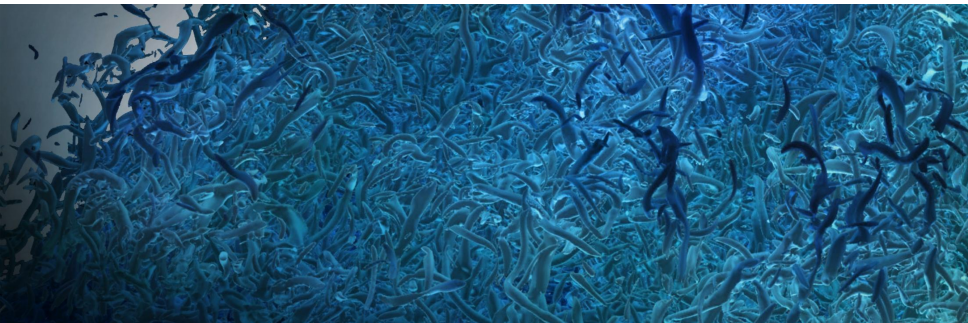


Towards a Parallel Algebraic Multigrid Solver Using PGAS

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The main goal of the EC Horizon 2020 funded ExaFLOW project is to address key algorithmic challenges in CFD to enable simulation at exascale,

- ▶ Accurate **error control**, adaptive mesh refinement
- ▶ Solver **efficiency**, scalable numerical methods and preconditioners
- ▶ Strategies to ensure **fault tolerance** and resilience
- ▶ **Input/output** for extreme data, data reduction
- ▶ **Energy awareness** in solver design

<http://www.exaflow-project.eu>

Why use PGAS languages?

- ▶ One-sided communication
- ▶ Allow for fine grained parallelism
- ▶ More productive languages

A typical two-sided distributed memory program

Data decomposition

while not done **do**

 Compute local part

 Send/Receive overlap

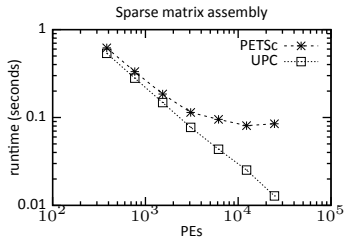
 Add contribution from overlap

end while

One sided comm. allows for novel parallelization algorithms

PGAS based Linear Algebra library

- ▶ Row wise distribution of matrices and column vectors
- ▶ Matrix/vector entries accessible by all
 - ▶ Easier to write solvers/preconditioners
 - ▶ Less synchronization points
- ▶ Implemented in Unified Parallel C
- ▶ Hybrid interface for use with MPI codes
- ▶ Low latency communication kernels
 - ▶ Reduce overhead cf. message passing
 - ▶ Improving fine grained parallelism
 - ▶ Sparse matrix assembly (FEM)
 - ▶ Allowing for less elements/core



A multilevel method for solving $Ax = b$, where A is an $n \times n$ matrix with entries a_{ij} and x, b are vectors of size n .

- ▶ Eliminated smooth errors by solving $Ae = r$, on a coarser problem
- ▶ Interpolate back, and correct the fine solution, $x = x + e$

if $k =$ coarsest level **then**

Solve $A^k x^k = f^k$

else

Relax μ_1 times on $A^k x^k = f^k$

Set $x^{k+1} = 0, f^{k+1} = I_k^{k+1}(f^k - A^k x^k)$

Apply V-cycle on level $k + 1$

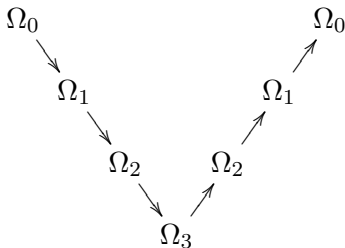
Correct the solution by

$x^k = x^k + I_{k+1}^k x^{k+1}$

Relax μ_2 times on $A^k x^k = f^k$

end if

- ▶ Asymptotically optimal complexity
- ▶ Notorious difficult to parallelize in an optimal way!



Let Ω^k be the set of components of x at level k

- ▶ Split Ω^k into two disjoint sets C and F
- ▶ No underlying geometry
- ▶ Need to pick coefficients in A^{k+1} related to a error freq.
- ▶ Classify if unknowns are strongly coupled to each other

$$-a_{ij} \geq \theta \max_{k \neq i} \{-a_{ik}\}$$

- ▶ Measure λ_i , number of points strongly influenced by i

Ruge-Stüben Coarsening

Related to interpolation quality

C1: For each point j that strongly influences a F -point i , j is either a C -point or it strongly depends on a C -point l that also strongly influences i .

Size of the coarser level

C2: C should be a maximal subset of all points with the property that no two C points are strongly connected to each other.

