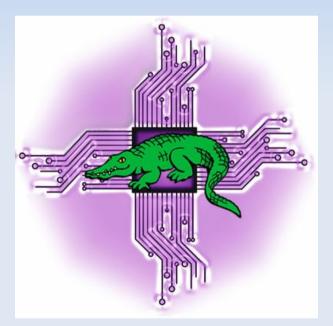
The Spartan HPC System at the University of Melbourne

COMP90024 Cluster and Cloud Computing



University of Melbourne, March 27, 2019

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Outline of Lecture

"This is an advanced course but we get mixed bag: students that have 5+ years of MPI programming on supercomputers, to students that have only done Java on Windows."

- Some background on supercomputing, high performance computing, parallel computing, scientific computing (there is overlap, but they're not the same thing).
- An introduction to Spartan, University of Melbourne's HPC/Cloud hybrid system
- Logging in, help, and environment modules.
- Job submission with Slurm workload manager; simple submissions, multicore, job arrays, job dependencies, interactive jobs.
- Parallel programming with shared memory and threads (OpenMP) and distributed memory and message passing (OpenMPI)
- Tantalising hints about more advanced material on message passing routines.

Supercomputers

'Supercomputer" arbitrary term. In general use it means any single computer system (itself a contested term) that has exceptional processing power for its time. One metric is the number of floating-point operations per second (FLOPS) such a system can carry out.

Supercomputers, like any other computing system, have improved significantly over time. The Top500 list is based on FLOPS using LINPACK - HPC Challenge is a broader, more interesting metric. The current number #1 system is Summit from the Oak Ridge National Laboratory in the United States.

1994: 170.40 GFLOPS 1996: 368.20 GFLOPS 1997: 1.338 TFLOPS 1999: 2.3796 TFLOPS 2000: 7.226 TFLOPS 2004; 70.72 TFLOPS 2005: 280.6 TFLOPS 2007: 478.2 TFLOPS 2008: 1.105 PFLOPS 2009: 1.759 PFLOPS 2010: 2.566 PFLOPS 2011: 10.51 PFLOPS 2012: 17.59 PFLOPS 2013: 33.86 PFLOPS 2014: 33.86 PFLOPS

2015: 33.86 PFLOPS 2016: 93.01 PFLOPS 2017: 93.01 PFLOPS

2018: 143.00 PFLOPS (200.795 Rpeak)



High Performance Computing

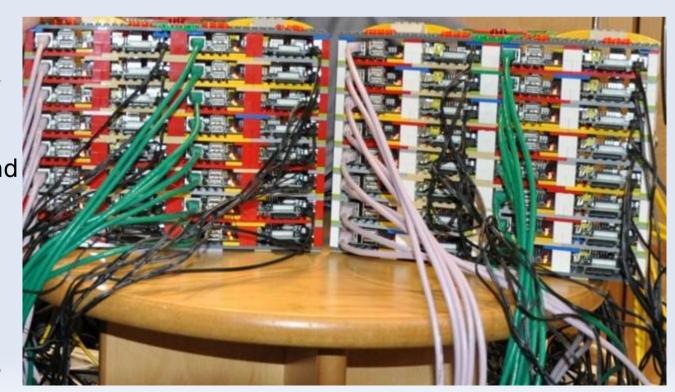
High-performance computing (HPC) is any computer system whose architecture allows for above average performance. A system that is one of the most powerful in the world, but is poorly designed, could be a "supercomputer".

Clustered computing is when two or more computers serve a single resource. This improves performance and provides redundancy; typically a collection of smaller computers strapped together with a high-speed local network (e.g., Myrinet, InfiniBand, 10 Gigabit Ethernet). Even a cluster of Raspberry Pi with

Lego chassis (University of Southampton, 2012)!

Horse and cart as a computer system and the load as the computing tasks. Efficient arrangement, bigger horse and cart, or a teamster?

The clustered HPC is the most efficient, economical, and scalable method, and for that reason it dominates supercomputing.



Parallel and Research Programming

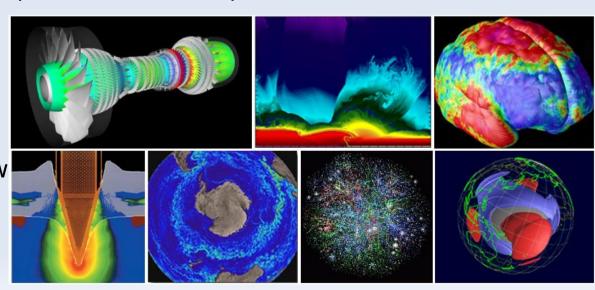
With a cluster architecture, applications can be more easily parallelised across them. Parallel computing refers to the submission of jobs or processes over multiple processors and by splitting up the data or tasks between them (random number generation as data parallel, driving a vehicle as task parallel).

Research computing is the software applications used by a research community to aid research. This skills gap is a major problem and must be addressed because as the volume, velocity, and variety of datasets increases then researchers will need to be able to process this data.

