Distributed Memory Parallelization in NGSolve

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From Shared to Distributed

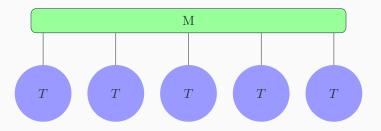
Memory

Shared Memory



Parallelization via **threads** (\rightarrow unit for **scheduling**).

 ${\bf Programming\ model:}$



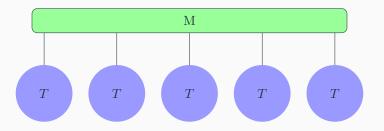
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Shared Memory



Parallelization via threads (\rightarrow unit for scheduling).

 ${\bf Programming\ model:}$



Assumption:

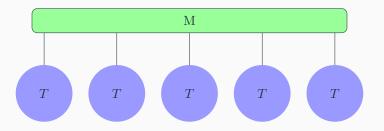
All threads can access all parts of the memory equally fast.

Shared Memory



Parallelization via threads (\rightarrow unit for scheduling).

 ${\bf Programming\ model:}$



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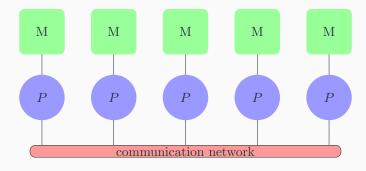
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1

Distributed Memory



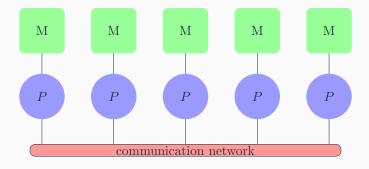
Programming model that better fits a cluster:



Distributed Memory



Programming model that better fits a cluster:



Parallelization via procs (\rightarrow unit for scheduling and memory management).

Outline



From Shared to Distributed Memory

MPI

NGS+MPI

Principles

Under the hood

Using MPI-parallel NGSolve

Solvers and Preconditioners

Experiences

Bonus slide; VTK + Paraview

MPI

Message Passing Interface



A standard for message-passing between independent processes; implemented in multiple libraries.

De facto standard for numerical computations.

It features functions for Point-to-Point communication between procs, by sending and receiving of "messages".

Portability is a huge aspect - write it once, run it on a laptop or a supercomputer!

Also: collective communication, one-sided comunication, ...

Point-to-Point communication



Process identified by rank, a number from 0 to NP.

To exchange message, one has to specify: pointer to data, amount of data, type of data, source/destination, ...

On the sending end:

```
 \begin{split} & \text{Array} < & \text{double} > \text{ send\_values(len);} \\ & \text{MPI\_Send(\&send\_values[0], len, MPLDOUBLE, recv\_proc, } \ldots); \end{split}
```

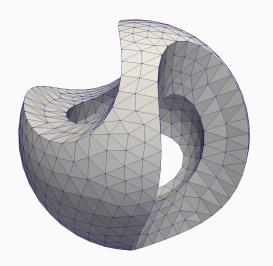
On the receiving end:

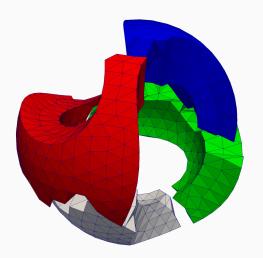
```
 \begin{split} & \text{Array} < & \text{double} > \text{ recv\_values (len)}; \\ & \text{MPI\_Recv(\&recv\_values [0], len, MPI\_DOUBLE, send\_proc, } \ldots); \end{split}
```

NGS+MPI

Distributing the Mesh



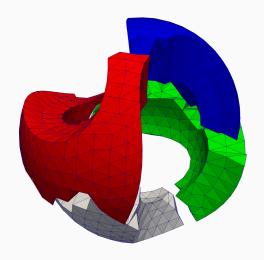




Distributing the Mesh



- non-overlapping partition (no ghost-elements)
- shared interface-nodes (edges in 2d, faces & edges in 3d)
- proc 0 (master proc) gets nothing!



Distributing the FESpace



A finite element consists of a space $V_{h,i}$ and a basis of its dual space $(\phi_{i,j})_j$.

The FESpace is then "glued together":

$$V_h = \left\{ u \in \Pi_i V_{h,i} : \phi(u_i) = \phi(u_k) \quad \forall \phi \in V'_{h,i} \cap V'_{h,j} \right\}$$

 \rightarrow Treat each subdomain as a "makro - Finite Element"!

Subdomain-spaces are glued together by functionals corresponding to nodes on interfaces.

Distributed LinAlg - Matrices and Vectors



Parallel Matrices

- Assemble element-matrices to submatrices A_i on each subdomain.
- The global Matrix $A = \sum_{i} E_i A_i E_i^T$ (with embeddings E_i) is never assembled.
- No communication needed to assemble matrices!

