	Introduction to High Performance Scientific Computing	
	Autumn, 2016	
	Lecture 16	
	Imperial College Prasun Ray	Ray
<u>[</u>	Imperial College Prasun Ray London 1 December 2016	
ſ	Last time: MPI collective data movement	7
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	P0 A Proadcast P1 A P2 A P2 A P3 A P4	
	P3	
	P1	
	P0 A P0 A B C D P1 B All gather P1 A B C D	
	P2 C P3 D P3 A B C D	
	P0 [A0 A1 A2 A3 P0 [A0 B0 G0 D0 P1 [B0 B1 B2 B3 All to All P1 [A1 B1 C1 D1 P2 [C0 C1 C2 C3 P2 [A2 B2 C2 D2 P3 [A2 B2 C2 D2 P3 P4 P4 P4 P4 P4 P4 P4	
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Today

- Coarse grain parallelization
- Send and Receive
- · Domain Decomposition
- Using gather and gatherv

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Today	
Coarse grain parallelization	
Send and Receive Domain Decomposition	
Using gather and gathery	
Part of course project	
Today's lecture All related Labs 8 (task 2)	
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Coarse-grain vs fine-grain parallelism	
With OpenMP, we've used fine-grain approach Look for a code segment (e.g. a loop) that can be parallelized Let OpenMP do the rest (just for that segment)	
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Coarse-grain vs fine-grain parallelism	
With OpenMP, we've used fine-grain approach Look for a code segment (e.g. a loop) that can be parallelized	
Let OpenMP do the rest (just for that segment) With MPI, typically take a coarse-grain approach	
At beginning of simulation, distribute data and tasks to processes Each process works on its own problem Occasionally communicating when necessary	
Can also use coarse-grain approach in OpenMP! (Project part 3.2)	
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Coarse-grain approach

· We have already seen a "sort-of" coarse grain approach with quadrature:

!set number of intervals per processor
Nper_proc = (N + numprocs - 1)/numprocs
!starting and ending points for processor
istart = myid * Nper_proc + 1
iend = (myid+1) * Nper_proc
if (iend>N) iend = N

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 iend = (myid+1) * Nper_proc
 if (iend>N) iend = N
!loop over intervals computing each interval's contribution to
 integral
 do il = istart,iend
 xm = dx*(i1-0.5) !midpoint of interval i1
 call integrand(xm,f)
 sum_i = dx*f
 sum_proc = sum_proc + sum_i !add contribution from interval
to total integral
 end do

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Coarse-grain approach

- More generally, at start of program we will:
 - Obtain myid and total number of processes, numprocs: call MPI_COMM_RANK(MPI_COMM_WORLD, myid, ierr) call MPI_COMM_SIZE(MPI_COMM_WORLD, numprocs, ierr)
 - 2. Use this information to distribute *Ntotal* points (or pieces of data) across *numprocs* processors

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Coarse-grain approach

- More generally, at start of program we will:
 - 1. Obtain *myid* and total number of processes, *numprocs*:

call MPI_COMM_RANK(MPI_COMM_WORLD, myid, ierr)
call MPI_COMM_SIZE(MPI_COMM_WORLD, numprocs, ierr)

- 2. Use this information to distribute Ntotal points (or pieces of data) across
- 3. We will use a simple fortran subroutine, MPE_DECOMP1D:

call MPE_DECOMP1D(Ntotal, numprocs, myid, istart, iend)

Nlocal = iend - istart + 1

- Simple subroutine which assigns istart and iend to each process
- If Ntotal=100, numprocs = 2:
 - myid = 0 → istart = 1, iend = 50
 myid = 1 → istart = 51, iend = 100

Complex parallelization

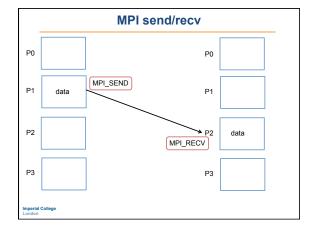
- MPE_DECOMP1D partitions data on a 'line'
- What about more complicated topologies or networks?
 e.g. simulation of 1e7 air molecules?

 - Advanced tools exist to do the partitioning for you
 - E.g. ParMETIS:

ParMETIS is an MPI-based parallel library that implements a variety of algorithms for partitioning unstructured graphs, meshes, and for computing fill-reducing orderings of sparse matrices. ParMETIS extends the functionality provided by METIS and includes routines that are especially suited for parallel AMR computations and large scale numerical simulations. The algorithms implemented in ParMETIS are based on the parallel multilevel k-way graph-partitioning, adaptive repartitioning, and parallel multi-constrained partitioning schemes developed in our lab.