Computational capacity does have a priority (the system must exist prior to use), in order for that capacity to realised in terms of usage a skill-set competence must also exist. The the core issue is that high performance compute clusters

is just speed and power but also usage, productivity, correctness, and reproducibility. (image from Lawrence Livermore National Laboratory)

There is nascent research that show a strong correlation between research output and availability of HPC facilities. (Apon et al 2010)



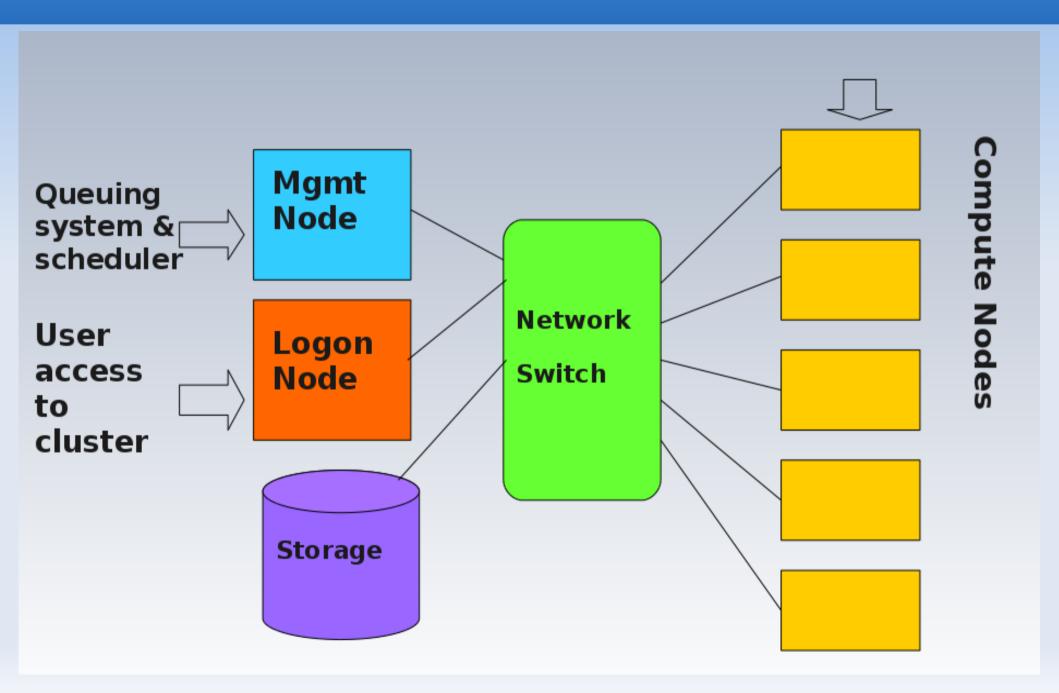
Some Local Examples

Researchers from Monash University, the Peter MacCallum Cancer Institute in Melbourne, the Birkbeck College in London, and VPAC in 2010 unravelled the structure the protein perforin to determine how pathogenic cells are attacked by white blood cells.

In 2015 researchers from VLSCI announced how natural antifreeze proteins bind to ice to prevent it growing which has important implications for extending donated organs and protecting crops from frost damage.

In 2016 CSIRO researchers successfully manipulated the behaviour of Metallic Organic Frameworks to control their structure and alignment which provides opportunities for real-time and implantable medical electric devices.

HPC Cluster Design



It's A GNU/Linux World

From November 2017 onwards of the Top 500 Supercomputers worldwide, every single machine used Linux.

The command-line interface provides a great deal more power and is very resource efficient.

GNU/Linux scales and does so with stability and efficiency. Critical software such as the Message Passing Interface (MPI) and nearly all scientific programs are designed to work with GNU/Linux.

The operating system and many applications are provided as "free and open source", which means that not only are there are some financial savings, were also much better placed to improve, optimize and maintain specific programs.

Free or open source software (not always the same thing) can be can be compiled from source for the specific hardware and operating system configuration, and can be optimised according to compiler flags. There is necessary where every clock cycle is important.

Flynn's Taxonomy and Multicore Systems

It is possible to illustrate the degree of parallelisation by using Flynn's Taxonomy of Computer Systems (1966), where each process is considered as the execution of a pool of instructions (instruction stream) on a pool of data (data stream).

Over time computing systems have moved towards multi-processor, multi-core, and often multi-threaded and multi-node systems.

The engineering imperative to these systems comes down to heat. From the mid-2000s clock speed on CPUs have largely stalled.

Some trends include GPGPU development, massive multicore systems (e.g., The Angstrom Project, the Tile CPU with 1000 cores) and massive network connectivity and shared resources (e.g., Plan9 Operating System).

