

Interacting with NERSC

- NERSC User Engagement Group
 - User Engagement, outreach
 - User Training and Documentation
- NERSC Consulting & Account Support
 - User Tickets
 - User Appointments
- NERSC Operations
- NERSC User Group (NUG)



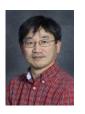




Consulting & Account Support Team





































































Day 1 Recap Programing Environments and Compilation Best Practices:

- Use module spider
- Use compiler wrappers –CC, cc, and ftn– with PrgEnv modules
- Verify CMake builds with ccmake
- man files are the source of truth, i.e. man intro_mpi
- See examples builds at https://github.com/NERSC/community-software
- More questions? Need help? ... http://help.nersc.gov/







Day 1 Recap Running Jobs (Scheduling) Best Practices/Tips:

- One job with a large allocation
 - Per node priority ageing is the highest
 - Can get scheduled first
- Shorter time length jobs
 - Easier to schedule as backfill
 - Use a workflow manager
- Choose the right time from Slurm
 - Balance between enough runtime
 - Waiting in the queue for a long job
- Utilize Workflow tools
 - GNU Parallel, Parsl, Fireworks, etc.
- Make Use of Job Script Generator
 - Provides sample option for slurm batch scripts







Python users, welcome to NERSC!

What we'll cover in this intro to Python at NERSC:





- Using Python at NERSC
 - python/conda modules
 - creating environments/installing packages
 - tips for parallel python
- Getting started with Python on GPUs!
 - high-level overview











Using Python at NERSC







How can I use Python at NERSC?

To get started, load the python module:

```
> module load python/3.11
(nersc-python) > python
Python 3.11.7 | packaged by conda-forge | (main, Dec 23 2023,
14:43:09) [GCC 12.3.0] on linux
Type "help", "copyright", "credits" or "license" for more
information.
>>> print("Welcome to NERSC New User Training Feb 2024!!!")
Welcome to NERSC New User Training Feb 2024!!!
                                  https://docs.nersc.gov/development/languages/python/
```







The NERSC Python Module

The NERSC python module provides python via a conda environment. It's convenient for simple use cases that only need relatively common Python packages in the scientific computing.

```
> module load python/3.11
(nersc-python)> conda list
 packages in environment at /global/common/software/.../nersc-python:
                          Version
 Name
                                                     Build
                                                            Channel
matplotlib
                          3.8.2
                                           py311h38be061 0
                                                               conda-forge
                           1.26.3
                                           py311h64a7726 0
                                                               conda-forge
numpy
                           1.11.4
                                           py311h64a7726 0
                                                               conda-forge
scipy
```







Other options for using Python at NERSC

Create a custom conda environment:

```
> module load conda
> conda create -n myenv python=3.11 numpy scipy
> conda activate myenv
(myenv)> python
Python 3.11.7 (main, Dec 23 2023, 14:43:09) [GCC 12.3.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>>
```

perlmutter> which python /usr/bin/python

This is not the Python you're looking for!

Use Python from a container:

```
> shifter --image=docker:library/python:latest python
Python 3.11.4 (main, Aug 16 2023, 19:58:34) [GCC 12.2.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
```









The NERSC Conda Module

conda is an environment and package management tool that is very popular in the scientific python community.

conda environments are great for creating isolated and reproducible software environments for your projects. The conda package manager is great for installing and resolving package dependencies for your projects.

Loading the conda module initializes conda. There is no need to run "conda init" or initialize conda in your ~/.bashrc (or similar shell startup file).







Package installation tips:

- Most packages installed via conda or pip should work at NERSC
 - packages installed via conda can come from different "channels". Channels are specified with "-c defaults" or "-c conda-forge".
 - In many cases it's fine to mix packages from different channels and/or pip but this can sometimes lead to version conflicts. Check the packages installed in your environment with "conda list"
- Some python packages should be compiled with the "compiler wrappers" available on the system. For example, mpi4py (see next slide) and potentially h5py (if you're using parallel IO features).
- cudatoolkit: module vs conda package:
 - Some GPU-enabled packages installed from conda-forge will install cudatoolkit into your conda environment. This may conflict with the cudatoolkit module that is loaded by default.







Building and using mpi4py

- mpi4py provides a Python interface to MPI
- mpi4py is available via module load python
- This mpi4py is with CUDA support in cray-mpich
- To install mpi4py with CUDA support in cray-mpich, follow this recipe:

```
> module load PrgEnv-gnu cudatoolkit craype-accel-nvidia80 conda
> conda create -n mpi4py-gpu python=3.9 -y
> conda activate mpi4py-gpu
> MPICC="cc -shared" pip install --force-reinstall --no-cache-dir --no-binary=mpi4py mpi4py
```

 Be aware that with any CUDA-aware mpi4py, you must have cudatoolkit loaded, even for code that does not use the GPU







Use pip with caution 1



- Be careful with pip!!! pip maintains a local package cache which, but sometimes you don't want this.
- Packages installed with --user are not confined to a particular environment
 - If you use pip install --user <package>, it will install packages to the location specified by **PYTHONUSERBASE**, which may be set to something like \$HOME/.local/perlmutter/python-3.11
- Best practices for pip:
 - Install packages inside of a conda environment, not outside (don't use pip install --user <package>)
 - Use pip install --no-cache-dir --force-reinstall <package> (Did you notice this in our mpi4py recipe?)







Running Python at scale at NERSC

Python startup/module imports can put significant load on shared global filesystems, especially when running in parallel.

