Concepts of patchwise mesh refinement in the context of DWR

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- Adaptivity meets HPC
- Reliability aspects of DWR
- Conclusion and discussion





observation 1: Goal-oriented adaptivity and local mesh refinement is mandatory for accurate and efficient computation.



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observation 2: The MFlop/s rates in adaptive FEM calculations today do by far not reach the peak performance.

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

Computer	#Unknowns	CM TL		STO]
	8,320	147	136	116]
DEC 21264	33,280	125	105	100	
(667 MHz)	133,120	81	71	58	
'EV67'	532,480	60	51	21	
\sim 1300 MFlop/s	2,129,920	58	47	13	
	8,519,680	58	45	10	

reasons for observation 2:

- 1. on modern computers: CPU speed \gg memory speed
- 2. current FEM-codes use
 - indirect addressing : many (unaligned) memory accesses
 - global data structures : effective caching prevented



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tensor product grids **allow direct adressing** and (seem to) **prevent adaptivity**.

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synthesis: FEAST

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example:

The global grid consists of "many" local generalized tensor product meshes ("macros"). example:



goal: adapted grids with local generalized tensor product structure

approach 1: macro wise adaptive refinement

- hanging nodes on macro edges are allowed
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MFLOP-rate preserving adaptivity II





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MFLOP-rate preserving adaptivity III

approach 2: grid deformation





MFLOP-rate preserving adaptivity III

approach 2: grid deformation

example



Grid deformation preserves the (local) logical structure of the grid.



Methods for r-adaptivity

1. location based methods:

- Winslow's method
- Brackbill's and Saltzman's method
- harmonic mapping

disadvantages:

(a) non-linear problems (demanding)

(b) interaction of monitor function and grid not clear

2. velocity based methods:

- MMPDE/GCL (Cao, Huang, Russell)
- Deformation method (Liao et al.)

advantages:

- (a) (several) Laplace problems on fixed mesh (fast)
- (b) monitor function "directly" from error distribution
- (c) mesh tangling prevented



Deformation method (Moser/Liao)

idea : construct transformation $\phi, x = \phi(\xi, t)$ with $\det \nabla \phi = f \Rightarrow$ local mesh area $\approx f$

1. compute monitor function $f(x,t) > 0, f \in C^1$ and $\int_{\Omega} f^{-1}(x,t) dx = |\Omega| \quad \forall t \in [0,1]$



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- **2.** solve $(t \in (0, 1])$

$$\Delta v(\xi,t) = -\frac{\partial}{\partial t} \left(\frac{1}{f(\xi,t)} \right), \quad \frac{\partial v}{\partial n} \bigg|_{\partial \Omega} = 0$$



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- **2. solve (** $t \in (0, 1]$ **)**

$$\Delta v(\xi,t) = -\frac{\partial}{\partial t} \left(\frac{1}{f(\xi,t)} \right), \quad \frac{\partial v}{\partial n} \bigg|_{\partial \Omega} = 0$$

3. solve the ODE system

$$\frac{\partial}{\partial t}\phi(\xi,t) = f\left(\phi(\xi,t),t\right)\nabla v\left(\phi(\xi,t),t\right)$$

new grid points: $x_i = \phi(\xi_i, 1)$





- 1. adaptive macrowise refinement
- 2. in the case of localised error sources: grid deformation on (sub)domains

The algorithmic aspects of the grid adaptation process are object of current research.





DWR-method:

$$(\nabla u, \nabla \varphi) = (f, \varphi), \quad (\nabla z, \nabla \varphi) = J(\varphi) \quad \forall \varphi \in V$$



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replace z by an approximation \tilde{z}

$$\begin{aligned} |J(u-u_h)| &\leq \left| \underbrace{(f,\tilde{z}-\varphi_h) - (\nabla u_h, \nabla(\tilde{z}-\varphi_h))}_{:=\eta(u_h)} \right| \\ &+ \left| \underbrace{(\nabla(u-u_h), \nabla(z-\tilde{z}))}_{:=\eta^*(u_h); \text{ usually neglected}} \right| \quad \forall \varphi_h \in V_h \end{aligned}$$

a-posteriori error correction with DWR (case of linear functional):

$$J(u) = J(u_h) + J(u - u_h) = J(u_h) + \eta(u_h) + \eta^*(u_h)$$

definition (corrected value)

$$J^*(u_h) := J(u_h) + \eta(u_h)$$

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Is this neglect admissible?



observation The neglect of η^* does not affect the grid steering process.