Ruge-Stüben Coarsening

Let $U = \Omega^k$

/* First phase */

while $U \neq \emptyset$ **do**

 Pick an $i \in U$ with maximal λ_i

 Set $U = U - \{i\}$, $C = C + \{i\}$

 Add all points j which strongly depends on i to F

 Increase the measure λ_l for all points l that are strongly dependent on j

 Decrease the measure λ_m for all points m that are strongly dependent on i

end while

/* Second phase */

for all $i \in F$ **do**

if i violates criteria **C1** **then**

$F = F - \{i\}$, $C = C + \{i\}$

end if

end for

A challenge for message passing is the restricted local view of the data

- ▶ A thread must be able to determine if a neighboring point is C or F
- ▶ Points will change between F and C during coarsening
- ▶ Explicit communication with neighbors
- ▶ Less optimal coarsening algorithms
 - ▶ Easier parallelization
 - ▶ Pay with more multigrid cycles
 - ▶ Set/Graph based (CLJP, PMIS HMIS)
 - ▶ Heuristics (RS3)

Restrictions comes from a two-sided message passing perspective

- ▶ Put C/F data in global memory
- ▶ Accessible by all threads
- ▶ Solves most of the parallelization issues

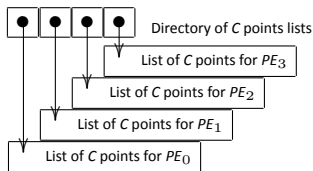
PGAS based RS

Most of the data structures are local

- ▶ Unordered set of C and F points
- ▶ Measure λ_i stored in red-black trees

Keep a list of C variables in global memory

- ▶ PE dependent block size
- ▶ Directory approach (arbitrary size)
- ▶ Protect the list with a set of variables
 - ▶ Declared as UPC `strict`
 - ▶ Less expensive than using locks



PGAS based RS

Let $U = \Omega^k$ and $Cg(:) = 0$

/* First phase */

while $U \neq \emptyset$ **do**

 Pick an $i \in U$ with maximal λ_i

 Set $U = U - \{i\}$, $C = C + \{i\}$ and $Cg(i) = 1$

 Add all points j which strongly depends on i to F

 Increase the measure λ_l for all points l that are strongly dependent on j

 Decrease the measure λ_m for all points m that are strongly dependent on i

end while

Barrier

/* Second phase */

for all $i \in F$ **do**

if i violates criteria **C1** **then**

 Wait **while** $S(i) \neq 0$

$S(i) = 1$ /* Protect variable i */

$F = F - \{i\}$, $C = C + \{i\}$ and $Cg(i) = 1$

$S(i) = 0$ /* Release variable i */

end if

end for

Load Balancing

- ▶ A^{k+1} becomes smaller and smaller for each coarsening level k
- ▶ Move operator towards a single core (easier coarse level solve)
- ▶ Use a load balanced linear distribution $N = PL + R$,

$$L = \left\lfloor \frac{N}{P} \right\rfloor,$$

$$R = N \bmod P,$$

$$n = \left\lfloor \frac{N + P - p - 1}{P} \right\rfloor$$

$$\begin{pmatrix} A^{k+1} & & & \\ & A^k & & \\ & & & \\ & & & A^0 \end{pmatrix} \begin{matrix} PE_0 \\ PE_1 \\ PE_2 \\ PE_3 \end{matrix}$$

- ▶ Multigrid Cycling
 - ▶ Matrix vector products
 - ▶ Redistribution routines (load balancing)
- ▶ Coarse Level Solver
 - ▶ Direct solver (single PE)
 - ▶ Krylov solver (multiple PEs)
- ▶ Smoother
 - ▶ Hybrid *CF* Gauss Seidel

$$x_i^k = \left(b_i - \sum_{j < i} a_{ij} x_j^k - \sum_{j > i} a_{ij} x_j^{k-1} \right) / a_{ii}$$

- ▶ Straightforward implementation (PGAS)
- ▶ Work across PE boundaries

Benchmark Problem

Poisson's equation on the unit square

$$-\Delta u(x, y) = f(x, y), \quad (x, y) \in \Omega,$$

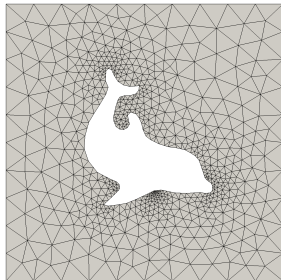
$$u(x, y) = 0, \quad (x, y) \in \Gamma_0,$$

$$\partial_n u(x, y) = g(x, y), \quad (x, y) \in \Gamma_1,$$

$$\partial_n u(x, y) = 0, \quad (x, y) \in \partial\Omega \setminus (\Gamma_0 \cup \Gamma_1),$$