Instruction Streams many one MISD SISD Streams traditional von maybe pipelined Neumann single CPU computers computer MIMD. Date Date SIMD vector processors multicomputers fine-grained data. multiprocessors parallel computers

(Image from Dr. Mark Meyer, Canisius College)

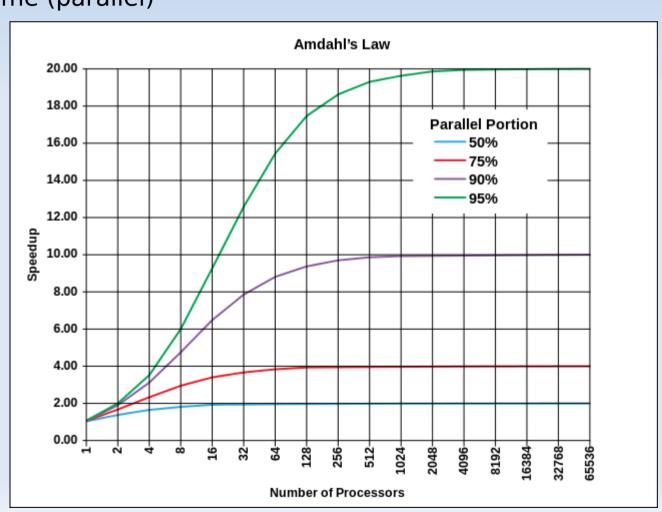
Limitations of Parallel Computation

Parallel programming and multicore systems should mean better performance. This can be expressed a ratio called speedup

Speedup (p) = Time (serial)/ Time (parallel)

Correctness in parallelisation requires synchronisation. Synchronisation and atomic operations causes loss of performance, communication latency.

Amdahl's law, establishes the maximum improvement to a system when only part of the system has been improved. Gustafson and Barsis noted that Amadahl's Law assumed a computation problem of fixed data set size.



UniMelb's HPC System: Spartan

A detailed review was conducted in 2016 looking at the infrastructure of the Melbourne Research Cloud, High Performance Computing, and Research Data Storage Services. University desired a 'more unified experience to access compute services'

Recommended solution, based on technology and usage, is to make use of existing NeCTAR Research cloud with an expansion of general cloud compute provisioning and use of a smaller "true HPC" system on bare metal nodes.

The 'bare metal' HPC component really will be laconi "Real" HPC is 1,356 cores, 21-43 GB per core. Intel E5-2643 v3 CPUs. 2x 40GbE network. "Cloud" partitions is almost 167 virtual machines with 2,004 cores. GPGPU partition has 1,752 cores and 4 P100 NVidIA GPUs per node (73 nodes, 3,584 CUDA Cores per card), plus departmental partitions (snowy, phi, mig, water and ashley).



Spartan is Small but Important

Spartan as a model of an HPC-Cloud Hybrid has been featured at Multicore World, Wellington, 2016, 2017; eResearchAustralasia 2016, Center for Scientific Computing (CSC) Goethe University Frankfurt, 2016, High Performance Computing Center (HLRS) University of Stuttgart, 2016, High Performance Computing Centre Albert-Ludwigs-University Freiburg, 2016; European Organization for Nuclear Research (CERN), 2016, Centre Informatique National de l'Enseignement Supérieur, Montpellier, 2016; Centro Nacional de Supercomputación, Barcelona, 2016, and the OpenStack Summit, Barcelona 2016. https://www.youtube.com/watch?v=6D1lobuCZqE

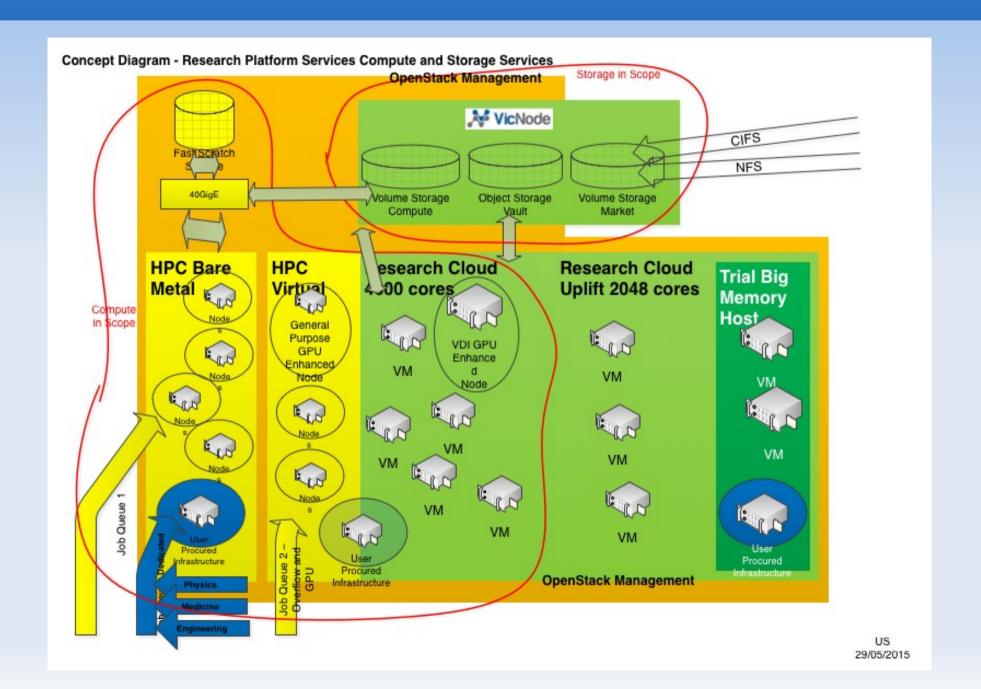
Also featured in OpenStack and HPC Workload Management in Stig Telfer (ed), The *Crossroads of Cloud and HPC: OpenStack for Scientific Research, Open Stack, 2016 http://openstack.org/assets/science/OpenStack-CloudandHPC6x9Booklet-v4online.pdf

