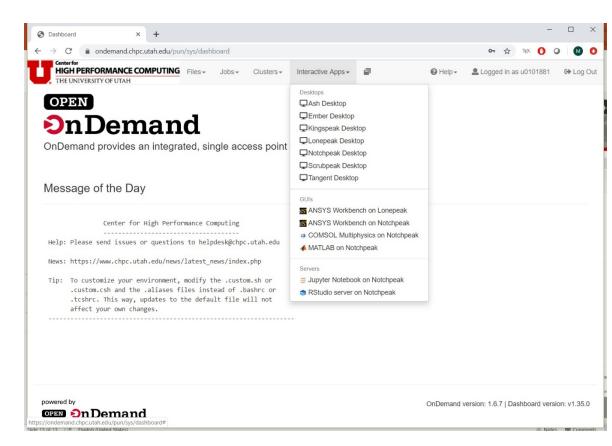


Open Ondemand Interactive apps - GUIs and servers



- Direct launch of a given application
- ANSYS, COMSOL, **MATLAB**
- Jupyter Notebook
- RStudio server
- Can set up others if needed

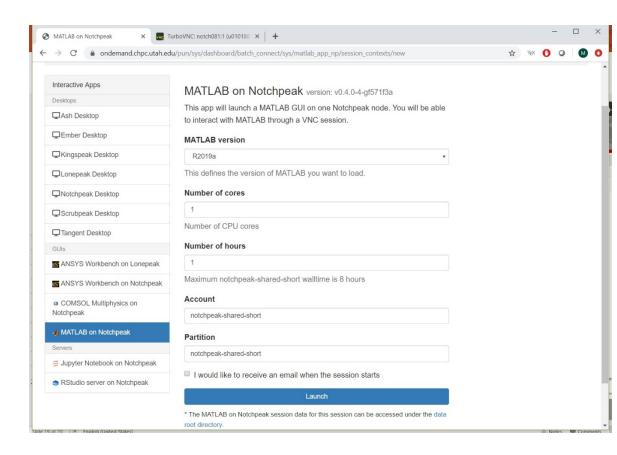




Interactive apps - MATLAB



- Same start parameters as in Interactive Desktop
- Plus option to choose MATLAB version
- Only on Notchpeak at the moment

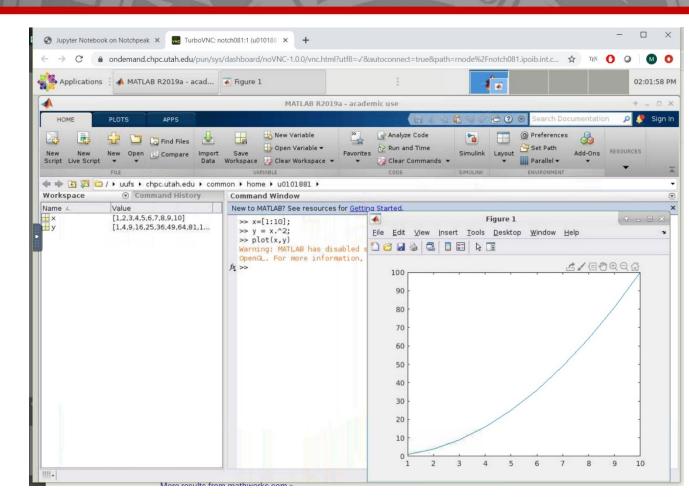




Interactive apps - MATLAB



- MATLAB GUI window
- Additional MATLAB windows appear over the GUI

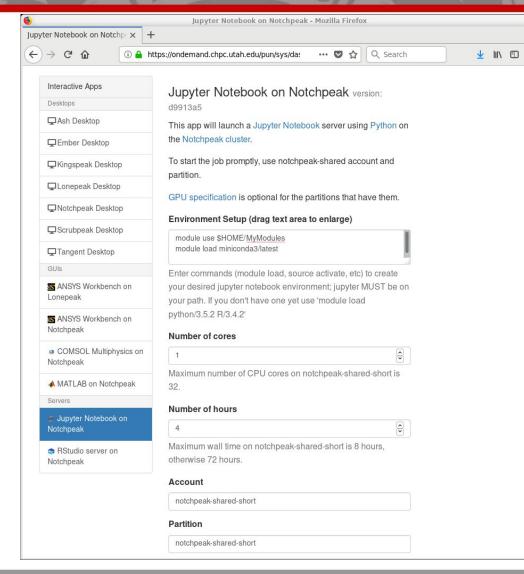




Interactive apps - Jupyter notebook



- Can also specify GPU but make sure to list the right account/partition
- Uses any Python module available including user installed miniconda
- Miniconda is recommended for user specific Python modules https://www.chpc.utah.edu/documentation/software/python-anaconda.php

