

## HPC for numerical methods and data analysis

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# Communication in Python and MPI

## Exercise I Reminder of a simple MPI code in Python

Given two vectors, b, c we want to compute d = 2b + c. Execute the following simple code on 2 processors several times.

```
from mpi4py import MPI
import numpy as np
b = np.array([1, 2, 3, 4])
c = np.array([5, 6, 7, 8])
a = np.zeros_like(b)
d = np.zeros_like(b)
comm = MPI.COMM_WORLD
rank = comm.Get_rank()
if rank == 0:
    for i in range(4):
        a[i] = b[i] + c[i]
    comm.Send(a, dest = 1, tag = 77)
else:
    comm.Recv(a, source = 0, tag = 77)
    for i in range(4):
        d[i] = a[i] + b[i]
print("I am rank = ", rank )
print("d: ", d)
```

Observe the order in which the prints take place and the value of d at the end.

## Exercise II Point to point communication - blocking and non-blocking communication

- a) Provide a brief definition of MPI. What is a communicator?
- b) Execute the following simple code on 4 processors.

```
from mpi4py import MPI
import numpy as np
# Initialize MPI
comm = MPI.COMM_WORLD
rank = comm.Get_rank()
if rank == 0:
    data = {'a': 7, 'b': 3.14}
    print("From process: ", rank, "\n data sent:", data, "\n")
    comm.send(data, dest=1, tag=11)
elif rank == 1:
    data = comm.recv(source=0, tag=11)
    print("From process: ", rank, "\n data received:", data, "\n")
elif rank == 2:
    data = np.array([1, 1, 1, 1, 1])
    print("From process: ", rank, "\n data sent:", data, "\n")
    comm.send(data, dest=3, tag = 66)
else:
    data = comm.recv(source = 2, tag = 66)
    print("From process: ", rank, "\n data received:", data, "\n")
```

In this case, why do we need to be careful when specifying the dest and tag parameters on both comm.send and comm.recv?

c) Describe the difference between blocking communication and non-blocking communication in MPI. Modify the code above such that it uses comm.isend instead of comm.send and comm.irecv instead of comm.recv while ensuring the messages are passed correctly.

### Exercise III Collective communication - scattering and broadcasting

a) Run the following script on 4 processors:

```
from mpi4py import MPI
import numpy as np

# Initialize MPI
comm = MPI.COMM_WORLD
rank = comm.Get_rank()
```

```
size = comm.Get_size()

# Define the vector
if rank == 0:
    vector = np.array([16, 62, 97, 25])
else:
    vector = None

data1 = comm.bcast(vector, root = 0)
data2 = comm.scatter(vector, root = 0)

print("rank: ", rank, " data1: ", data1, " data2: ", data2)
```

What is the difference in MPI between scattering and broadcasting?

- b) Consider the multiplication of a matrix  $A \in \mathbb{R}^{m \times n}$  with a vector  $v \in \mathbb{R}^n$ . Write a Python file containing a script that:
  - Creates a matrix of dimension  $m \times n$
  - $\bullet$  Creates a vector of dimension n
  - ullet Makes sure that the dimensions of the matrix and the vector agree in such way that we can compute Av
  - Computes Av using MPI's scattering, make sure you execute your code on the right amount of processors (*Hints: you'll need to use* comm.gather. What are the entries of Av?)

### Exercise IV Collective communication - all-to-all and reduce

• Run the following code on 4 processors:

```
from mpi4py import MPI
import numpy as np

comm = MPI.COMM_WORLD
size = comm.Get_size()
rank = comm.Get_rank()

senddata = rank*np.ones(size, dtype = int)

recvdata = comm.alltoall(senddata)

print(" process ", rank, " sending ", senddata, " receiving ", recvdata )

What is comm.alltoall doing? Compare it to comm.scatter.
```

• In this exercise we are going to use reduction operations on MPI. Run the following code on 4 processors:

What is a reduction operation? What is the difference between this and comm.gather?

• In the previous code, change comm.reduce to comm.allreduce. What is the difference between the two? (Note, comm.allreduce doesn't use the argument root).

#### Exercise V Deciding what to use - Mid point rule

Numerical integration describes a family of algorithms for calculating the value of definite integrals. One of the simplest algorithms to do so is called the Mid Point Rule. Assume that f(x) is continous on [a, b]. Let n be a positive integer and h = (b - a)/n. If [a, b] is divided into n subintervals,  $\{x_0, x_1, ..., x_{n-1}\}$ , then if  $m_i = (x_i + x_{i+1})/2$  is the midpoint of the i-th subinterval, set:

$$M_n = \sum_{i=1}^n f(m_i)h.$$

Then:

$$\lim_{n \to \infty} M_n = \int_a^b f(x) dx.$$

Thus, for a fixed n, we can approximate this integral as:

$$\int_{a}^{b} f(x)dx \approx \sum_{i=1}^{n} f(m_i)h$$

Set n = s \* 500,  $f(x) = \cos(x)$ , a = 0,  $b = \pi/2$ . Write a Python script such that:

- Defines a function that given  $x_i, h, n$  first calculates 500 mid points on a subinterval  $[x_i, x_{i+1}]$  and returns the approximation of the integral on this subinterval.
- ullet Using MPI approximates the integral of f on [a,b]
- $\bullet$  Run your script on s processors

to approximate the integral of f. (Hints: there are many ways of doing this, one approach is using comm.bcast and comm.reduce).