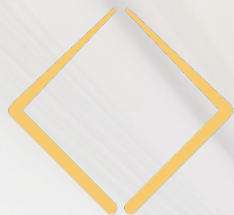




This project has received funding from the European Union's Horizon 2020 research and innovation program through grant agreement 801101.



**MAESTRO**  
DATA ORCHESTRATION

# Dynamically Provisioning Cray DataWarp Storage

**François Tessier**, Maxime Martinasso, Matteo Chesi, Mark Klein, Miguel Gila

*Swiss National Supercomputing Centre, ETH Zurich, Lugano, Switzerland*

Cray User Group Meeting 2019  
Montréal, Canada



# Context

Complex workflows or frameworks in various scientific domains have increasing I/O needs

Institution	Scientific domain	Workflows	Data size (real & projection)
European Centre for Medium-Range Weather Forecasts (ECMWF)	Weather Forecast	Ensemble forecasts, data assimilation,...	12PB/year
Paul Scherrer Institute (PSI)	Synchrotron imaging	X-ray spectroscopy, high resolution microscopy,...	10-20PB/year
Cherenkov Telescope Array (CTA)	Astrophysics	Gamma Rays & Cosmic Sources,...	25PB/year

- Workloads with specific needs of data movement
  - Big data analysis, machine learning, checkpointing, in-situ, co-located processes, ...
  - Multiple data access pattern (model, layout, data size, frequency)

# Context

- But I/O performance is decreasing!

Criteria	2007	2017	Relative Inc./Dec.
Name, Location	BlueGene/L, USA	Sunway TaihuLight, China	N/A
Theoretical perf.	596 TFlops	125,436 TFlops	× 210
#Cores	212,992	10,649,600	× 50
I/O bw	128 GBps	288 GBps	× 2.25
I/O bw/core	600 kBps	27 kBps	÷ 22.2
I/O bw/TFlop	214 MBps	2.30 MBps	÷ 93.0

- Mitigating the I/O bottleneck from an hardware perspective leads to an increasing complexity and a diversity of the architectures
  - Node-local storage (PCIe, SATA)
  - Burst buffers like Cray DataWarp, DDN Infinite Memory Engine

# Context

- But I/O performance is decreasing!



System Specs	TITAN	SUMMIT	FRONTIER
Peak Performance	27 PF	200 PF	>1.5 EF (X 7.5)
Storage	32 PB, 1 TB/s Lustre file-system	250 PB, 2.5 TB/s GPFS	<b>2-4x</b> performance and capacity of Summit's I/O subsystem. Frontier will have near node storage like Summit.

Source: <https://www.olcf.ornl.gov/frontier/>

- Mitigating the I/O bottleneck from an hardware perspective leads to an increasing complexity and a diversity of the architectures
  - Node-local storage (PCIe, SATA)
  - Burst buffers like Cray DataWarp, DDN Infinite Memory Engine

# Context

Scientific domains require more and more often varied data managers (object-based storage, database, ...)

- Data management inside a workflow usually relies on a global shared parallel file system
  - Unique data access semantic (POSIX)
  - Performance variability
- Workflow specific data managers are installed on a use case basis


Limited support and  
reduced capacity

OR

Specialized and  
expensive

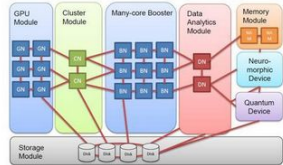
# Context


- On the HPC center side, not feasible to support a large variety of data management systems
- ... and hard to provide dedicated storage resources
  - Usually, data resources are shared while compute resources are exclusive
  - Shared storage resources are subject to contention and high unexpected performance decrease

 **James Lin**  
@JamesLinHPC Following

any possible to build an exascale machine with the modular supercomputer design?  
[@HPC\\_Guru](#)

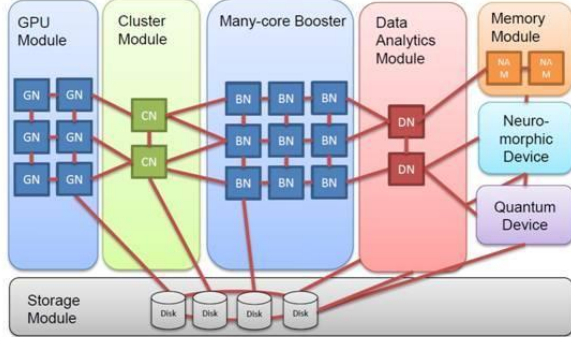
**... AND EVOLUTION TO A MODULAR SUPERCOMPUTING ARCHITECTURE**



Mitglied der Helmholtz-Gemeinschaft 

1:21 PM - 6 Mar 2019 from Shanghai, People's Republic of China


## ... AND EVOLUTION TO A MODULAR SUPERCOMPUTING ARCHITECTURE



The diagram illustrates a modular supercomputing architecture with the following components and connections:


- GPU Module:** A 3x3 grid of GPU nodes (GN).
- Cluster Module:** Two central Cluster Node (CN) nodes.
- Many-core Booster:** A 3x3 grid of Many-core Booster Node (BN) nodes.
- Data Analytics Module:** Two Data Node (DN) nodes.
- Memory Module:** Two Memory Module (MM) nodes.
- Neuromorphic Device:** A light blue box.
- Quantum Device:** A purple box.
- Storage Module:** A base layer containing four Disk nodes.

Connections are shown as red lines: GN nodes connect to CN nodes; CN nodes connect to BN nodes; BN nodes connect to DN nodes; DN nodes connect to MM nodes; MM nodes connect to Neuromorphic and Quantum devices; and all modules connect to the Storage Module.

Mitglied der Helmholtz-Gemeinschaft 

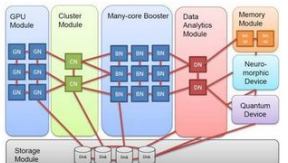
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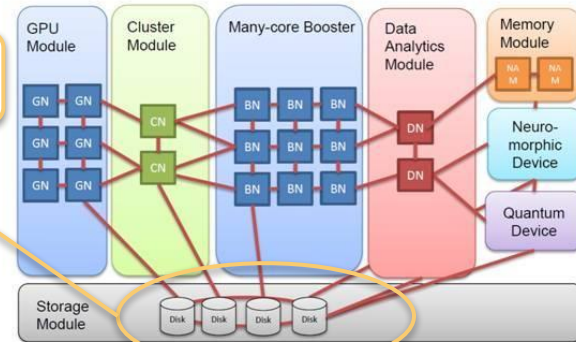
**JÜLICH**  
Forschungszentrum

1:21 PM - 6 Mar 2019 from Shanghai, People's Republic of China

Raise your hand, please

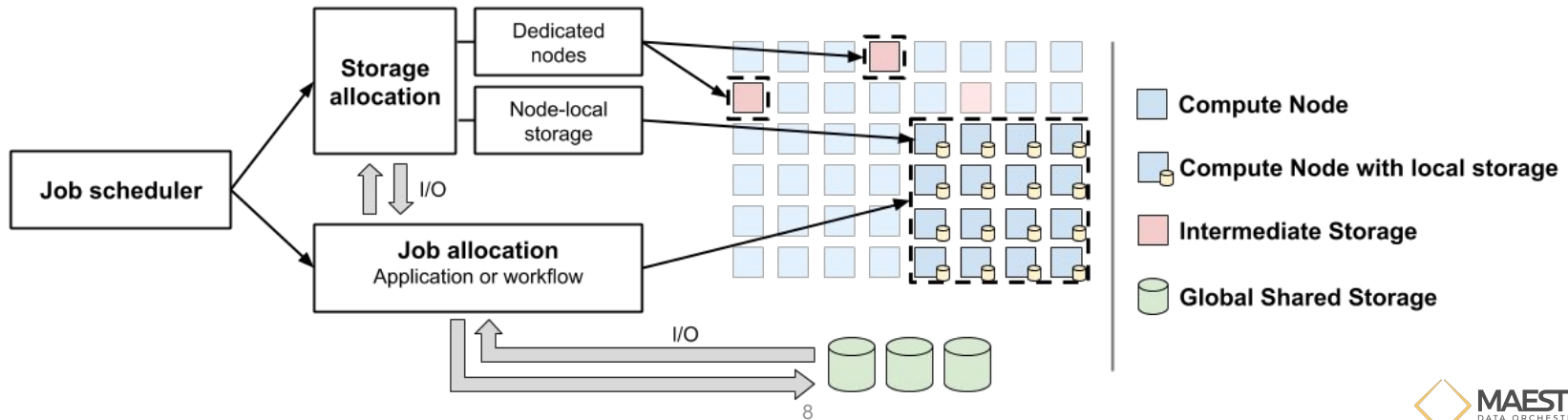
## ... AND EVOLUTION TO A MODULAR SUPERCOMPUTING ARCHITECTURE

Contention!