Are reliability and efficiency of DWR affected ?

definition of the efficiency index $I_{\rm eff}$:

$$I_{\text{eff}} = \left| \frac{\eta}{J(u) - J(u_h)} \right|$$



Example 1

$$\Omega = [0,1]^2$$
, ansatz space : \mathcal{Q}^1
 $\Delta u = f, \quad u|_{\partial\Omega} = 0$
 $u(x,y) = x(x-1)(1-y)y^2 \sin(x+2y)$
 $J(\varphi) = \varphi(0.5,0.5)$



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NEL	$J(u-u_h)$	$I_{\text{eff}}, \mathcal{Q}_1, intp.$	$I_{ ext{eff}}, \mathcal{Q}_2$	$ J_{ref} - J^*(u_h) $
64	$4.05 \cdot 10^{-4}$	2.14	0.99	$1.27 \cdot 10^{-6}$
256	$1.01 \cdot 10^{-4}$	2.10	1.00	$7.87 \cdot 10^{-8}$
1024	$2.52 \cdot 10^{-5}$	2.09	1.00	$4.90 \cdot 10^{-9}$
4096	$6.31 \cdot 10^{-6}$	2.08	1.00	$3.06 \cdot 10^{-10}$
16384	$1.58 \cdot 10^{-6}$	2.08	1.00	$1.91 \cdot 10^{-11}$
65536	$3.94 \cdot 10^{-7}$	2.08	1.00	$1.20 \cdot 10^{-12}$

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no problems

Example 2, I

harder problem (convection, edge singularities)

$$-\Delta u + (10,0)^{\top} \cdot \nabla u = 10, \quad u|_{\partial\Omega} = 0$$



$$J(\varphi) = \int_{\partial S} \frac{\partial \varphi}{\partial n} \, ds$$

ansatz space : \mathcal{Q}_1



Example 2, II

reference value J_{ref} by "brute force comp.": 8 regular + 5 local ref. around the square, reference calculation with Q_2



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dual problem with Q_1 and biquadratic interpolation, same mesh

NEL	$ J_{\rm ref} - J(u_h) $	$I_{\rm eff}$	$ J_{\rm ref} - J^*(u_h) $
1088	$1.1386 \cdot 10^{-2}$	1.83	9.4884 $\cdot 10^{-3}$
4352	$1.3793 \cdot 10^{-2}$	0.54	$2.1288 \cdot 10^{-2}$
17408	$1.9698 \cdot 10^{-2}$	0.07	$2.1116 \cdot 10^{-2}$
69632	$1.8048 \cdot 10^{-2}$	0.05	$1.7155 \cdot 10^{-2}$
278528	$1.4222 \cdot 10^{-2}$	0.22	$1.2735 \cdot 10^{-2}$
1114112	$1.0395 \cdot 10^{-2}$	0.21	$8.9900 \cdot 10^{-3}$



Example 2, III

dual problem with \mathcal{Q}_1 and biquadratic interpolation, dual mesh one level finer

NEL	$ J_{\rm ref} - J(u_h) $	$I_{ m eff}$	$ J_{\rm ref} - J^*(u_h) $
1088	$1.1386 \cdot 10^{-2}$	3.31	$2.6340 \cdot 10^{-2}$
4352	$1.3793 \cdot 10^{-2}$	0.80	$2.4863 \cdot 10^{-2}$
17408	$1.9698 \cdot 10^{-2}$	0.00	$1.9706 \cdot 10^{-2}$
69632	$1.8048 \cdot 10^{-2}$	0.20	$1.4404 \cdot 10^{-2}$
278528	$1.4222 \cdot 10^{-2}$	0.29	$1.0057 \cdot 10^{-2}$
1114112	$1.0395 \cdot 10^{-2}$	0.34	$6.8202 \cdot 10^{-3}$



Example 2, IV

dual problem with \mathcal{Q}_2 , same mesh

NEL	$ J_{\mathrm{ref}} - J(u_h) $	$I_{ m eff}$	$ J_{\rm ref} - J^*(u_h) $
1088	$1.1386 \cdot 10^{-2}$	2.84	$2.0999 \cdot 10^{-2}$
4352	$1.3793 \cdot 10^{-2}$	0.09	$1.0802 \cdot 10^{-2}$
17408	$1.9698 \cdot 10^{-2}$	0.62	$7.5091 \cdot 10^{-3}$
69632	$1.8048 \cdot 10^{-2}$	0.75	$4.5457 \cdot 10^{-3}$
278528	$1.4222 \cdot 10^{-2}$	0.80	$2.7832 \cdot 10^{-3}$
1114112	$1.0395 \cdot 10^{-2}$	0.83	$1.7166 \cdot 10^{-3}$