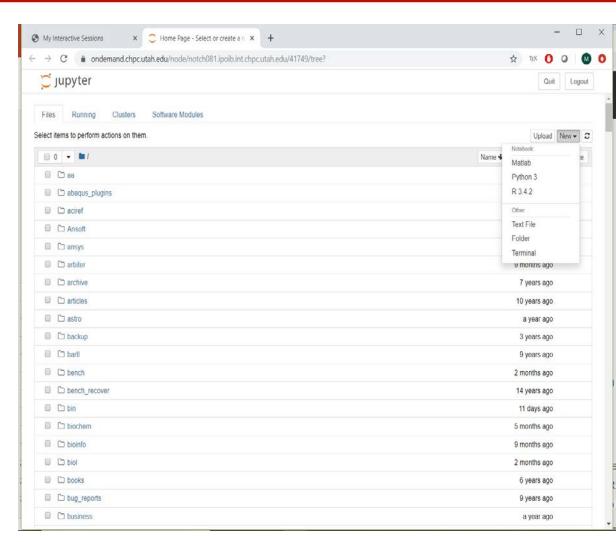




Interactive apps - Jupyter notebook



- Can open existing notebooks or create new ones
- CHPC's python/3.5.2 includes Matlab and R notebook support

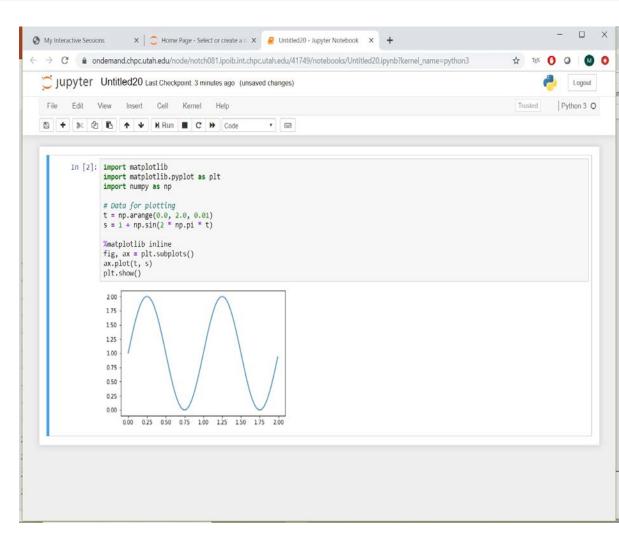




Interactive apps - Jupyter notebook



 The notebook is launched in another browser tab





Running Matlab



Interactive - Open OnDemand Matlab app

UNIVERSITY OF UTAH		
Home / My Interactive Sessions / Ma	ATLAB on Notchpeak	
Interactive Apps	MATLAB on Notchpeak	
Desktops		
☐ Ash Desktop	This app will launch a MATLAB GUI on one Notchpeak node. You will be able to interact with MATLAB through a VNC session.	
☐ Ember Desktop	MATLAB version	
☐ Kingspeak Desktop	R2019a v	
☐ Lonepeak Desktop	This defines the version of MATLAB you want to load.	
☑Notchpeak Desktop	Number of cores	
☐Scrubpeak Desktop	1	
☐ Tangent Desktop	Number of CPU cores	
GUIs	Number of hours	
MANSYS Workbench on Lonepeak	Ĭ	
ANSYS Workbench on Notchpeak	Maximum notchpeak-shared-short walltime is 8 hours	
COMSOL Multiphysics on	Account	
Notchpeak	notchpeak-shared-short	
MATLAB on Notchpeak	Partition	
Servers	notchpeak-shared-short	
5 Jupyter Notebook on Notchpeak	■ I would like to receive an email when the session starts	
n RStudio server on Notchpeak		
	Launch	



Running Matlab



Batch - serial job script

```
#!/bin/tcsh
#SBATCH --time=1:00:00 # walltime, abbreviated by -t
                    # number of cluster nodes, abbreviated by -N
#SBATCH --nodes=1
#SBATCH -o slurm-%j.out-%N # name of the stdout, using the job number (%j) and the
first node (%N)
                  # number of MPI tasks, abbreviated by -n
#SBATCH --ntasks=1
# additional information for allocated clusters
#SBATCH --partition=notchpeak-shared-short # partition, abbreviated by -p
# just in case purge the old modules and load Matlab module
module purge
module load matlab
# run matlab program via the run_matlab script
matlab -nodisplay -r run_matlab_serial -logfile matlab_serial.log
```

