





CENTRE EUROPÉEN DE RECHERCHE ET DE FORMATION AVANCÉE EN CALCUL SCIENTIFIQUE



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Abstract: Multigrid methods play an important role in the numerical approximation of partial differential equations. As long as only a moderate number of processors is used, many alternatives can be used as solver for the coarsest grid. However, when the number of processors increases, then standard coarsening will stop while the problem is still large and the communication overhead for solving the corresponding coarsest grid problem may dominate. In this case, the coarsest grid must be agglomerated to only a subset of the processors. This article studies the use of sparse direct methods for solving the coarsest grid problem as it arises in a multigrid hierarchy. We use as test case a Stokes-type model and solve algebraic saddle point systems with up to $O(10^{11})$ degrees of freedom on a current peta-scale supercomputer. We compare the sparse direct solver with a preconditioned minimal residual iteration and show that the sparse direct method can exhibit better parallel efficiency.

Project TerraNeo:

Multigrid framework : HHG

Scalability issue: agglomeration



(Hierarchical Hybrid Grids)

The framework:

 \Rightarrow Structured refinement of unstruct. tetrahedral meshes \rightarrow Matrix-free, stencil-based kernels

 \Rightarrow Native PMINRES solver on the coarse grid



<u>The MG</u>: All-at-once Uzawa MG method \triangle Potential convergence issue on the coarse grid Depending on the problem size/hardness



based on the multifrontal scheme.

Three phases

Analysis: ordering, scaling,





Proc.	DoFs coarse	Visco.	PMINRES coarse(s)	MUMPS coarse(s)	DOFs: $5.37 \cdot 10^9$ $4.29 \cdot 10^{10}$ $1.21 \cdot 10^{11}$						(a) Strong and weak interactions between clus- ters in the geometric domain. (b) Corresponding blow clusters in the matrix. (clusters in the matrix.						
40	9.22E+04	iso	1.0	0.16								0 -					
		jump	3.1	0.16		Proc.	PMINRES		MUMPS				PMINRES		MUMPS-BLR-SP		
160	6.96E + 05	iso	2.9	2.32			Coarse(s)	Par. Eff.	Coarse(s)	Par. Eff.		Proc.	fine(s)	Coarse(s)	fine(s)	Coarse(s)	-
		jump	21.0	2.32		1 920	3.1	1.00	0.16	1.00		1 920	75.5	3.56	75.8	0.18	
225	1.94E + 06	iso	3.4	11.51		15 360	21.0	0.73	2.32	0.87		15 360	8/10		82.1	1 79	-
		jump	18.3	12.08		43 200	28.3	0.66	12.08	0.76		13 300	887	21.44	85.26	5.0	_
												45 200	00.1	00.01	09.20	0.9	

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