Example 2, V

dual problem with Q_3 , same mesh

NEL	$ J_{\rm ref} - J(u_h) $	$I_{\rm eff}$	$ J_{\rm ref} - J^*(u_h) $
1088	$1.1386 \cdot 10^{-2}$	1.60	$6.7949 \cdot 10^{-3}$
4352	$1.3793 \cdot 10^{-2}$	0.73	$3.7562 \cdot 10^{-3}$
17408	$1.9698 \cdot 10^{-2}$	0.89	$2.1758 \cdot 10^{-3}$
69632	$1.8048 \cdot 10^{-2}$	0.93	$1.2972 \cdot 10^{-3}$
278528	$1.4222 \cdot 10^{-2}$	0.95	$7.8469 \cdot 10^{-4}$
1114112	$1.0395 \cdot 10^{-2}$	0.95	$4.7639 \cdot 10^{-4}$



Example 2, V

dual problem with Q_3 , same mesh

NEL	$ J_{\rm ref} - J(u_h) $	$I_{\rm eff}$	$ J_{\rm ref} - J^*(u_h) $
1088	$1.1386 \cdot 10^{-2}$	1.60	$6.7949 \cdot 10^{-3}$
4352	$1.3793 \cdot 10^{-2}$	0.73	$3.7562 \cdot 10^{-3}$
17408	$1.9698 \cdot 10^{-2}$	0.89	$2.1758 \cdot 10^{-3}$
69632	$1.8048 \cdot 10^{-2}$	0.93	$1.2972 \cdot 10^{-3}$
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 \Rightarrow suffi cient approximation of dual problem absolutely necessary



Remedy (?)

 \Rightarrow

idea (Suttmeier): if \tilde{z} is FEM-solution:

$$|\eta^*| = |(\nabla(u - u_h), \nabla(z - \tilde{z}))| \le \eta_p \eta_d,$$

where η_p, η_d are energy error estimators

• reliable error control

$$|J(u-u_h)| \le |\eta| + \eta_p \eta_d$$

reliable error control of a-posteriori corrected values

$$|J(u) - J^*(u_h)| \le \eta_p \eta_d$$



Remedy

example: Laplace-eq., $u(r, \varphi) = r^{2/3} \sin((2\varphi - \pi)/3)$ on L-shaped domain; η_p, η_d : ZZ-estimator, primal and dual problem with Q_1 on same grid

$$J(\varphi) = \int_{S} \varphi(x) \, dx, \quad S = \{(x, y) \in \Omega \mid x = 0\}$$

remark: Galerkin-orthogonality : $\eta = 0 \Rightarrow J(u_h) = J^*(u_h)$



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NEL	$ J_{ref} - J(u_h) $	$I_{ m eff}$
48	$4.02 \cdot 10^{-3}$	3.17
192	$1.62 \cdot 10^{-3}$	3.73
768	$6.53 \cdot 10^{-4}$	4.14
3072	$2.61 \cdot 10^{-4}$	6.48
12288	$1.04 \cdot 10^{-4}$	4.60
49152	$4.15 \cdot 10^{-5}$	7.14
196608	$1.65 \cdot 10^{-5}$	6.81



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reliable



same example, but with adaptive steering of

- primal grid according to η_p , η
- dual grid according to η_d

NEL_p	NEL_d	$ J(u) - J^*(u_h) $	$I_{\rm eff}$
483	846	$8,68 \cdot 10^{-4}$	3.35
1101	2034	$3.85 \cdot 10^{-5}$	31.1
2541	4833	$6.27 \cdot 10^{-5}$	5.36
5829	11502	$7.31 \cdot 10^{-5}$	1.54
13146	27060	$2.30 \cdot 10^{-5}$	1.51
29202	62826	$1.17 \cdot 10^{-5}$	0.96
62184	145032	$4.36 \cdot 10^{-6}$	1.03



Conclusion

- for reliable estimation: Suttmeier's method with 2 different grids: primal and dual grid
- MFLOP-rate preserving grid adaptivity by deformation and macro wise refinement

agenda

- extension of Suttmeier's method to more complex problems
- efficient implementation in FEAST
- efficient incorporation of grid deformation in adaptation process



Thank you for your attention!