• Submit as sbatch run_matlab_serial.slr



Running Matlab



Batch - helper script

```
% path where is the matlab program we want to run
% addpath path_to_my_matlab_script
% execute the matlab program
loop_serial
% exit from Matlab - if we don't exit Matlab the job sits around idle and wasting time till it runs out of walltime
exit
```

Actual serial Matlab code

```
tic
n = 400;
A = 1000;
a = zeros(1,n);
for i = 1:n
        a(i) = max(abs(eig(rand(A))));
end
toc
```

 NOTE - code like this does not run in parallel so it may not run efficiently - more on this later





3F.B. 28TH

Center for High-Performance Computing

Introduction to Parallel Computing

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Overview



- Types of parallel computers.
- Parallel programming options.
- OpenMP, OpenACC, MPI
- Debugging, profiling and libraries
- Higher level languages
- Summary, further learning.



How to compute faster



- Faster CPU clock speed
 - Higher voltage = more heat not sustainable
- Work distribution
 - Vectorization process more than one value at a time
 - Parallelization spread work over multiple processing elements
 - Specialization application specific processors (ASIC), programmable logic (FPGA)



Single processor:

SISD – single instruction single data.

Multiple processors:

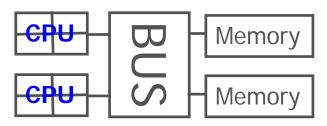
- SIMD single instruction multiple data.
- MIMD multiple instruction multiple data.
 - Shared Memory
 - Distributed Memory
- Current processors combine SIMD and MIMD
 - Multi-core CPUs w/ SIMD instructions (AVX, SSE)
 - GPUs with many cores and SIMT



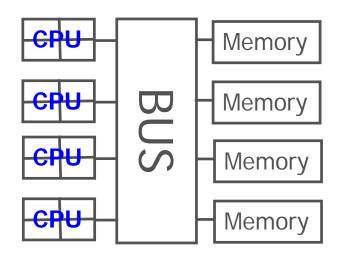


- All processors have access to local memory
- Simpler programming
- Concurrent memory access
- More specialized hardware
- Representatives: Linux clusters nodes 8-32 core GPU nodes

Dual quad-core node



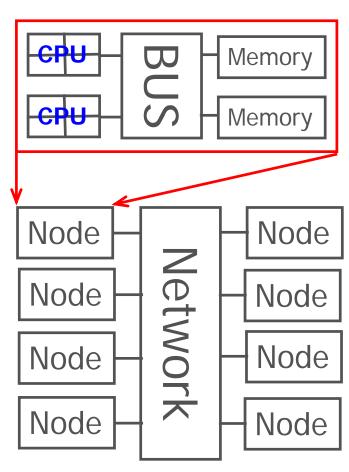
Many-CPU node (e.g. SGI)







- Process has access only to its local memory
- Data between processes must be communicated
- More complex programming
- Cheap commodity hardware
- Representatives:
 Linux clusters



8 node cluster (64 cores)



Ways of program execution

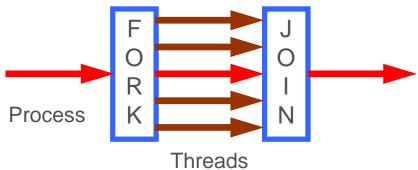


Process (task)

Entity that executes a program – has its own memory space, execution sequence, is independent from other processes

Thread

Has own execution sequence but shares memory space with the original process - a process may have many threads





Parallel programming options



Shared Memory

- Threads
 - POSIX Pthreads, OpenMP (CPU, MIC), OpenACC, CUDA (GPU)
- Processes
 - message passing, independent processes

Distributed Memory

- Independent processes
- Message passing libraries
 - General MPI, PVM, language extensions (Co-array Fortran, UPC. ...)

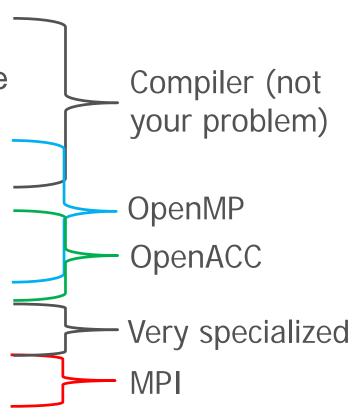
Higher level programming languages (Python, R, Matlab) do a combination of these approaches under the hood.



