# Various parallel programmes using MPI in Fortran

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# 1 Speed-Up due to Pipelining

We define the speed-up as  $s = T/T_{pipe}$ . In the case where the is no pipelining the time it would take to do n operations would be T = 4n+1 (the last operation is used only to flush the last result out of the buffer; this is largely a question of definitions) while when using pipelining it is  $T_{pipe} = 4 + n + 1$  meaning that the speed-up can be written as follows;

$$S = \frac{4n+1}{n+5} \tag{1}$$

where the latency, the cycles taken before results start coming out, would be 4.

# 2 Timing and Optimization

In this small exercise, wall-clock times for different types of loops are to be tested. The types are normally coded loops, unrolled loops, interchanged loops and split loops. Also different compiler optimizations are tested.

**a**, Here a normally coded loop and a partly unrolled loop are compared. The source code can be seen in appendix A in the subroutine **ex2**. Table 1 shows the results from the normally coded version and the partly unrolled version compiled with different compiler optimizations. For the version with no optimization, the

	-O0	-01	-O2	-03
Normally coded	3.589	2.223	2.016	0.004
Unrolled	1.716	1.223	0.856	0.049
Time reduction	-52%	-45%	-58%	+965%

Table 1: Timing Results (in seconds). Normal loop vs. Unrolled loop.

unrolled version of the loop is much faster than the normally coded version. With the O1 option on for the compiler, the unrolled version is again faster than the normal version, but less so than in the case with no optimization. This could indicate that the compiler itself creates some kind of unrolled version when compiling. With option O2, the unrolled version is the fastest, while for the O3 option the normally coded loop becomes the fastest. Evidently the normally coded loop is simpler for the compiler to figure out, so it is capable of optimizing it more aggressively.

**b**, Here a triply nested loop is being timed and compared to a version where the loop nesting order is reversed. The loop in its ordinary version calls the Fortran built-in function **real** on the *outermost* loop variable, but stores this in a 3D array with leading dimension corresponding *also* to the outermost loop. Thus, in the ordinary version, it is easy to reduce the number of integer-to-real conversions, but the array is accessed contrary to its memory layout. The reversed version of the loop accesses the array in memory-contiguous order, but needs to call the **real** function many more times.

The source code can be seen in appendix A in the subroutine ex3a. The results are shown in table 2. The reversed version is by far the fastest version.

	-00	-01	-02	-03
Nested loop	1.851	1.791	1.819	0.148
Reversed	0.609	0.473	0.450	0.074
Time reduction	67%	-74%	-75%	-50%

Table 2: Timing Results (in seconds). Nested loop vs its reversely nested loop.

It can be seen that the reversed version becomes more efficient when compiler optimizations are used. However the O3 optimization considerably reduces the computation time for both the nested loop and the reverse.

As a final part of the question, a loop containing an IF-statement and its split version, without an IF-statement are to be compared. The source code can be seen in appendix A in the subroutine ex3b. The results are shown in table 3. The split version is the fastest version when only compiler option O0 or O1

	-00	-01	-O2	-03
Loop with IF-statement	2.412	1.916	1.463	0.0067
Split version	1.753	1.238	1.551	0.267
Time reduction	27%	-35%	+6%	+3885%

Table 3: Timing Results (in seconds). Loop with an IF-statement vs its split version.

is used, while the loop with the IF-statement is the fastest if compiler option O2 and O3 are used.

## **3** Loop Vectorization

Here we consider the possibility of vectorizing two versions of a DO-loop. See exercise text. The two do loops do not perform the same operation and they do not give the same result. The second DO-loop can be vectorized as follows;

$$a(2:n) = b(1:n-1) + d(2:n)$$
(2)

$$d(2:n) = a(1:n-1) - d(2:n)$$
(3)

The first DO-loop can not be vectorized, as values in one iteration depend directly on values from the immediately preceding iteration.

For the last part of the question we should explain how the Intel compiler can vectorize a particular loop with the cyclic dependency, where the dependency goes p iterations back. Our guess is that the Intel compiler does the **n** operations in vectors of size p. If one would show the execution time versus the parameter



Figure 1: Principle

p, we think it would have the form;

$$t = \alpha \frac{1}{p} \tag{4}$$

# 4 Parallel World with MPI

Here our first MPI program saying "Hello World" is created. The source code can been seen in appendix B.

