{\mathcal{T}}_H$  such that

$$\operatorname{osc}(f, \hat{\mathcal{T}}_H) \geq \hat{\theta} \operatorname{osc}(f, \mathcal{T}_H).$$

2. Mark all elements in  $\hat{\mathcal{T}}_H$  for refinement.

#### Convergent Algorithm C

Choose parameters  $0 < \theta, \hat{\theta} < 1$ .

- 1. Pick up any initial mesh  $\mathcal{T}_0$  such that *A* is piecewise constant over  $\mathcal{T}_0$ .
- **2**. Solve the system on  $T_0$  for the discrete solution  $u_0$ .

**3**. Let k = 0.

- 4. Compute the local indicators  $\eta_S$ .
- 5. Construct  $\hat{\mathcal{T}}_k$  by **Marking Strategy D** and parameter  $\hat{\theta}$ .
- 6. Let  $\mathcal{T}_{k+1}$  be a refinement of  $\mathcal{T}_k$  such that each element of  $\hat{\mathcal{T}}_k$ , as well as each of its sides, contains a node of  $\mathcal{T}_{k+1}$  in its interior.
- 7. Solve the system on  $\mathcal{T}_{k+1}$  for the discrete solution  $u_{k+1}$ .
- 8. Let k = k + 1 and go to 4.

# The MAIN RESULT I

## Theorem (CONVERGENCE)

For  $0 < \theta, \hat{\theta} < 1$ , let  $0 < \alpha < 1, \mu > 0$  be given by the "error decreas theorem" and  $0 < \hat{\alpha} < 1$  by the previous lemmata. Algorithm *C* produces a convergent sequence  $(u_k)_{k \in \mathbb{N}_0}$  with

$$\begin{aligned} \|u - u_k\|_{A,\Omega} &\leq C_0 \beta^k, \\ \beta &= \max\{\alpha, \hat{\alpha}\}, \\ C_0 &= \max\{\|u - u_0\|_{A,\Omega}, \frac{\textit{osc}(f, \mathcal{T}_0)}{\alpha \mu}\}. \end{aligned}$$

# The MAIN RESULT II

## Proof.

INDUCTION. IA  $k = 0 \checkmark$ .

IS Case study

- 1.  $||u u_k||_{A,\Omega} > C_0 \beta^{k+1}$ 2.  $||u u_k||_{A,\Omega} \le C_0 \beta^{k+1}$ .

# The MAIN RESULT III

Proof continued.

1. Marking Strategy D gives

 $\operatorname{osc}(f, \mathcal{T}_k) \leq \hat{\alpha}^k \operatorname{osc}(f, \mathcal{T}_0) \leq \beta^k \operatorname{osc}(f, \mathcal{T}_0)$ 

Hence for  $\epsilon := C_0 \beta^{k+1}$ 

$$\operatorname{osc}(f, \mathcal{T}_k) \le \mu C_0 \alpha \beta^k \le \mu C_0 \beta^{k+1} = \mu \epsilon.$$

Since, per assumption,  $||u - u_k||_{A,\Omega} > \epsilon$  use IH and Error Reduction THEOREM

$$||u - u_{k+1}||_{A,\Omega} \le \beta ||u - u_k||_{A,\Omega} \le C_0 \beta^{k+1}.$$

# The MAIN RESULT IV

Proof continued.

2. Since  $T_{k+1}$  is refinement of  $T_k$ , error cannot increase,

$$||u - u_{k+1}||_{A,\Omega} \le ||u - u_k||_{A,\Omega} \le C_0 \beta^{k+1}$$

#### Practical method?

**Algorithm C** only needs  $\theta$ ,  $\hat{\theta}$ . The unknown constants  $\alpha$ ,  $\hat{\alpha}$ ,  $\mu$  are not needed (but give convergence rate).

# **EXAMPLE: Crack problem**

$$\Omega = \{ |x| + |y| < 1 \} \setminus \{ 0 \le x \le 1, y = 0 \}$$
$$u(r, \theta) = r^{1/2} \sin \frac{\theta}{2} - \frac{1}{4}r^2.$$
$$A = I, f = 1.$$

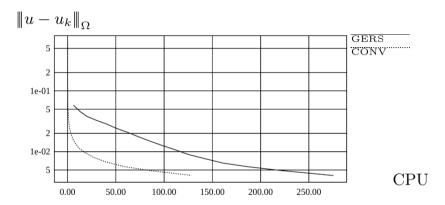


FIG. 5.2. Comparison of CPU time for GERS and CONV.

