### **MATH 676**

## Finite element methods in scientific computing

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### **Lecture 41:**

# Parallelization on a cluster of distributed memory machines

### **Part 1: Introduction to MPI**

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### **Shared memory**

#### In the previous lecture:

- There was a single address space
- All parallel threads of execution have access to all data

#### Advantage:

• Makes parallelization simpler

#### **Disadvantages:**

- Problem size limited by
  - number of cores on your machine
  - amount of memory on your machine
  - memory bandwidth
- Need synchronisation via locks
- Makes it too easy to avoid hard decisions

### **Shared memory**

#### Example:

- Only one Triangulation, DoFHandler, matrix, rhs vector
- Multiple threads work in parallel to
  - assemble linear system
  - perform matrix-vector products
  - estimate the error per cell
  - generate graphical output for each cell
- All threads access the same global objects

For examples, see several of the step-xx programs and the "Parallel computing with multiple processors accessing shared memory" documentation module

### **Shared vs. distributed memory**

#### This lecture:

- Multiple machines with their own address spaces
- No direct access to remote data
- Data has to be transported explicitly between machines

#### Advantage:

- (Almost) unlimited number of cores and memory
- Often scales *better* in practice

#### **Disadvantages:**

- Much more complicated programming model
- Requires entirely different way of thinking
- Practical difficulties debugging, profiling, ...

### **Distributed memory**

#### Example:

- One Triangulation, DoFHandler, matrix, rhs vector object per processor
- Union of these objects represent global object
- Multiple programs work in parallel to
  - assemble *their part of the* linear system
  - perform *their part of the* matrix-vector products
  - estimate the error on their cells
  - generate graphical output for each of their cells
- Each program only accesses their part of global objects

See step-40/32/42 and the "Parallel computing with multiple processors using distributed memory" module

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### **Distributed memory**

## There are many ways to do distributed memory computing:

- Message passing interface (MPI)
- Remote procedure calls (RPC)
- Partitioned global address space (PGAS) languages:
  - Unified Parallel C (UPC an extension to C)
  - Coarray Fortran (part of Fortran 2008)
  - Chapel, X10, Titanium

#### **MPI's model is simple:**

- The "universe" consists of "processes"
- Typically:
  - One single-threaded process per core
  - One multi-threaded process per machine
- Processes can send "messages" to other processes...
- ...but nothing happens if the other side is not listening

Mental model: Sending letters through the mail system

#### **MPI's model implies:**

- You can't "just access" data of another process
- Instead, option 1:
  - you need to send a request message
  - other side has to pick up message
  - other side has to know what to do
  - other side has to send a message with the data
  - you have to pick up message
- Option 2:
  - depending on phase of program, I know when someone else needs my data → send it
  - I will know who sent me data  $\rightarrow$  go get it

#### **MPI's model implies:**

- You can't "just access" data of another process
- Instead...

#### This is bothersome to program. However:

- It exposes to the programmer what is happening
- Processes can do other things between sending a message and waiting for the next
- Has been shown to scale to >1M processes

#### **MPI** implementations:

- MPI is defined as a set of
  - functions
  - data types
  - constants
     with bindings to C and Fortran
- Is not a language on its own
- Can be compiled by a standard C/Fortran compiler
- Is typically compiled using a specific compiler wrapper: mpicc -c myprog.c -o myprog.o mpiCC -c myprog.cc -o myprog.o mpif90 -c myprog.f90 -o myprog.o
- Bindings to many other languages exist

#### **MPI's bottom layer:**

- Send messages from one processor to others
- See if there is a message from any/one particular process
- Receive the message

#### Example (send on process 2 to process 13):

#### **MPI's bottom layer:**

- Send messages from one processor to others
- See if there is a message from any/one particular process
- Receive the message

#### Example (query for data from process 13):

#### Note: One can also specify "anywhere"/"any tag".

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#### **MPI's bottom layer:**

- Send messages from one processor to others
- See if there is a message from any/one particular process
- Receive the message

#### Example (receive on process 13):

#### Note: One can also specify "anywhere"/"any tag".

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#### **MPI's bottom layer:**

