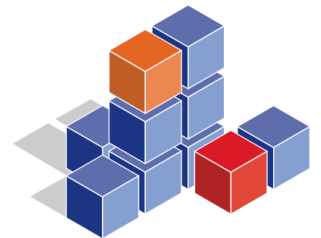

A fast and accurate method for grid deformation

Matthias Grajewski

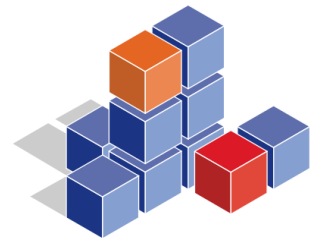
`Matthias.Grajewski@math.uni-dortmund.de`

University of Dortmund



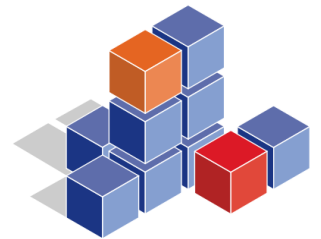
Overview

- Motivation
- Grid deformation: derivation and convergence aspects
- Multilevel deformation
- r-adaptivity



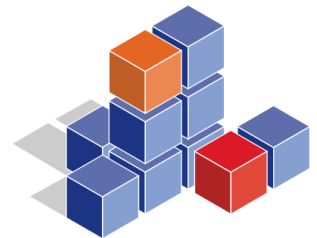
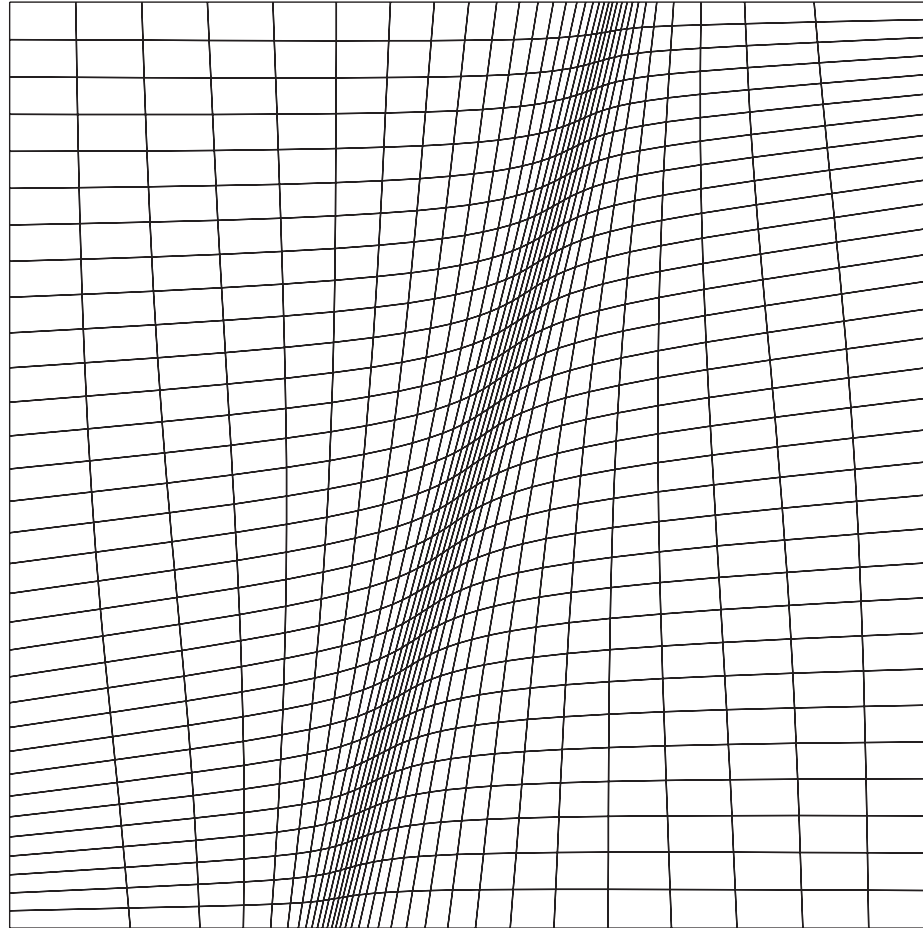
Why Grid Deformation?

main reason: building block for r-adaptivity



Why r -Adaptivity?

1. reason: flexibility

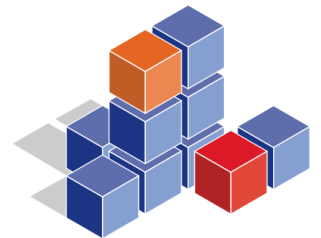


Why r -Adaptivity?

2. reason: **SPEED**

FEM example:

- tensor product mesh
- Q_1 FE, Laplace eq., equidistant grid
- for lexicographical ordering: 9 nonzero bands



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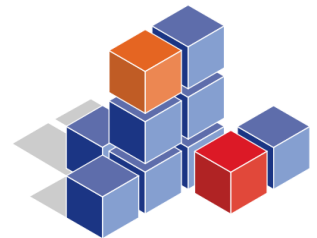
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MV-multiplication in **FEATFLOW** (F77-code):

	N	SP-MOD	SP-STO	SBB-V
AMD Opteron 852	4.225	557	561	1805
2,6 Ghz	66.049	395	223	660
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from: PhD thesis Chr. Becker 2007



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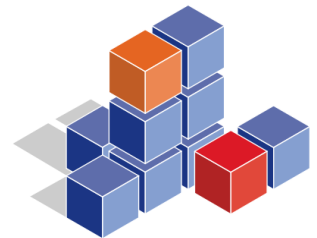
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observation: AFEM: MFlop/s-Rate \ll peak performance

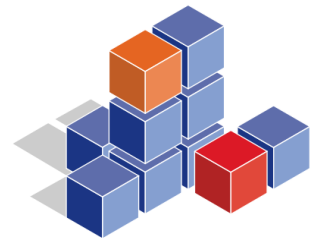


Why r -Adaptivity?

Reference machine: AMD Opteron 852

- peak performance: 4.3 GFlop/s
- peak memory bandwidth: 5.96 GB/s

⇒ peak performance, if ≈ 6 flops per memory access



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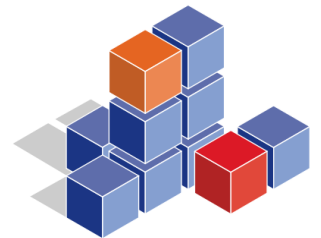
\Rightarrow peak performance, if ≈ 6 flops per memory access

$y = Ax$, N unknowns:

	classical CSR	FEAST (bands)
flops	$17 N$	$17 N$
loads	$27 N$	$18 N$
stores	$1 N$	$1 N$

arithmetic intensity:

- CSR: $17/28 \ll 1 \Rightarrow \approx 9\%$ of peak performance
- FEAST: $17/19 \leq 1 \Rightarrow \approx 15\%$ of peak performance