Types of parallelism in a nutshell



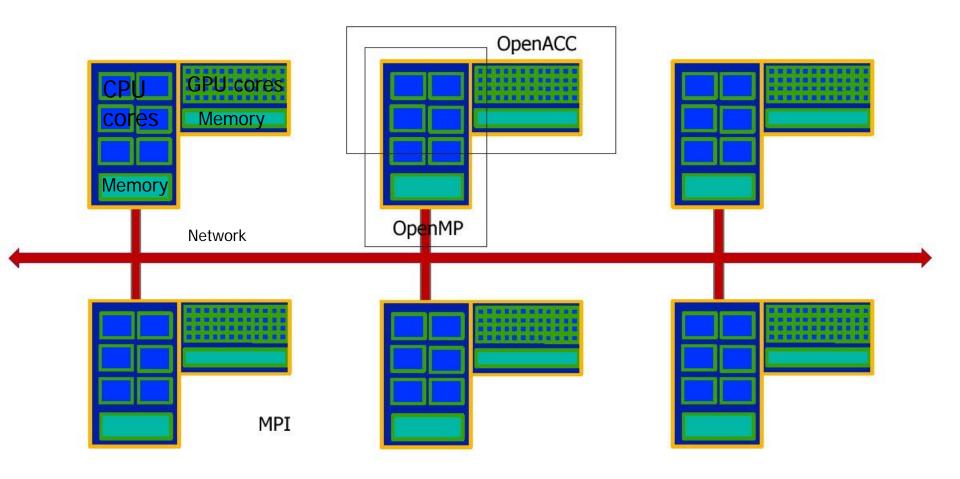
- Instruction level (ILP)
 - Instruction pipelining, speculative execution, branch prediction, ...
- Vector (SIMD)
- Multi-core/Multi-socket SMP
- Accelerators (GPU, MIC)
- FPGA, ASIC
- Distributed clusters





Mapping onto the hardware







OpenMP basics



- Compiler directives to parallelize (CPU or GPU)
- Fortran source code comments
 !\$omp parallel/!\$omp end parallel
- C/C++ #pragmas

 #pragma omp parallel
- Small set of subroutines
- Degree of parallelism specification
- OMP_NUM_THREADS or omp_set_num_threads(INTEGER n)





- Compiler directives to offload to GPU
- Fortran source code comments !\$acc kernels/!\$acc end kernels
- C/C++ #pragmas #pragma acc kernels
- Small set of subroutines
- Degree of parallelism specification
- OMP_NUM_THREADS or omp_set_num_threads(INTEGER n)



MPI Basics



- Communication library
- Language bindings:
- C/C++-int MPI_Init(int argv, char* argc[])
- Fortran MPI_Init(INTEGER ierr)
- Quite complex (100+ subroutines)
 but only small number used frequently
- User defined parallel distribution





- Complex to code
- Slow data communication
- Ported to many architectures
- Many tune-up options for parallel execution

- Easy to code
- Fast data exchange
- Memory access (thread safety, data dependencies)
- Limited usability
- Limited user's influence on parallel execution



- saxpy vector addition:
- $\overline{z} = a\overline{x} + \overline{y}$
- simple loop, no cross-dependence, easy to parallelize

```
subroutine saxpy_serial(z, a, x, y, n)
integer i, n
real z(n), a, x(n), y(n)

do i=1, n
  z(i) = a*x(i) + y(i)
enddo
return
```



OpenMP program example



```
subroutine saxpy_parallel_omp(z, a, x, y, n)
integer i, n
real z(n), a, x(n), y(n)
!$omp parallel do
                                        FORK
do i=1, n
  z(i) = a*x(i) + y(i)
enddo
                                        JOIN
return
$ gcc -fopenmp saxpy.f
$ export OMP NUM THREADS=16
```

\$./a.out



OpenMP caveats



- Data dependencies
 - Private (thread-local) variables

$$x = a(i)$$

b(i) = c + x

Flow dependence – rearrangement

$$a(i) = a(i+1) + x$$

Reduction (sum over threads)

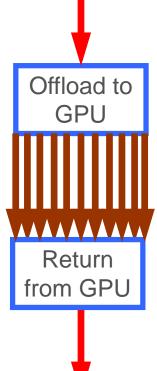
- Scheduling
 - What runs on what thread schedule, task,...
- Advanced features
 - Thread affinity (to CPU core)
 - Vectorization
 - Accelerator offload