$$f(x, y) = 500 \exp(-((x - 0.5)^2 + (y - 0.5)^2)/0.02)$$

$$g(x, y) = 25 \sin(5\pi y).$$

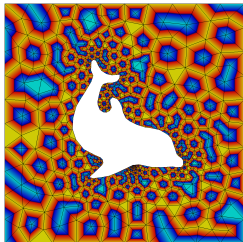


Benchmark Problem

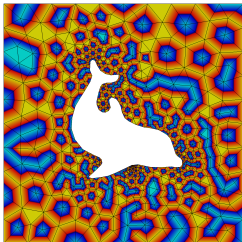
- ▶ Discretize PDE by FEM
- ▶ Use the FEM framework FEniCS
 - ▶ FEniCS assembles the stiffness matrix
 - ▶ Linear system solved by external libraries
- ▶ Hybrid MPI + PGAS
- ▶ Use PETSc as a reference krylov solver
- ▶ All experiments performed on the Cray XC40 Beskow at PDC/KTH

Parallel Coarsening

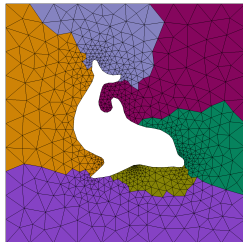
- No artifacts from boundary between partitions



Serial coarsening



Parallel coarsening

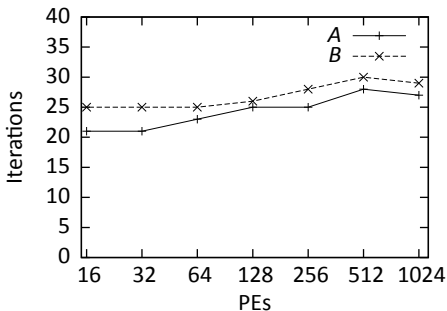


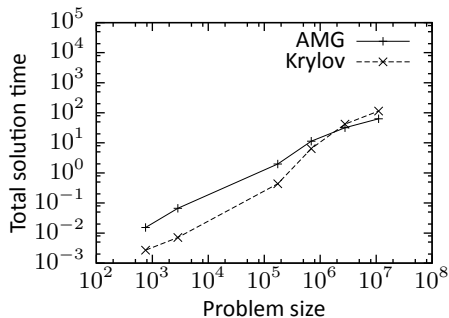
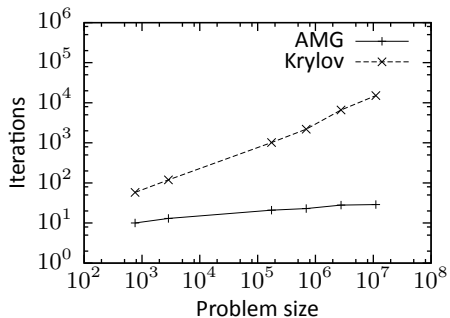
Mesh partitions

Asymptotically optimal complexity

- ▶ Number of V-cycles independent of problem size
- ▶ Number of V-cycles independent of PEs

Benchmark solved for two matrix sizes and diff. numbers of PEs





PEs	n	C_{op}	C_g	l	AMG				Krylov	
					Iters	t_{setup}	t_{solve}	t_{tot}	Iters	t_{tot}
1	759	1.821	1.480	2	10	0.008	0.008	0.015	58	0.003
4	2868	2.137	1.552	4	13	0.030	0.037	0.067	119	0.007
32	174144	2.265	1.529	7	21	0.712	1.265	1.977	1023	0.436
128	693888	2.098	1.495	8	23	3.779	7.635	11.414	2196	6.432
512	2770176	2.065	1.486	9	28	8.190	23.601	31.791	6624	42.117
1024	11069952	2.044	1.480	10	29	11.476	51.787	63.263	15139	114.841

- ▶ AMG overhead costs too high for small matrices
- ▶ AMG setup costs (t_{setup}) less than solve (t_{solve})
- ▶ Operator (C_{op}) and grid complexity (C_g) doesn't grow
- ▶ AMG iterations doesn't grow too much with large numbers of PEs

- ▶ New parallel formulation of Ruge-Stüben
 - ▶ Not possible to formulate using MPI
- ▶ Retains similar properties as the serial algorithm
- ▶ Easier implementation due to the PGAS abstraction

Future work

- ▶ Reduce AMG overhead
- ▶ Optimized collective operations
 - ▶ No subset collectives in std. UPC
 - ▶ Handwritten versions not optimized
- ▶ Investigate UPC atomics
 - ▶ Our Cray compiler didn't support it
 - ▶ Alternative instead of strict variables