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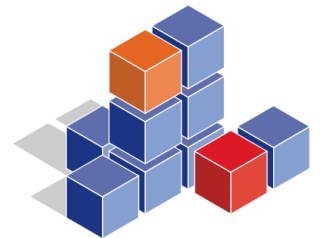
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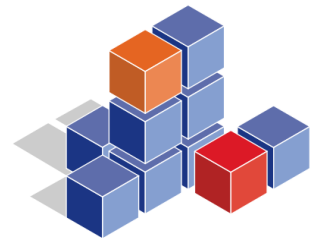
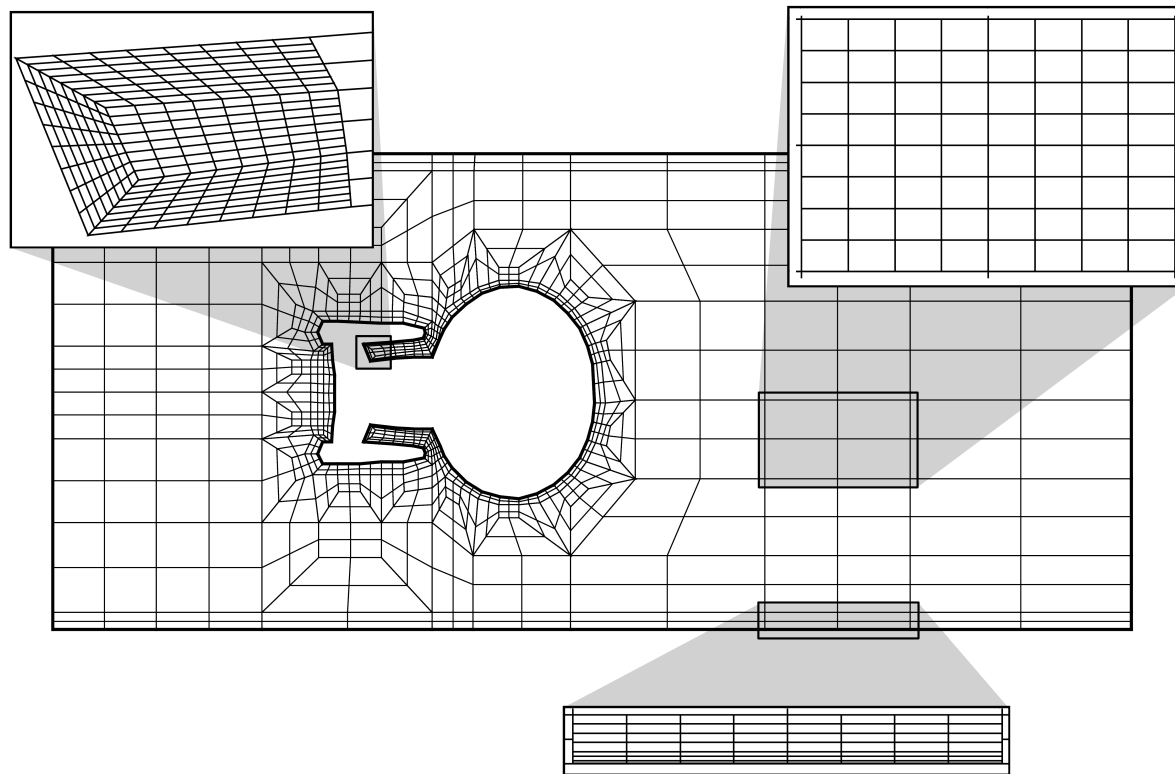
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Avoid unstructured meshes!



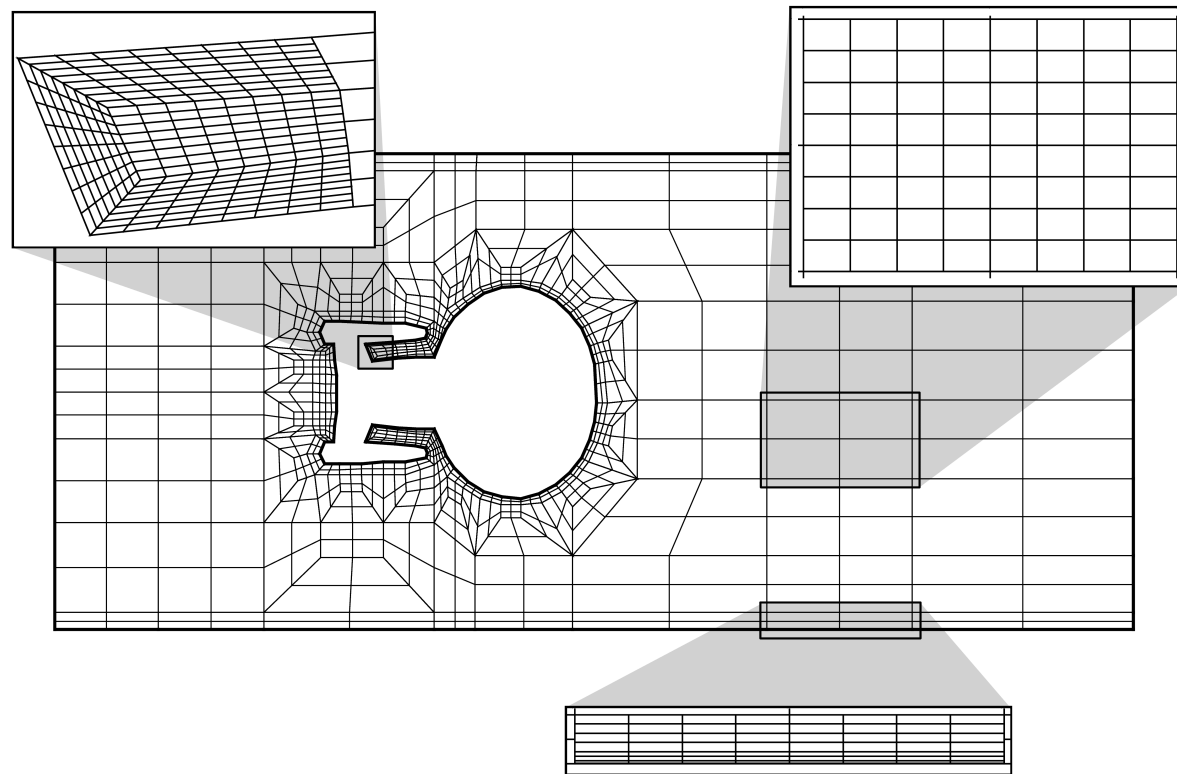
FEAST-Concept (Grid-Related Part)

global grid: “many” local generalised tensor product meshes (“macros”).

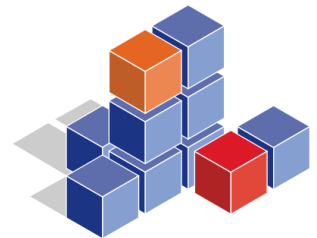


FEAST-Concept (Grid-Related Part)

global grid: “many” local generalised tensor product meshes (“macros”).



r-adaptivity



Derivation

- domain Ω
- triangulation \mathcal{T} , quads T
- “monitor function” $0 < \varepsilon < f \in \mathcal{C}^1(\bar{\Omega})$: **desired area distribution**
- “weighting function” $0 < \varepsilon < g \in \mathcal{C}^1(\bar{\Omega})$: **current area distribution**

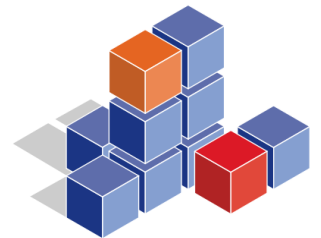
goal: transformation $\Phi : \Omega \rightarrow \Omega$ with

$$g(x)|J\Phi(x)| = f(\Phi(x)) \quad \forall x \in \Omega$$

and

$$\Phi : \partial\Omega \rightarrow \partial\Omega.$$

$$\mathcal{T}^d = \Phi(\mathcal{T}), \quad X := \Phi(x)$$



Derivation

$$m(\Phi(T)) := \int_{\Phi(T)} 1 \, dx = \int_T |J\Phi(x)| dx,$$

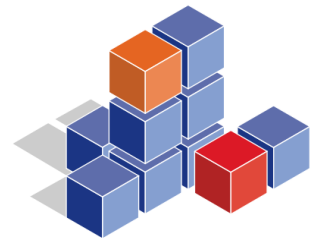


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$$m(\Phi(T)) := \int_{\Phi(T)} 1 \, dx = \int_T |J\Phi(x)| dx,$$

1×1 Gauss-formula:

$$g(x_c) \frac{m(\Phi(T))}{m(T)} = f(\Phi(x_c)) + \mathcal{O}(h).$$



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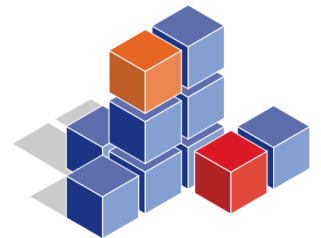
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If

$$g(x) = c(h) m(T) + \mathcal{O}(h), x \in T$$

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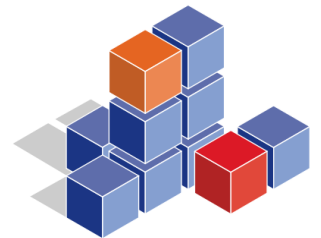
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Grid Deformation Method

Deformation(f, \mathcal{T})

compute $\tilde{f} - \tilde{g}$, $\tilde{f} := c/f, \tilde{g} = C/g, \int \tilde{f} \stackrel{!}{=} \int \tilde{g}$

solve

$$-\operatorname{div}(v(x)) = \tilde{f}(x) - \tilde{g}(x), \quad x \in \Omega, \quad v(x) \cdot \mathbf{n} = 0, \quad x \in \partial\Omega$$