Architecture also featured in:

Spartan and NEMO: Two HPC-Cloud Hybrid Implementations. 2017 IEEE 13th International Conference on e-Science, DOI: 10.1109/eScience.2017.70

The Chimera and the Cyborg, Hybrid Compute Advances in Science, Technology and Engineering Systems Journal Vol. 4, No. 2, 01-07, 2019

Moving Towards A New System



Spartan's Performance

Service	Network [Device I	Network	Protocol		Latency
(usecs)						
UoM HPC Traditional	Mellanox	56Gl	b In	nfiniband FD	R 1.17	
Legacy Edward HPC	Cisco Nexu	s I	10Gbe	TCP/IP	19	
Spartan Cloud nodes	Cisco Nexu	s 1	10Gbe	TCP/IP	60	
Spartan Bare Metal Me	llanox	40Gbe	TCP/II	P 6	.85	
Spartan Bare Metal Me	llanox	25Gbe	RDMA	A Ethernet 1	.84	
Spartan Bare Metal Me	llanox	40Gbe	RDMA	A Ethernet 1.	.15	
Spartan Bare Metal Me	llanox	56Gbe	RDMA	A Ethernet 1	.68	
Spartan Bare Metal Me	llanox	100Gbe	RDMA	A Ethernet 1	.3	

Job Cloud	Task	Resources	Contro	ol	HPC	Spartan
	Dial.	O sava Ciranla Nada	1.10.40	1.02.56	1.1	0.01
BWA	Disk	8 core Single Node		1:02:56		0:21
GROMACS	Compute	128 core Multinode	0:30:02	0:30:10	0:3	0:32
NAMD	Compute, I/O	128 core Multinode	1:11:41	1:00:46	1:5	5:54

These are figures from 2016; recently we've had a significant upgrade to network and processors - about 25% improvement.

Setting Up An Account and Training

Spartan uses its own authentication that is tied to the university Security Assertion Markup Language (SAML). The login URL is `https://dashboard.hpc.unimelb.edu.au/karaage`

Users on Spartan must belong to a project. Projects must be led by a University of Melbourne researcher (the "Principal Investigator") and are subject to approval by the Head of Research Compute Services.

Participants in a project can be researchers or research support staff from anywhere.

The University, through Research Platforms, has an extensive training programme for researchers who wish to use Spartan. This includes day-long courses in "Introduction to Linux and HPC Using Spartan", "Advanced Linux and Shell Scripting for High Performance Computing", "Parallel Programming On Spartan", "GPU Programming with OpenACC and CUDA". Coming soon "Linux Regular Expressions" and "From Spartan to NCI".

University of Melbourne is a key contributor to the International HPC Certification Program.

Logging In and Help

To log on to a HPC system, you will need a user account and password and a Secure Shell (ssh) client. Most HPC cluster administrators do not allow connections with protocols such as Telnet, FTP or RSH as they insecurely send passwords in plain-text over the network, which is easily captured by packet analyser tools (e.g., Wireshark). Linux distributions almost always include SSH as part of the default installation as does Mac OS 10.x, although you may also wish to use the Fugu SSH client. For MS-Windows users, the free PuTTY client is recommended. To transfer files use scp, WinSCP, Filezilla, and especially rsync.

Logins to Spartan are based on POSIX identity for the system

ssh your-username@spartan.hpc.unimelb.edu.au or

For help go to http://dashboard.hpc.unimelb.edu.au or check `man spartan`. Lots of example scripts at `/usr/local/common`

Need more help? Problems with submitting a job, need a new application or extension to an existing application installed, if job generated unexpected errors etc., an email can be sent to: `hpc-support@unimelb.edu.au`

The Linux Environment and Modules

Assumption here is that everyone has had exposure to the Linux command line. If not, you'd better get some! At least learn the twenty or so basic environment commands to navigate the environment, manipulate files, manage processes. Plenty of good online material available (e.g., my book "Supercomputing with Linux", https://github.com/VPAC/superlinux)

Environment modules provide for the dynamic modification of the user's environment (e.g., paths) via module files. Each module contains the necessary configuration information for the user's session to operate according according to the modules loaded, such as the location of the application's executables, its manual path, the library path, and so forth.

Modulefiles also have the advantages of being shared with many users on a system and easily allowing multiple installations of the same application but with different versions and compilation options. Sometimes users want the latest and greatest of a particular version of an application for the feature-set they offer. In other cases, such as someone who is participating in a research project, a consistent version of an application is desired. Having multiple version of applications available on a system is essential in research computing.

Modules Commands

Some basic module commands include the following:

module help

The command module help, by itself, provides a list of the switches, subcommands, and subcommand arguments that are available through the environment modules package.

module avail

This option lists all the modules which are available to be loaded.

module whatis <modulefile>
This option provides a description of the module listed.

module display <modulefile>

Use this command to see exactly what a given modulefile will do to your environment, such as what will be added to the PATH, MANPATH, etc. environment variables.