# Dynamic Resource Provisioning

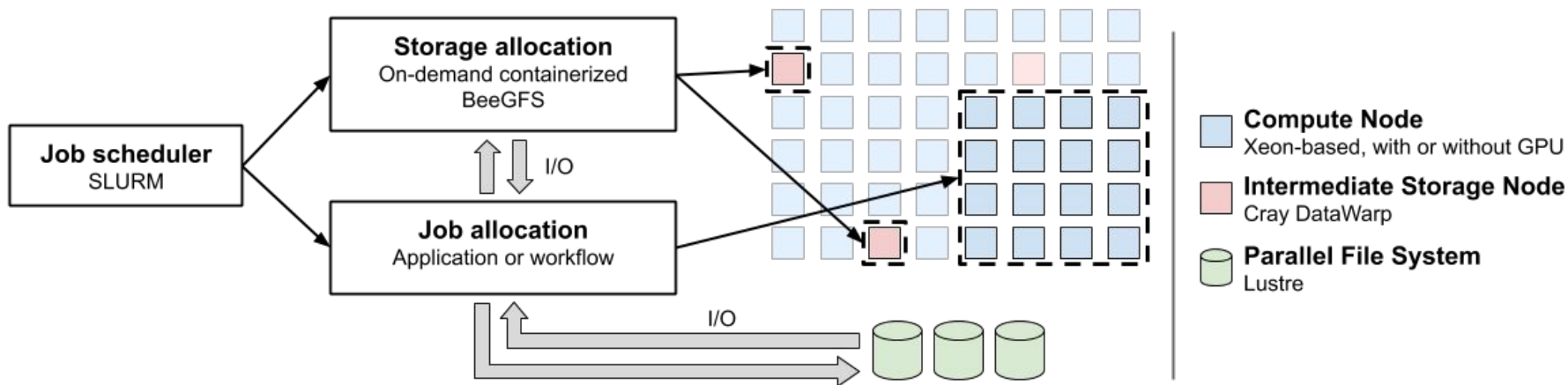
- Provisioning of storage system at job level:
  - Storage available during the job lifetime
  - Storage resources dedicated to a job (isolation)
- Dynamically supply a data management system on top of those resources
  - Several types supported: file system, object-based storage, database
  - Containerized data management services
  - Deployment fully integrated at a job scheduler level





# Our Approach

- Repurposing Cray DataWarp nodes
- Get an allocation of intermediate storage nodes along with compute nodes
- Deploy a well-sized BeeGFS across disks on DataWarp nodes
- Configure the compute nodes to act as clients of the BeeGFS instance



# Accessing DataWarp Nodes

## Standard implementation of DataWarp

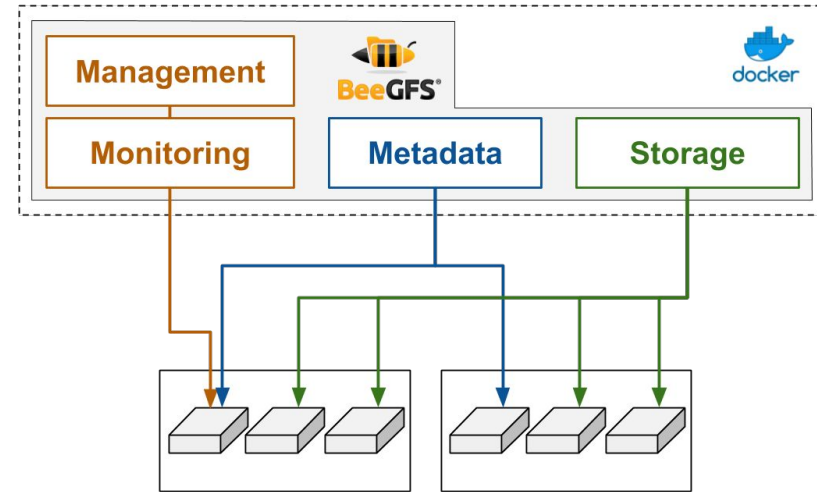
- Projection of DataWarp storage onto the compute node (through DVS)

## Repurposing

- System customization to reconfigure the nodes
  - From hidden **service** nodes to standard **compute** nodes
  - Mapping of a compute node image to boot with
- Setup the local NVMe storage
  - XFS file system
  - Mount point with permissions granted to any user
- New SLURM constraint: `storage`

# On-demand containerized BeeGFS

- BeeGFS: POSIX-compliant parallel file system based on a client-server architecture
  - Server-side: management, monitoring, metadata, storage
  - Client-side: kernel-space client, monitoring visualization
- Servers bundled in a Docker container and deployed with Sarus, a container runtime system
  - 1 metadata and 2 storage servers per DataWarp node
- Mount point on clients (compute nodes)
  - Kernel module required
  - Special privileges to `mount BeeGFS`