The MPI library is used by specifying the include ''mpif.h''. Then the parallel environment is established with the MPI\_init command. To get the number of processors and the *rank* of each processor the commands MPI\_rank and MPI\_size is used. The command MPI\_Barrier is used to ensure that all threads has done the do before the another cycle begins. To finalize the MPI program the command MPI\_Finalize is used.

The results is shown below.

Hello world 0 4 Hello world 1 4 Hello world 2 4 Hello world 3 4

# 5 MPI Ping-Pong

In this question a MPI ping-pong program is written. A message is sent from process 0 to process 1. When the process 1 have received the message, its sends back the message to process 0 (ping-pong). This process is timed when completed a 1000 times.

First a single message is sent having a size of 16 byte. The time its takes to send this message from process 0 to process 1 is 8.188E - 3s (Ping) and the time its takes to send the message back from process 1 to process 0 is 7.907E - 3s(Pong). For the record it should be mentioned that the ping-pong is run on the Bohr-cluster. The estimated bandwidth is given as (message size/transfer time);

Bandwidth = 
$$\frac{\text{Messagesize}}{\text{transfertime}} = \frac{16byte}{0.008s \cdot 1000} \approx 2MB/sek.$$
 (5)

From the preceding equation an estimated bandwidth of approximately 2MB/sek. can be found.

Next the message size is increased incrementally up to a final message size of approximately 2MB. The results is shown in Fig. 2.



Figure 2: Timing results.

It can be seen that the wall clock time scales nearly linear with a slope of 2.1 which means that the approximate bandwidth is 2.1MB/sek..

The source code can be seen in appendix C.

# 6 Non-blocking communication

In this exercise the sum of the ranks of a set of processes enclosed by a ring is to be calculated. In order to avoid deadlocking the method **Issend-Recv-Wait** is used.

The output from a run on 8 processors produces the following output.

2 issendrecvwait 28
3 issendrecvwait 28
5 issendrecvwait 28
4 issendrecvwait 28
1 issendrecvwait 28
6 issendrecvwait 28
0 issendrecvwait 28
7 issendrecvwait 28

From the preceeding that the result is 28 from all processors as expected.

The send-receive method is now replaced with the MPI\_Sendrecv. the result is shown in the following;

Again the result is 28. Finally the MPI\_Sendrecv\_Replace is used. The advantage is here that one do not have to copy the receive buffer to the send buffer. The results is shown in the following;

```
3 sendrecv_replace 28
4 sendrecv_replace 28
5 sendrecv_replace 28
6 sendrecv_replace 28
7 sendrecv_replace 28
0 sendrecv_replace 28
1 sendrecv_replace 28
2 sendrecv_replace 28
```

The source code can be seen in appendix D.

# 7 Collective communication

Here the sum of all teh ranks is again computed. To do this, the command MPI\_Allreduce is used. In the present, the program is written such that is processor has its rank stored in a variable. The MPI\_Allreduce reads this variable from each processor and sum all the values, since it is given the argument MPI\_SUM. The results is then distributed to all processors. The output is shown below.

```
4 allreduce 28
6 allreduce 28
0 allreduce 28
2 allreduce 28
1 allreduce 28
5 allreduce 28
3 allreduce 28
7 allreduce 28
```

Result is 28.

To produce the i! of i = 1 the MPI\_Scan is used. The function basiccally scans each processor for a given value and the carries that value with it to the next processor where it calculates the product using the MPI\_PROD. The result is shown below;

It can be seen that processor 0 returns 1, processor 1 returns 2 and processer 2 returns 6 as expected. Finally the results from each processor is collected on a single processor which do an I/O.

Collected 1 2 6 24 120 720 5040 40320

## 8 Monte Carlo computation of $\pi$

In the exercise, the quantity  $\pi$  is to be found by the Monte Carlo method of integration. A unit circle is tightly enclosed by a square width is 2 meaning that the area of the square is 4, while the area of the unit circle is  $\pi$ .