OpenACC program example



```
subroutine saxpy_parallel_oacc(z, a, x, y, n)
integer i, n
real z(n), a, x(n), y(n)
!$acc kernels datain(x,y) dataout(z)
do i=1, n
  z(i) = a*x(i) + y(i)
enddo
return
$ pgcc -acc -Minfo=accel saxpy.f
```



- \$ pgaccelinfo To verify that GPU is available
- \$./a.out



OpenACC caveats



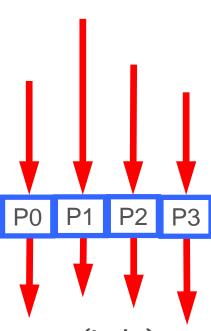
- Data dependencies (Like in OpenMP)
- Data locality
 - Transfers from host to GPU and back take time
 - need to minimize them

```
#pragma acc data [copyin, copyout, create,...]
```

- Parallel regions
 - More explicit execution control (warps, threads)
 #pragma acc parallel
- Procedure calls
 - If procedure is executed on the GPU #pragma acc routine



```
subroutine saxpy_parallel_mpi(z, a, x, y, n)
integer i, n, ierr, my_rank, nodes, i_st, i_end
real z(n), a, x(n), y(n)
call MPI_Init(ierr)
call MPI Comm rank(MPI COMM WORLD, my rank, ierr)
call MPI_Comm_size(MPI_COMM_WORLD, nodes, ierr)
i st = n/nodes*my rank+1
i end = n/nodes*(my rank+1)
do i=i st, i end
  z(i) = a*x(i) + y(i)
enddo
call MPI Finalize(ierr)
```



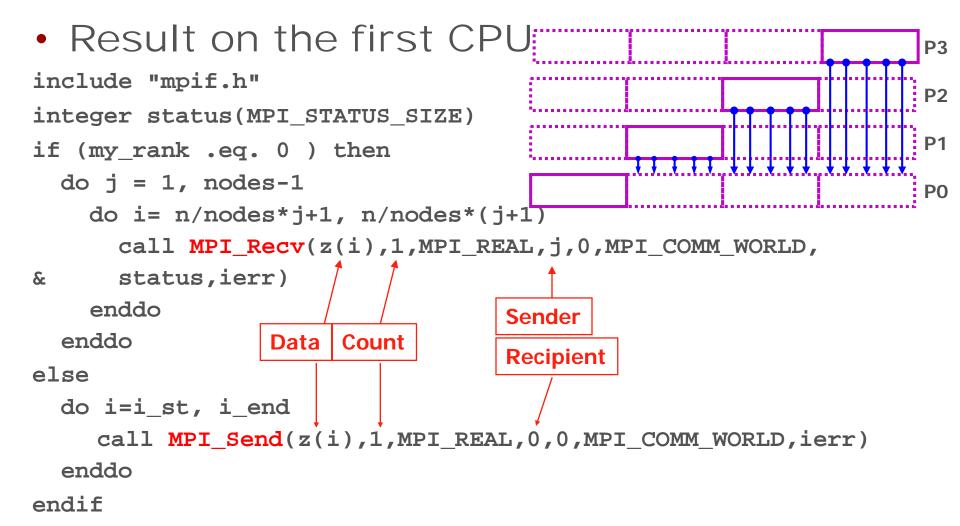
z(i) operation on 4 processes (tasks)

z(1	z(n/4+1	z(2*n/4+1	z(3*n/4+1
	` .		_(0
n/4)	2*n/4)	3*n/4)	n)
	=,	,	,

return











zi(i)

zi(i)

zi(i)

zi(i)

Process 0

Process 1

Process 2

Process 3

Collective communication

```
real zi(n)
j = 1
do i=i_st, i_end
  zi(j) = a*x(i) + y(i)
  j = j +1
```

enddo Send data

Receive data

z(i)

Root process

Result on all nodes

No root process



MPI distributions



- Different networks
 - Ethernet
 - InfiniBand
 - Intel OmniPath
 - most MPI distributions now come with multiple networks support
- Several distributions follow the MPI standard
 - MPICH, MVAPICH2
 - Intel MPI, Cray MPI,...
 - OpenMPI
 - Ensure that build and run is done with the same distribution (ABI compatibility)



But wait, my program is not in C or Fortran

Interpreted languages are popular

Matlab, Python, R

Each has some sort of parallel support, but most likely it will not perform as well as using OpenMP or MPI with C/Fortran.

Try to parallelize (and optimize) your Matlab/Python/R code and if it's still not enough consider rewriting in C++ or Fortran.