- Send messages from one processor to others
- See if there is a message from any/one particular process
- Receive the message

#### **Notes:**

- MPI\_Send blocks the program: function only returns when the data is out the door
- MPI\_Recv blocks the program: function only returns when

   a message has come in
   the data is in the final location
- There are also non-blocking start/end versions (MPI\_Isend, MPI\_Irecv, MPI\_Wait)

#### **MPI's higher layers: Collective operations**

- Internally implemented by sending messages
- Available operations:
  - Barrier
  - Broadcast (one item from one to all)
  - Scatter (many items from one to all),
  - Gather (from all to one), AllGather (all to all)
  - Reduce (e.g. sum from all), AllReduce

**Note:** Collective operations lead to deadlocks if some processes do not participate!

**Example:** Barrier use for timing (pseudocode)

std::duration global time = end global - start;

```
... do something ...
MPI_Barrier (MPI_COMM_WORLD);
std::time_point start = std::now(); // get current time
foo(); // may contain MPI calls
std::time_point end_local = std::now(); // get current time
MPI_Barrier (MPI_COMM_WORLD);
std::time_point end_global = std::now(); // get current time
std::duration local time = end local – start;
```

#### **Note:** Different processes will compute different values.

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#### **Example:** Reduction

```
parallel::distributed::Triangulation<dim> triangulation; ... create triangulation ...
```

```
unsigned int my_cells = triangulation.n_locally_owned_cells();
unsigned int global_cells;
```

```
MPI_Reduce (&my_cells, &global_cells, MPI_UNSIGNED, 1,
/*operation=*/MPI_SUM,
/*root=*/0,
MPI_COMM_WORLD);
```

Note 1: Only the root (processor) gets the result.

**Note 2:** Implemented by (i) everyone sending the root a message, or (ii) hierarchical reduction on a tree

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#### **Example:** AllReduce

```
parallel::distributed::Triangulation<dim> triangulation; ... create triangulation ...
```

unsigned int my\_cells = triangulation.n\_locally\_owned\_cells(); unsigned int global\_cells;

MPI\_Allreduce (&my\_cells, &global\_cells, MPI\_UNSIGNED, 1, /\*operation=\*/MPI\_SUM, MPI\_COMM\_WORLD);

**Note 1:** All processors now get the result. **Note 2:** Can be implemented by MPI\_Reduce + MPI\_Broadcast

#### **MPI's higher layers: Communicators**

- MPI\_COMM\_WORLD denotes the "universe" of all MPI processes
- Corresponds to a "mail service" (a communicator)
- Addresses are the "ranks" of each process in a communicator
- One can form subsets of a communicator
- Forms the basis for collective operations among a subset of processes
- Useful if subsets of processors do different tasks

#### **MPI's higher layers: I/O**

- Fact: There is a bottleneck if 1,000 machines write to the file system at the same time
- MPI provides ways to make this more efficient

#### Also in MPI:

- "One-sided communication": directly writing into and reading from another process's memory space
- Topologies: mapping network characteristics to MPI
- Starting additional MPI processes

