

# SCA/HPCAsia 2026 Tutorial: Accelerating HPC Application I/O with Fast Node-Local Storage

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# Schedule

## Tutorial Link:

<https://wangchen.io/unifyfs-dyad-tutorial>

Time	Event
13:30 - 14:00	Introduction and Motivation
14:00 - 14:30	Deep Dive: UnifyFS
14:30 - 14:50	Break
14:50 - 15:20	Deep Dive: DYAD
15:20 - 15:50	DYAD Walkthrough
15:50 - 16:15	UnifyFS Walkthrough
16:15 - 16:30	Wrap-up and Q&A

## Cluster account request:

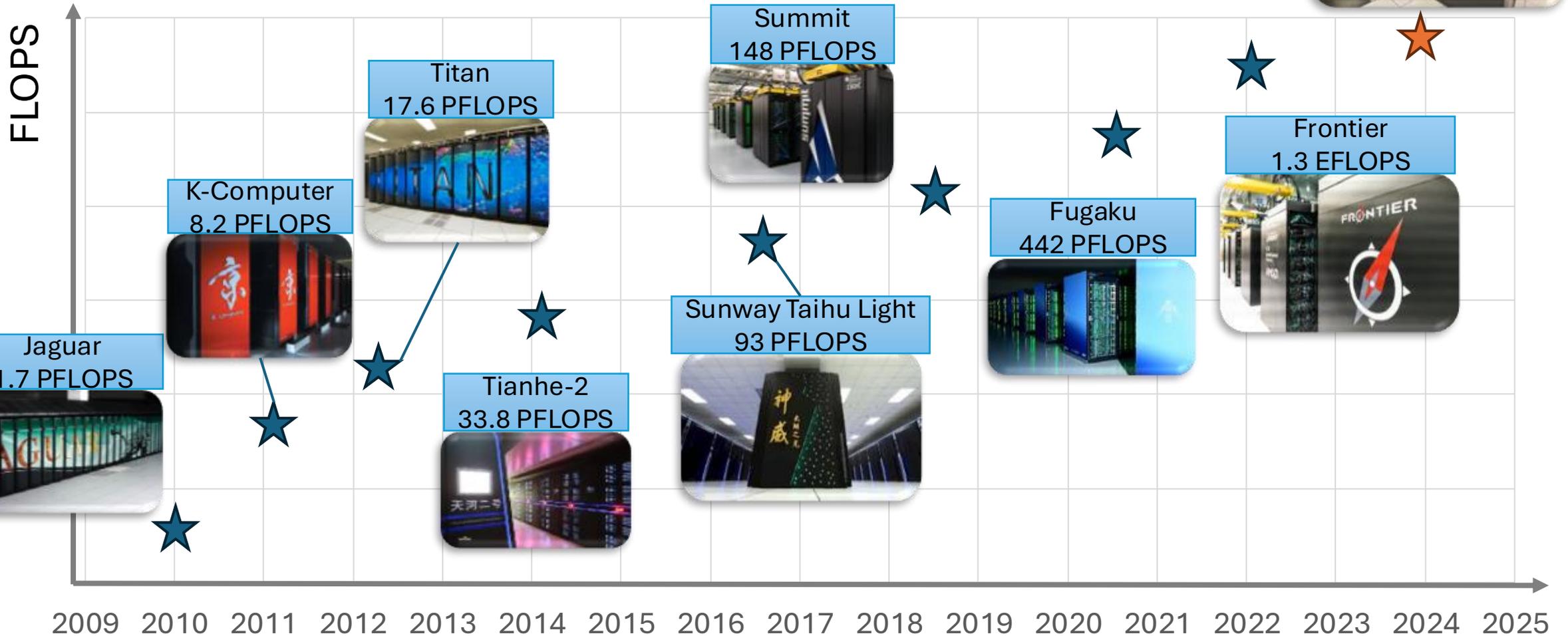
<https://forms.gle/Ya1WRVn9cxdRSMmXA>



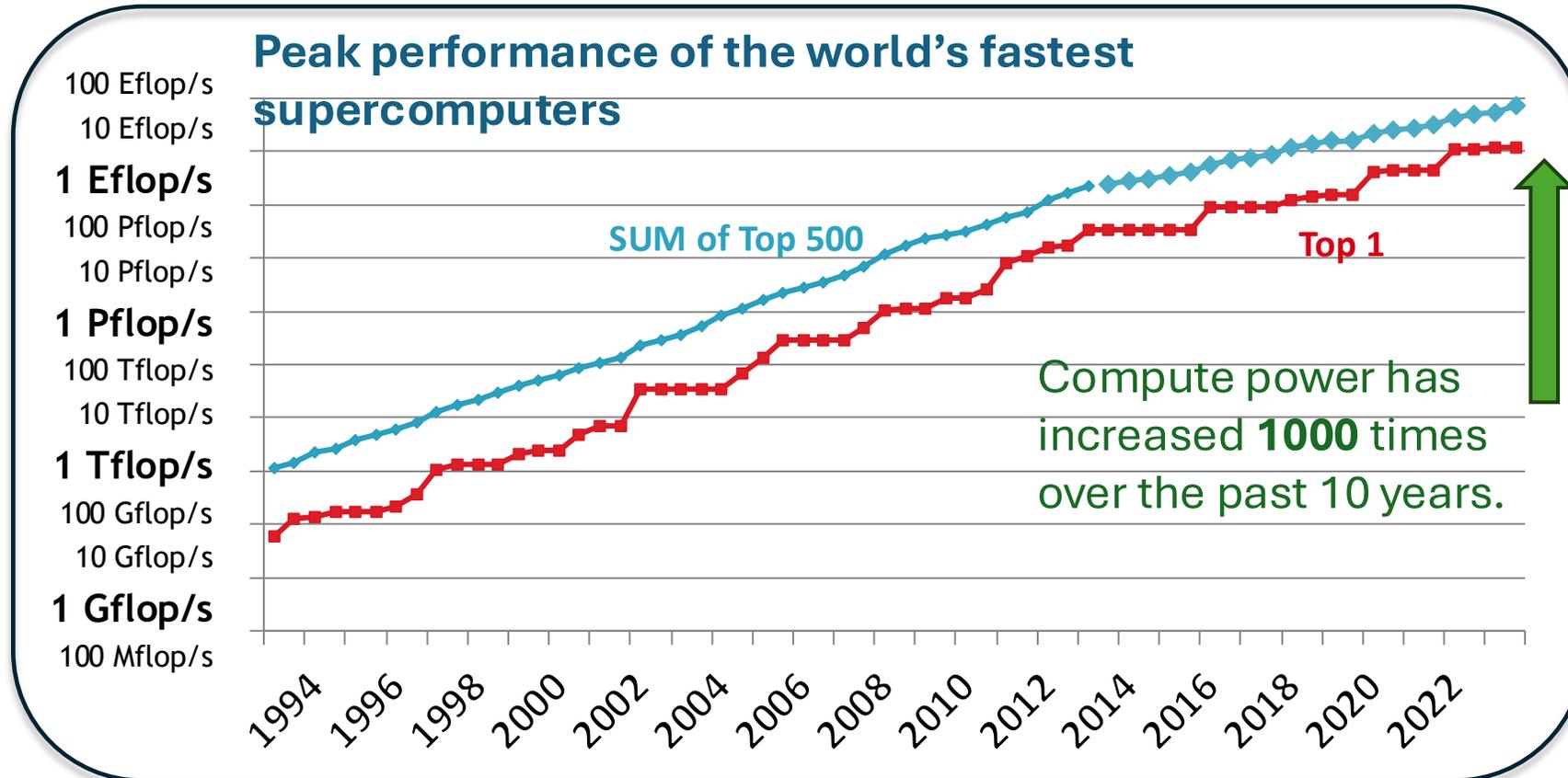
```
ssh -p 2223 username@13.215.163.223  
password: P@ssw0rd
```

# Top 1 Supercomputers

We have been in the exascale era since 2022.

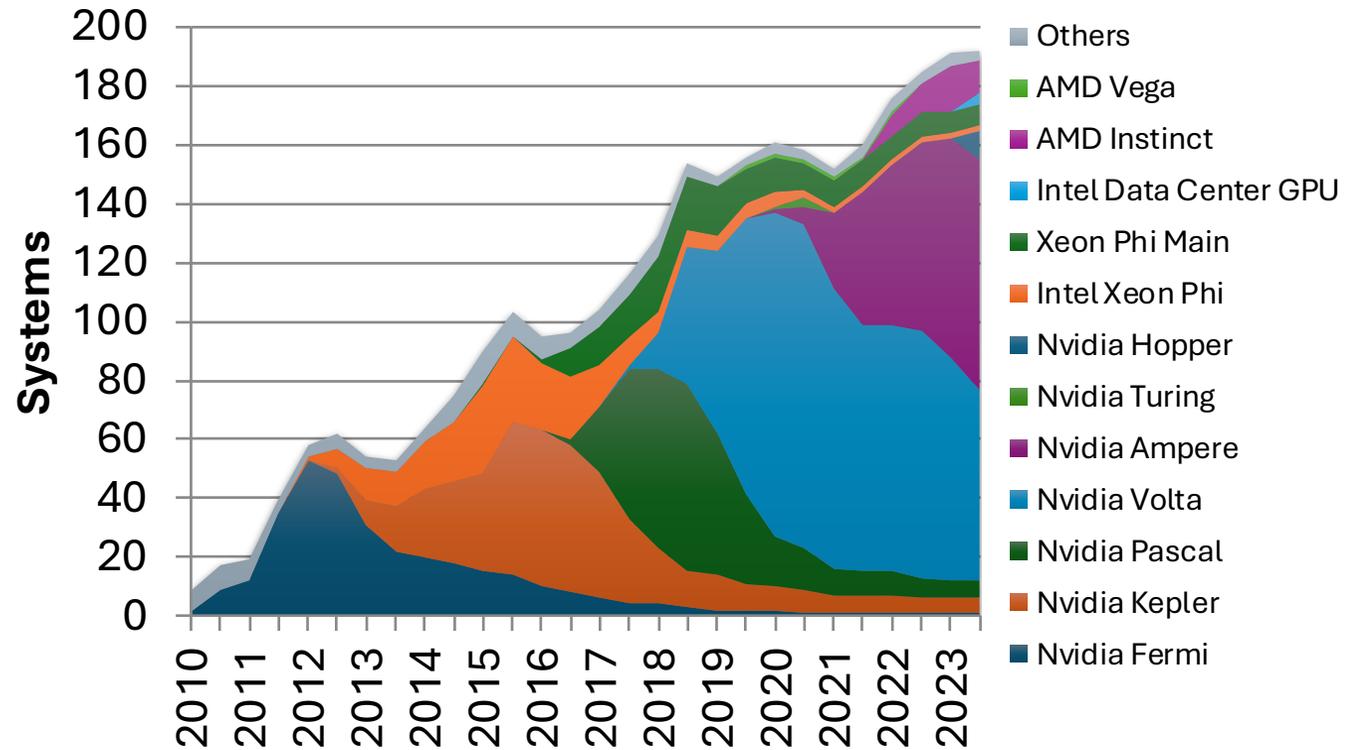
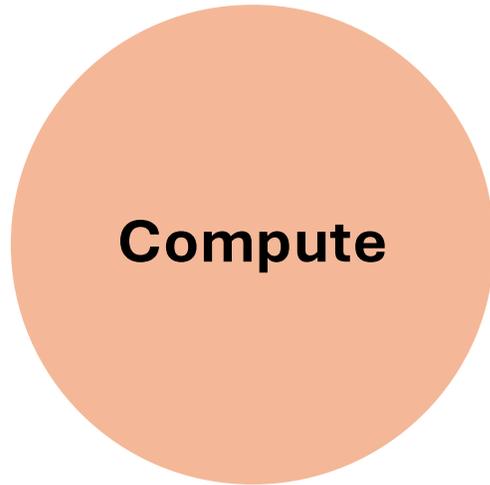