Cluster running options for Matlab, Python, R

- Using parallelization in the program run through interactive or batch job
 - multi-threading and/or multi-processing packages (parfor, mpi4py, R parallel, Rmpi, ...)
- Using built in job submission
 - Matlab Parallel Server, rslurm, python Dask, snakemake
- Independent calculations in parallel
 - launching concurrent calculations in a job



Matlab



Threads

 Built in Matlab functions. Vector/matrix operations threaded (and vectorized) through Intel MKL library, many other functions also threaded

Tasks (processes)

- Parallel Computing Toolbox allows for task based parallelism
- Parallel Server can distribute tasks to multiple nodes
- Great for independent calculations, when communication is needed uses MPI under the hood

https://www.chpc.utah.edu/documentation/software/matlab.php



Matlab tasks



Parallel program

```
function t = parallel_example
parfor idx = 1:16 
A(idx) = idx;
end
Will launch loop iterations on multiple workers
```

Parallel worker pool on a single machine

```
poolobj=parpool('local',8); ← Starts multiple workers pool
parallel_example;
delete(poolobj);
```

Parallel pool on a cluster



Matlab examples



- Parallel worker pool on a single node
 - best run from a SLURM job
 loop_parallel_onenode.m, run_matlab_onenode.m,
 run_matlab_onenode.slr https://github.com/CHPC-UofU/USU-lectures
 - sbatch run_matlab_onenode.slr
- Parallel worker pool on a multiple nodes
 - must run from inside of Matlab
 - start Matlab on interactive node inside of a FastX session

```
ml matlab matlab &
```

loop_parallel.m, parallel_multinode.m

```
parallel_multinode
```





Threads

- Under the hood threading with specially built (or Microsoft) R for vector/matrix operations using MKL
- parallel R library

Tasks (processes)

- parallel R library (uses multicore for shared and snow for distributed parallelism)
- Parallelized *apply functions, e.g. mclapply
- Rmpi library provides MPI like functionality
- Many people run multiple independent R instances in parallel

https://www.chpc.utah.edu/documentation/software/r-language.php



Parallel R on a cluster



Load libraries

```
library(parallel)
library(foreach)
library(doParallel)
```

```
hostlist.txt comes from a job script
srun -n $SLURM_NTASKS > hostlist.txt
```

Start R cluster

Run parallel loop

```
r <- foreach(icount(trials), .combine=rbind) %dopar% {}
```

Stop R cluster

stopCluster(cl)



R examples



- Parallel R on one node
 - best run from a SLURM job
 parallel-onenode-iris.R, R-parallel-onenode-iris.slr-https://github.com/CHPC-UofU/USU-lectures
 - sbatch R-parallel-onenode-iris.slr
- Parallel R multiple nodes
 - must specify list of nodes where R workers run parallel-multinode-iris.R, R-parallel-multinode-iris.slr-
 - sbatch R-parallel-onenode-iris.slr
- Submit SLURM job directly from R rslurm
 - SLURM-aware apply function, some issues with results collection
 - <u>rslurm-example.R</u>



Python



Threads

- No threads in Python code because of GIL (Global Intepreter Lock)
- C/Fortran functions can be threaded (e.g. NumPy -Anaconda)

Tasks (processes)

- Several libraries that use MPI under the hood, most popular is mpi4py
- More-less MPI function compatibility, but slower communication because of the extra overhead
- Also many other data-parallel libraries, e.g. Dask

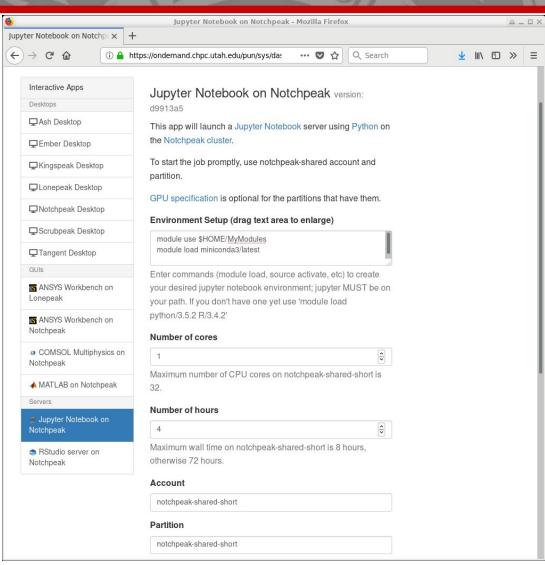
https://www.chpc.utah.edu/documentation/software/python.php