**Cray DataWarp**  
3 SSD per node

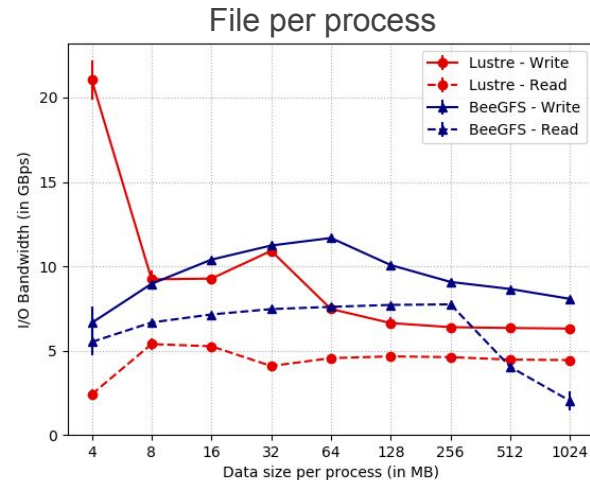
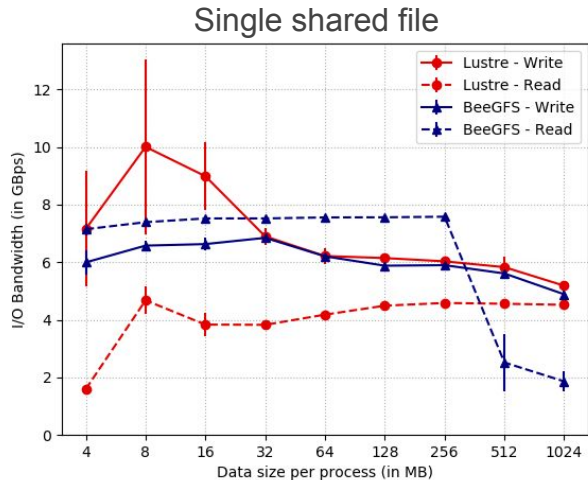


# Limitations

- Kernel-space file system such as BeeGFS implies special privileges
  - Load/unload kernel module: `modprobe [-r] beegfs`
  - Mount BeeGFS on compute nodes: `mount -t beegfs [...] $HOME/beegs [...]`
  - Module pre-installed on nodes?
  - Prolog script for file-system creation and mount point?
- Fresh data manager provisioned meaning no data available
  - Stage-in/stage-out phase, such as on native DataWarp?
  - Should this step be counted in the allocation time?
- Trade-off between capacity and capability
  - Better I/O bandwidth implies more disks and possibly capacity wasted

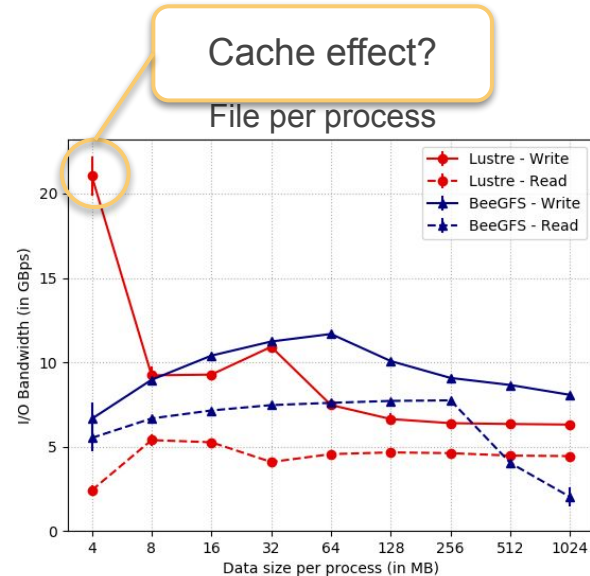
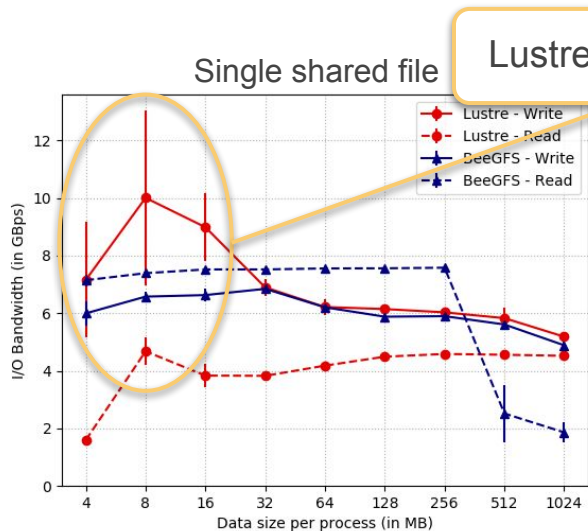
# Performance Evaluation