More Modules Commands

module load <modulefile>
This adds one or more modulefiles to the user's currer environment (some modulefiles load other modulefile)

module unload <modulefile>
This removes any listed modules from the user's current environment.

module switch <modulefile1> <modulefile2> This unloads one modulefile (modulefile1) and loads another (modulefile2).

module purge
This removes all modules from the user's environmen



In the Imod system as used on Spartan there is also "module spider" which will search for all possible modules and not just those in the existing module path and provide descriptions.

(Image from NASA, Apollo 9 "spider module")

Batch Systems and Workload Managers

The Portable Batch System (or simply PBS) is a utility software that performs job scheduling by assigning unattended background tasks expressed as batch jobs among the available resources.

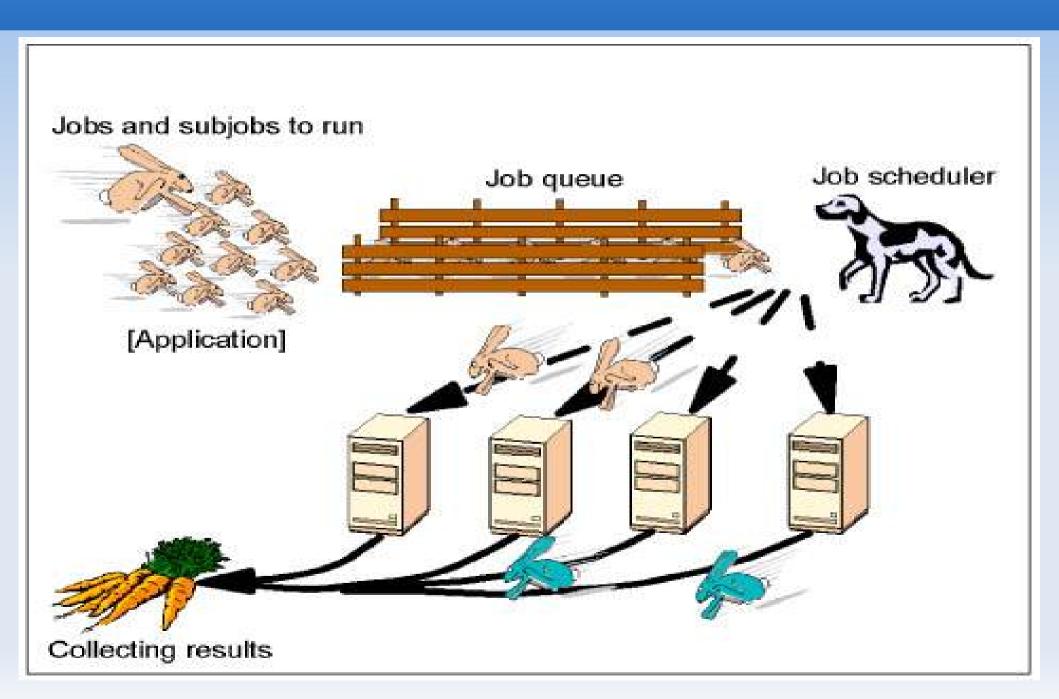
The original Portable Batch System was developed by MRJ Technology Solutions under contract to NASA in the early 1990s. In 1998 the original version of PBS was released as an open-source product as OpenPBS. This was forked by Adaptive Computing (formally, Cluster Resources) who developed TORQUE (Terascale Opensource Resource and QUEue Manager). Many of the original engineering team is now part of Altair Engineering who have their own version, PBSPro.

In addition to this the popular job scheduler Slurm (originally "Simple Linux Utility for Resource Management"), now simply called Slurm Workload Manager, also uses batch script where are very similar in intent and style to PBS scripts.

Spartan uses the Slurm Workload Manager. A job script written on one needs to be translated to another (handy script available pbs2slurm https://github.com/bjpop/pbs2slurm)

In addition to this variety of implementations of PBS different institutions may also make further elaborations and specifications to their submission filters (e.g., site-specific queues, user projects for accounting). (Image from the otherwise dry IBM 'Red Book' on Queue Management)

Submitting and Running Jobs



Submitting and Running Jobs

Submitting and running jobs is a relatively straight-forward process consisting of:

- 1) Setup and launch
- 2) Job Control, Monitor results
- 3) Retrieve results and analyse.

Don't run jobs on the login node! Use the queuing system to submit jobs.

1. Setup and launch consists of writing a short script that initially makes resource requests and then commands, and optionally checking queueing system.

Core command for checking queue: squeue | less

Alternative command for checking queue: showq -p cloud | less

Core command for job submission: sbatch [jobscript]

2. Check job status (by ID or user), cancel job.

Core command for checking job in Slurm: squeue -j [jobid]

Detailed command in Slurm: scontrol show job [jobid]

Core command for deleting job in Slurm: scancel [jobid]

3. Slurm provides an error and output files They may also have files for post-job processing. Graphic visualisation is best done on the desktop.