The square is filled with randomly placed seeds. Some will be placed outside

the unit circle, and some will be placed within the unit circle. The ratio  $\hat{\pi}$  will be given as follows;

$$\hat{\pi} = 4 \cdot \frac{n_{ni}}{n} \tag{6}$$

where  $\hat{\pi} \xrightarrow{n \to \infty} \pi$  as the number of seeds is increased. To produce a set of random number from 0 to 1, the Fortran random number generator is used. A total of 2000 seeds is used for the computation.

When run on multiple processors, the command MPI\_reduce is used to collect the numbers from each processor to a single value. An output example is shown below for a MPI run on 8 processors.

0 3.104 1 3.09 3 3.146 4 3.138 5 3.16 2 3.136 6 3.214 7 3.152 Total 3.1425

where the first column is the rank and the second column is the value  $\pi$  from each processor. The source code can be seen in appendix E.

# 9 Poisson solver using red-black Gauss-Seidel method

We consider the Poisson equation

$$\nabla^2 u(\mathbf{r}) = f(\mathbf{r}) \tag{7}$$

in the 2D square domain  $\mathbf{r} \in ]0; 1[\times]0; 1[$  with some arbitrary boundary conditions, say u(0, y) = u(1, y) = 1 and u(x, 0) = u(x, 1) = 0. We wish to implement a parallel Fortran routine using MPI which solves this on a real-space grid  $u_{i,j}$ by repeatedly performing the operation

$$u_{i,j} \leftarrow \frac{1}{4} (u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + u_{i,j+1} + h^2 f_{i,j}).$$
(8)

Depending on the order in which neighbouring elements are updated in this fashion, the above scheme can be applied in several different ways. Here we shall use the red-black Gauss-Seidel method.

Points are classified as "red" or "black" in a checkerboard pattern, such that each point has four neighbours of the opposite colour. First all red points are updated as per Eq. (8), requiring knowledge of only the black points. Then the black points are updated, using the newly calculated red values. Each point is thus updated once by performing two separate *sweeps* of the array.

To do this in parallel, we assign each CPU to be responsible for a distinct subdomain, such that the subdomains are square (for simplicity) and equally large. The edge points for one CPU will then, during a given sweep, depend on values for which the neighbouring CPUs are responsible. Every CPU therefore has to send and receive information to and from each of its four neighbours in the grid, with the exception of those CPUs that are on the domain boundary.

#### 9.1 Implementations and performance

We have implemented a main loop which performs one "red" sweep and one "black" sweep per overall iteration, where each sweep is governed by the following code:

```
call recv_buffers_nonblocking(rbuf, ranks, M, comm, reqs)
call update_local_array(U, M, M, h2, eps, color)
call copy_array_to_buffers(U, M, M, sbuf)
call send_buffers_blocking(ranks, sbuf, M, comm)
call wait_for_requests(ranks, reqs)
call copy_buffers_to_array(U, M, M, rbuf)
```

- First a non-blocking receive is started with a reception buffer rbuf (M is the size of the array which must for simplicity be square).
- Then the local array slice U is updated according to the usual procedure (the color parameter determines the checkerboard offset, h2 is the squared grid-spacing and eps is the residual which is integrated dynamically).
- After the update, the array edges are copied out into a send buffer (sbuf) and sent to the neighbours in a blocking manner (which is far from optimal; the parallelization only helps to the extent that the processes do not finish the previous step equally fast. This will be improved below).
- Finally the receive buffer is copied back into the array.

Figure 3 shows a speedup plot for the Poisson solver.

It scales to some extent, although less than perfectly due to the communication. We have previously theorized about another reason for bad scaling, namely that the test computer is in fact a shared-memory machine, which means that a sufficiently large number of CPUs can exhaust the main memory bandwidth if they are loading large arrays (that do not fit in the caches) at the same time.

The latter we cannot do anything about, but for smaller matrices in particular, performance should improve by using a better non-blocking method. Our plan is to update only the boundary points at first, then send the boundary values in a non-blocking manner, *then* update the remaining interior points, and finally copy the by hopefully received buffers back into the array.

