# MPI4PY on Compute Canada Clusters

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#### Outline I

Who Am I

Python

How to Speed up Python

#### MPI4Y

Caveats
Hello to the World
MPI4PY Communications

#### Compute Canada Clusters

Submitting Jobs

Live Demo

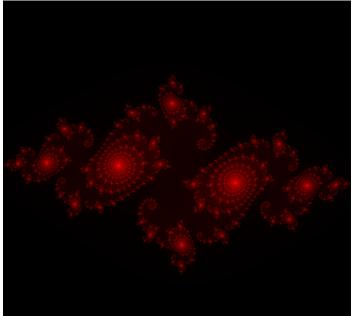
Summary

References

#### Who Am I

- BSc in Computer Science from Wilfirid Laurier University
  - ► Majors in Computer Electronics Applied Mathematics
  - Minor in Biochemistry
- MSc in Biology from University of Waterloo
  - Petabase-scale Data Mining Identifies Novel Clostridial Species and Neurotoxins Associated with Ancient Human DNA
- Experience programming in several languages
  - Python, Bash, C/C++, Java/Groovey, PIC Assembly, MATLAB/R
- Over ten years of experience using various HPC systems
  - Carbon Compute Cluster : Argonne National Labs
  - ► Sharcnet : University of Waterloo, University of Toronto
  - Compute Canada (now the Digital Research Alliance of Canada): BC, Ontario, Quebec
- Thousands of compute hours running
  - Black white boxes
- custom scripts / pipelines / visualizations
- ► Run Linux as my daily OS
- ► Used Python with MPI4PY for my CP431 projects with Julia sets for my final project

# Julia Sets



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All this can make developing new code "easier" and faster.

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You should use the right tool for the job. Sometimes that's Python and sometimes that's C/C++ or even Assembly.

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See Dr. Pawel Pomorski Slides Numpy, Cython, Multiprocessing and Numba.

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- Multi-threading/processing doesn't scale beyond one computer

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What you lose in performance, you gain in shorter development time, and potentially fewer lost neurons.

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  - ► MPI\_Init() is called when you import the module
  - MPI\_Finalize() is called before the Python process ends

#### Hello to the World

```
#!/usr/bin/env python3
11 11 11
Parallel Hello World
11 11 11
from mpi4py import MPI
size = MPI.COMM WORLD.Get size()
rank = MPI.COMM_WORLD.Get_rank()
name = MPI.Get_processor_name()
print('Greetings. I am process {} of {} on {}'.format
```

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See Texas tutorials for more options

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- MPI4Py handles the packing and unpacking for user defined data types
- ► Two types of communication: Blocking and non Blocking

### Collective Communications

Used to send messages to multiple processes at once. Eg. Broadcast, Scatter, Gather, Reduction

See Texas tutorials for examples and specifics

# Transferring Python Data

```
#!/usr/bin/env python3
11 11 11
Send Python Data
11 11 11
from mpi4py import MPI
comm = MPI.COMM WORLD
rank = comm.Get rank()
if rank == 0:
   data = \{ 'a': 7, 'b': 3.14 \}
   comm.send( data , dest=1, tag=11)
   print( 'Message set, data is : ', data )
elif rank == 1:
   data = comm.recv( source=0, tag=11)
   print( 'Message Received, data is : ', data )
```

# Transferring Numpy Data I

```
#!/usr/bin/env python3
11 11 11
Send Numpy Data
11 11 11
from mpi4py import MPI
import numpy
comm = MPI.COMM_WORLD
rank = comm.Get_rank()
# pass explicit MPI data types
if rank == 0:
   data = numpy.random.randint(0,100, size=(2, 4),
       → dtype='i')
   comm.Send( [ data, MPI.INT ] , dest=1, tag=77)
```

### Transferring Numpy Data II

```
elif rank == 1:
   data = numpy.empty ( (2,4), dtype='i' )
   comm.Recv( [ data, MPI.INT ], source=0 , tag=77)
   print(data)
# automatic MPI data type discovery
if rank == 0:
   data = numpy.random.randint(0,100, size=(2, 3, 4)
       → , dtvpe='i')
   comm.Send( data, dest=1, tag=13)
elif rank == 1 :
   data = numpy.empty((2,3,4), dtype='i')
   comm.Recv(data, source=0, tag=13)
   print(data)
```

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- Use comm.lprobe(source=target, tag=11) to check for incoming if you wanted to use irecv

### Compute Canada Clusters

- Documentation at https://docs.computecanada.ca/wiki/Graham
- Python documentation at https://docs.computecanada.ca/wiki/Python
- MPI documentation at https://docs.computecanada.ca/wiki/MPI

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```
#!/bin/bash
#SBATCH --account=def-some-user
#SBATCH --nodes=2
#SBATCH --ntasks=8 # number of MPI processes
#SBATCH --mem-per-cpu=512M # memory; default unit is
   \hookrightarrow megabytes
#SBATCH --time=0-00:15 # time (DD-HH:MM)
module load StdEnv/2020 intel/2020.1.217 openmpi
   \hookrightarrow /4.0.3 scipy-stack/2022a mpi4py/3.1.2 python
   \hookrightarrow /3.9
#export OMP_NUM_THREADS=$SLURM_CPUS_PER_TASK #Use

    → this with multi-threaded code

echo 'Hello World Example'
srun ./hello_world.py # mpirun or mpiexec also work
echo 'Send Python Data Example'
srun ./send_python_data.py
echo 'Send Numpy Data Example'
srun ./send_numpy_data.py
                                       4 D > 4 B > 4 B > 4 B > 9 Q P
```

### Live Demo

What every could go wrong:)

start with from mpi4py import MPI

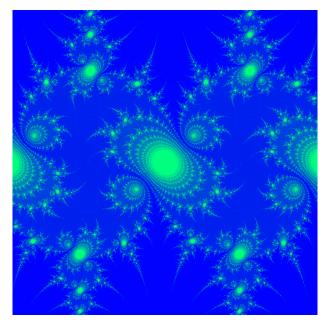
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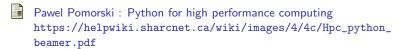
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- sbatch run\_job.sh or salloc

# Questions



### References



Pawel Pomorski: Speeding up Python code with Numba https: //helpwiki.sharcnet.ca/wiki/images/4/4e/Numba\_webinar.pdf

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