

CRUK cluster introduction (III of III) Some advanced topics



Working with Lustre

- I. Revisit architecture
- 2. Stripes
- 3. Avoiding I/O Bottlenecks
- 4. Using System Cache

Lustre: Quick Review

Lustre is a massively parallel distributed file system

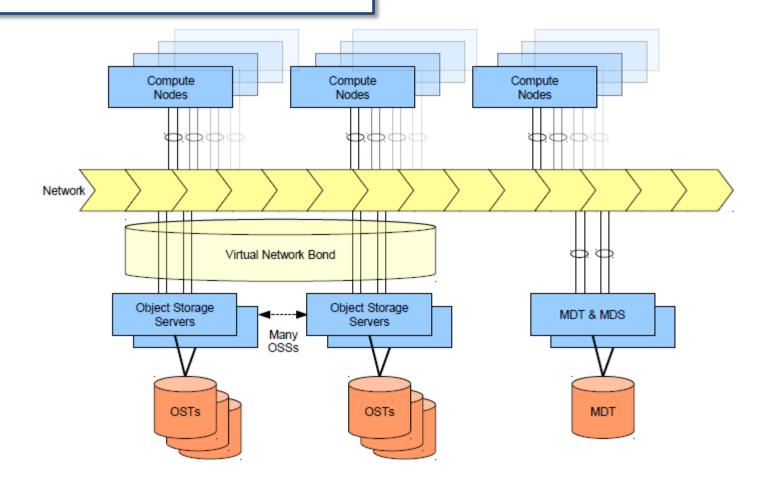
- Deployed in 7 out of 10 most powerful supercomputers
- POSIX compliant

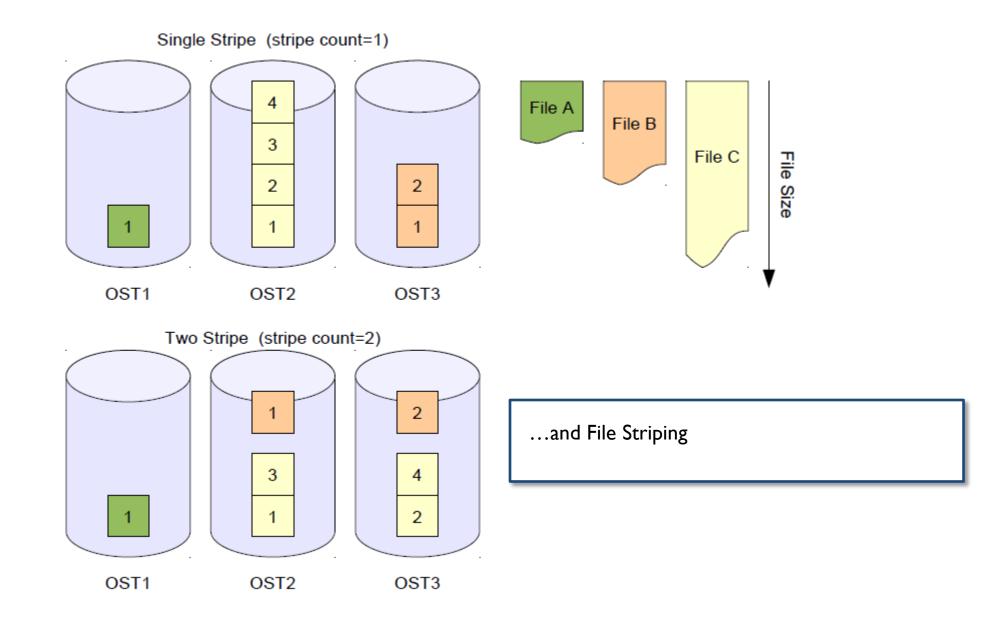
Lustre design paradigm concepts

- Separation of file meta-data and storage allocation
- Scalable data serving through parallel data striping
- Aggregates network bandwidth
- Distributed operation

'scratch' storage we (deliberately) don't back it up

Lustre Architecture ...





File Striping Large Files

Striping allows file size to exceed single OST size

- Currently 8TB LUNs

Performance benefit

- Aggregates I/O bandwidth to single large file
- In general, more strips improves performance
- Small overhead associated with open/closing striped files

Many jobs reading single file

- For example blastdb and maq reference data
- /lustre/reference_data/genomes

Many jobs reading & writing multiple large files

- Requires benchmarking

Many jobs writing to single file

- High bandwidth but requires careful coding (can be disastrous)

Set Stripe Information

Set per file or directory
Default is not to stripe
Only newly created files will be stripes

- Use cp (not mv) to migrate existing files

```
uk-cri-lcst01 ~ $ lfs setstripe <file|dir> --size <stripe_size>
--count <count> --index <index>
```

Where,
size = stripe size specified in k, m or g (0 default IMB)
count = OST stripe count (0 defaults 4 OST and -I over all OSTs)
index = OST index of first stripe (-I indicating default)

Read Stripe Information

Inspect file and directory stripe information with **lsf getstripe**

Using System Cache

Disk access is slow (no escape from this!)