To avoid this we recommend:

- Use a container (run with shifter or podman)
- Use /global/common/software/<project>
 - use the -p/--prefix option when creating conda environments:

```
> conda create -p /global/common/software/cproject>/envs/myenv python=3.11
```

- > conda activate /global/common/software/project>/envs/myenv
- Avoid \$HOME
- Avoid \$CFS







Other common parallel Python pitfalls

- Unexpected oversubscription due to indirect parallelism.
 - numpy uses OpenMP threading under the hood.
 - When using multiple processes, make sure num_processes_per_node x OMP_NUM_THREADS does not exceed the number of physical CPU cores per node.
 - Default worker pool size in multiprocessing.
 - It's common for Python applications to use the value of `os.cpu_count()` to set a default value for the number of processes / workers. This does not account for cpu-binding.
 - For nested parallel applications, you should specify the number of workers to use and not trust the default.







Using Python at NERSC Summary

- Use conda environments (or containers!) for customizable Python sandboxes.
- Use the /global/common/software/oject> filesystem (or containers!)
 for better performance when running in parallel.
- Use the compiler wrappers to build packages such as mpi4py.
- Avoid running "conda init" which will hardcode conda initialization in your shell startup file (\$HOME/.bashrc)
- Be careful using pip.
- Avoid using the system python from /usr/bin!
- Watch out for defaults which may unexpectedly lead to oversubscription of resources.









Using Python on GPUs







Getting started with GPUs in Python

- NumPy and SciPy do not utilize GPUs out of the box
- There are many Python GPU frameworks out there:
 - o "drop in" replacements for numpy, scipy, pandas, scikit-learn, etc
 - CuPy, RAPIDS
 - "machine learning" libraries that also support general GPU computing
 - PyTorch, TensorFlow, JAX
 - "I want to write my own GPU kernels"
 - Numba, CUDA Python
 - multi-gpu / multi-node / distributed memory:
 - mpi4py+X, dask, cuNumeric
- Many of these GPU libraries have adopted the <u>CUDA Array</u> <u>Interface</u> which makes it easier to pass array-like objects stored in GPU memory between the libraries
- There is also effort in the community to standardize around a common Python array API



```
numpy:
               mean(a, axis=None, dtype=None, out=None, keepdims=<no value>)
dask.array:
               mean(a, axis=None, dtype=None, out=None, keepdims=<no value>)
               mean(a, axis=None, dtype=None, out=None, keepdims=False)
cupy:
               mean(a, axis=None, dtype=None, out=None, keepdims=False)
jax.numpy:
               mean(a, axis=None, dtype=None, out=None, keepdims=False)
mxnet.np:
               s.mean(axis=None, keepdims=False, dtype=None, out=None)
sparse:
torch:
               mean(input, dim, keepdim=False, out=None)
               reduce_mean(input_tensor, axis=None, keepdims=None, name=None,
tensorflow:
                           reduction_indices=None, keep_dims=None)
```







cudatoolkit dependency via module

> module load conda

Note: cudatoolkit module is loaded by default Current default version is cudatoolkit/11.7

```
> conda create --name cupy-demo python=3.11 numpy scipy
```

```
> conda activate cupy-demo
```

```
> pip install cupy-cuda11X -
```

> python

```
>>> import cupy as cp
```

>>> print(cp.array([1, 2, 3]))

[1 2 3]

Check your package documentation to see cudatoolkit compatibility requirements

See documentation at https://docs.nersc.gov/development/languages/python/using-python-perlmutter/







cudatoolkit dependency via conda

```
cupy conda-forge package will pull cudatoolkit
                                        dependencies into conda env
> module load conda
> module unload cudatoolkit
> conda create --name cupy-demo python=3.11 numpy scipy
> conda activate cupy-demo
> conda install -c conda-forge cupy
> python
                                              cupy conda-forge package will pull cudatoolkit
>>> import cupy as cp
                                              dependencies into conda env
>>> print(cp.array([1, 2, 3]))
[1 2 3]
```

See documentation at https://docs.nersc.gov/development/languages/python/using-python-perlmutter/







Is my code a good fit for a GPU?

CPUs → low latency GPUs → high throughput

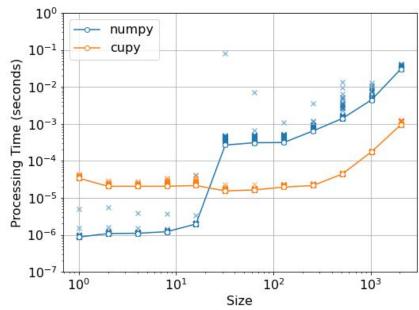
GPUs are likely a good fit if the following are true for your application:

- Performs computation using large arrays, matrices, or images
- Dataset can fit in GPU memory
 - (40GB for Perlmutter's A100 GPUs)
- IO is not a bottleneck

For more help choosing a GPU-accelerated Python framework:

https://docs.nersc.gov/development/languages/python/perlmutter-prep/

```
a = xp.random.rand(size, size)
b = xp.random.rand(size, size)
def f(a, b):
    return xp.dot(a, b)
```











Wrap Up





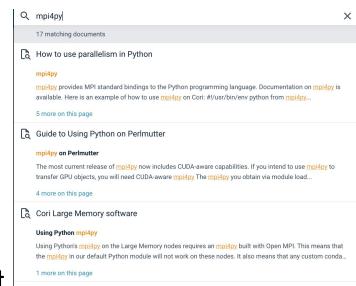


Best Practices & Where to get Python information

Utilize Conda

•

- Check out Python in NERSC docs:
 - Python at NERSC
 - Python on Perlmutter
 - Jupyter at NERSC
 - Try the search bar at <u>docs.nersc.gov</u>, it's pretty good!
- Can't find the answer? Submit a ticket at help.nersc.gov









Summary

- Welcome to NERSC!
- We are here to help you use Python productively on Perlmutter

If you have questions, please check our <u>docs.nersc.gov</u>

or file a ticket at help.nersc.gov







I has a question...