DO FORALL $x \in \mathcal{T}$

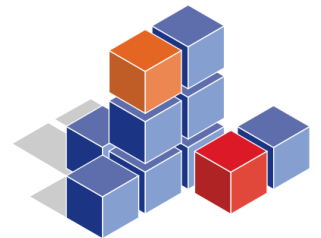
solve

$$\partial_t \varphi(x, t) = \frac{v(\varphi(x, t), t)}{t\tilde{f}(\varphi(x, t)) + (1-t)\tilde{g}(\varphi(x, t))}, \quad 0 \leq t \leq 1, \quad \varphi(x, 0) = x$$

$$\Phi(x) := \varphi(x, 1)$$

ENDDO

END Deformation



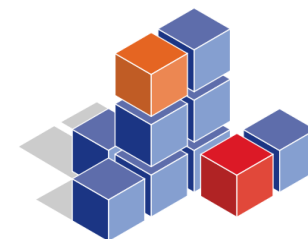
Theoretical Results

Theorem(Moser) Let $0 \leq k \in \mathbb{N}$, $\alpha > 0$. Let $\Omega \subset \mathbb{R}^n$ a domain with $\mathcal{C}^{3+k,\alpha}$ -smooth boundary. Let $f, g \in \mathcal{C}^{k,\alpha}(\bar{\Omega})$ with $\int_{\Omega} f = \int_{\Omega} g$. Then there is a $\mathcal{C}^{k+1,\alpha}$ -diffeomorphism $\Phi : \bar{\Omega} \rightarrow \mathbb{R}^n$ with

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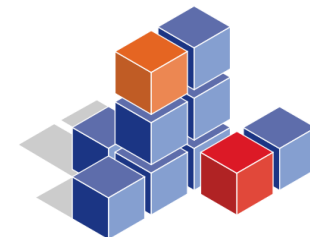
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Theorem Let be Ω as above. If $\Phi : \Omega \rightarrow \Omega$ exists, it fulfills the aforementioned conditions.

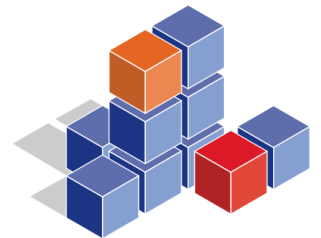


Measuring the Error and Convergence

situation: Let $(\mathcal{T}_i)_{i \in I}$, $N_i < N_{i+1}$ with

$$h_i := \max_{e \in \mathcal{E}_i} |e| = \mathcal{O}(N_i^{-0.5}) \quad \forall i \in I \text{ (edge-length regularity)}$$

$$\exists 0 < c, C : ch_i^2 \leq m(T) \leq Ch_i^2 \quad \forall T \in \mathcal{T}_i \quad \forall i \in I \text{ (size regularity)}$$



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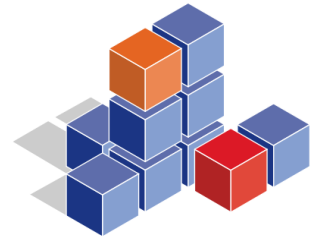
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similarity condition: $\exists 0 < g_{\min} < g < g_{\max} < \infty$ with

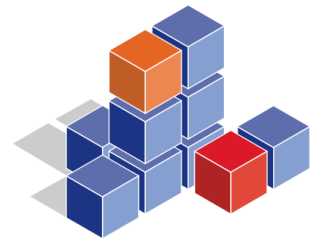
$$\frac{1}{h_i^2} c_i m(T) = g(x) + \mathcal{O}(h_i) \quad \forall x \in T \quad \forall T \in \mathcal{T}_i \forall i \in I, \quad c_s \leq c_i \leq C_s.$$



Measuring the Error and Convergence

1. approach: comparison with “reference deformation”:

$$||\Phi_h - \Phi|| \rightarrow 0$$



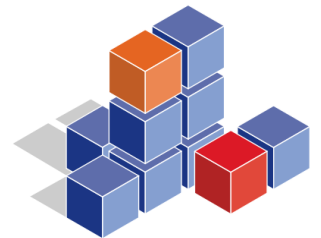
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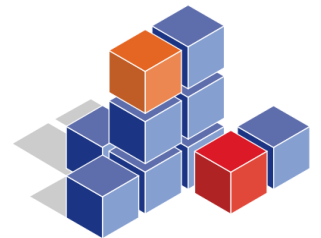
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$$Q_0 := ||q||_{L^2(\Omega)}, \quad Q_\infty := ||q||_{L^\infty(\Omega)}$$



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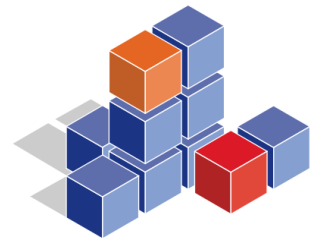
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convergence $:\Leftrightarrow Q_0 \rightarrow 0, Q_\infty \rightarrow 0, h \rightarrow 0$



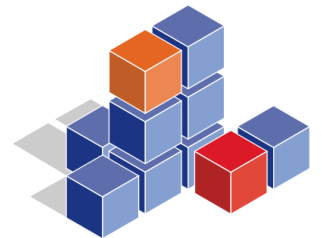
Convergence Theorem

Let $(\mathcal{T}_i)_{i \in I}$ be edge-length regular and fulfill the similarity condition, $0 < \varepsilon < f \in \mathcal{C}^1(\bar{\Omega})$. Furthermore,
 $\|\nabla w - G_h w_h\|_\infty = \mathcal{O}(h^{1+\delta})$, $\delta > 0$ and $\|X_h - \tilde{X}\| = \mathcal{O}(h^{1+\delta})$.

Then:

- a) $(\tilde{\mathcal{T}}_i)_{i \in I}$ is edge-length regular.
- b) $(\tilde{\mathcal{T}}_i)_{i \in I}$ is size regular.
- c) $\exists c > 0$:

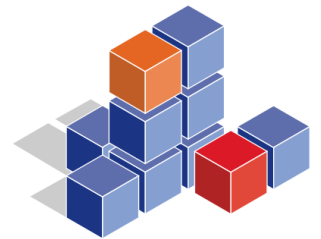
$$Q_0 \leq ch^{\min\{1,\delta\}}, \quad Q_\infty \leq ch^{\min\{1,\delta\}}.$$



Test Problem

$\Omega = [0, 1]^2$, tensor product mesh

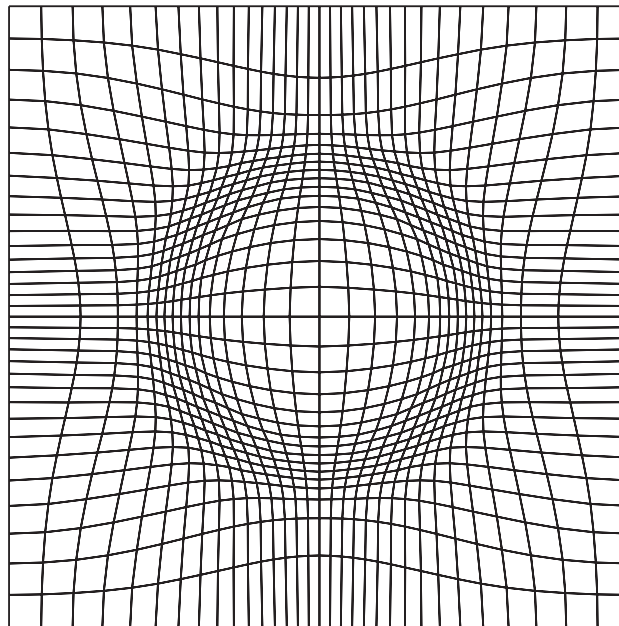
$$f(x) = \min \left\{ 1, \max \left\{ \frac{|d - 0.25|}{0.25}, \varepsilon \right\} \right\}, \quad d := \sqrt{\left(x_1 - \frac{1}{2}\right)^2 + \left(x_2 - \frac{1}{2}\right)^2}$$



