Storage Lessons from HPC
Extreme Scale Computing Driving High Performance Data Storage

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Abstract

In this tutorial, we will introduce the audience to the lunatic fringe of extreme high-performance computing and its storage systems. The most difficult challenge in HPC storage is caused by millions (soon to be billions) of simultaneously writing threads. Although cloud providers handle workloads of comparable, or larger, aggregate scale, the HPC challenge is unique because the concurrent writers are modifying shared data.

We will begin with a brief history of HPC computing covering the previous few decades, bringing us into the petaflop era which started in 2009. Then we will discuss the unique computational science in HPC so that the audience can understand the unavoidability of its unique storage challenges. We will then move into a discussion of archival storage and the hardware and software technologies needed to store today’s exabytes of data forever. From archive we will move into the parallel file systems of today and will end the lecture portion of the tutorial with a discussion of anticipated HPC storage systems of tomorrow. Of particular focus will be namespaces handling concurrent modifications to billions of entries as this is what we believe will be the largest challenge in the exascale era.
Eight Decades of Production Weapons
Computing to Keep the Nation Safe

Maniac
IBM Stretch
CDC
Cray 1
Cray X/Y
CM-2

CM-5
SGI Blue Mountain
DEC/HP Q
IBM Cell Roadrunner
Cray XE Cielo

Cray Intel KNL Trinity
Ising DWave
Cross Roads

Los Alamos
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Simple View of our Computing Environment

- **Premier Machine**: 2PB Dram
  - Private IO Nodes
  - General IO Nodes
- **Capacity machines**: ~50-300 TB Dram
  - /sitescratch(s)
  - /home
  - /project
- **General IO Nodes
- **Capacity machines**: ~50-300 TB Dram
  - /sitescratch(s)
  - /home
  - /project
- **Private IB**
  - Local Scratch
    - 100 PB
    - 1 TB/sec
    - 1-4 Weeks
- **Site Scratch**
  - 10’s PB
  - 100 GB/sec
  - 1-4 Weeks
- **Site Scratch**
  - 10’s PB
  - 100 GB/sec
  - 1-4 Weeks
- **HPSS 100 PB**
  - 10 GB/sec
  - Forever
  - Parallel Tape with Disk Cache
- **NFS**
  - /home
  - /project
- **Analytics machine potentially disk full/big memory HDFS?**
  - /sitescratch(s)
  - /analytics (HDFS other)
  - Use HDFS – POSIX Shim for access to POSIX resources
- **Parallel load balanced movers**
  - /localscratch(s)
  - /sitescratch(s)
  - /home
  - /project
  - HPSS
  - /analytics (HDFS other)

100(s) Gbits/sec
Large Machines and Infrastructure

Trinity

- Haswell and KNL
- 20,000 Nodes
- Few Million Cores
- 2 PByte DRAM
- 4 PByte NAND Burst Buffer
  - ~ 4 Tbyte/sec
- 100 Pbyte Scratch PMR Disk File system
  - ~1.2 Tbyte/sec
- 30PByte/year Sitewide SMR Disk Campaign Store
  - ~ 1 Gbyte/sec/Pbyte (30 Gbyte/sec currently)
- 60 PByte Sitewide Parallel Tape Archive
  - ~ 3 Gbyte/sec

Pipes for Trinity Cooling

- 30-60MW
- Single machines in the 10k nodes and > 18 MW
- Single jobs that run across 1M cores for months
- Soccer fields of gear in 3 buildings
- 20 Semi’s of gear this summer alone
HPC Driving Industry

Flash Burst Buffers
- DDN
- Infinite Memory
- IBM
- EMC
- HPSS
- Data Warp

Parallel File Systems
- lustre
- panasas
- pNFS
- ceph

Parallel Archives
- HPSS
- DataTree
- CFS
- Other
- HDF
- Tokutek

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Ceph begins

We propose to improve both file system performance and functionality by building a storage system from object-based storage devices (OBSDs) connected by high-speed networks. The key advantage of OBSDs in a high-performance environment is the ability to delegate low-level block allocation and synchronization for a given segment of data to the device on which it is stored, leaving the file system to decide only on which OBSD a given segment should be placed. Since this decision is quite simple and allows massive parallelism, each OBSD need only manage concurrency locally, allowing a file system built from thousands of OBSDs to achieve massively parallel data transfers. Additionally, OBSDs can each manage their own storage consistency, removing the need to run a system-wide consistency check that could take days on a petabyte-scale traditional file system.
HPC Simulation Background