- Dom, Cray XC50 system with DataWarp at CSCS
  - Test and development system of Piz Daint (27PFlops)
  - 8 nodes with two 18-cores Intel Broadwell CPU and 64GB of DRAM
  - 4 DataWarp nodes each with three 5.9TB PCIe SSD
- On demand-BeeGFS (2 DW nodes) VS Lustre file system (Sonexion 1600, 2 OSTs)
- IOR benchmark: independent I/O, 10 runs



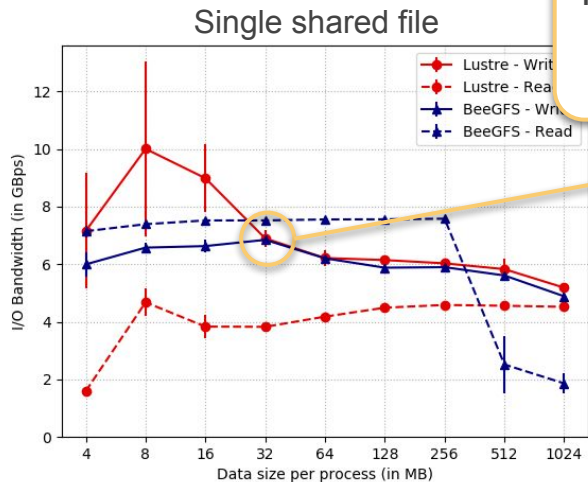
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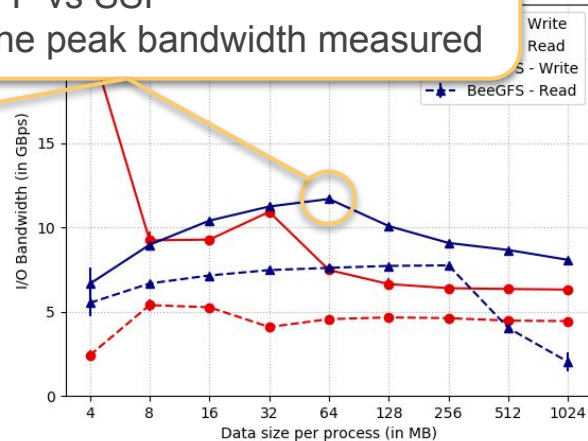
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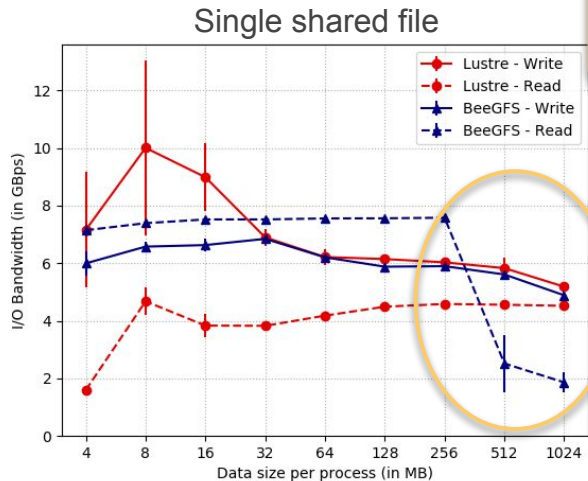
Peak write bandwidth:

- **+70% FPP vs SSF**
- **93% of the peak bandwidth measured**

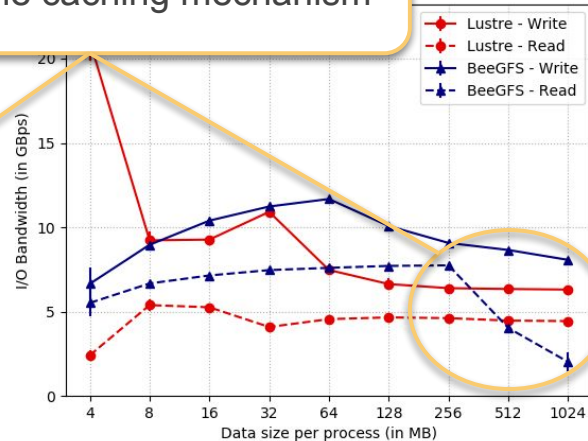


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Not enough memory on DW nodes (64GB) for the caching mechanism