Simple Script Example

```
#!/bin/bash
#SBATCH --partition=cloud
#SBATCH --time=01:00:00
#SBATCH --nodes=1
#SBATCH --ntasks-per-node=1
module load my-app-compiler/version
my-app data
```

The script first invokes a shell environment, followed by the partition the job will run on (the default is 'cloud' for Spartan). The next four lines are resource requests, specifically for one compute node, one task.

After these requests are allocated, the script loads a module and then runs the executable against the dataset specified. Slurm also automatically exports your environment variables when you launch your job, including the directory where you launched the job from. If your data is a different location this has to be specified in the path!

After the script is written it can be submitted to the scheduler.

```
[lev@spartan]$ sbatch myfirstjob.slurm
```

Multithreaded, Multicore, and Multinode Examples

Modifying resource allocation requests can improve job efficiency.

For example shared-memory multithreaded jobs on Spartan (e.g., OpenMP), modify the --cpusper-task to a maximum of 8, which is the maximum number of cores on a single instance.

```
#SBATCH --cpus-per-task=8
```

For distributed-memory multicore job using message passing, the multinode partition has to be invoked and the resource requests altered e.g.,

```
#!/bin/bash
#SBATCH -partition=physical
#SBATCH --nodes=2
#SBATCH --ntasks-per-node=4
module load my-app-compiler/version
srun my-mpi-app
```

Note that multithreaded jobs *cannot* be used in a distributed memory model across nodes. They can however exist be conducted on distributed memory jobs which include a shared memory component (hybrid OpenMP-MPI jobs).

Arrays and Dependencies

Alternative job submissions include specifying batch arrays, and batch dependencies.

In the first case, the same batch script, and therefore the same resource requests, is used multiple times. A typical example is to apply the same task across multiple datasets. The following example submits 10 batch jobs with myapp running against datasets dataset1.csv, dataset2.csv, ... dataset10.csv

```
#SBATCH --array=1-10
myapp ${SLURM_ARRAY_TASK_ID}.csv
```

In the second case a dependency condition is established on which the launching of a batch script depends, creating a conditional pipeline. The dependency directives consist of `after`, `afterok`, `afternotok`, `before`, `beforeok`, `beforenotok`. A typical use case is where the output of one job is required as the input of the next job.

```
#SBATCH --dependency=afterok:myfirstjobid mysecondjob
```

Interactive Jobs

For real-time interaction, with resource requests made on the command line, an interactive job is called. This puts the user on to a compute node.

This is typically done if they user wants to run a large script (and shouldn't do it on the login node), or wants to test or debug a job. The following command would launch one node with two processors for ten minutes.

```
[lev@spartan interact]$ sinteractive --nodes=1 --ntasks-per-node=2
srun: job 164 queued and waiting for resources
srun: job 164 has been allocated resources
[lev@spartan-rc002 interact]$
```

X-Windows Forwarding

In almost all cases it is much better to do computation on the cluster and visualisation on a local system. In some cases however it is unavoidable to require x-windows forwarding.

It is best to login with the -Y option for security and then to login with -X to the login node. The compute node with then pass through the graphics via the login node to the desktop system.

Please note that you will need an x-windows client on your desktop for the visualisation.

[lev@cricetomys ~]\$ ssh lev@spartan.hpc.unimelb.edu.au -X

[lev@spartan]\$ sinteractive --nodes=1 --ntasks-pernode=2 -X

srun: job 602795 queued and waiting for resources

srun: job 602795 has been allocated resources

[lev@spartan-rc002 ~]\$ xclock



PBS, SLURM Comparison

User Commands	PBS/Torque	SLURM
Job submission	qsub [script_file]	sbatch [script_file]
Job submission	qdel [job_id]	scancel [job_id]
Job status (by job)	qstat [job_id]	squeue [job_id]
Job status (by user)	qstat -u [user_name]	squeue -u [user_name]
Node list	pbsnodes -a	sinfo -N
Queue list	qstat -Q	squeue
Cluster status	showq, qstatus -a	squeue -p [partition]
Environment		
Job ID	\$PBS_JOBID	\$SLURM_JOBID
Submit Directory	\$PBS_O_WORKDIR	\$SLURM_SUBMIT_DIR
Submit Host	\$PBS_O_HOST	\$SLURM_SUBMIT_HOST
Node List	\$PBS_NODEFILE	\$SLURM_JOB_NODELIST
Job Array Index	\$PBS_ARRAYID	\$SLURM_ARRAY_TASK_ID

PBS and SLURM Comparison

Job Specification	PBS	SLURM
Script directive	#PBS	#SBATCH
Queue	-q [queue]	-p [queue]
Job Name	-N [name]	job-name=[name]
Nodes	-l nodes=[count]	-N [min[-max]]
CPU Count	-l ppn=[count]	-n [count]
Wall Clock Limit	-l walltime=[hh:mm:ss]	-t [days-hh:mm:ss]
Event Address	-M [address]	mail-user=[address]
Event Notification	-m abe	mail-type=[events]
Memory Size	-I mem=[MB]	mem=[mem][M G T]
Proc Memory Size	-l pmem=[MB]	mem-per-cpu=[mem][M G T]

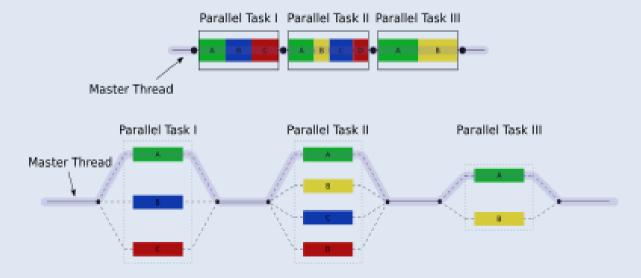
Shared Memory Parallel Programming

One form of parallel programming is multithreading, whereby a master thread forks a number of sub-threads and divides tasks between them. The threads will then run concurrently and are then joined at a subsequent point to resume normal serial application.