# Compute Power Has Increased Significantly



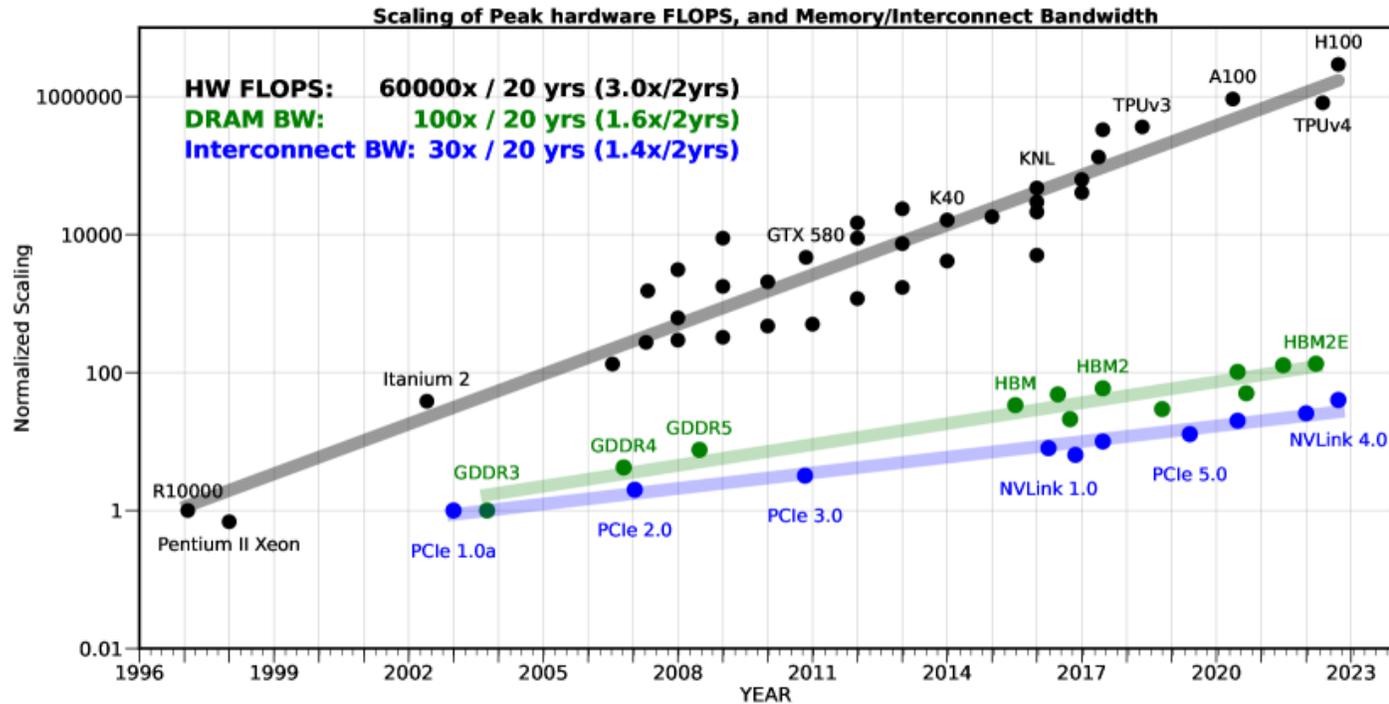
Data Source: Top500.org

# GPUs Have Become the Primary Workhorse



Over 1/3 of top 500 systems have accelerators

# AI and Memory Wall

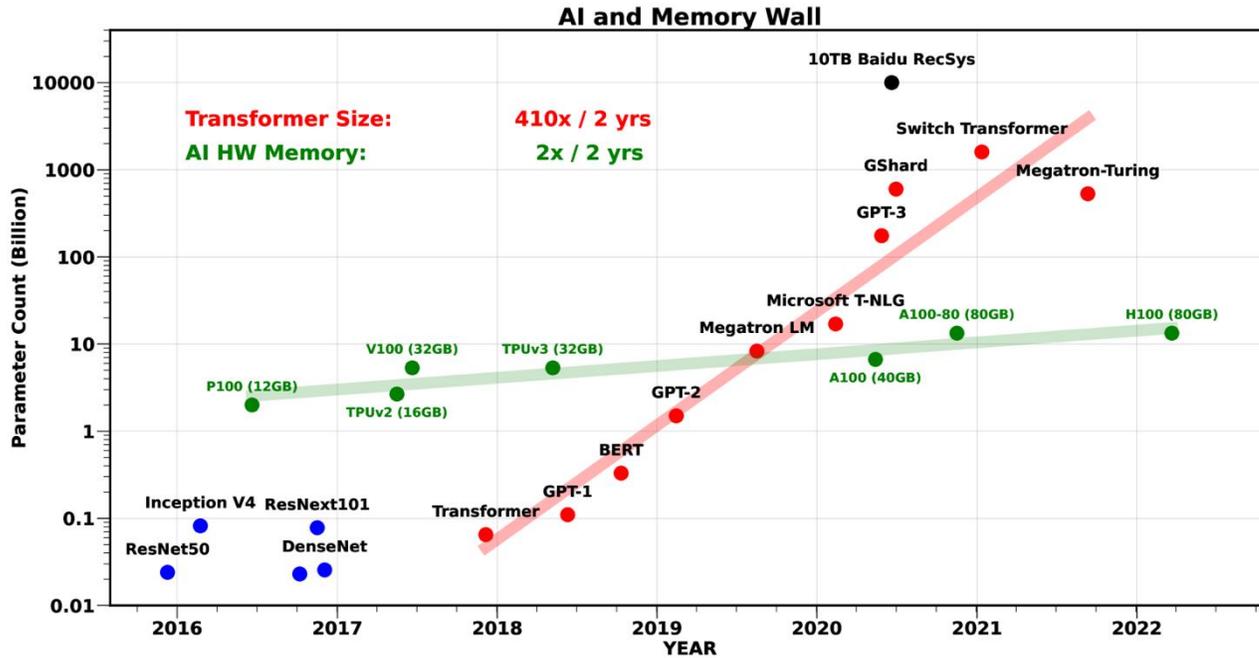


## GPU FLOPS vs. Memory Bandwidth

Gholami, Amir, Zhewei Yao, Sehoon Kim, Coleman Hooper, Michael W. Mahoney, and Kurt Keutzer. "Ai and memory wall." *IEEE Micro* 44, no. 3 (2024): 33-39.