Link scales

Meshes

Methods

Structured Unstructured Resolutions

1D 2D 3D

Meshes

Lagrangian Eulerian

ALE

AMR

http://eng-cs.syr.edu/research/mathematical-and-numerical-analysis

http://media.archnumsoft.org/10305/

http://web.cs.ucdavis.edu/~ma/VolVis/amr_mesh.jpg
Scaling and Programming

Scaling

http://media.archnumsoft.org/10305/

Strong scaling: fixed total problem size
Weak scaling: fixed work per task

Process Parallel Dominates the Past

Collectives

Point to Point
Current Programming Model Directions
Combining Process Parallel, Data Parallel, Async Tasking, and Threading

MPI+ Threads

MPI+ Data Parallel

Async Tasking

Domain Specific Languages

Others: OpenShmem, CAF, Chapel, X10 are possible as well
Figure 1: An example of an APEX simulation science workflow.
Key observation: a grind (for strong-scaling apps) traverses all of memory.
Not Just Computation

failure
HPC environment

- Big difference from cloud: parallel, tightly coupled, extremely simple nodes to lower jitter and job failure due to tightly coupled behavior (one code syncs between all neighbors every 1 millisecond [comm of grind crunch])

- Low-latency interconnect (IB, vendor proprietary torus)
- Burst Buffers
- SAN
- Routing nodes (IO nodes)
- Paralle File System (Lustre, GPFS)
HPC IO Patterns

- Million files inserted into a single directory at the same time
- Millions of writers into the same file at the same time
- Jobs from 1 core to N-Million cores
- Files from 0 bytes to N-Pbytes
- Workflows from hours to a year (yes a year on a million cores using a PB DRAM)
A Simple collective 2D layout as an example

Collective I/O for 2-Dimensional Data

- 2-Dimensional data accesses by 4 processes

Apps can map their 1,2,3,N dimensional data onto files in any way they think will help them. This leads to formatting middleware like HDF5, NetCDF, MPI-IO.
Why $N \rightarrow 1$ strided can be problematic

- Process 1
  - RAID Group 1
- Process 2
  - RAID Group 2
- Process 3
  - RAID Group 3
- Process 4
  - RAID Group 4

RAID Group 1

RAID Group 2

RAID Group 3

PSC

Lustre

LANL

GPFS
PLFS: A Checkpoint Filesystem for Parallel Applications

John Bent; Garth Gibson; Gary Grider; Ben McClelland; Paul Nowoczenski; James Nunez; Milo Polte; Meghan Wingate

ABSTRACT

Parallel applications running across thousands of processors must protect themselves from inevitable system failures. Many applications insulate themselves from failures by checkpointing. For many applications, checkpointing into a shared single file is most convenient. With such an approach, the size of writes are often small and not aligned with file system boundaries. Unfortunately for these applications, this preferred data layout results in pathologically poor performance from the underlying file system which is optimized for large, aligned writes to non-shared files. To address this fundamental mismatch, we have developed a virtual parallel log structured file system, PLFS. PLFS remaps an application’s preferred data layout into one which is optimized for the underlying file system. Through testing on PanFS, Lustre, and GPFS, we have seen that this layer of indirection and reorganization can reduce checkpoint time by an order of magnitude for several important benchmarks and real applications without any application modification.

1: Summary of our results. This graph summarizes our results which will be explained in detail in Section 4. The key observation here is that our technique has improved checkpoint bandwidths for all seven studied benchmarks and applications by up to several orders of magnitude.
Decouples Logical from Physical

- N to 1 is fixed by PLFS
- But won’t scale to exascale
  - (N-N unscalable (billions of files)
Tightly Coupled Parallel Application

Concurrent, unaligned, interspersed IO

Parallel File System

Concurrent, aligned, interleaved IO

Tape Archive

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Economics have shaped our world
The beginning of storage layer proliferation 2009

- Economic modeling for archive shows bandwidth/capacity better matched for disk

Economic modeling for large burst of data from memory shows bandwidth/capacity better matched for solid state storage near the compute nodes.
Tightly Coupled Parallel Application

Burst Buffer

Parallel File System

Tape Archive
HPC Storage Stack, 2016-2020

- Tightly Coupled Parallel Application
- SCM
- Burst Buffer
- Parallel File System
- Object Store
- Tape Archive