One implementation of multithreading is OpenMP (Open Multi-Processing). It is an Application Program Interface that includes directives for multi-threaded, shared memory parallel programming. The directives are included in the C or Fortran source code and in a system where OpenMP is not implemented, they would be interpreted as comments.

There is no doubt that OpenMP is an easier form of parallel programming, however it is limited to a single system unit (no distributed memory) and is thread-based rather than using message passing. Many examples in `/usr/local/common/OpenMP`.

(image from: User A1, Wikipedia)



Shared Memory Parallel Programming

```
#include <stdio.h>
#include "omp.h"
int main(void)
   int id:
   #pragma omp parallel num threads(8) private(id)
   int id = omp get thread num();
   printf("Hello world %d\n", id);
return 0;
program hello2omp
   include "omp lib.h"
   integer :: id
   !$omp parallel num threads(8) private(id)
     id = omp_get_thread num()
          print *, "Hello world", id
   !$omp end parallel
end program hello2omp
```

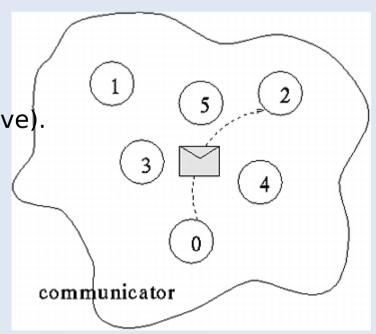
Distributed Memory Parallel Programming

Moving from shared memory to parallel programming involves a conceptual change from multi-threaded programming to a message passing paradigm. In this case, MPI (Message Passing Interface) is one of the most well popular standards and is used here, along with a popular implementation as OpenMPI.

The core principle is that many processors should be able cooperate to solve a problem by passing messages to each through a common communications network.

The flexible architecture does overcome serial bottlenecks, but it also does require explicit programmer effort (the "questing beast" of automatic parallelisation remains somewhat elusive).

The programmer is responsible for identifying opportunities for parallelism and implementing algorithms for parallelisation using MPI.



Distributed Memory Parallel Programming

```
#include <stdio.h>
#include "mpi.h"
int main( argc, argv)
int argc;
char **argv;
  int rank. size:
  MPI_Init( &argc, &argv );
  MPI Comm size (MPI COMM WORLD, &size);
  MPI_Comm_rank( MPI_COMM_WORLD, &rank );
  printf( "Hello world from process %d of %d\n", rank, size );
  MPI Finalize();
  return 0;
    Fortran MPI Hello World
    program hello
    include 'mpif.h'
    integer rank, size, ierror, tag, status(MPI STATUS SIZE)
    call MPI INIT(ierror)
    call MPI COMM SIZE(MPI COMM WORLD, size, ierror)
    call MPI_COMM_RANK(MPI_COMM_WORLD, rank, ierror)
    print*, 'node', rank, ': Hello world'
    call MPI FINALIZE(ierror)
    end
```

```
MyProgram: 0

MyProgram: 0

MyProgram: 0

MyProgram: 2

MyProgram: 3

(communication over network)
```

MPI Compilation and Job Scripts

The OpenMP example needs to be compiled with OpenMP directives. The OpenMP example cannot run across compute nodes; therefore it is best run on the "cloud" partition. The OpenMPI compilation needs to call the MPI wrappers.

```
module load OpenMPI/1.10.0-GCC-4.9.2
gcc -fopenmp helloomp.c -o helloompc
mpigcc mpihelloworld.c -o mpihelloworld
#!/bin/bash
#SBATCH -p cloud
#SBATCH --nodes=1
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=16
export OMP NUM THREADS=16
module load GCC/4.9.2
mpiexecu helloompc
#!/bin/bash
#SBATCH -p physical
#SBATCH --nodes=2
#SBATCH --ntasks=16
module load OpenMPI/1.10.2-GCC-4.9.2
mpiexec mpi-helloworld
```

MPJ Express for Java

Java can be compiled with MPI bindings; we have done this with OpenMPI/3.0.0 only. A more common option is to use MPJ-Express. The following "HellowWorld.java" program is compiled and executed.

```
import mpi.*;
public class HelloWorld {
public static void main(String args[]) throws Exception {
        MPI.Init(args);
        int me = MPI.COMM WORLD.Rank();
        int size = MPI.COMM WORLD.Size();
        System.out.println("Hi from <"+me+">");
        MPI.Finalize();
sinteractive --time=1:00:00 --nodes=1 --ntasks=2
module load MPJ-Express
javac -cp .:/usr/local/easybuild/software/MPJ-Express/0.44-goolf-
2015a-Java-9.0.4/lib/mpj.jar HelloWorld.java
mpjrun.sh -np 2 HelloWorld
```