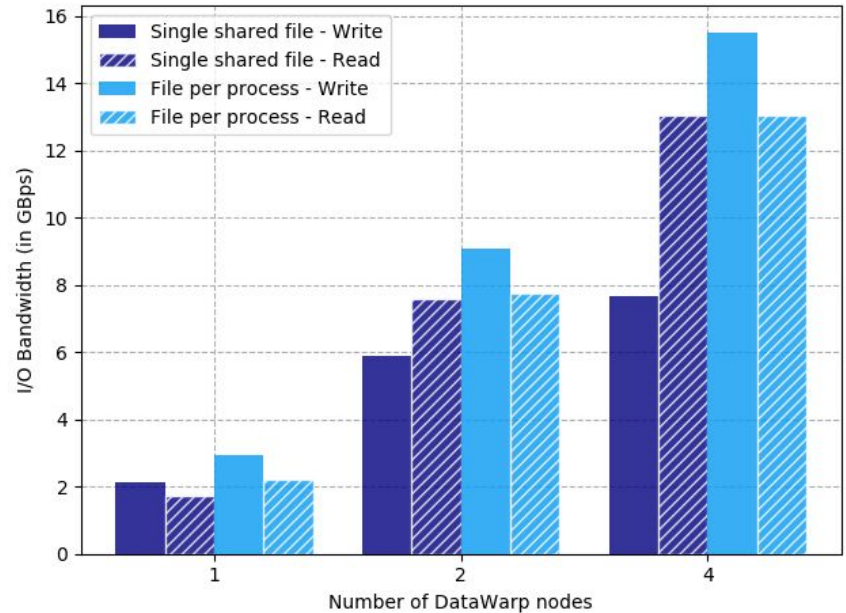
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- On demand-BeeGFS (2 DW nodes) versus global Lustre file system (2 OSTs)
- *mdtest* benchmark

Target	Operation	BeeGFS	Lustre	L/B
		Ops		
Directory	Creation	8276.43	37222.57	× 4.5
	Stat	5301788.76	182330.42	÷ 29.1
	Removal	12967.02	38732.00	× 3.0
File	Creation	6618.37	22916.15	× 3.5
	Stat	144410.46	169140.32	× 1.2
	Read	22541.08	45181.55	× 2.0
	Removal	8431.71	35985.96	× 4.3
Tree	Creation	2183.40	3310.42	× 1.5
	Removal	125.23	1298.55	× 10.4

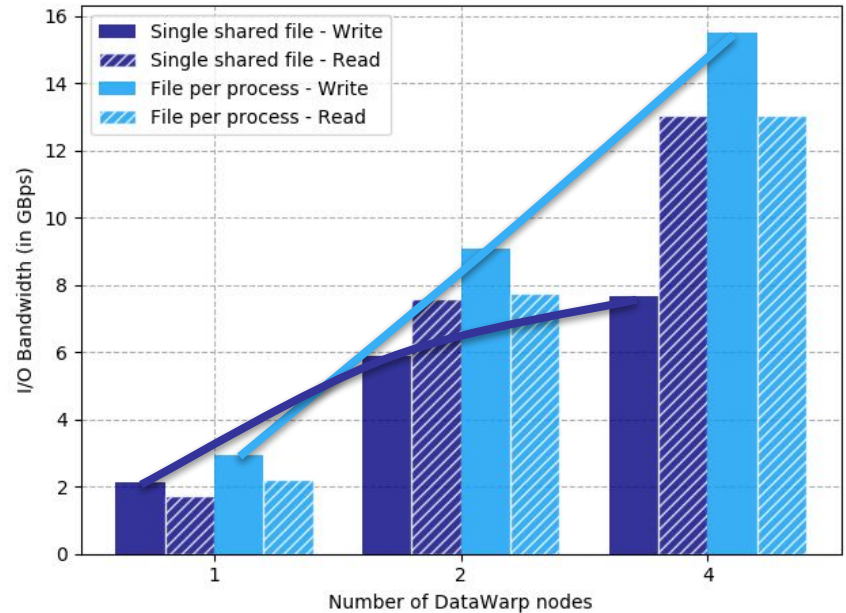
# Performance Evaluation

- Small-scale study of... scalability
- IOR from 8 compute nodes (36 ppn)
  - 256MB written/read per process
- Dynamically provisioned BeeGFS
  - From 1 to 4 nodes
  - Ratio metadata:storage server per node kept to 1:2
- Reasonable scalability overall
  - Except SSF - write