The performance gap is expected to grow at 50% per year.

# AI and Memory Wall



Transformer Size vs. Memory Capacity

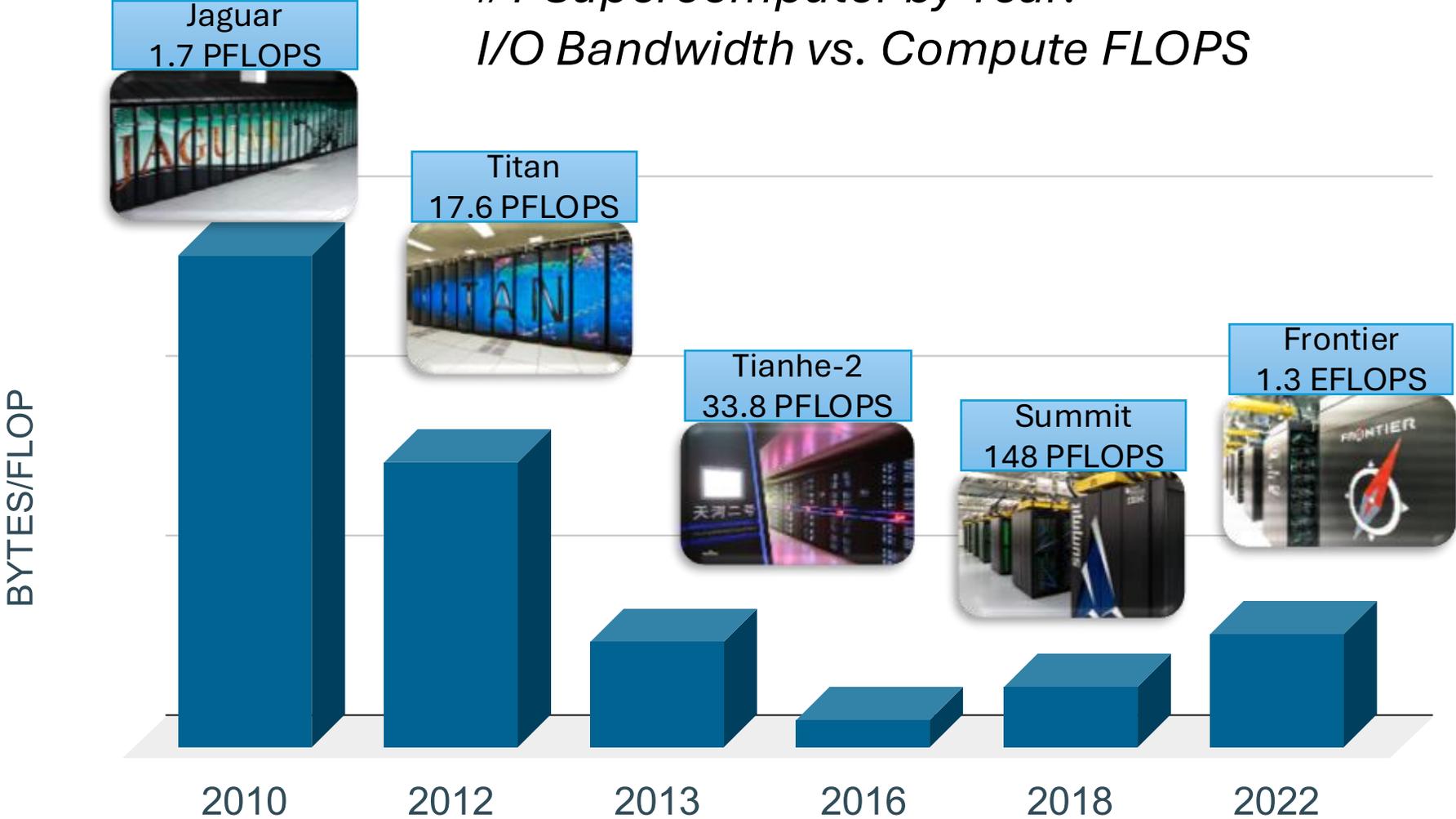


The Evolution of GPT Models

Gholami, Amir, Zhewei Yao, Sehoon Kim, Coleman Hooper, Michael W. Mahoney, and Kurt Keutzer. "Ai and memory wall." *IEEE Micro* 44, no. 3 (2024): 33-39.