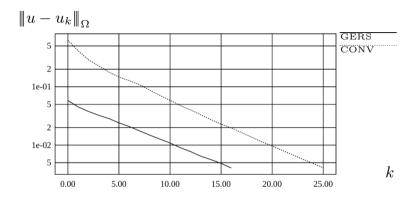


FIG. 5.3. Comparison of reduction rate  $\alpha^k$  for GERS, CONV.

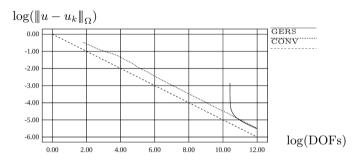
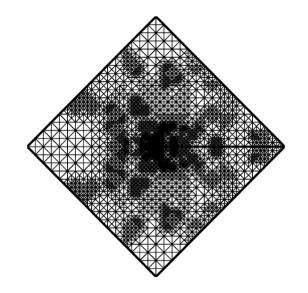
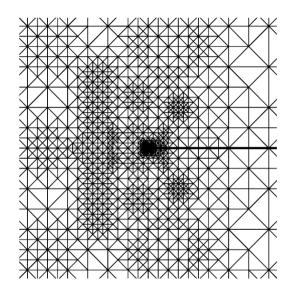
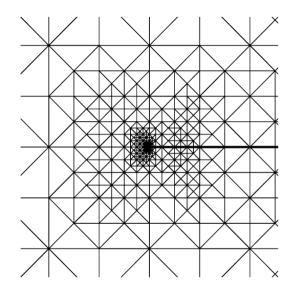
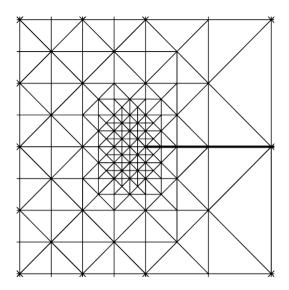


FIG. 5.4. Quasioptimality of GERS and CONV. The optimal decay is indicated by the dashed line with slope -1/2.









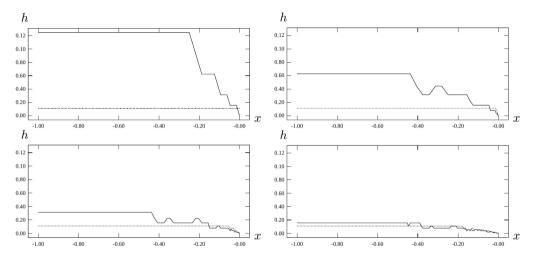


FIG. 5.6. Comparison of local meshsizes h on the line y = 0 for GERS (dotted line) and CONV (solid line) on meshes with approximately same errors  $|||u - u_k|||_{\Omega}$ .

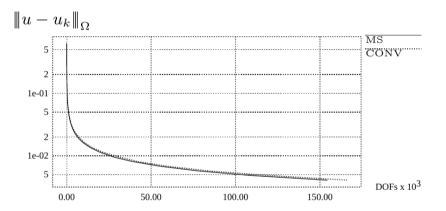


FIG. 5.7. Comparison of CONV and MS.

## **EXAMPLE:** Discontinuous coefficients

 $\square \Omega = (-1,1)^2$ 

- $A = a_1 I$  in the first and third quadrants
- $A = a_2 I$  in the second and fourth quadrants

Exact weak solution of u for  $f \equiv 0$  is given by  $u(r, \theta) = r^{\gamma} \mu(\theta)$  with

$$\mu(\theta) = \begin{cases} \cos((\pi/2 - \sigma)\gamma) \cdot \cos((\theta - \pi/2 + \rho)\gamma) & \text{if } 0 \le \theta \le \pi/2, \\ \cos(\rho\gamma) \cdot \cos((\theta - \pi + \sigma)\gamma) & \text{if } \pi/2 \le \theta \le \pi, \\ \cos(\sigma\gamma) \cdot \cos((\theta - \pi - \rho)\gamma) & \text{if } \pi \le \theta < 3\pi/2, \\ \cos((\pi/2 - \rho)\gamma) \cdot \cos((\theta - 3\pi/2 - \sigma)\gamma) & \text{if } 3\pi/2 \le \theta \le 2\pi \end{cases}$$

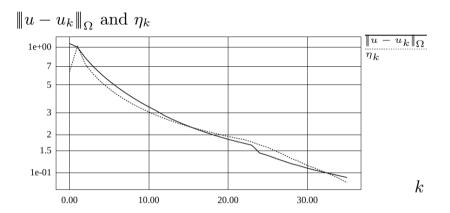


FIG. 5.8. Error reduction: estimate and true error.