Improved main loop for one sweep:



Figure 3: Speedup of first Poisson solver for different CPU counts and matrix sizes

```
call update_boundaries_and_copy_to_buffers(U, M, h2, eps, color, sbuf)
call send_buffers_nonblocking(sbuf, M, ranks, comm, reqs)
call update_interior(U, M, h2, eps, color)
call recv_buffers_blocking(rbuf, M, ranks, comm)
call wait_for_requests(ranks, reqs)
call copy_buffers_to_array(U, M, M, rbuf)
```

Figure 4 shows the speedup for the improved non-blocking scheme. It turns out that this does scale better: for 16 CPUs, the previous method had a speedup of 8 to 12 depending on matrix size, whereas the new one reaches 9.5 to 13. The tendency is more favourable for smaller matrix sizes, consistently with the theory of memory bandwidth exhaustion. We have observed that the variation in serial speed due to the extra looping over boundary points does not contribute measurably to calculation time.

### 10 Parallel matrix multiplication

We would like to perform the matrix multiplication AB = C in a distributed manner using the Fox algorithm with MPI.

Let us say that we have p processes and, for simplicity, that p is a square number such that we can have a quadratic process grid of  $b \times b$ . We would then like to allocate the matrices **A**, **B** and **C** in a blocked manner: Each CPU has one block of **A** and **B**, and is responsible for calculating one block of **C** (which means it will have to obtain data from the other blocks of **A** and **B** by communicating with the other processes). We denote the (p, q)'th block by  $\mathbf{A}^{pq}$ 



Figure 4: Speedup of improved Poisson solver. Consistently better scaling than the previous implementation.

and its (i, j)'th element by  $a_{ij}^{pq}$ . The blocks contain, in total, all elements of **A** exactly once.

Suppose we have a 2 by 2 process grid. As an example, the element  $c_{23}$  of a 4 by 4 square matrix product can be written down in the usual ("global") way, or in terms of 2 by 2 blocks of 2 by 2 elements, respectively:

$$c_{23} = a_{20}b_{03} + a_{21}b_{13} + a_{22}b_{23} + a_{23}b_{33} \tag{9}$$

$$c_{23} = c_{01}^{11} = a_{00}^{10}b_{01}^{01} + a_{01}^{10}b_{11}^{01} + a_{00}^{11}b_{01}^{01} + a_{01}^{11}b_{11}^{01}.$$
 (10)

As an example, for 2 by 2 processes, block (1, 1) of the product is

$$\mathbf{C}^{11} = \mathbf{A}^{10}\mathbf{B}^{01} + \mathbf{A}^{11}\mathbf{B}^{11}$$
(11)

While process (1, 1) already has the (1, 1) slices of **A** and **B**, it will need to receive the (1, 0) and (0, 1) slices of **A** and **B**, respectively, to finish the calculation.

As a more complete example of Fox' algorithm, this demonstrates the multiplication of a 3 by 3 matrix distributed on 9 processors:

Thus, processes i0, i1, and i2 all need slice ii of **A**. Process ii will have to broadcast its slice to those three processes before the step is performed. Aside

from this, only the local slices are used. Borrowing the notation "+ =" from C/C++, the next step is:

This time process (i, i + 1) (modulo b) broadcasts its slice of **A**. Also each proces has exchanged its slice of **B** with another process. Specifically process (i, j) must have sent its slice to process (i, j - 1) (modulo b). To generalize: at each step, the next processor to the "west" in the process grid broadcasts **A** along the east-west direction. Then each process performs a block multiplication with the currect local slices. Last, the local slice of **B** is sent back "north" in the grid for the next step. For completeness, the final step in a 3 by 3 block multiplication is:

$$\begin{aligned}
 C_{00} + &= \mathbf{A}_{02} \mathbf{B}_{20} & \mathbf{C}_{01} + &= \mathbf{A}_{02} \mathbf{B}_{21} & \mathbf{C}_{02} + &= \mathbf{A}_{02} \mathbf{B}_{22} \\
 C_{10} + &= \mathbf{A}_{10} \mathbf{B}_{00} & \mathbf{C}_{11} + &= \mathbf{A}_{10} \mathbf{B}_{01} & \mathbf{C}_{12} + &= \mathbf{A}_{10} \mathbf{B}_{02} \\
 C_{20} + &= \mathbf{A}_{21} \mathbf{B}_{10} & \mathbf{C}_{21} + &= \mathbf{A}_{21} \mathbf{B}_{11} & \mathbf{C}_{22} + &= \mathbf{A}_{21} \mathbf{B}_{12}
 \end{aligned} \tag{14}$$