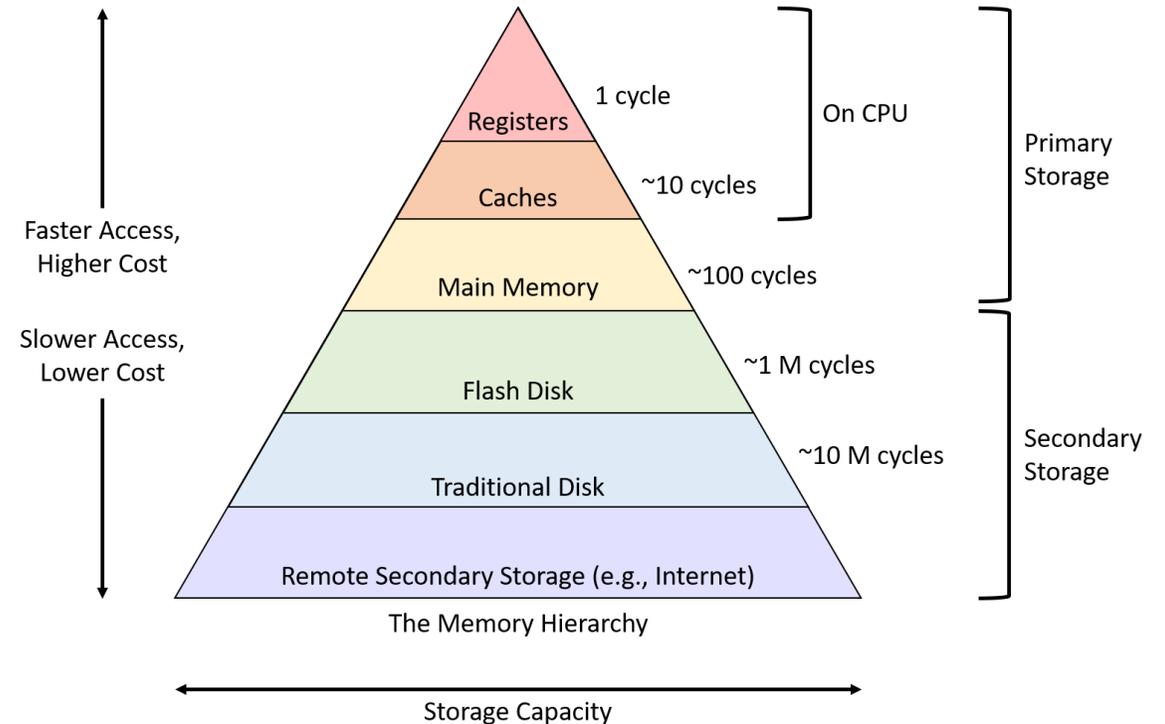
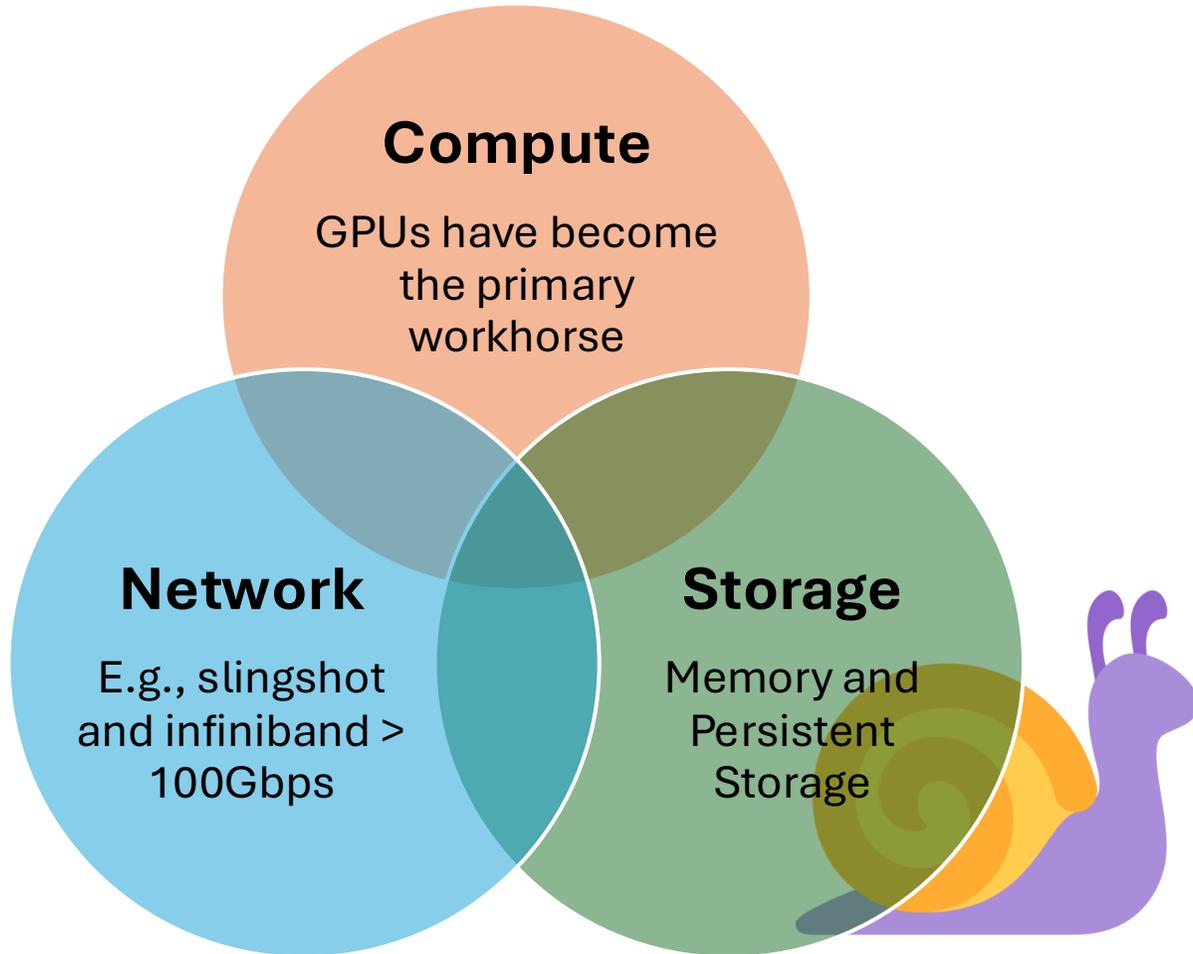
# I/O is even Slower