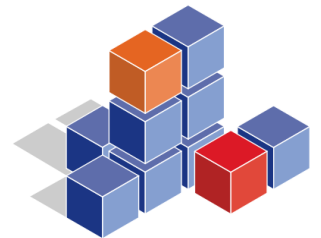
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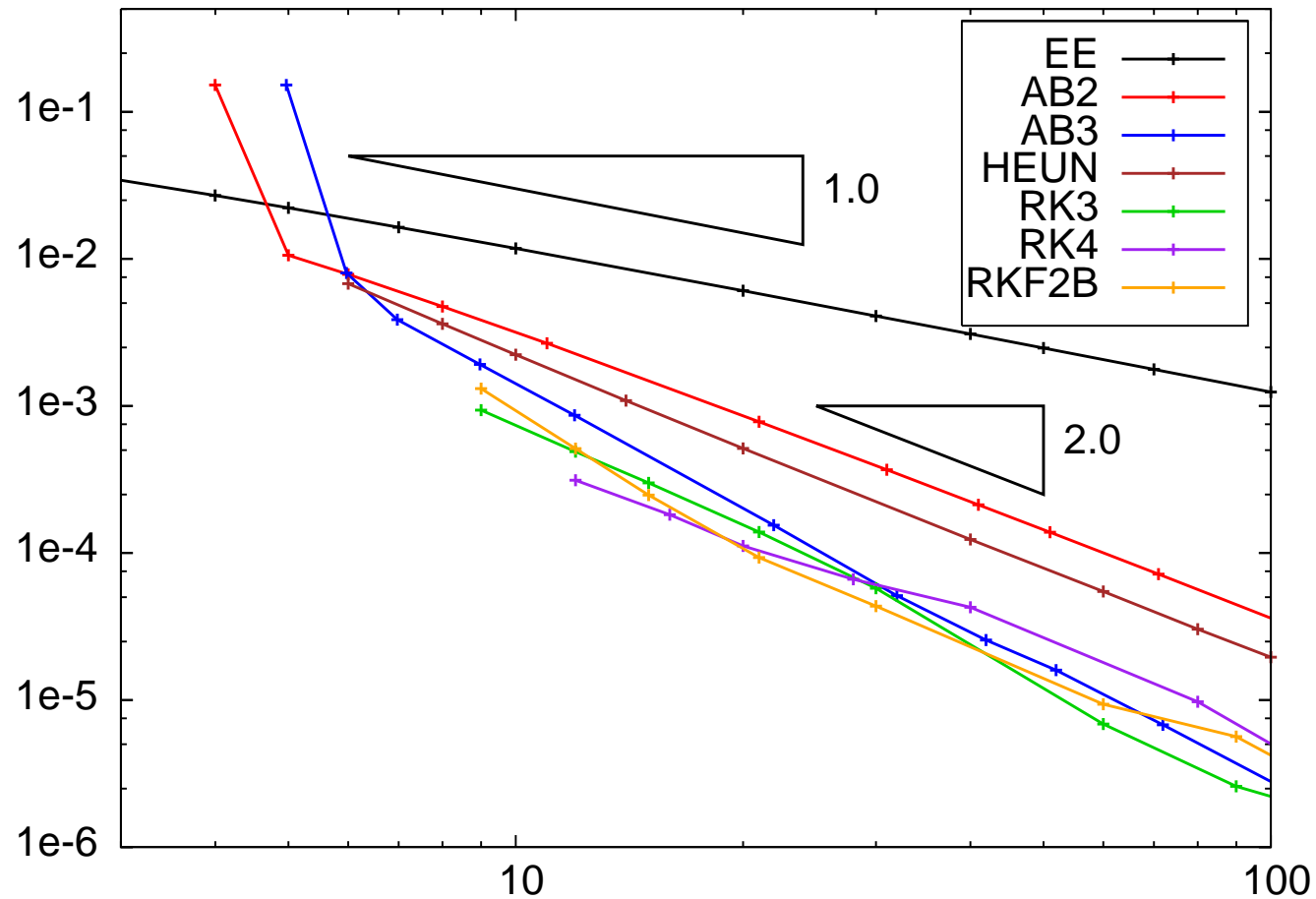
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$$\varepsilon = 0.1$$

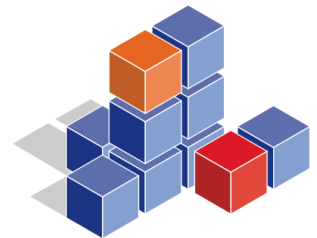


Convergence of ODE-Solvers



$NEL = 65536$

ODE-error: $\mathcal{O}(\Delta t^2)$

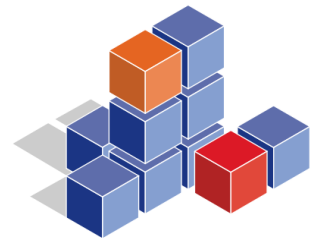


convergence for the Test Problem

Corollary Let us assume that

$$\|\nabla w - G_h w_h\|_{L^\infty} = \mathcal{O}(h^2), \quad \Delta t = \mathcal{O}(h).$$

$$\Rightarrow \quad Q_0 = \mathcal{O}(h), \quad Q_\infty = \mathcal{O}(h)$$

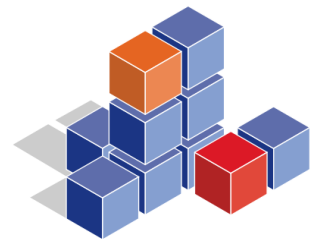
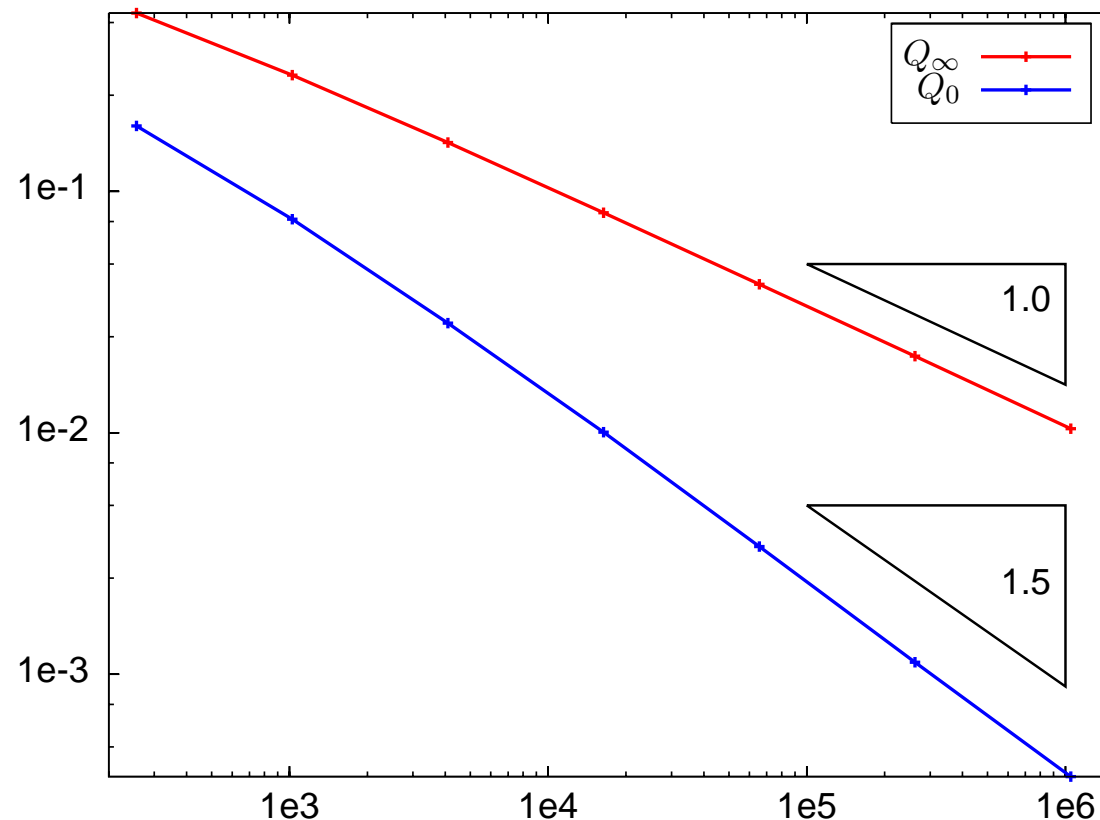