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why HBM?
CPU \(\longrightarrow\) DRAM

why SSD?
DRAM \(\longrightarrow\) HDD

why NVMe?
SCSI \(\longrightarrow\) SSD

why SMR?
HDD \(\longrightarrow\) Tape

why SCM?
DRAM \(\longrightarrow\) SSD

SSD exposed pent-up demand for storage IOPs
and desire for finer and finer IOPs

why GenZ/OPA/etc?
NVMe \(\longrightarrow\) SCM

SSD exposed pent-up demand for shared IOPs to giant datasets
which physical will survive?
grider’s bent crystal ball

compute servers
  hbm
  scm

performance storage
  dram
  ssd
  (performance hdd)

capacity storage
  dram
  capacity hdd
how many logical tiers will survive?
the number is 2

of physical, there may be many
the number needed for economic efficiency
of logical, there should be two
the number needed to balance site, human, and workload efficiency

human in loop to make difficult decisions
one storage system focused on performance, one on capacity
capacity should be site-wide, performance can be machine-local
better resilience, users should always have at least one available
HPC Storage Stack, 2020

- Tightly Coupled Parallel Application
- SCM
- Burst Buffer
- Object Store
What about the Capacity Tier: Won’t cloud technology provide the capacity solution?

- Erasure to utilize low cost hardware
- Object to enable massive scale
- Simple minded interface, get put delete

- Problem solved → NOT

- Works great for apps that are newly written to use this interface
- Doesn’t work well for people, people need folders and rename and …
- Doesn’t work for the $trillions of apps out there that expect some modest name space capability (parts of POSIX)
How about a Scalable Near-POSIX Name Space over Cloud style Object Erasure: MarFS

- Best of both worlds
  - Objects Systems
    - Provide massive scaling and efficient erasure techniques
    - Friendly to applications, not to people. People need a name space.
    - Huge Economic appeal (erasure enables use of inexpensive storage)
  - POSIX name space is powerful but has issues scaling

- The challenges
  - Mismatch of POSIX an Object metadata, security, read/write semantics, efficient object/file sizes.
  - No update in place with Objects
  - How do we scale POSIX name space to trillions of files/directories
What it is

- 100-1000 GB/sec, Exabytes, Billion files in a directory, Trillions of files total
- Near-POSIX global scalable name space over many POSIX and non POSIX data repositories (Scalable object systems - CDMI, S3, etc.)
  - (Scality, EMC ECS, all the way to simple erasure over ZFS’s)
- It is small amount of code (C/C++/Scripts)
  - A small Linux Fuse
  - A pretty small parallel batch copy/sync/compare/ utility
  - A moderate sized library both FUSE and the batch utilities call
- Data movement scales just like many scalable object systems
- Metadata scales like NxM POSIX name spaces both across the tree and within a single directory
- It is friendly to object systems by
  - Spreading very large files across many objects
  - Packing many small files into one large data object

What it isn’t

- No Update in place! It’s not a pure file system, Overwrites are fine but no seeking and writing.
Scaling test on our retired Cielo machine:
835M File Inserts/sec Stat single file < 1 millisecond
> 1 trillion files in the same director

Striping across 1 to X Object Repos
MarFS Internals Overview Uni-File

UniFile - Attrs: uid, gid, mode, size, dates, etc.
Xattrs - objid repo=1, id=Obj001, objoffs=0, chunksize=256M, Objtype=Uni, NumObj=1, etc.
MarFS Internals Overview Multi-File (striped Object Systems)

/MarFS  top level namespace aggregation

/GPFS-MarFS-md1

Dir1.1

trashdir

/GPFS-MarFS-mdN

MultiFile - Attrs: uid, gid, mode, size, dates, etc.
Xattrs - objid repo=S, id=Obj002., objoffs=0, chunksize=256M, ObjType=Multi, NumObj=2, etc.

Metadata

Object System 1

Obj002.1

Object System X

Obj002.2
MarFS Internals Overview Packed-File

Metadata

Data

UniFile - Attrs: uid, gid, mode, size, dates, etc.

Xattrs - objid repo=1, id=Obj003, objoffs=4096, chunksize=256M, Objtype=Packed, NumObj=1, Obj=4 of 5, etc.

Object System 1

Object System X

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Pftool – parallel copy/rsync/compare/list tool

- Walks tree in parallel, copy/rsync/compare in parallel.
- Parallel Readdir’s, stat’s, and copy/rsync/compare
  - Dynamic load balancing
  - Restart-ability for large trees or even very large files
  - Repackage: breaks up big files, coalesces small files
  - To/From NFS/POSIX/parallel FS/MarFS

Diagram:
- Load Balancer Scheduler
- Dirs Queue
- Stat Queue
- Cp/R/C Queue
- Readdir
- Stat
- Copy/Rsync/Compare
- Done Queue
- Reporter
How does it fit into our environment