#1 Supercomputer by Year:  
*I/O Bandwidth vs. Compute FLOPS*



Tutorial link: <https://wangchen.io/unifyfs-dyad-tutorial>

# The Deep Storage Hierarchy

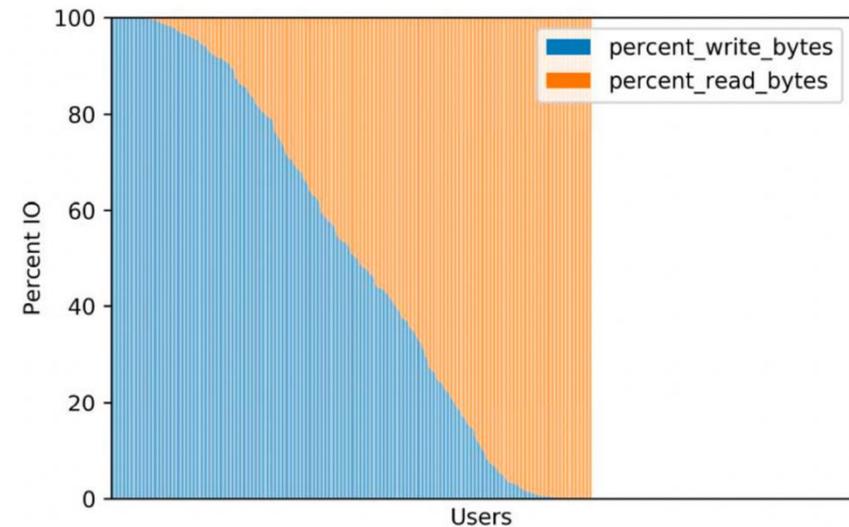


*The entire storage hierarchy is getting deeper and more complex, and the boundary between memory and storage is steadily blurring.*

# I/O Subsystem Inefficiency

A study of 4 million jobs over four years on two LLNL systems shows that

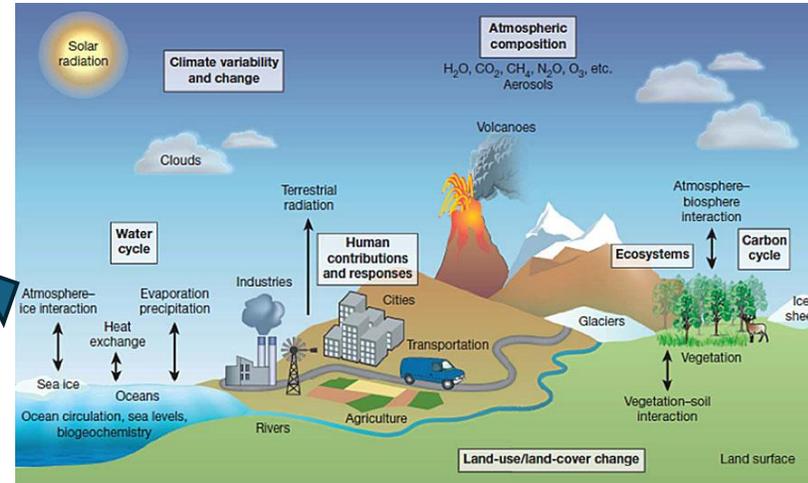
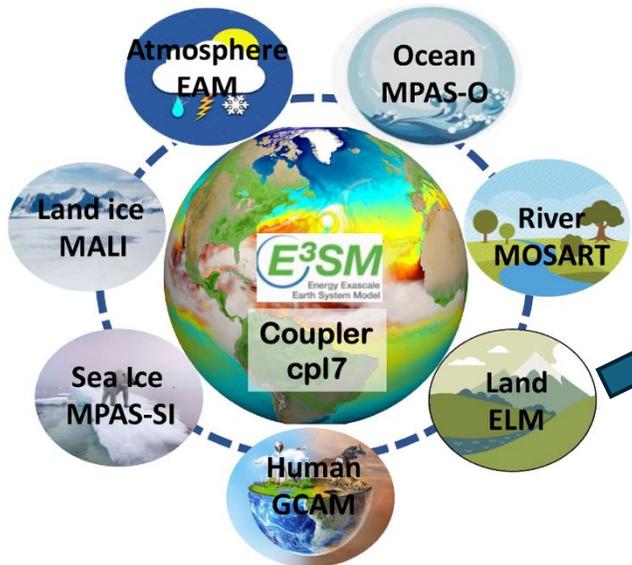
- on average, jobs which performing I/O spread I/O activities across 78.8% of their runtime.
- less than 22% write-intensive jobs perform efficient writes.
- HPC jobs are no longer write dominated



**Percentage I/O (Write vs. Read) by User**

*Paul, Arnab K., et al. "Understanding HPC Application I/O behavior using system level statistics." 2020 IEEE HiPC.*

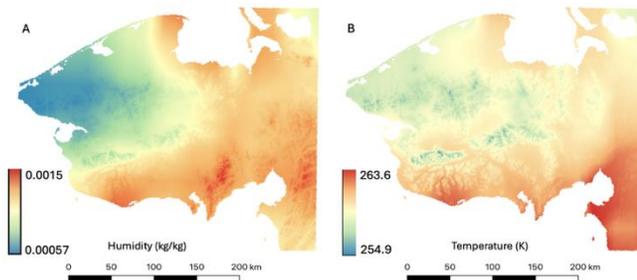
# Case Study: Energy Exascale Earth System Model (E3SM)