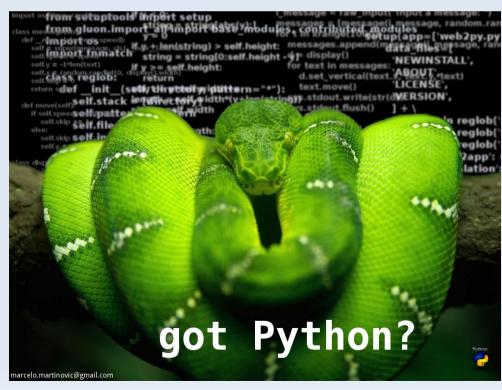
MPI4Py Express for Python

Python too has various MPI bindings available. The most common used is MPI4Py. Sample assignment data is provided from the 2016 and 2015 in the Spartan directory, `/usr/local/common`. As a package it can be simply imported (e.g., `from mpi4py import MPI`).

But remember! With environment modules with extensions you do not necessarily get all the packages/libraries/extensions that you might expect. See the README file for an explanation of how to review the extensions already installed.

Examples provided in the directory of various Python jobs with MPI bindings with Slurm submissions scripts for single-core, dual core, eight-core, sixteen-core (different partition), and two-nodes with four cores each.

The actual Python script for the above is `twitter_search_541635.py`



MPI Communication: A Game of Ping-Pong

A very popular and basic use of MPI Send and Recv routines is a ping-ping program. Why? Because it can be used to test latency within and between nodes and partitions if they have different interconnect (like on Spartan).

An example is given in `/usr/local/common/MPI` as `mpi-pingpong.c` with a job submission script that can be modified accord to the test case `mpi-pingpong.slurm`. There are some routines here which manage the communication in the ping-pong activity.

MPI_Status() MPI_Status is not a routine, but rather a data structure and is typically attached to an MPI_Recv() routine.

MPI_Request() A wrapper for MPI Requests such as wait, waitany, waitall, waitsome, start, cancel, startall.

MPI_Barrier() Enforces synchronisation between MPI processes in a group by placing a barrier on communication between groups. An MPI barrier completes after all group members have entered the barrier.

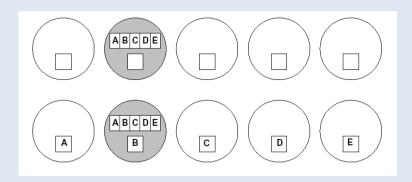
MPI_Wtime() - Returns an elapsed time as a floating-point number of seconds on the calling processor from an arbitrary time in the past.

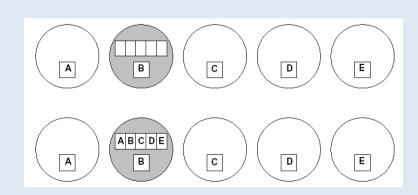
MPI Collective Communications

There are *many* other MPI routines which are *not* going to explored here! However just as a little taste one of the most popular is MPI collective communications and reduction operations. Collective communications include MPI_Broadcast, MPI_Scatter, MPI_Gather, MPI_Reduce, and MPI_Allreduce.

MPI_Bcast Broadcasts a message from the process with rank "root" to all other processes of the communicator, including itself. It is significantly more preferable than using a loop.

MPI_Scatter sends data from one task to all tasks in a group; the inverse operation of MPI_Gather. The outcome is as if the root executed n send operations and each process executed a receive. MPI Scatterv scatters a buffer in parts to all tasks in a group.

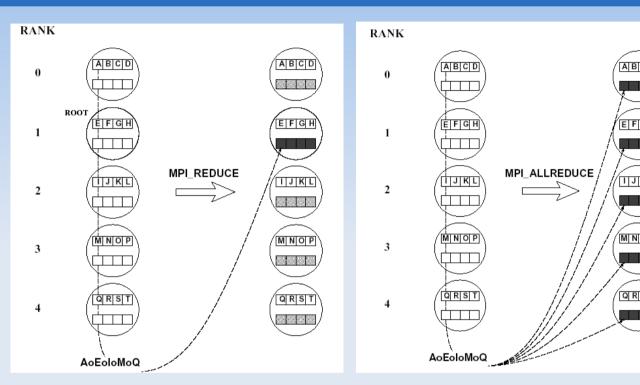




MPI Reduction Communications

MPI_Reduce performs a reduce operation (such as sum, max, logical AND, etc.) across all the members of a communication group.

MPI_All reduce conducts the same operation but returns the reduced result to all processors.



The general principle in Reduce and All Reduce is the idea of reducing a set of numbers to a small set via a function. If you have a set of numbers (e.g., [1,2,3,4,5]) a reduce function (e.g., sum) can convert that set to a reduced set (e.g., 15). MPI_Reduce takes in an array of values as that set and outputs the result to the root process. MPI_AllReduce outputs the result to all processes.