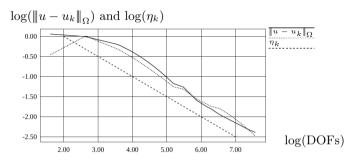
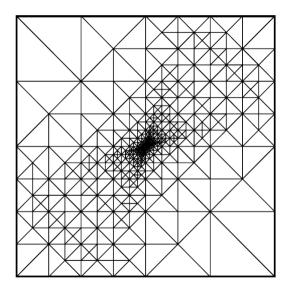
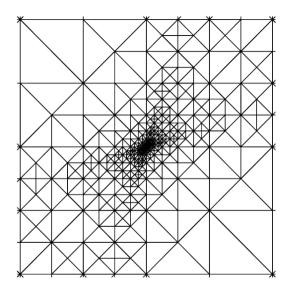
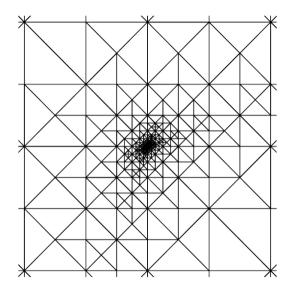
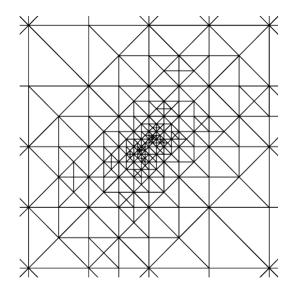


FIG. 5.9. Quasioptimality of CONV: estimate and true error. The optimal decay is indicated by the dashed line with slope -1/2.









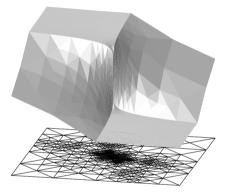


FIG. 5.11. Graph of the discrete solution and underlying grid.

## **EXAMPLE:** Variable source

$$\Omega = (-1, 1)^d, d = 2, 3$$
$$u(x) = e^{-10|x|^2}$$

• 
$$A = I$$
 and nonconstant  $f = -\Delta u$ .

f exhibits large variations in Ω, forcing "additional" refinement due to oscillation.

## TABLE 5.1

Total number and number of marked elements per iteration in two dimensions (left) and three dimensions (right): est.: marked elements due to error estimator, osc.: additionally marked elements to data oscillation.

iter.	elements	est.	OSC.
0	4	8	0
1	64	16	16
2	704	56	8
3	2256	80	0
4	4208	96	8
5	6624	112	24
6	8752	344	0
7	17512	432	0
8	28368	608	0
9	42896	768	16
10	60216	2192	0
11	113040	2296	24
12	160592	3816	24

iter.	elements	est.	OSC.
0	6	6	0
1	384	48	0
2	7776	48	48
3	15936	576	0
4	112320	5040	0
5	860592	5136	720
6	1693536	30144	0

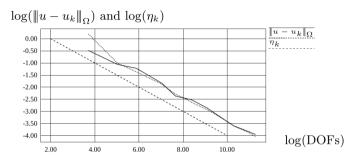


FIG. 5.12. Quasioptimality of CONV: estimate and true error in two dimensions. The optimal decay is indicated by the line with slope -1/2.

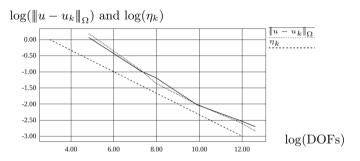


FIG. 5.13. Quasioptimality of CONV: estimate and true error in three dimensions. The optimal decay is indicated by the line with slope -1/3.

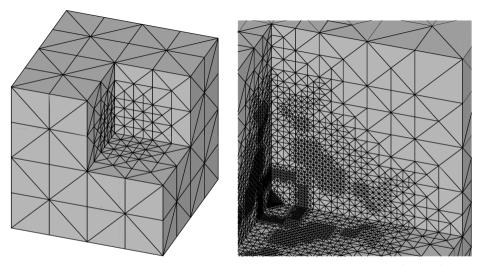


FIG. 5.14. Adaptive grids of the three-dimensional simulation on  $\partial((-1,1)^3\setminus(0,1)^3)$ : full grid of the 2nd iteration (left), zoom into the grid of the 4th iteration (right).