#### More information on MPI:

http://www.mpi-forum.org/

http://www.dealii.org/

#### Situation:

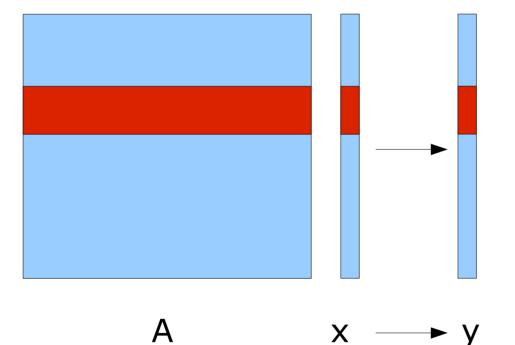
- Multiply a large NxN matrix by a vector of size N
- Matrix is assumed to be dense
- Every one of *P* processors stores *N*/*P* rows of the matrix
- Every processor stores *N/P* elements of each vector
- For simplicity: *N* is a multiple of *P*

```
struct ParallelVector {
  unsigned int size;
  unsigned int my elements begin;
  unsigned int my elements end;
  double *elements;
  ParallelVector (unsigned int sz,MPI Comm comm) {
    size = sz;
    int comm size, my rank;
    MPI Comm size (comm, &comm size);
    MPI Comm rank (comm, &my rank);
    my elements begin = size/comm size*my rank;
    my elements end = size/comm size*(my rank+1);
    elements = new double[my elements end-my elements begin];
```

```
struct ParallelSquareMatrix {
  unsigned int size;
  unsigned int my rows begin;
  unsigned int my rows end;
  double *elements;
  ParallelSquareMatrix (unsigned int sz,MPI Comm comm) {
    size = sz;
    int comm size, my rank;
    MPI Comm size (comm, &comm size);
    MPI Comm rank (comm, &my rank);
    my rows begin = size/comm size*my rank;
    my_rows_end = size/comm_size*(my_rank+1);
    elements = new double[(my rows end-my rows begin)*size];
```

#### What does processor *P* need:

• Graphical representation of what *P* owns:



 To compute the *locally owned* elements of y, processor P needs all elements of x

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```
void vmult (A, x, y) {
   int comm size=..., my rank=...;
   for (row block=0; row block<comm size; ++row block)
     if (row block == my rank) {
        for (col_block=0; col_block<comm_size; ++col_block)</pre>
           if (col block == my rank) {
             for (i=A.my_rows_begin; i<A.my_rows_end; ++i)
              for (j=A.size/comm size*col block; ...)
                y.elements[i-y.my rows begin] = A[...i,j...] * x[...j...];
           } else {
             double *tmp = new double[A.size/comm_size];
             MPI_Recv (tmp, ..., row_block, ...);
             for (i=A.my_rows_begin; i<A.my_rows_end; ++i)
              for (j=A.size/comm size*col block; ...)
                y.elements[i-y.my_rows_begin] = A[...i,j...] * tmp[...j...];
             delete tmp;
     } else {
        MPI_Send (x.elements, ..., row_block, ...);
```

#### Analysis of this algorithm

- We only send data right when we need it:
  - receiving processor has to wait
  - has nothing to do in the meantime
  - A better algorithm would:
  - send out its data to all other processors
  - receive messages as needed (maybe already here)
- As a general rule:
  - send data as soon as possible
  - receive it as late as possible
  - try to interleave computations between sends/receives
- We repeatedly allocate/deallocate memory should set up buffer only once

```
void vmult (A, x, y) {
   int comm size=..., my rank=...;
   for (row block=0; row block<comm size; ++row block)
     if (row block != my rank)
        MPI Send (x.elements, ..., row_block, ...);
  col block = my rank;
  for (i=A.my rows begin; i<A.my rows end; ++i)
    for (j=A.size/comm size*col block; ...)
      y.elements[i-y.my_rows_begin] = A[...i,j...] * x[...j...];
  double *tmp = new double[A.size/comm_size];
   for (col_block=0; col_block<comm_size; ++col_block)</pre>
     if (col block != my rank) {
        MPI Recv (tmp, ..., row block, ...);
        for (i=A.my_rows_begin; i<A.my_rows_end; ++i)
           for (j=A.size/comm size*col block; ...)
              y.elements[i-y.my_rows_begin] = A[...i,j...] * tmp[...j...];
   delete tmp;
```

#### Notes on using MPI:

- Usually, algorithms need data that resides elsewhere
- Communication needed
- Distributed computing lives in the conflict zone between
  - trying to keep as much data available locally to avoid communication
  - not creating a memory/CPU bottleneck
- MPI makes the flow of information explicit
- Forces programmer to design data structures/algorithms for communication
- Typical programs have relatively few MPI calls

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#### **Alternatives to MPI:**

- boost::mpi is nice, but doesn't buy much in practice
- Partitioned Global Address Space (PGAS) languages like Co-Array Fortran, UPC, Chapel, X10, ...:

#### **Pros:**

- offer nicer syntax
- communication is part of the language

#### Cons:

- typically no concept of "communicators"
- communication is implicit
- encourages poor data structure/algorithm design

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