#### 10.1 Implementation

This is the interesting part of the source code for the multiplication function:

```
call copy(A, buf2, M)
do step=0,nblocks - 1
   call copy(buf2, A, M)
   root = mod(step + coords(1), nblocks)
   call MPI_Bcast(A, MM, dtype, root, rcomm, err)
   call multiply_add(A, B, C, M)
   call MPI_Sendrecv( &
      B, MM, dtype, ranks(NORTH), 0, &
      buf, MM, dtype, ranks(SOUTH), 0, &
      comm, status, err)
   call copy(buf, B, M)
```

enddo

We perform as many steps as there are blocks, which is why there is a loop. Each step consists of the following operations:

- A, B and C are the M by M matrices involved in the product. First A is copied into a backup buffer (copy is a small utility subroutine), whereafter it is broadcast from a particular rank on rcomm, a communicator for the processor row.
- The subroutine call multiply\_add then adds the local slice product of A and B to C.



Figure 5: Speedup of Fox algorithm for various matrix sizes.

• Finally the local slice of B is propagated along the process grid to the "north" by calling MPI\_Sendrecv. Another temporary buffer is used for this.

## 10.2 Timings

Figure 5 shows the speedup of the matrix multiplication algorithm for different matrix sizes and CPU counts. In most cases the speedup is superlinear as the matrix slices fit better into the 8MB L2 cache when using many CPUs. For the 800 by 800 matrix, performance decreases with 16 CPUs as there is little work compared to the amount of synchronization. For larger matrices the multiplication scales very well. The superlinear speedup disappears for the 2400 by 2400 matrix as the matrices never fit well into the cache.