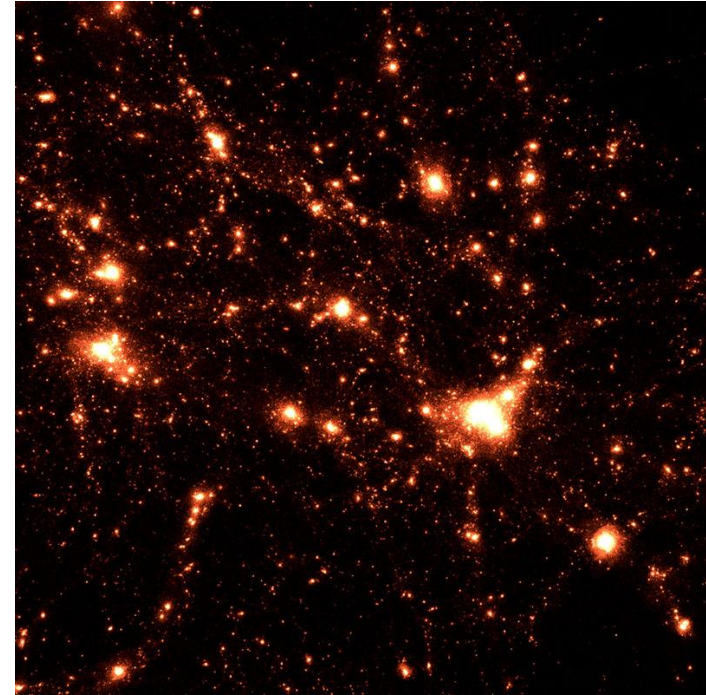
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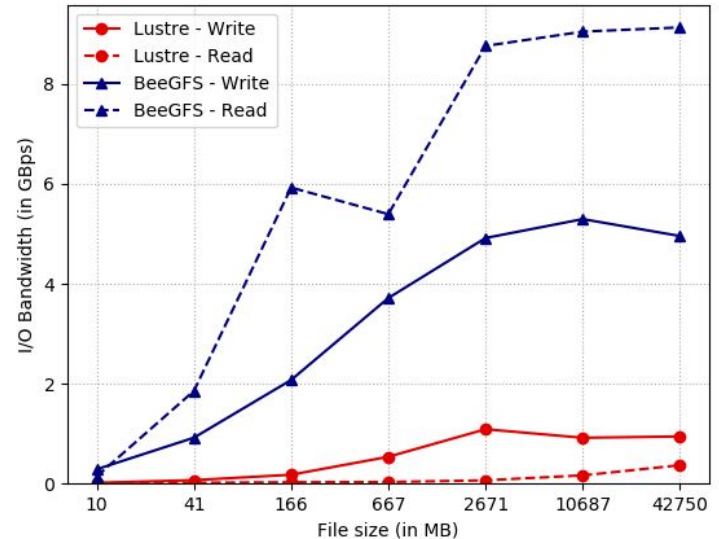
- I/O part of a large-scale cosmological application simulating the mass evolution of the universe with particle-mesh techniques
- Each process manages particles defined by 9 variables (38 bytes)
  - XX, YY, ZZ, VX, VY, VZ, phi, pidandmask
- Single shared checkpointing file with data in an array of structure data layout
- Average and standard deviation on 10 runs



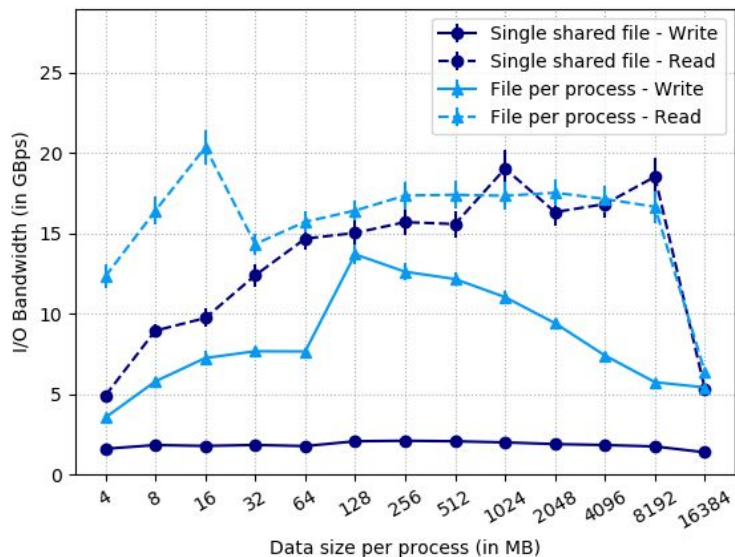
*Credits: Silvio Rizzi and Joe Insley, Argonne National Laboratory*