File	Size
Surface (I)	188 GB
Forcing (I)	1.4 TB
History (O)	134 GB
Restart (O)	4.2 TB

I:Input O:Output

A high-resolution (1kmx1km, previously 10kmx10km) land simulation over Alaska (21.6 Million land grid cell)  
 Used three supercomputers: Perlmutter, Summit (#1 from 2018-2020), and Frontier (#1 2022-2023, #2 now)

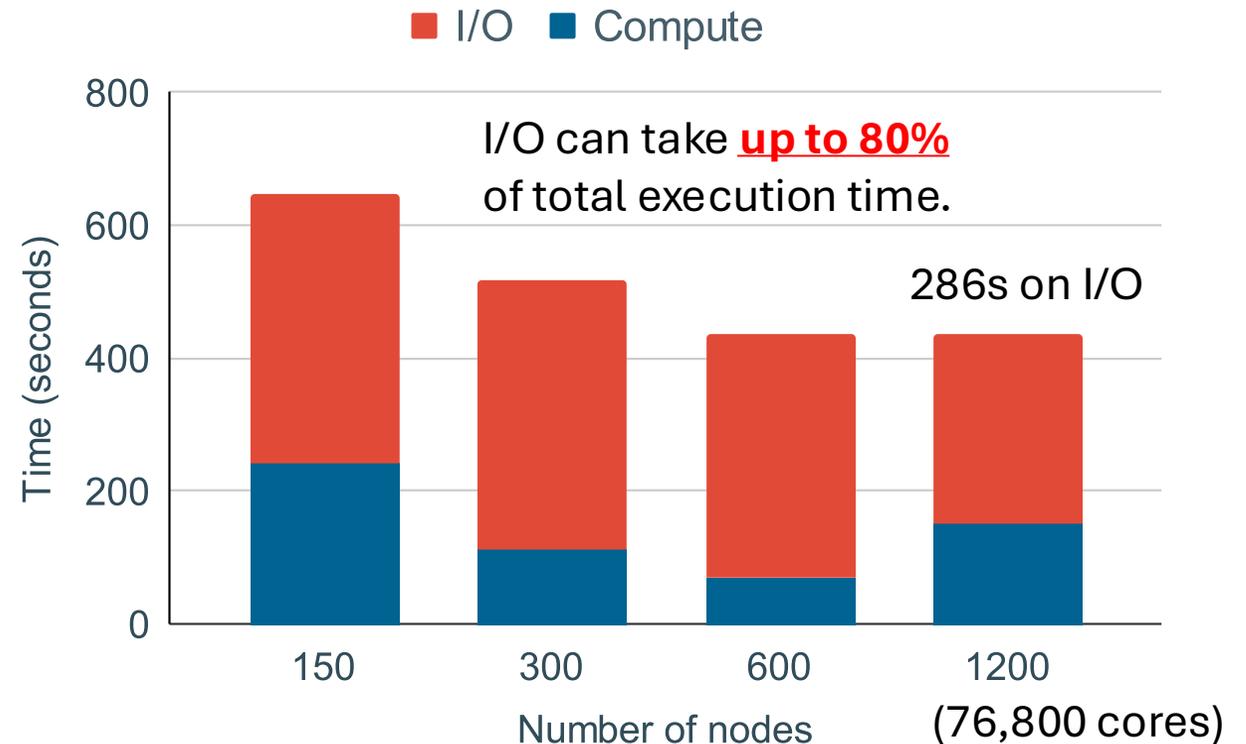


2025 CCGRID SCALE Challenge Finalist: Dali Wang, Chen Wang, Qinglei Cao, and et.al. "Scaling  
 Tutorial link: <https://wangchen.io/unifyfs-dvnd/tutorial>  
 Ultrahigh-Resolution E3SM Land Model for Leadership-Class Supercomputers".

# Case Study: Energy Exascale Earth System Model (E3SM)

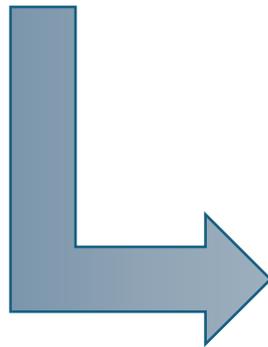
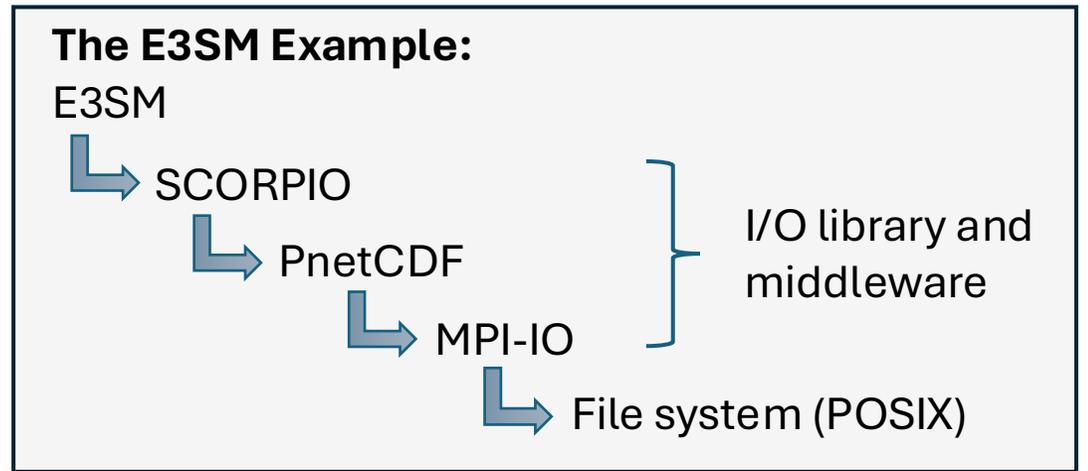
Strong scaling results on Frontier.