Parallel Vectors

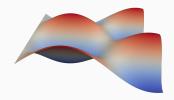
- Local vector on each subdomain
- Can be CUMULATED (consistent values) or DISTRIBUTED (partial values)

$$x_j^{(C)} = E_j^T \sum_i E_i x_i^{(D)}$$

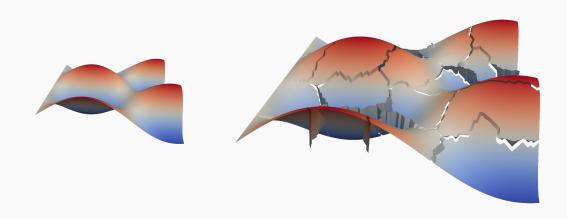
• DISTRIBUTED → CUMULATED requires communication!

Parallel Vectors - Visualized



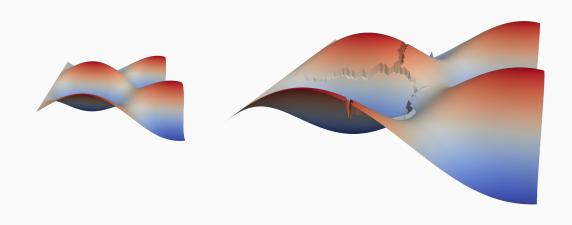






Parallel Vectors - Visualized





Distributed LinAlg - Matrix-Vector-Product



$$(Ax)_{j}^{(C)} = E_{j}^{T} \sum_{i} E_{i} A_{i} x_{i}^{(C)} = E_{j}^{T} \sum_{i} E_{i} (Ax)_{i}^{(D)}$$
$$A_{i} x_{j}^{(C)} = (Ax)_{j}^{(D)}$$

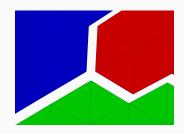
Multiplication with local matrices takes a C-vector and returns a D-vector, to make it a C-vector again, we need communication!



We want to **apply** a (possibly nonlinear) operator involving dg-jumps, e.g:

```
a = BilinearForm(V)
a += SymbolicBFI( (u-u.Other())*(v-v.Other()),
skeleton=True)
a.Apply(u.vec, res)
```

remember: **no ghost elements** \rightarrow what is ".Other()" on a subdomain interface??





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- We know "both sides" have the same expression
- Need trace values in IPs of all trial-proxies "from the other side"!
- "Other side" needs trace values of all trial-proxies from me!
- \rightarrow Exchange trial-proxy traces!



```
Table < double > tv_send; //np rows
Table < double > tv_recv; //np rows
for (facet : mpi_facet)
        CalcTraceValues(facet, tv_send[p][facet_part])
for(p : Range(0, np))  {
        MPI_Isend(tv_send[p],p,...) // send trace values
        MPI_Irecv(tv_recv[p],p,...) // recv trace values
MPI_Waitall(...) // wait for communication to finish
for (facet : mpi_facet)
        ApplyFromTraceValues(my_el, tv_recv[p][facet_part])
```

Under the hood - bottom line



Typical for programing with MPI:

- ullet understand the problem
- understand the exact data dependencies
- $\bullet\,$ simplify & reformulate to minimize those
- implement

Using MPI-parallel NGSolve

Setting up MPI-parallel NGSolve



Ideally, only need to switch on MPI with CMake option -DUSE_MPI=ON.

In fact, on clusters, this can be a bit more messy.

Using MPI-parallel NGSolve



Generally, we try to hide MPI as well as possible, however, in fact, the user has to do some stuff differently.

Things to consider:

- meshing is sequential only
- the netgen-ui is shm-parallel only (\rightarrow use VTK+ParaView)
- you have to take care of parallelizing anything that happens outside the ngsolve-libraries (e.g shell in- and output)
- any other considerations for running large jobs (e.g workload-manager)
- not quite as feature rich as shm-parallel version



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```
#!/usr/bin/bash
#

#SBATCH — job-name=ngs_phi
#SBATCH — partition=phi
#SBATCH — ntasks 640
#SBATCH — ntasks-per-node=64
#SBATCH — ntasks-per-core=1

mpirun ngspy mpi_poisson.py
```

Python Files



Solvers and Preconditioners



Own solvers/preconditioners :

- BDDC
- "Soon": Own AMG (\rightarrow Bernd Schwarzenbacher, tomorrow)

Interfaces for:

- MUMPS
- HYPRE BoomerAMG
- HYPRE AMS

Switch on with -DUSE_MUMPS=ON / -DUSE_HYPRE=ON; superbuild downloads and builds the libraries automatically.



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Does definitely NOT work (yet): multigrid, Block-jacobi, Block-GS



Experiences

Experiences



Works well on 2000 cores across 100 nodes.

• 3d-poisson, p2, 450M NDOF, 40 sec. setup + 13 sec solve (CG+AMG)

Works on 10 xeon phi's (640 mpi-procs).

• 3d-poisson, p1, 36M NDOF, 28 sec. setup + 6.5 sec solve (CG+BoomerAMG)

Bonus slide; VTK + Paraview