# HPC Python Programming Hands-On Exercises

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These are the instructions for the hands-on for the Python for HPC sesion of the Petascale Computing Institute 2019. There are 4 hands-on exercises, each of which should take about 10 minutes. The first two exercises have optional parts for students that finish the main part of that exercise early.

### Setting Up On Blue Waters

To get set up for the session using the Blue Waters supercomputer at NCSA, perform the following steps.

- 1. Login to Blue Waters:
- \$ ssh -Y USERNAME@bwbay.ncsa.illinois.edu
- 2. Copy code for this session:
- \$ cp -r /u/training/instr041/hpcpython/ \$HOME
- 3. Request resources on a compute node (if you want accurate timing results):
- \$ qsub -I -X -lnodes=1:ppn=16,walltime=2:00:00
- 4. Setup the environment:
- \$ source /u/training/instr041/hpcpyenv/activate

#### Setting Up On Cori

To get set up for the session using the Cori supercomputer at NERSC, perform the following steps.

- 1. Login to Blue Waters:
- \$ ssh -Y USERNAME@cori.nersc.gov
- 2. Copy code for this session:
- \$ cp -r /global/homes/r/rzon/hpcpython/ \$HOME
- 3. Request resources on a compute node (if you want accurate timing results):
- \$ salloc -N 1 -n 16 -C haswell -q interactive -t 02:00:00
- 4. Setup the environment:
- \$ source /global/homes/r/rzon/hpcpyenv/activate

#### Setting Up On Your Own Computer

Alternatively, you can use the code of the exercises on your own computer by downloading https://support.scinet.utoronto.ca/~rzon/petapy.zip .

You will need to have Python 3 installed with the following packages:

```
numpyscipynumexprmatplotlibpsutilline_profilermemory_profilermpi4pycythonnumba
```

The examples are aimed for a Linux environment with g++/gfortran/make, but if you do not have that, you can still use the Python code in the zip file if the above packages are installed.

The code also comes with a setup script, so if you're on Linux, open a terminal, go to the code directory and type the command 'source setup'. This will try to setup a virtual environment with the above packages and will try to compile the rarray and pgplot libraries.

# 1. Hands-on 1: Profiling

### Profile the auc\_serial.py code

- Consider the Python code for computing the area under the curve in auc\_serial.py.
- Put the code in a wrapper function. Make sure it still works!
- Add @profile to the main function.
- Run this through the line profiler and see what line(s) cause the most cpu usage.

# Optional (if you finish early): Profile the diff2d.py code

- Reduce the resolution and runtime in diff2dparams.py, i.e., increase dx to 0.5, and decrease runtime to 2.0.
- In the same file, ensure that graphics=False.
- Add @profile to the main function in diff2d.py
- Run this through the line profiler and see what line(s) cause the most cpu usage.

# 2. Hands-on 2: Vectorizing

#### Vectorize the auc serial.py code

- Copy the Python code in auc\_serial.py to a new file auc\_vector.py
- Remove any Oprofile decorators.
- Reexpress the code using numpy arrays.
- Make sure you are using vectorized operations.
- Measure the speed-up (if any) with /usr/bin/time.

## Optional (if you finish early): Vectorize the slow numpy code

If you are done with the auc example, try this:

- Copy the file diff2d\_slow\_numpy.py to diff2d\_numpy.py.
- Replace the indexed loops with whole-array vector operations.
- Measure the speed-up (if any) with /usr/bin/time.

## 3. Hands-on 3: Parallelize Area-under-the-curve

- Use the numexpr package to parallelize the auc\_vector.py code.
- Measure the speed-up using /usr/bin/time with up to 16 threads.

## 4. Hands-on 4: MPI area under the curve

The exercise is for you to parallelize auc\_vector.py with the mpi4py package:

- Take your numpy-vectorized code in auc\_vector.py.
- Divide the computation over MPI tasks.
- Measure speed-up for up to 8 processes.