Python - Jupyter



- Several options
 listed at
 https://www.chpc.utah.
 edu/documentation/soft
 ware/jupyterhub.php
- The easiest is to use Open
 OnDemand





Python tasks



- Our personal favorite is to ignore all the Python parallel efforts, divide the data into independent parts and run multiple Python processes on parts of the data concurrently
- Only works if data can be split
- Use various approaches for independent parallel calculations listed at https://www.chpc.utah.edu/documentation/software/seriall-jobs.php
- More on this later



Python- Dask



- With relatively small effort one can use Dask
- Install Miniconda

```
wget https://repo.continuum.io/miniconda/Miniconda3-latest-Linux-
x86_64.sh
bash ./Miniconda3-latest-Linux-x86_64.sh -b -p
$HOME/software/pkg/miniconda3
mkdir -p $HOME/MyModules/miniconda3
cp
/uufs/chpc.utah.edu/sys/installdir/python/modules/miniconda3/latest.lua
$HOME/MyModules/miniconda3
```

Use own miniconda and install Jupyter and Dask

```
module use $HOME/MyModules
module load miniconda3/latest
conda install jupyter dask
```

- Start Open OnDemand Jupyter notebook
 - log into ondemand.chpc.utah.edu with CHPC credentials



Python- Dask



- Go to Interactive Apps Jupyter Notebook on notchpeak
- In the Environment Setup text box, put:

```
module use $HOME/MyModules
module load miniconda3/latest
```

- Use notchpeak-shared-short for account and partition, and select your choice of CPU cores and walltime hours (within the listed limits). Then hit Launch to submit the job.
- Once the job starts, hit the blue Connect to Jupyter button
- Open one of the following notebooks: <u>dask_embarrass.ipynb</u>, <u>dask_slurmcluster.ipynb</u>, <u>dask_slurm_xarray.ipynb</u>
- DASK also allows to submit jobs to SLURM (last 2 examples)



- Different approaches based on the nature of the calculations
 - Runtime length, variability, number of calculations
- Similar runtime, small calculation count
 - Shell script in a SLURM job

```
#!/bin/bash
for (( i=0; i < $SLURM_NTASKS ; i++ )); do
   /path_to/myprogram $i &
done
wait.</pre>
```

- srun -multi-prog

```
srun --multi-prog my.conf
cat my.conf
0-11 ./example.sh %t
```

https://www.chpc.utah.edu/documentation/software/serial-jobs.php



Variable runtime



- Mini-scheduler inside of a job
 - to launch calculations till all are done
 - GNU Parallel https://www.gnu.org/software/parallel/
 - TACC Launcher https://www.tacc.utexas.edu/research-development/tacc-software/the-launcher
- Workflow managers
 - More on this later
 - Makeflow, Swift, Snakemake, Pegasus
- Distributed computing resources
 - Open Science Grid https://opensciencegrid.org/



Single executable across desktops and clusters



- MPICH, MVAPICH2 and Intel MPI are cross-compatible using the same ABI
 - Can e.g. compile with MPICH on a desktop, and then run on the cluster using MVAPICH2 and InfiniBand
- Intel and PGI compilers allow to build "unified binary" with optimizations for different CPU platforms
 - But in reality it only works well under Intel compilers
- On a desktop

```
module load intel mpich
mpicc -axCORE-AVX512,CORE-AVX2,AVX program.c -o program.exe
mpirun -np 4 ./program.exe
```

On a cluster

```
srun -N 2 -n 24 ...
module load intel mvapich2
mpirun -np $SLURM_NTASKS ./program.exe
```

 https://www.chpc.utah.edu/documentation/software/singleexecutable.php



Debuggers



- Useful for finding bugs in programs
- Several free
 - gdb GNU, text based, limited parallel
 - ddd graphical frontend for gdb
- Commercial that come with compilers
 - pgdbg PGI, graphical, parallel but not intuitive
 - pathdb, idb Pathscale, Intel, text based
- Specialized commercial
 - totalview graphical, parallel, CHPC has a license
 - ddt Distributed Debugging Tool
 - Intel Inspector XE memory and threading error checker
- How to use:
- http://www.chpc.utah.edu/docs/manuals/software/par_ devel.html

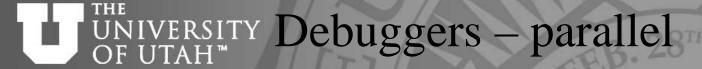


- Parallel debugging more complex due to interaction between processes
- DDT is the debugger of choice at CHPC
- Expensive but academia get discount
- How to run it:
 - compile with -g flag
 - run ddt command
 - fill in information about executable, parallelism, ...
- Details:

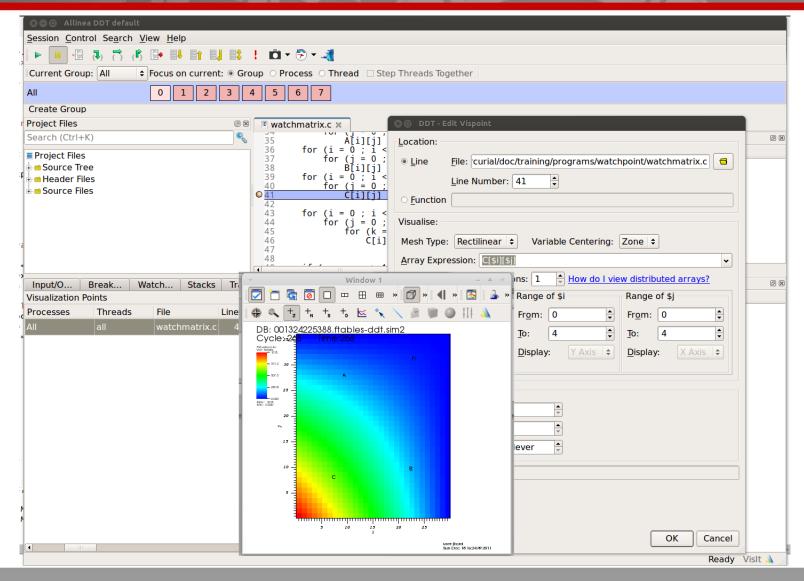
https://www.chpc.utah.edu/documentation/software/debugging.php

Further information

https://www.allinea.com/products/ddt









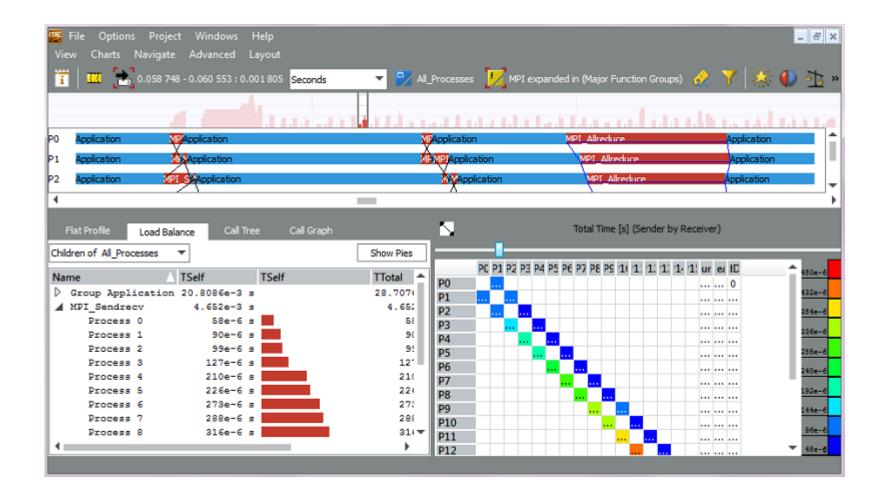


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- Measure performance of the code
- Serial profiling
 - discover inefficient programming
 - computer architecture slowdowns
 - compiler optimizations evaluation
 - gprof, pgprof, pathopt2, Intel tools
- Parallel profiling
 - target is inefficient communication
 - Intel Trace Collector and Analyzer, AdvisorXE,
 VTune











- Use libraries for common operationsSerial
 - BLAS, LAPACK linear algebra routines
 - MKL, ACML hardware vendor libraries
- Parallel
 - Scalapack, Petsc, Fftw
 - MKL dense and sparse matrices
- Design a new code around existing library
 - PETSc, Trilinos,...



Summary



- Shared vs. Distributed memory parallelism
- OpenMP, OpenACC and MPI for low level parallelism
- There are tools for debugging, profiling
- Different approaches for higher level languages
- Many ways to run independent calculations in parallel



To learn more



- Videos of recent workshops
- XSEDE HPC Summer Boot Camp
 - OpenMP, OpenACC, MPI
 - https://www.youtube.com/XSEDETraining
- Petascale Computing Institute
 - Wide range of parallel programming topics
 - https://bluewaters.ncsa.illinois.edu/bw-petascalecomputing-2019/agenda