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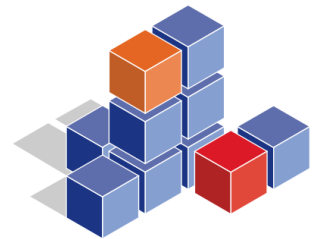
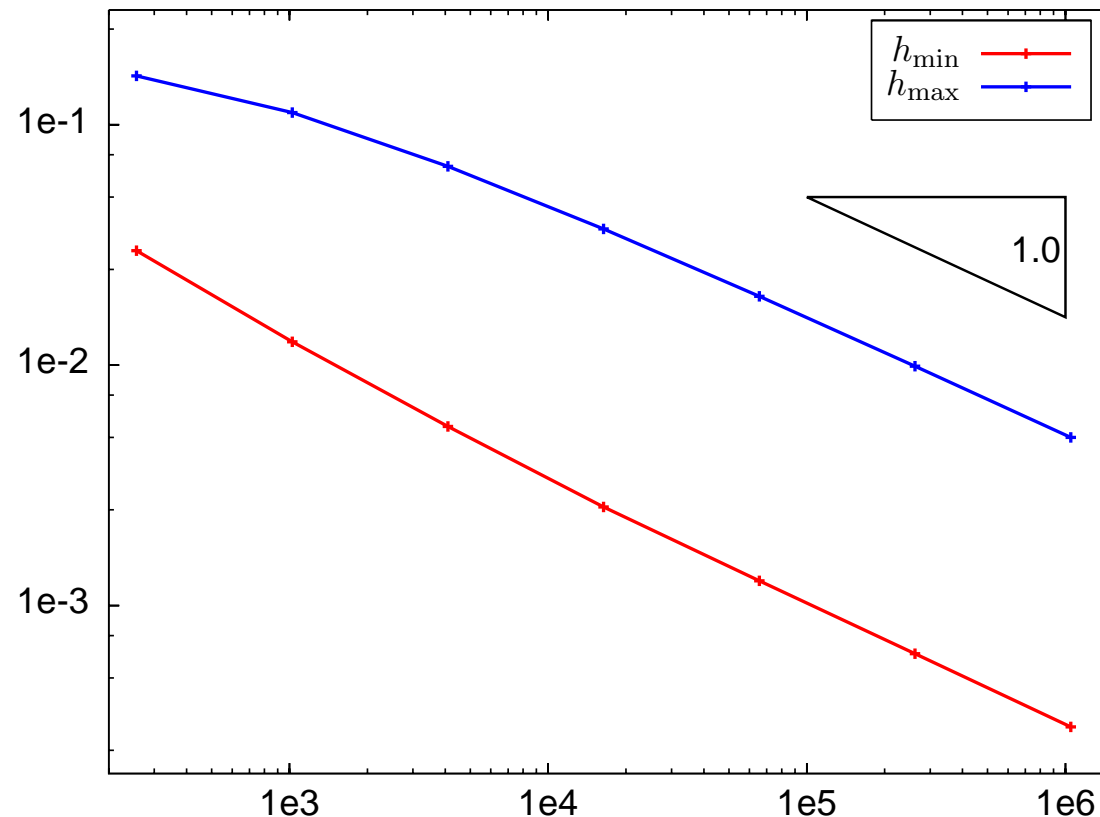


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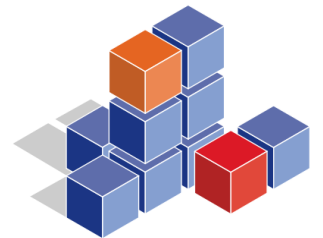
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Runtime

convergence: time step size $\Delta t = \mathcal{O}(h) = \mathcal{O}(N^{-1/2})$

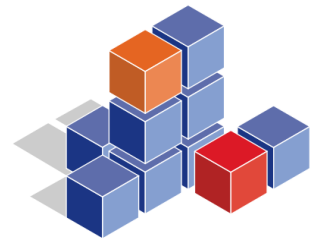
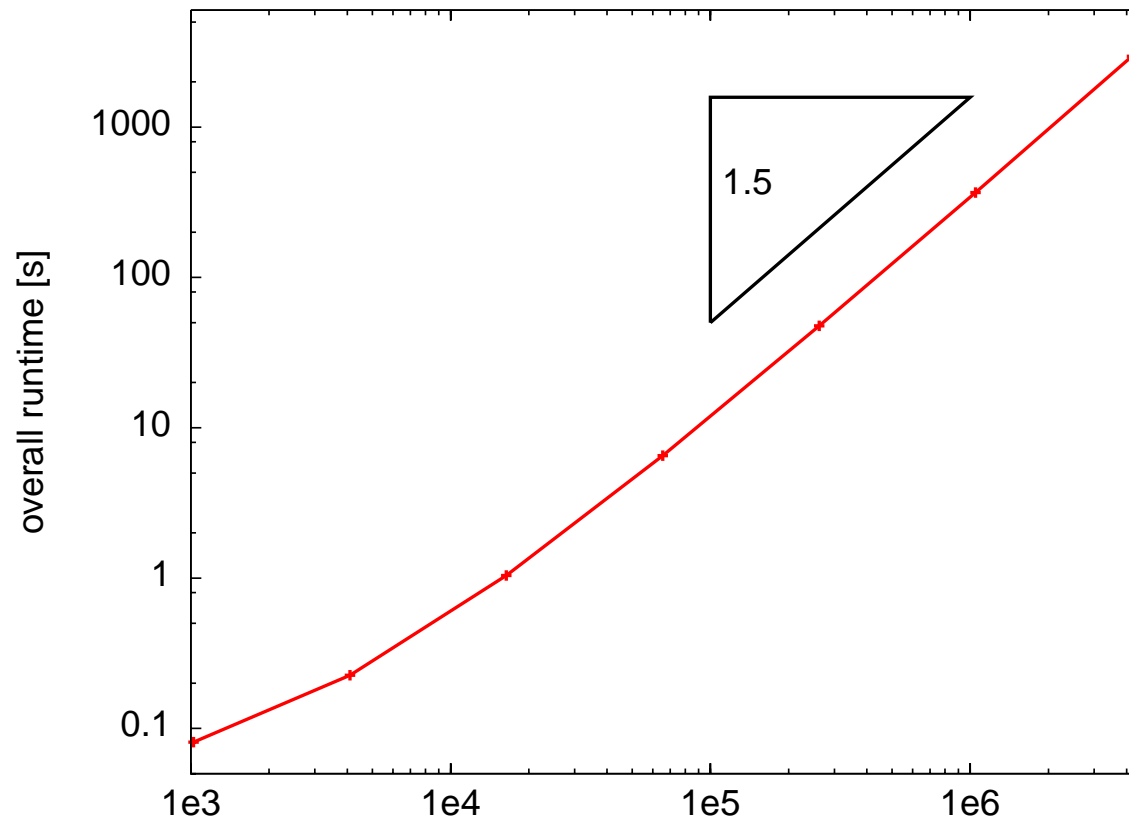
complexity: $\mathcal{O}(N^{3/2})$



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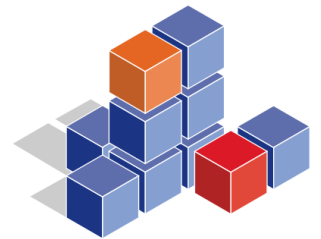
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Multilevel Deformation

goal: fixed time step size + convergence

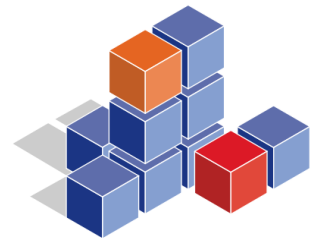
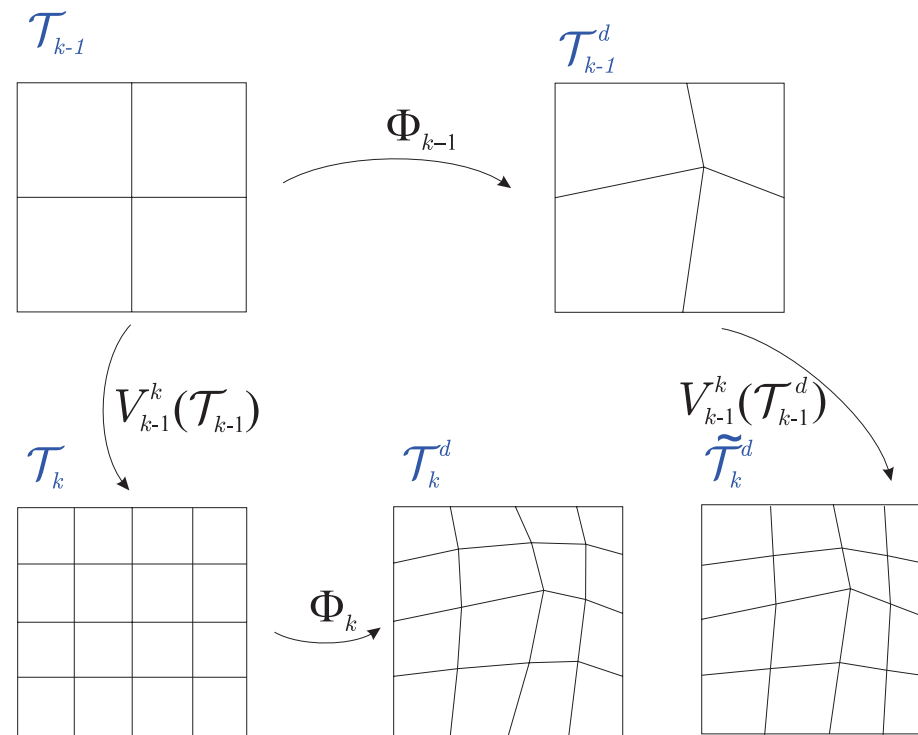
in practical computations: sequence of grids by successive regular refinement