MPI Send/Recv

- · Bcast and Reduce are examples of collective communication
- · Point-to-point communication carried out by send and recv
- · Probably the most basic and most important MPI commands



MPI Send/Recv

- Bcast and Reduce are examples of *collective* communication
- · Point-to-point communication carried out by send and recv
- Probably the most basic and most important MPI commands
- · Can send data between any two processors.
- Both send and recv are needed for data transfer.
- E.g. for previous figure need: if myid==1, send data to P2 and If myid==2 receive data from P1

MPI Send/Recv

If (myid==1) call MPI_SEND(n, 1, MPI_INTEGER,0, tag, MPI_COMM_WORLD,ierr)

If (myid==0) call MPI_RECV(n, 1, MPI_INTEGER,1, tag, MPI_COMM_WORLD, status, ierr)

These will send the integer ${\bf n}$ which has size ${\bf 1}$ from processor 1 to processor ${\bf 0}$.

- New code: sendExample.f90
- Now, array1 is only computed on P0, and we want to send the $3^{\rm rd}$ component to P1 and store it in P1's (empty) array1

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MPI Send/Recv: example

- f90example2.f90: compute array1 = sin(i1), i1=1,2,...
- New code: sendExample.f90
- Now, array1 is only computed on P0, and we want to send the 3rd component to P1 and store it in P1's (empty) array1
- Compute array1 on P0 and send it to P1:

 = 3 (myid==0) then	
 call calculations(N,array1) !fill in array1	
call MPI_SEND(array1(i1),1,MPI_DOUBLE_PRECISION,1,	i1,MPI_COMM_WORLD,
	ierr)
Destination	
Tag	J

MPI Send/Recv: example Compute array1 on P0, and send it to P1: i1 = 3 if (myid==0) then call calculations(N,array1) !fill in array1 call MPI_SEND(array1(i1),1,MPI_DOUBLE_PRECISION,1,i1,MPI_COMM_WORLD, ierr) Must also have MPI_RECV on P1: Notes: MPI_ANY_TAG: The destination will accept a message with any tag status(MPI_TAG) = 3; we have used the tag to send/set the array index

Comments on send/recv

- · Send/Recv are blocking operations

 - Code waits at send until the data has been received
 But what if all processes are trying to send data to each other?

 Can degrade performance or freeze the code

Comments on send/recv

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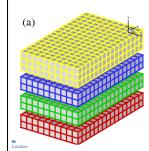
 - Code waits at send until the data has been received
 But what if all processes are trying to send data to each other?

 Can degrade performance or freeze the code
- · Solutions:
 - Combined send/recv: MPI_SENDRECV
 - Non-blocking send/recv: MPI_ISEND, MPI_IRECV
 - Usually used with MPI_WAIT or MPI_TEST
 - Buffered send: MPI_BSEND

 - Sender sends message and moves on
 Message is stored in buffer until receiver is ready

Send/Recv and domain decomposition

A parallel computation computes a potential field, f(x,y,z,t) on four processors. P0, P1, P2, P3 solve for *f* in separate subdomains



How would you compute the gradient?

$$\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}\right)$$

Computation of derivative

Equispaced grid: $x = x_1, x_2, x_3, ...$

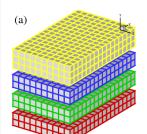
 $x_{i+1} - x_i = h = constant$

Then,
$$\frac{df_i}{dx} \approx \frac{f_{i+1} - f_{i-1}}{2h}$$

Send/Recv and domain decomposition

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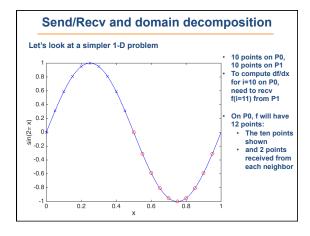
processors. P0, P1, P2, P3 solve for *f* in separate subdomains



How would you compute the gradient?

$$\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial x}, \frac{\partial f}{\partial x}\right)$$

- No problems with x and z directions
- But what about y?



gradient_p.f90: compute df/dx given f=sin($2\pi x$) distributed across processors

Code outline:

- I. Initialize MPI
 Read Ntotal from data.in
 Construct domain decomposition assign Nlocal points from istart to 3. Construct domain decomposition – assign Niocal points i lend to each processor.
 4. Make grid and field, f=sin(2*pi*x), in the local subdomain
 5. Compute derivative
 6. Output error

Parallel differentiation example

Key parts:

• Domain decomposition (subroutine from MPE library)

!construct decomposition
 call MPE_DECOMP1D(Ntotal, numprocs, myid, istart, iend)
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Key parts: • Domain decomposition (subroutine from MPE library) !construct decomposition call MPE_DECOMPIO(Ntotal, numprocs, myid, istart, iend) Nlocal = iend - istart + 1 • Make local grid and field !make grid and field call make_grid(Ntotal,Nlocal,istart,iend,x) dx = x(2)-x(1) print *, 'proc', myid, ' has been assigned the interval x=', x(1),x(Nlocal) call make_field(Nlocal,x,f(2:Nlocal+1)) !note: f(1) and f(Nlocal+2) must be obtained from neighboring processors