- **Premier Machine 2PB Dram**
  - 4 PB
  - BB 2
  - TB/S
- **Private IO Nodes**
  - Local Scratch
  - 100 PB
  - 1 TB/sec
  - 1-4 Weeks
- **General IO Nodes**
  - Site Scratch
  - 10's PB
  - 100 GB/sec
  - 1-4 Weeks
  - HPSS 100 PB
  - 10 GB/sec
  - Forever
- **Capacity machines ~50-300 TB Dram**
  - /localscratch(s)
  - /sitescratch(s)
  - /home
  - /project
- **Capacity machines ~50-300 TB Dram**
  - Parallel load balanced movers
  - /localscratch(s)
  - /sitescratch(s)
  - /home
  - /project
  - /campaign
- **General IO Nodes**
  - Interactive FTA(s)
  - WAN FTA(s)
  - Special Security Rules
  - Batch File Transfer Agents
  - /localscratch(s)
  - /sitescratch(s)
  - /home
  - /project
  - /campaign
  - HPSS
  - /analytics (HDFS other)
- **Analytics machine potentially disk full/big memory HDFS?**
  - /sitescratch(s)
  - /campaign
  - /analytics (HDFS other)
  - Use HDFS – POSIX Shim for access to POSIX resources
  - Parallel Tape with Disk Cache
  - Campaign MarFS 100's PB
  - 100's GB/sec
  - Few Years (erasuere)
  - 100(s) Gbists/sec
Not Just LANL Developing This Tier

Spectra Logic Announces Lustre Archive Campaign Storage Solution

Posted on Tuesday, November 13th, 2016 at 6:00 am.
Written by Spectra Logic

Spectra’s Lustre archive solution, powered by Campaign Storage HSM, offers new architecture for scalable high-performance archive based on industry standard workflows and technologies.

Salt Lake City, UT, SC-16, #1401 – November 15, 2016 – Spectra Logic, the deep storage experts, today announced a new archive solution for Lustre file systems widely deployed in high-performance computing (HPC), education and government supercomputers, and data centers. The solution, which includes embedded hierarchical storage management (HSM) software by Campaign Storage LLC, works with Spectra’s BlackPearl® Deep Storage Gateway, Intel’s Lustre file system, and an industry-standard policy manager. It offers seamless Lustre Archive functionality with a high-performance archive search and HSM management tool allowing nearly unlimited scalability, performance, monitoring and flexibility.

Seagate ClusterStor A200

Spectra Logic Campaign Storage LLC
Serving Data to the Lunatic Fringe: The Evolution of HPC Storage

Authors: John Benz, Brad Settlemyer, and Gary Grider

Article Section: STORAGE

Before the advent of Big Data, the largest storage systems in the world were found almost exclusively within high performance computing centers such as those found at US Department of Energy national laboratories. However, these systems are now dwarfed by large datacenters such as those run by Google and Amazon. Although HPC storage systems are no longer the largest in terms of total capacity, they do exhibit the largest degree of concurrent writes access to shared data. In this article, we will explain why HPC applications must necessarily exhibit this degree of concurrency and the unique HPC storage architectures required to support them.
DOE Exascale Computing and Future Considerations

- R&D and integration required to deploy Applications on Exascale computers in 2023+
- Partnership involving: Government, Computer industry, DOE laboratorie, Academia
- Target System Characteristics
  - 1-10 Billion degrees of concurrency
  - 20-30 MW Power requirement for one machine (machine only)
  - <300 cabinets
  - Development and execution time productivity improvements
  - 100 PB working sets
  - Checkpoint times < 1 minute (constant failure)
  - Storage systems need to be very reliable as the machine wont be
  - Leverage->Exploit new technology, dense flash/SCM/low latency high bandwidth byte addressable networks
Transactional/Versioning Coupled with Async Programming Models

Save every microsecond

Spread the load over as much time as the app will allow
Exploiting New Technology (dense flash, SCM, one sided networks)

- Today's storage stacks do not allow full exploitation of even flash latencies, never the less SCM – trapped IOP's in software.
- Must move away from thin client to heavy server to block storage (Lustre, Ceph, MongoDB).
- These stacks are incredibly thick. Client KV/Object -> MDS/Access Server->KV/Object Server->Kernel File System->block or network block.
- Head towards embedding more of this in the client and provide light weight one side enabled servers.
- Applications use Middleware (HDF5 for example) to form Objects that go directly to light weight object servers, no real server just light weight kvs/object on the hardware.
  - HDF5 (HDF Group) Vol interface is an example
  - MDHIM (LANL) MultiDimensional Hierarchical Middleware is a user space distributed KVS middleware
  - DeltaFS (CMU/LANL) is a user space file system Middleware
https://github.com/mar-file-system/marfs
https://github.com/pftool/pftool

Thank You For Your Attention