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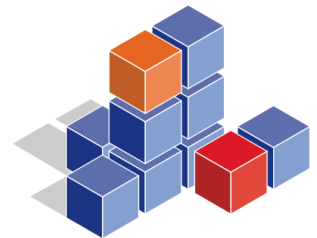
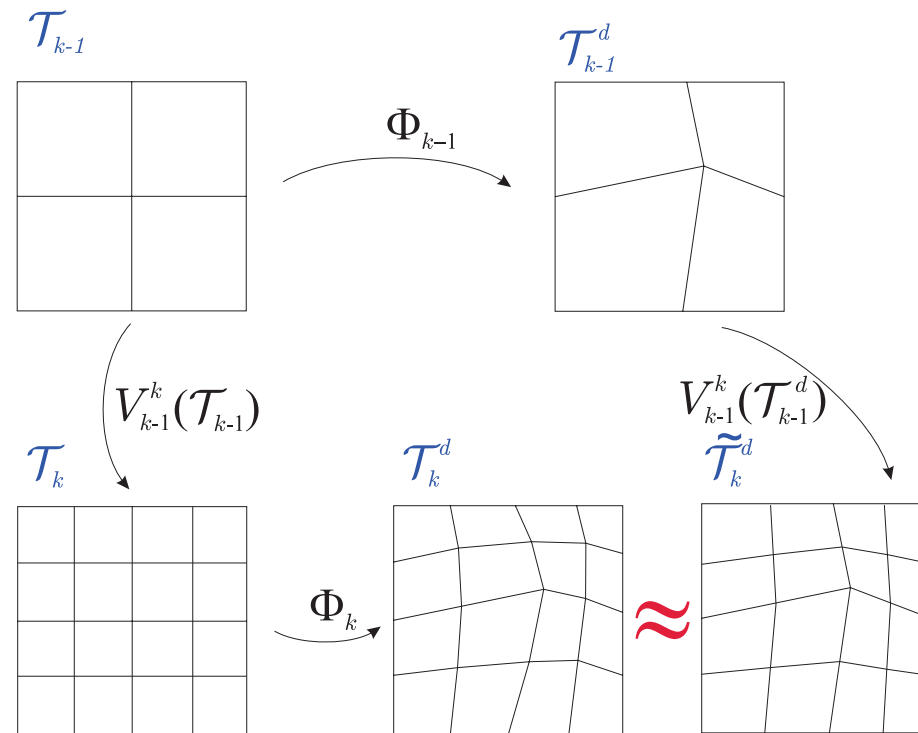
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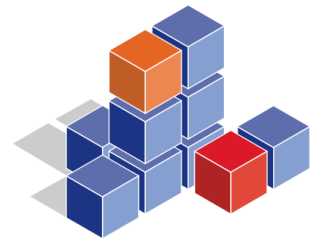
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Multilevel Deformation

idea:

- deformation on coarse grid
- regular refinement
- deformation on fine grid (correction step)



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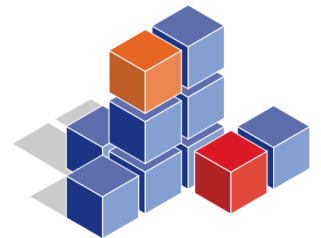
- deformation on coarse grid
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assumption 1:

$$d_k := \max_{x \in \mathcal{X}_k} \|x - \Phi(x)\| \stackrel{?!}{=} \mathcal{O}(h^2)$$

assumption 2:

$$\frac{\|X_h - \tilde{X}\|}{\|x - \Phi(x)\|} \leq c$$



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- regular refinement
- deformation on fine grid (correction step)

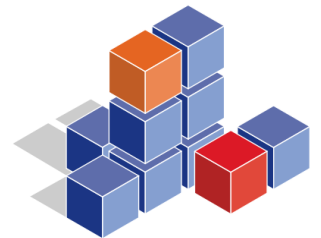
assumption 1:

$$d_k := \max_{x \in \mathcal{X}_k} \|x - \Phi(x)\| \stackrel{?!}{=} \mathcal{O}(h^2)$$

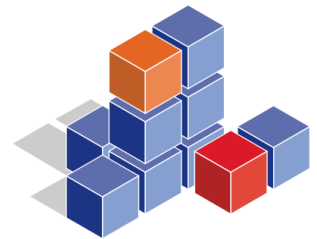
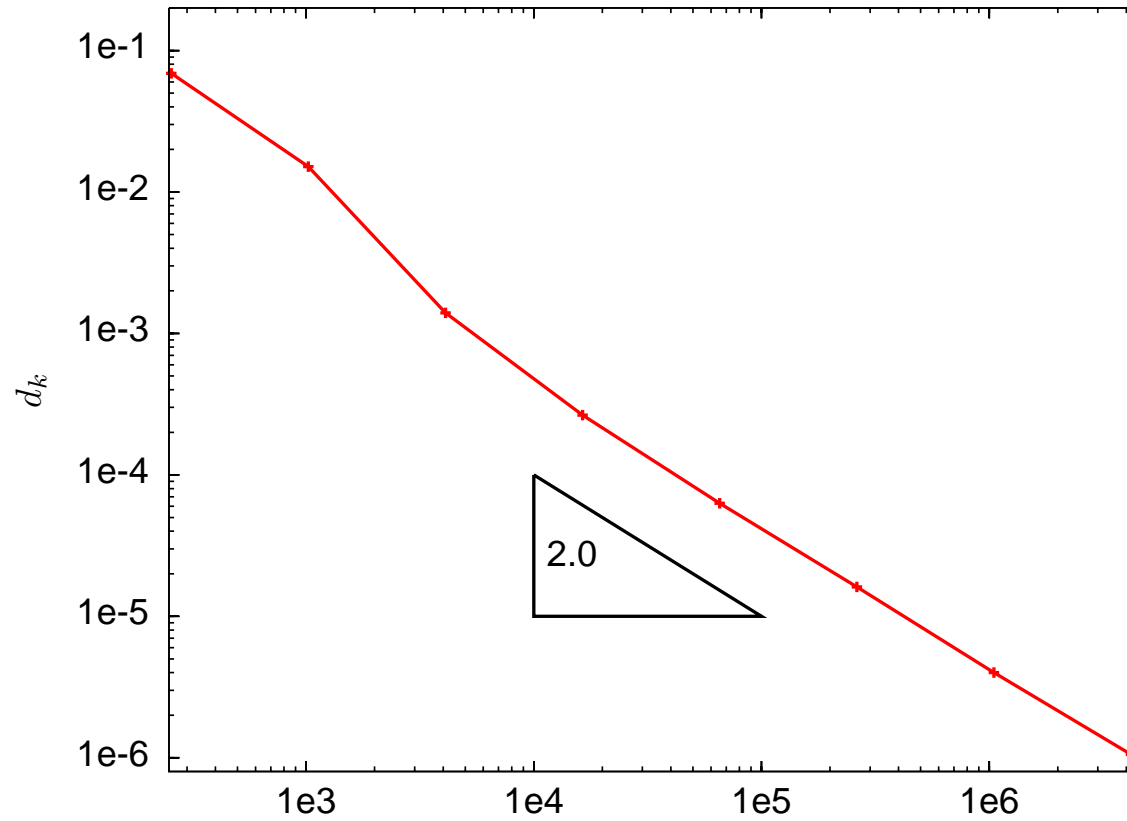
assumption 2:

$$\frac{\|X_h - \tilde{X}\|}{\|x - \Phi(x)\|} \leq c$$

$$\Rightarrow \|X_h - \tilde{X}\| = \mathcal{O}(h^2)$$



Multilevel Deformation



Multilevel Deformation (Algorithm)

MultilevelDef($f, \mathcal{T}, N_{\text{pre}}$) : \mathcal{T}

$\mathcal{T}_{i_{\min}} := R(\mathcal{T}, i_{\min})$

DO $i = i_{\min}, i_{\max}, i_{\text{incr}}$

$\mathcal{T}_i := \text{PreSmooth}(\mathcal{T}_i, N_{\text{pre}}(i))$

$\mathcal{T}_i := \text{Deformation}(f, \mathcal{T}_i)$

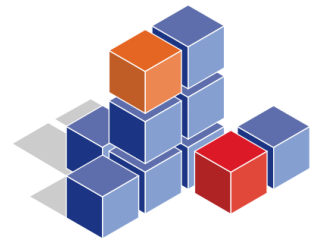
IF ($i < i_{\max}$) $\mathcal{T}_{i+1} := V(\mathcal{T}_i)$

