Week13-mpi-with-python

% Basic MPI Concepts in Python

1. Ranks

Each process in MPI gets a unique ID called a "rank".

rank == 0 is usually the **master** process.

2. Size

Total number of processes running.

3. Communication

Use send(), recv(), or collective operations like bcast() and gather().

🔁 Example: Hello World with mpi4py

```
# hello_mpi.py
from mpi4py import MPI

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

print(f"Hello from process {rank} of {size}")

Run it with 4 processes:

mpiexec -n 4 python hello_mpi.py
```

mpi4py

is a **Python wrapper** for the **MPI** (**Message Passing Interface**) standard, allowing Python programs to perform **parallel processing** across multiple CPUs, cores, or even multiple machines — just like traditional HPC (High Performance Computing) applications written in C or Fortran using MPI.

What is mpi4py?

- mpi4py is a Python package that binds the MPI C API to Python.
- It enables you to:
 - o Launch multiple processes (mpiexec -n N python your script.py)
 - Send and receive messages (send(), recv())
 - Use collective communication (bcast(), scatter(), gather(), reduce())
- It supports **NumPy arrays**, making it efficient for large data communication.

Core Concepts in mpi4py

Concept	MPI API	Description
World Comm	COMM_WORLD	The default communicator (all processes)
Rank	Get_rank()	Each process has a unique rank ID
Size	Get_size()	Total number of processes in the group
Send/Recv	send(), recv()	Point-to-point messaging
Broadcast	bcast()	Send data from one to all
Scatter	scatter()	Split data across processes
Gather	gather()	Collect data from all processes
Reduce	reduce()	Aggregate data (e.g., sum, max)
Barrier	barrier()	Synchronize all processes

Why Use mpi4py?

- Run the same code across multiple processes (not just threads).
- Avoid Python's Global Interpreter Lock (GIL) for CPU-bound tasks.
- Scale programs across multiple nodes in an HPC cluster.
- Works great with NumPy for numerical and matrix-heavy applications.

1. Point-to-Point Communication: send() and recv()

```
# point to point.py
from mpi4py import MPI
comm = MPI.COMM WORLD
rank = comm.Get rank()
if rank == 0:
    data = "Hello from rank 0"
    comm.send(data, dest=1)
   print(f"[{rank}] Sent data to rank 1")
elif rank == 1:
   data = comm.recv(source=0)
   print(f"[{rank}] Received data: {data}")
```

2. Broadcast: bcast()

```
# broadcast.py
from mpi4py import MPI
comm = MPI.COMM WORLD
rank = comm.Get rank()
data = None
if rank == 0:
    data = "Broadcast message"
data = comm.bcast(data, root=0)
print(f"[{rank}] Got: {data}")
```



3. Scatter: scatter()

```
# scatter.py
from mpi4py import MPI
comm = MPI.COMM_WORLD
rank = comm.Get rank()
size = comm.Get_size()
if rank == 0:
   data = [i for i in range(size)] # One value per process
   data = None
recv data = comm.scatter(data, root=0)
print(f"[{rank}] Received {recv data}")
```



4. Gather: gather()

```
# gather.py
from mpi4py import MPI

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

send_data = rank * 2
gathered_data = comm.gather(send_data, root=0)

if rank == 0:
    print(f"[{rank}] Gathered_data: {gathered_data}")
```

5. Reduce: reduce()

```
# reduce.py
from mpi4py import MPI

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

send_value = rank + 1  # 1 for rank 0, 2 for rank 1, etc.
total = comm.reduce(send_value, op=MPI.SUM, root=0)

if rank == 0:
    print(f"[{rank}] Sum of ranks+1 = {total}")
```

Example: Distributed Array Sum

```
# sum_mpi.py
from mpi4py import MPI
import numpy as np

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

# Split array among processes
if rank == 0:
    data = np.arange(100, dtype='i')
    chunks = np.array_split(data, size)
else:
    chunks = None

# Scatter chunks to all processes
local_data = comm.scatter(chunks, root=0)
```

```
# Each process computes local sum
local sum = np.sum(local data)
# Gather all local sums at root
total sum = comm.reduce(local sum, op=MPI.SUM, root=0)
if rank == 0:
   print("Total sum:", total sum)
```

♦ When to Use MPI in Python?

- Large-scale data processing
- Scientific simulations
- Parallel numerical computations
- Where shared memory models (like multithreading) won't scale across machines

Things to Keep in Mind

- MPI excels on clusters; not always efficient on a single machine.
- Python has overhead—use compiled extensions (NumPy, Numba) for heavy lifting.
- Debugging is harder—consider writing unit tests for components before running them distributed.

How to setup in windows?

Microsoft MPI v10.1.3

Stand-alone, redistributable and SDK installers for Microsoft MPI

https://www.microsoft.com/en-us/download/details.aspx?id=105289

✓ To fix this on Windows, follow these steps:

1. Install Microsoft MPI

Download the Microsoft MPI (MS-MPI) from the official site:

https://www.microsoft.com/en-us/download/details.aspx?id=105289

- Install both:
 - MS-MPI Redistributable Package
 - o MS-MPI SDK

This will install mpiexec.exe and necessary DLLs.

2. Add MPI to System PATH (if needed)

If after install it still fails, manually add this to your PATH:

C:\Program Files\Microsoft MPI\Bin

3. Verify Installation

Open a new Command Prompt, and run:

mpiexec

You should see usage info instead of an error.

4. Install mpi4py (if not done yet)

pip install mpi4py