# A Source code - Timing and Optimaztion (main.f90)

```
1
    program timing
 \mathbf{2}
       implicit none
 3
      call ex3b
 4
 \mathbf{5}
    end program timing
 6
 \overline{7}
    subroutine ex3b()
 8
      implicit none
 9
      integer :: i, n, j, m=500
real(kind=8), dimension(:), allocatable :: a
10
11
12
      real :: t1, t2
13
      n = 100000
14
15
       allocate(a(n))
16
17
       do i=1, 100000
         if (mod(i, 10) == 0) then
a(i) = 0.0
18
19
20
           else
          a(i) = real(i, 8)
endif
21
22
23
       enddo
24
25
26
       call cpu_time(t1)
      do j=1, m
do i=1, n
27
28
             if (mod(i, 10) == 0) then
29
                 a(i) = 0.0
30
31
              else
32
                a(i) = real(i, 8)
33
              endif
34
          enddo
      enddo
35
      call cpu_time(t2)
print*, t2 - t1, a(17)
36
37
38
39
       call cpu_time(t1)
      do j=1, m
do i=1, n
a(i) = real(i, 8)
40
41
42
43
          enddo
         do i=10, n, 10
a(i) = 0.0
44
45
          enddo
46
47
       enddo
48
       call cpu_time(t2)
      print*, t2 - t1, a(17)
49
50
51
    end subroutine ex3b
52
    subroutine ex3a()
    implicit none
53
54
55
       integer :: m, i, j, k, n, n0
56
57
       real :: t1, t2
58
      real(kind=8), dimension(:, :, :), allocatable :: f
59
60
      m = 200
61
62
       allocate(f(m, m, m))
63
64
       ! warmup
65
       do i=1,m
```

```
do j=1,m
    do k=1,m
    f(i, j, k) = real(i, 8)
 66
 67
 68
 69
 70 \\ 71 \\ 72
            enddo
         enddo
 73
 74
75
        call cpu_time(t1)
 76
        do i=1,m
         do 1-1,m
    do j=1,m
    do k=1,m
        f(i, j, k) = real(i, 8)
 77
78
 79
 80
 81
            enddo
        enddo
call cpu_time(t2)
 82
 83
        print*, t2 - t1
print*, f(17, 17, 17)
 84
 85
 86
 87
         call cpu_time(t1)
 88
         do k=1,m
          io k=1,m
    do j=1,m
        do i=1,m
        f(i, j, k) = real(i, 8)
 89
 90
 91
 92
 93
           enddo
 94
         enddo
 95
        call cpu_time(t2)
 96
        print*, t2 - t1
print*, f(17, 17, 17)
 97
 98
 99
100
      end subroutine ex3a
101
102
      subroutine ex2()
implicit none
103
104
105
        real :: t1, t2
106
        real :: t
        integer :: it, i, n
107
108
        real(kind=8), dimension(:), allocatable :: a
109
110
111
        n = 100000
112
113
         allocate(a(n))
114
        ! warmup loop
do it=1, 100
    do i=1,n
115
116
117
118
               a(i) = real(i, 8)
119
           enddo
         enddo
120
121
122
        call cpu_time(t1)
123
124
        do it=1, 1024
         do i=1,n
a(i) = real(i, 8)
125
126
127
            enddo
128
        enddo
129
130
        call cpu_time(t2)
131
132
        print*, t2 - t1
```

```
134
135
136
                        a(i) = real(i, 8)
a(i + 1) = real(i + 1, 8)
a(i + 2) = real(i + 2, 8)
a(i + 3) = real(i + 3, 8)
137
138
139
140
                         a(i + 5) = real(i + 5, 8)
a(i + 4) = real(i + 4, 8)
a(i + 5) = real(i + 5, 8)
a(i + 6) = real(i + 6, 8)
a(i + 7) = real(i + 7, 8)
141
142
143
144
145
                   enddo
             enddo
146
             call cpu_time(t2)
print*, t2 - t1
147
148
149
         end subroutine ex2
```

# B Source code - Parallel World with MPI (question4.f90)

```
1
     program helloworld
 2
       include 'mpif.h'
 3
      integer :: ierror
integer :: rank, size,r
 4
 \mathbf{5}
 6
       call MPI_Init(ierror)
 7
       call MPI_Comm_rank(MPI_Comm_World, rank, ierror)
call MPI_Comm_size(MPI_Comm_World, size, ierror)
 8
 9
10
11
       do r=0,size
12
         if (r.EQ.rank) then
           print*,"Hello world",rank,size
endif
13
14
          call MPI_Barrier(MPI_Comm_World,ierror)
15
       enddo
16
       call MPI_Finalize(ierror)
17
18
19 end program helloworld
```