ENDDO

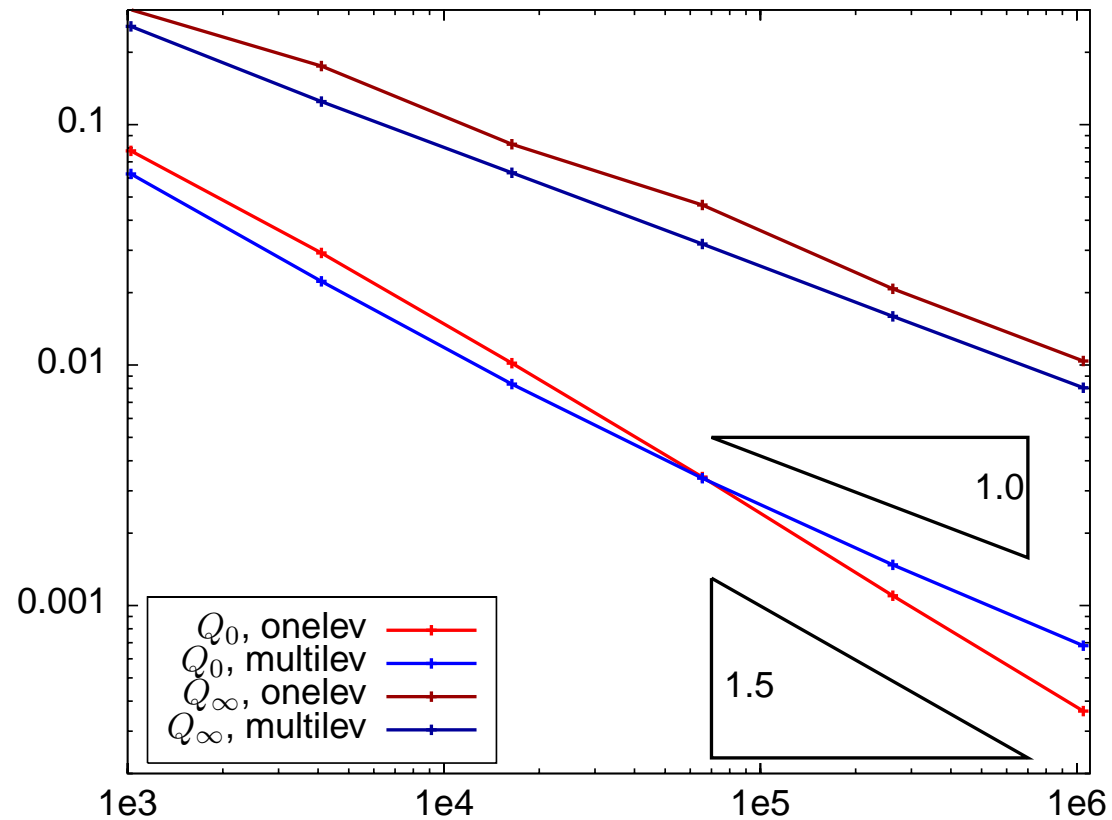
$\mathcal{T} := \mathcal{T}_{i_{\max}}$

RETURN \mathcal{T}

END MultilevelDef

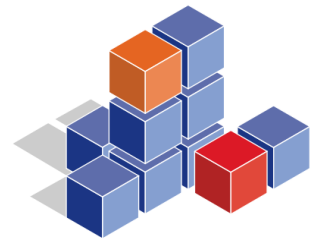


Multilevel Deformation

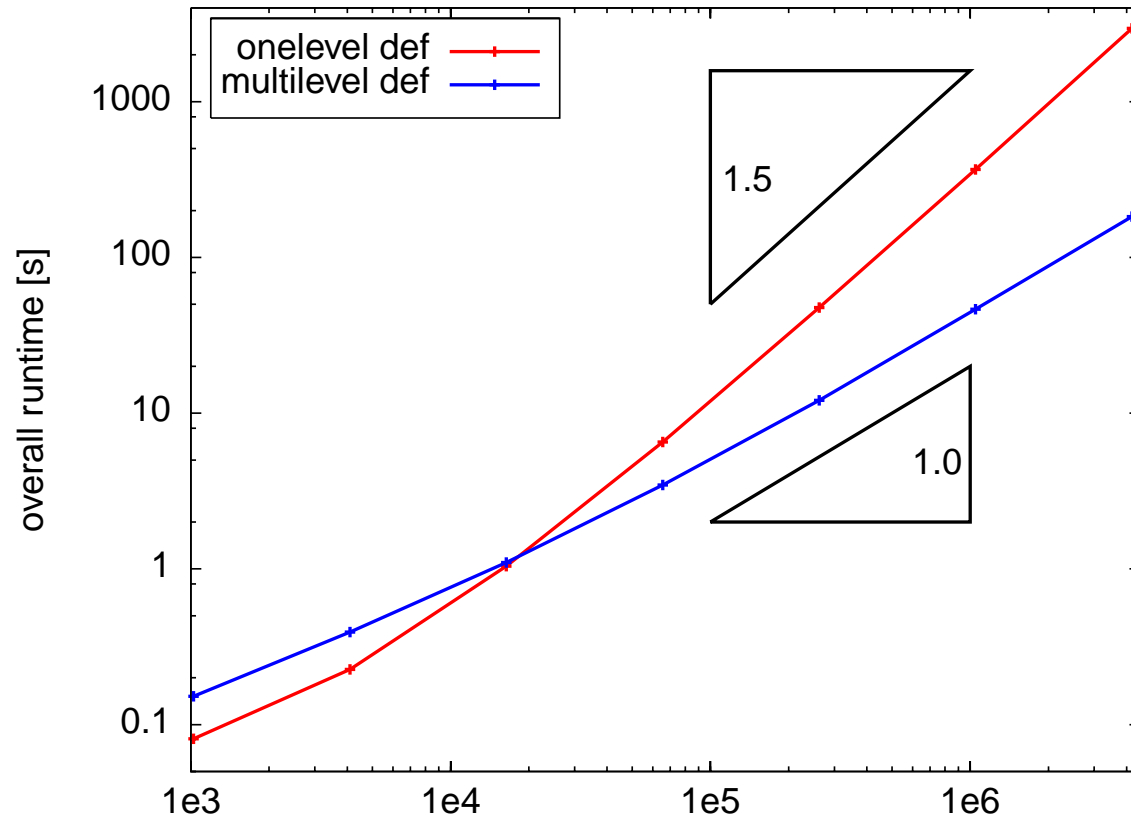


convergence despite of fixed time step size

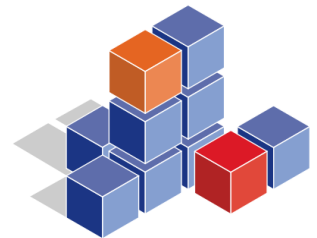
$$i_{\min} = 3, i_{\text{incr}} = 1, N_{\text{Pre}} = 2$$