- Memory access measured in a few nanoseconds
- Disc access measured in 10s of milliseconds

Linux uses free memory as cache

 Memory reclaimed as least used files expunged

"Pre-warming" cache

- Can dramatically increase I/O performance

```
uk-cri-lcst01 ~ $ cat largefile > /dev/null
uk-cri-lcst01 ~ $ grep searchString largefile
```

Avoiding Cache misses

Say you want to compare 3 sequences against 3 large databases:

"Out of the box" example

- sequence I vs database I No cache: disk read required
- sequence I vs database 2 Cache miss:
 disk read required
- sequence I vs database 3 Cache miss: disk read required
- sequence 2 vs database | No cache: data expunged from cache
- sequence 2 vs database 2 Cache miss:
 disk read required

Re-ordering to avoid cache misses

- sequence I vs database I No cache: disk read required
- sequence 2 vs database | Cache hit: data in cache
- sequence 3 vs database | Cache hit: data in cache
- sequence I vs database 2 No cache: disk read required
- sequence 2 vs database 2 Cache hit: data in cache

Avoiding Bad Performance

- · Interactive use
- Statting files can be slow (common with shared file systems)
- Avoid directly editing small files on lustre (keep to /home)
- Turn off "color Is" (stat required for each file/directory)

Random seeks

- Small random I/O extremely slow on lustre
- Avoid, as much as possible, running databases on lustre

e.g. mysql, sqlite, Berkeley DB etc

Limit number of files in directory

- 100,000s files in single directory bad (avoid, if possible)
- Use Ifs setstripe to confine all files to single OST

uk-cri-lcst01 ~ \$ lfs setstripe --count 1 directory

Parallel Workloads

What happens when you increase your dataset size?

- O(n), $O(n \log n)$, $O(n^2)$, $O(n^9)$
- Runtime & memory increase (O) with problem size (n)

Calculations are carried out in parallel

- Operating on principle that large calculations can often be divided into N smaller tasks
- These tasks are solved concurrently
- Time reduces to a function of $O(n^x)/N$

Parallel overheads

- communication, concurrency, parallel I/O introduce new overheads
- O(N), $O(N \log N)$, $O(N^2)$

Amdahl's Law

- Speed up of parallel application is ultimatley limited by the fixed runtime of any sequence sections
- There's no magic bullet for this, you may have to change your algorithm, or mix parallel and single node sections.

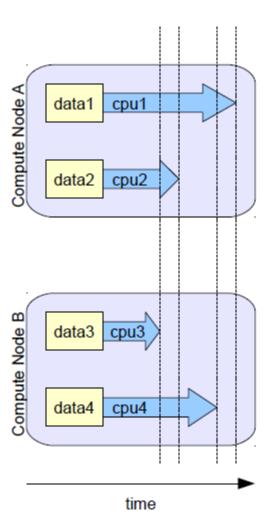
Simple Parallel

('Embarassing' or 'Trivial' in the computing science literature)

Solving many similar and independent tasks

- Analysis split into tasks
- Task assigned to one cpu
- No inter-task communication
- More throughput by running more tasks
- Task runtime varies

90% of bioinformatics codes fall into this model



Shared Memory

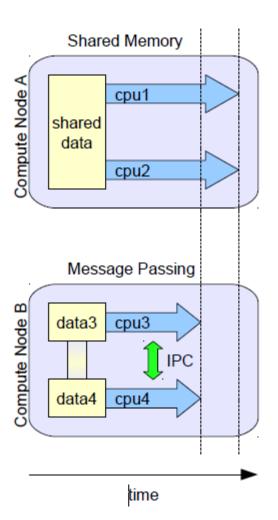
Shared tasks and memory

- Tasks assigned to cpus or cores
- Inter-task communication via shared memory
- Runtime decreases by adding more threads

Message Passing - local

Multiple processes communicate using O/S level systems. Code must be specifically written to exploit parallelism

- OpenMP/OpenMPI/etc



Message passing over network

Single task split across many compute nodes

- inter-machine communication (IMC)
- through MPI/OpenMP libraries again

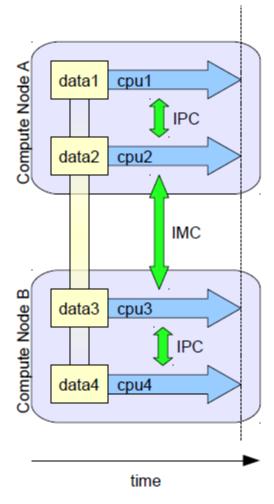
Hybrid models

Single task split across many compute nodes

- Mix SMP, local MP, network MP

Can be tricky to predict performance.

Your code may get quite complex...





Acknowledgements

Richard Bemrose Marc O'Brien Peter Maccallum Mark Dunning Ann Pajon