# C Source code - MPI Ping-Pong (mpi\_pingpong.f90)

```
1
    program pingpong
       include "mpif.h"
implicit none
 \mathbf{2}
 3
 4
 \mathbf{5}
       integer :: n, npings
 \mathbf{6}
       integer :: count
 \overline{7}
       integer :: err
 8
       call MPI_Init(err)
 9
10
11
       do n=2, 19
12
          count = 2**n
13
          npings = 1000! * 32 / count
14
          if (npings.le.4) then
15
             npings = 4
          endif
16
17
          call runpingpong(count, npings)
18
      enddo
19
20
      call MPI_Finalize(err)
21
22
    end program pingpong
23
    subroutine runpingpong(count, npings)
include "mpif.h"
24
25
26
       implicit none
27
28
       integer :: err ! should really check error state but can't be bothered
       integer :: rank, size
29
       integer :: iping, npings
integer :: dest ! target rank (0 or 1)
integer :: tag = 0
30
31
32
33
       integer :: count
34
       integer :: comm
35
36
       !character, dimension(count) :: sendbuf, recvbuf
37
       integer, dimension(:), allocatable :: sendbuf, recvbuf
integer, dimension(MPI_STATUS_SIZE) :: stat
38
39
40
       double precision :: t1, t2
41
       allocate(sendbuf(count), recvbuf(count))
42
43
       sendbuf = 0
       recvbuf = 0
44
45
46
       !character(count) :: sendbuf ! dimension(4)?
47
48
       !character(count) :: recvbuf
49
50
       call MPI_Comm_size(MPI_COMM_WORLD, size, err)
51
       call MPI_Comm_rank(MPI_COMM_WORLD, rank, err)
52
       comm = MPI_COMM_WORLD
53
54
       if(size.ne.2) then
55
56
          print*, 'Please use size 2'
57
          call MPI_Finalize(err)
58
          stop
59
       endif
60
       t1 = MPI_Wtime()
61
62
63
       if (rank.eq.0) then
          sendbuf(1) = 0!'p'
sendbuf(2) = 2!'i'
64
65
```

```
sendbuf(3) = 5!'n'
sendbuf(4) = 1!'g'
66
67
68
            !sendbuf = 'ping
69
            dest = 1
           !print*, 'Rank:', rank, 'Send:', sendbuf
call MPI_Send(sendbuf, count, MPI_INTEGER, dest, tag, MPI_COMM_WORLD,
70 \\ 71
                  err)
72
        else
            !sendbuf = 'pong'
73
            sendbuf(1) = 1
74
            sendbuf(2) = 42
75
\frac{76}{77}
           sendbuf(3) = 17
            sendbuf(4) = 37
78
            dest = 0
79
           !call MPI_Recv(recvbuf, count, MPI_INTEGER, dest, tag, MPI_COMM_WORLD, &
80
                   stat, err)
81
        endif
82
        do iping = 1, npings - 1
    call MPI_Recv(recvbuf, count, MPI_INTEGER, dest, tag, comm, stat, err)
    !print*, 'Rank:', rank, 'Recv:', recvbuf, 'Send:', sendbuf, 'Count:',
83
84
85
                 iping
86
            call MPI_Send(sendbuf, count, MPI_INTEGER, dest, tag, comm, err)
87
        enddo
88
89
        if (rank.eq.1) then
            (running, find)
call MPI_Recv(recvbuf, count, MPI_INTEGER, dest, tag, comm, stat, err)
!print*, 'Rank:', rank, 'Recv:', recvbuf, 'Send:', sendbuf
90
91
92
        endif
93
94
        t2 = MPI Wtime()
95
       print*, 'Rank', rank, 'Time', t2 - t1, 'Size', count, 'Number', npings
96
97
98
99
     end subroutine runpingpong
```