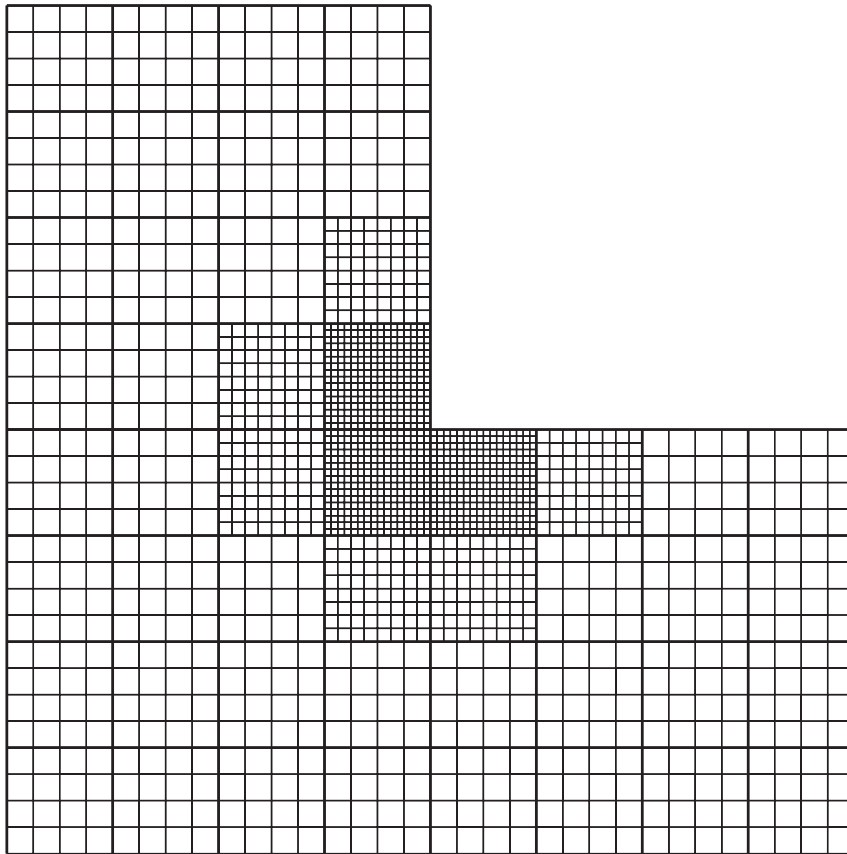
Runtime Comparison



almost optimal complexity



Test Problem: L-domain



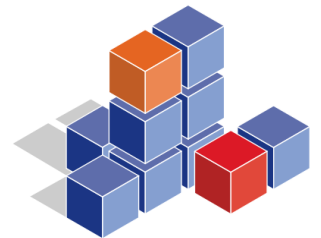
Poisson equation

$$\Omega = [-0.5, 0.5]^2 / [0, 0.5]^2$$

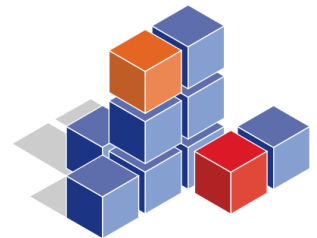
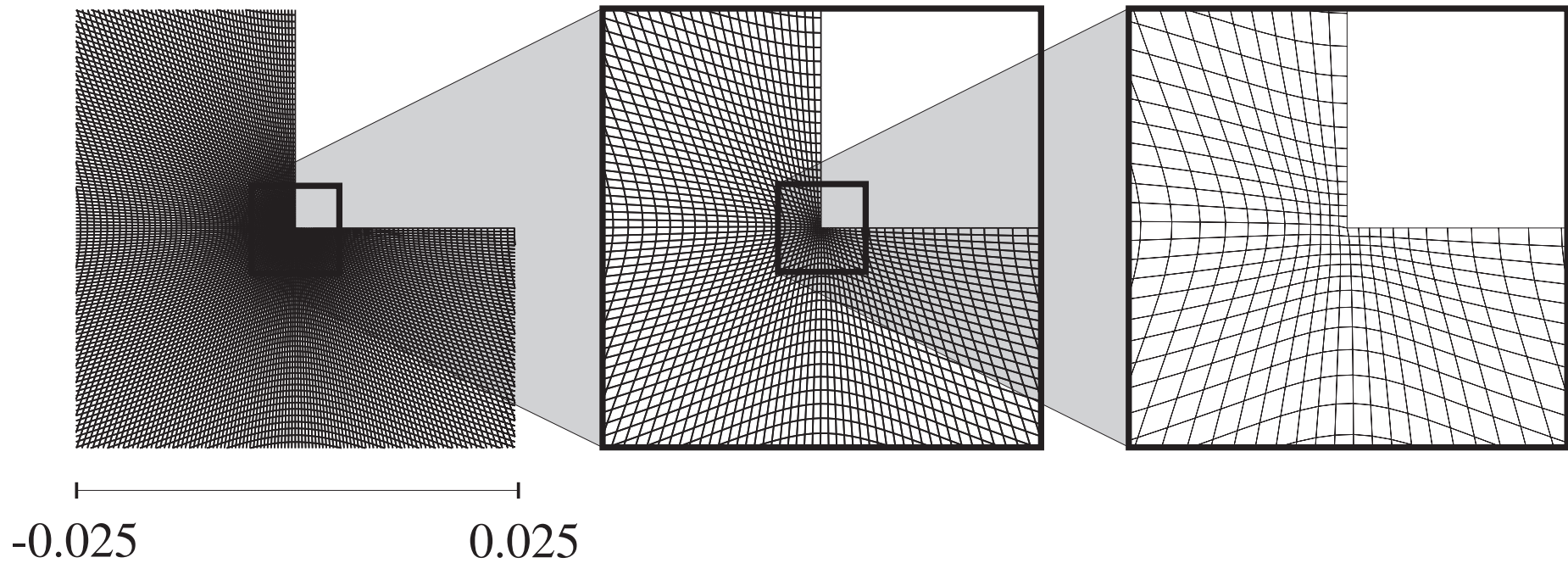
$$u(r, \varphi) = r^{2/3} \sin(2/3\varphi)$$

$$f(r) = \min \left\{ 1, \max \{ c_0 h, \sqrt{2} |r| \} \right\}$$

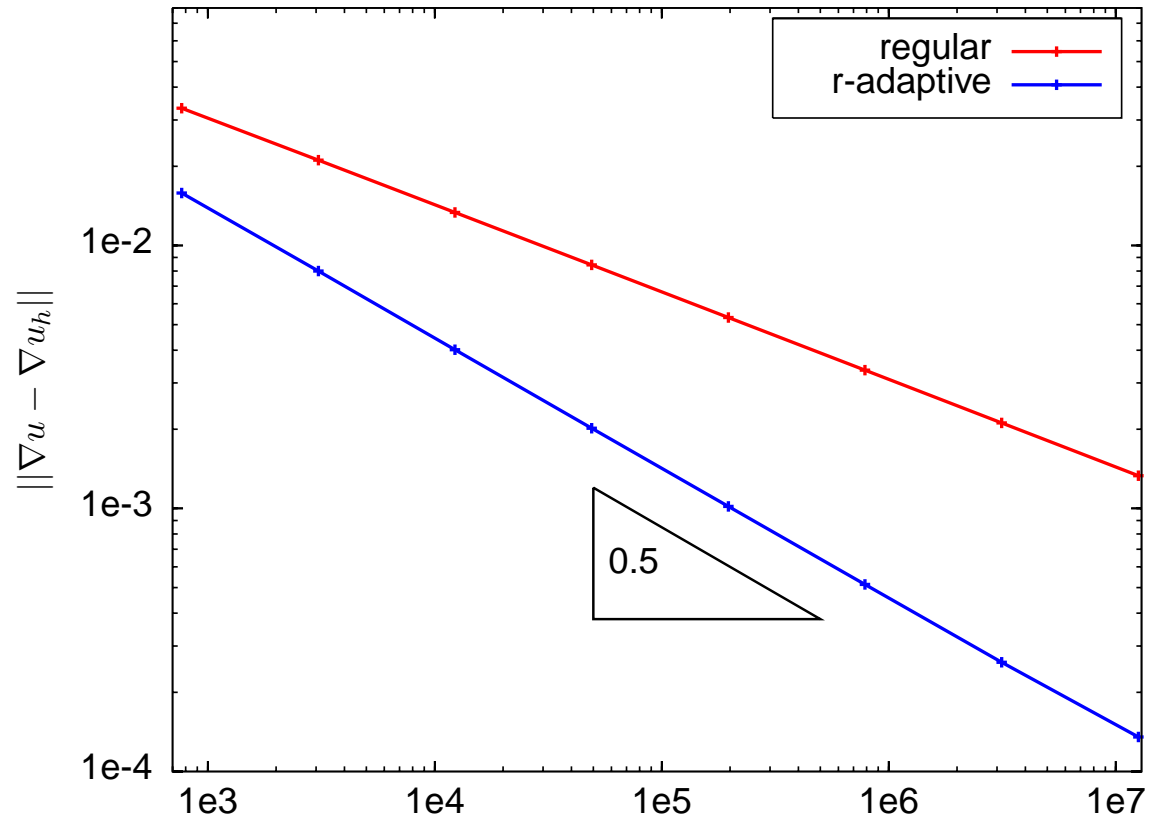
desired: gradient error



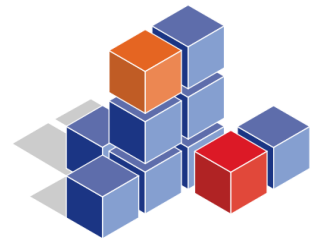
Grid at Reentrant Corner



Discretisation Error

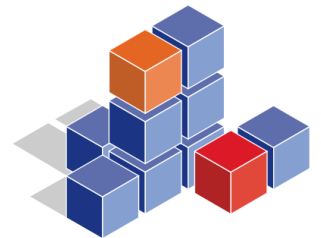


optimal convergence rate by deformed grids



Conclusion

- HPC: locally structured mesh
- Deformation method: derivation and convergence aspects
- Multilevel deformation
- L-domain: r-adaptivity



Thank you for your attention!