# Performance Evaluation - HACC-IO

- HACC-IO from 8 compute nodes, 36 ppn
- BeeGFS (2 DW) vs Lustre (2 OSTs)
  
- BeeGFS peak **write** bandwidth: **5.3GBps**  
**read** bandwidth: **9.1GBps**
  
- As expected (previous work), BeeGFS highly outperforms Lustre
  - Single shared file and array of structure data layout is a bad combination on Lustre



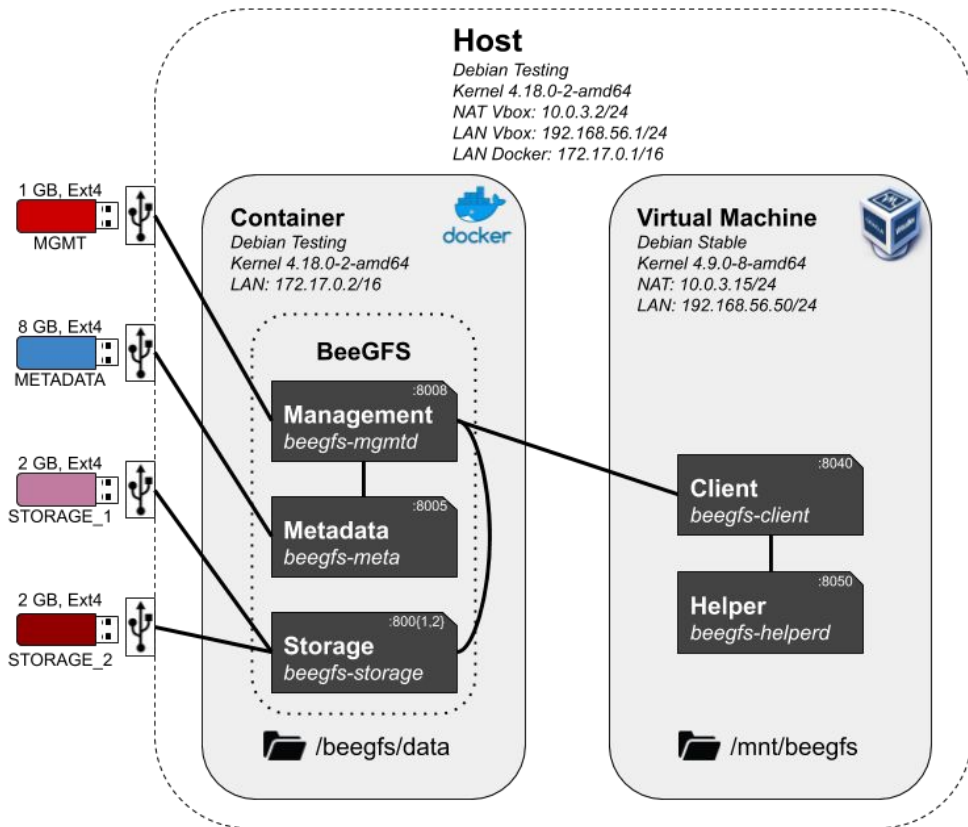
# Portability



- Ault, testbed platform at CSCS allowing for prototyping experimental services and platforms
  - Various types of hardware
  - Safe privileged-access level for researchers
- Ault11, compute node with a 22-core Intel Xeon Gold 6152 CPU
  - 16 3D NAND NVMe disks
- Dynamically provisioned BeeGFS
  - 1 disk for management and monitoring
  - 2 disks for metadata
  - 5 disks for storage
- Peak **read** bandwidth: **20.36GBps**
- Peak **write** bandwidth: **13.70GBps**
- In line with values communicated by the vendor



# Portability For Fun



How to give a second lease of life to HPC conference USB Keys?

# Conclusion

- Proof of concept of a mechanism to dynamically provision data managers on top of intermediate storage resources
  - Focused on containerized BeeGFS + DataWarp
- Promising performance and scalability with IOR and the I/O kernel of a real application
- Portability on different types of hardware and systems
- **Next steps**
  - Integration within the job scheduler (prolog/epilog scripts)
  - Configurable system for deployment: architecture's description, data manager-specific settings, ...
  - Extends to other data managers packaged in a unique container

## Acknowledgment

- This work is part of the MAESTRO EU Project
- 3-year European project, started in September 2018
- **Middleware library that automates data movement across diverse memory systems**
- <https://www.maestro-data.eu/>





# Conclusion

Thank you for your attention!

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# MAESTRO

DATA ORCHESTRATION