Parallel differentiation example

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call make_grid(Ntotal,Nlocal,istart,iend,x)
dx = x(2)-x(1)
print *, 'proc', myid, ' has been assigned the interval x=', x(1),x(Nlocal)

call make_field(Nlocal,x,f(2:Nlocal+1)) !note: f(1) and f(Nlocal+2) must be obtained from neighboring processors

Compute derivative (with send/recv at subdomain boundaries)...
```

Parallel differentiation example

Parallel differentiation example

- · At end of computation, each process has it's own part of df/dx
- It is sometimes useful to gather the data onto one process (e.g. for writing data to a file)
- Easy to do if gathering ten numbers from ten processors and storing them in an an array

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Parallel differentiation example

- At end of computation, each process has it's own part of df/dx
- It is sometimes useful to gather the data onto one process (e.g. for writing data to a file)
- Easy to do if gathering ten numbers from ten processors and storing them in an an array

!gather Nlocal from each proc to array Nper_proc on myid=0
call MPI_GATHER(Nlocal,1,MPI_INT,Nper_proc,1,MPI_INT 0,MPI_COMM_WORLD,
ierr)

Nlocal (size 1, type int) is sent into Nper_proc (rank 1 array, type int) on myid = 0

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- Nlocal (size 1, type int) is sent into Nper_proc (rank 1 array, type int) on myid = 0
- Trickier to do when gathering arrays into larger array. Then need array of locations where to place sub-arrays
 - e.g., disps = [1, 1+Nper_proc(1), 1+Nper_proc(1)+Nper_proc(2), ...]

Parallel differentiation example

- Trickier to do when gathering arrays into larger array. Then need array
- of locations where to place sub-arrays
 e.g., disps = [1, 1+Nper_proc(1), 1+Nper_proc(1)+Nper_proc(2), ...]
- Then use mpi_gatherv with disps as input:

• df (size Nlocal) is gathered from each processor and stored in df_total (in locations determined from Nper_proc and disps)

Notes on method of lines

Lecture 14: Solving N ODEs:

$$\frac{dT_i}{dt} = S_i(t) + \frac{T_{i+1} - 2T_i + T_{i-1}}{\Delta x^2}, \ i = 1, 2, ..., N$$

- Can solve ODEs with odeint which adjusts the time step error criteria are satisfied
- · Can also use time-step methods (see lab 8)
- · Simplest is explicit Euler:

$$\frac{dT_i}{dt} pprox \frac{T_i(t + \Delta t) - T_i(t)}{\Delta t}$$

$$\begin{split} \frac{dI_i}{dt} &\approx \frac{I_1(t+\Delta t) - I_1(t)}{\Delta t} \\ T_i(t+\Delta t) &= T_i(t) + \Delta t \left[S_i(t) + \frac{T_{i+1} - 2T_i + T_{i-1}}{\Delta x^2} \right], \ i = 1, 2, ..., N \end{split}$$

Notes on method of lines

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$$\begin{split} \frac{dT_i}{dt} &\approx \frac{T_i(t+\Delta t) - T_i(t)}{\Delta t} \\ T_i(t+\Delta t) &= T_i(t) + \Delta t \left[S_i(t) + \frac{T_{i+1} - 2T_i + T_{i-1}}{\Delta x^2} \right], \ i = 1, 2, ..., N \end{split}$$

error ~ dt, very poor stability propertis \rightarrow requires very small time step

- Fourth-order Runge-Kutta (RK4) is a much better fixed-time step method
- In course project, you are provide with routines for both methods
 Will have to modify euler

 - But will only have to provide appropriate RHS for RK4 (as RHS was provided for *odeint*)

Timing code

1. Lazy approach:

\$ time mpiexec -n 2 midpointpt real 0m0.073s user 0m0.081s sys 0m0.030s

Timing code

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\$ time mpiexec -n 2 midpointpt real 0m0.073s user 0m0.081s sys 0m0.030s

2. Use MPI_WTIME to time particular parts of code:

starttime = MPI_WTIME() !***START TIMER*** !code... !
endtime = MPI_WTIME() !***STOP TIMER***
print *, 'time= ',endtime - starttime, 'seconds'

Timing code 1. Lazy approach: \$ time mpiexec -n 2 midpointpt real 0m0.073s user 0m0.081s sys 0m0.081s sys 0m0.030s 2. Use MPI_WTIME to time particular parts of code: starttime = MPI_WTIME() !***START TIMER*** !code... ! endtime = MPI_WTIME() !***STOP TIMER*** print *, 'time= ', endtime - starttime, 'seconds' 3. But to get detailed information, use a profiler: VampirTrace, IPM, ... Imperial College London