# D Source code - Non-Blocking communication (ranksum.f90)

```
1
    program ranksum
      include "mpif.h"
implicit none
2
3
 4
5
      integer :: rank, size, err
 6
      integer :: comm
 7
      integer :: sum = 0
8
      integer :: sendrq
9
      integer :: src, dst
integer :: i
10
11
12
13
      integer, dimension(1) :: sendbuf, recvbuf
14
      integer, dimension(MPI_STATUS_SIZE) :: stat
15
      integer, dimension(:), allocatable :: collectivebuf
16
17
      call MPI_Init(err)
      comm = MPI_COMM_WORLD
18
19
20
      call MPI_Comm_rank(comm, rank, err)
21
      call MPI_Comm_size(comm, size, err)
22
23
      allocate(collectivebuf(size))
24
      collectivebuf = 0
25
```

```
dst = mod(rank + 1, size)
src = modulo(rank - 1, size)
26
27
28
29
       sendbuf(1) = rank
30
       sum = 0
31
       do i=1, size
32
          call MPI_Issend(sendbuf, 1, MPI_INTEGER, dst, i, comm, sendrq, err)
call MPI_Recv(recvbuf, 1, MPI_INTEGER, src, i, comm, stat, err)
33
34
           sum = sum + recvbuf(1)
sendbuf(1) = recvbuf(1)
35
36
           call MPI_Wait(sendrq, stat, err)
37
38
       enddo
39
       if (rank.ne.recvbuf(1)) then
40
           print*, 'BAD'
       endif
41
42
       print*, rank, 'issendrecvwait', sum
43
44
       call MPI_Barrier(comm, err)
45
46
       sum = 0
47
       do i=1, size
48
          call MPI_Sendrecv( &
                 sendbuf, 1, MPI_INTEGER, dst, i, &
recvbuf, 1, MPI_INTEGER, src, i, &
49
50
          comm, stat, err)
sum = sum + recvbuf(1)
sendbuf(1) = recvbuf(1)
51
52
53
54
       enddo
55
       if (rank.ne.recvbuf(1)) then
56
          print*, 'BAD'
57
       endif
       print*, rank, 'sendrecv', sum
58
59
60
       flush(6)
61
       call MPI_Barrier(comm, err)
62
63
       sum = 0
64
       do i=1, size
65
          call MPI_Sendrecv_Replace(sendbuf, 1, MPI_INTEGER, dst, i, src, i, comm,
                 &
66
                 stat, err)
67
          sum = sum + sendbuf(1)
68
       enddo
       if (rank.ne.sendbuf(1)) then
    print*, 'BAD'
69
70
71
       endif
72
       print*, rank, 'sendrecv_replace', sum
73
       ! If we want to use blocking Recv/Send, then we'll have to get
! e.g. even ranks to start with Send, while uneven ones start with Recv.
\frac{74}{75}
76
       !enddo
77
78
       sum = 0
79
       recvbuf(1) = 0
80
       call MPI_Allreduce(sendbuf, recvbuf, 1, MPI_INTEGER, MPI_SUM, comm, err)
81
       sum = recvbuf(1)
82
83
       print*, rank, 'allreduce', sum
84
       sendbuf(1) = rank + 1
85
       call MPI_Scan(sendbuf, recvbuf, 1, MPI_INTEGER, MPI_PROD, comm)
call MPI_Barrier(comm, err)
86
87
88
89
       print*, rank, recvbuf, size
90
       call MPI_Gather( &
91
              recvbuf, 1, MPI_INTEGER, &
              collectivebuf, 1, MPI_INTEGER, &
92
```

```
0, comm, err)
if(rank.eq.0) then
93
94
95
         print*, 'Collected', collectivebuf
96
       endif
97
       call MPI_Barrier(comm, err)
98
99
       call MPI_Finalize(err)
100
       deallocate(collectivebuf)
101
     end program ranksum
```

# E Source code - Monto Carlo computation of $\pi$ (montecarlo.f90)

```
program montecarlo
 1
\mathbf{2}
       include "mpif.h"
3
       implicit none
4
       integer :: ni = 0, nx = 0, n, i, nmax = 2000
5
6
      real :: rx, ry
real :: pi, tmp
real :: sum = 0.0
 \overline{7}
 8
9
      integer :: seedsize = 7
integer, dimension(:), allocatable :: seed
10
11
12
13
       integer :: err
      integer :: rank, size
14
15
16
       real, dimension(1) :: sendbuf
      real, dimension(1) :: recvbuf
17
18
19
       call MPI_Init(err)
20
21
       call MPI_Comm_rank(MPI_COMM_WORLD, rank, err)
22
       call MPI_Comm_size(MPI_COMM_WORLD, size, err)
23
24
       call random seed(size=seedsize)
25
       allocate(seed(seedsize))
26
      do i=1, seedsize
27
          seed(i) = 42 + rank * 17
28
       enddo
29
      call random_seed(put=seed)
30
      do n=1, nmax
31
32
         call random_number(rx)
33
          call random_number(ry)
          tmp = rx**2 + ry**2
if (tmp.le.(1.0)) then
34
35
36
             ni = ni + 1
          else
37
38
             nx = nx + 1
39
          endif
40
       enddo
41
       pi = 4.0 * real(ni) / (real(ni) + real(nx))
42
      print*, rank, pi
43
44
45
       sendbuf(1) = pi
46
       call MPI_Reduce( &
47
48
            sendbuf, recvbuf, 1, &
            MPI_REAL, MPI_SUM, O, MPI_COMM_WORLD, err)
49
50
      if (rank.eq.0) then
    print*, 'Total', recvbuf(1) / size
51
52
```

53 endif 54 55 call MPI\_Finalize(err) 56 57 end program montecarlo