- Up to 1200 nodes with I/O.
  - **Bottleneck:** 76,800 processes concurrently write to a single file.
- Up to 4000 nodes (nearly half of the Frontier) without I/O.
  - Bottleneck: initialization phase
- *Note this is only a 5-day test simulation.*
- *We encountered both the scalability issue and the I/O bottleneck.*

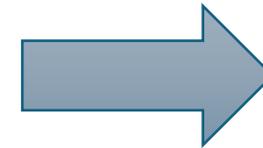


# How I/O Works in HPC

## HPC Application



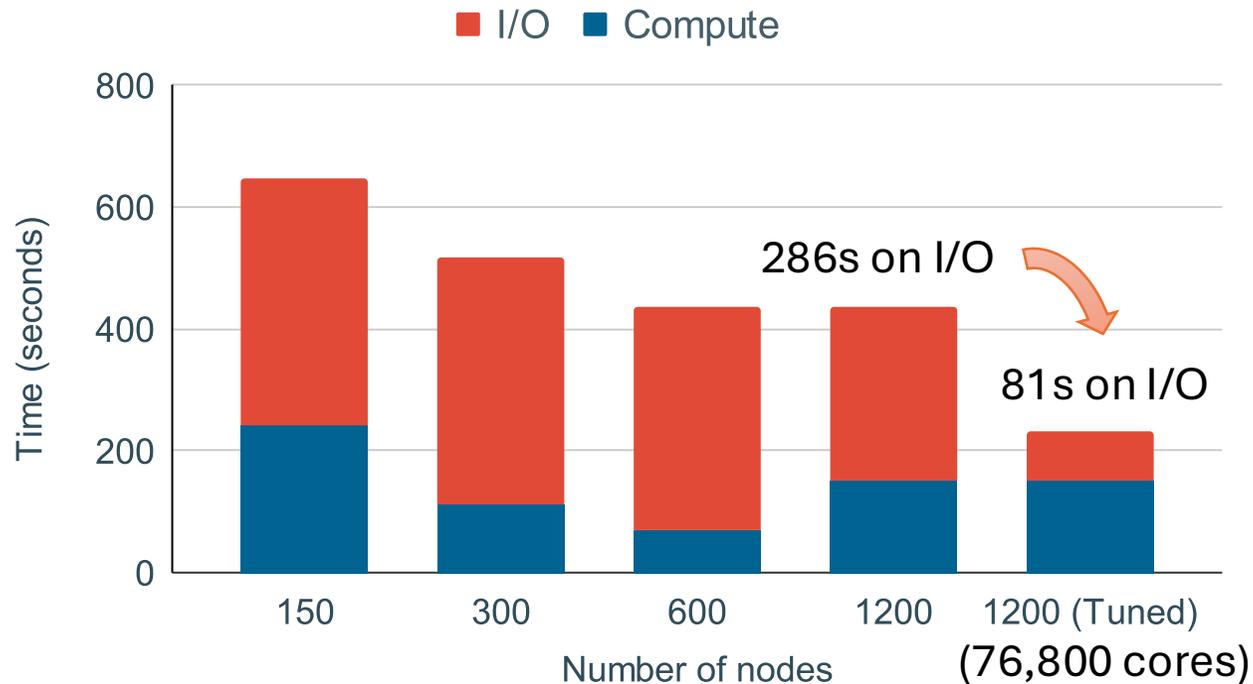
## I/O Library and middleware



## HPC File System



# Case Study: Energy Exascale Earth System Model (E3SM)



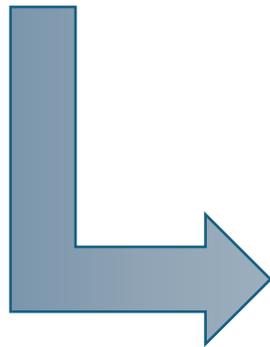
76,800 processes concurrently open/read/write a single file, causing significant congestions.  
→ Delegate all I/O to one aggregator per node. Operate on one file per node.

After tuning:

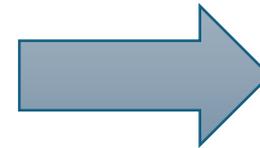
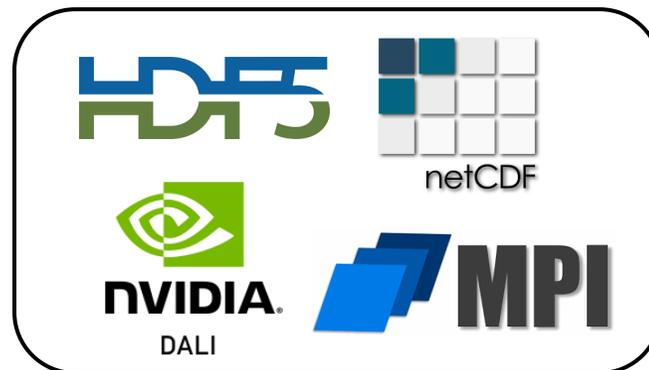
- I/O time: 286s → 81s. (3.5x)
- Write bandwidth → ~300GB/s. This is still far away from the system peak performance (5TB/s).

# Optimizing HPC File Systems?

## HPC Application



## I/O Library and middleware



## HPC File System



# Optimizing HPC File Systems?

Traditional HPC file systems are **global resources shared by all users and jobs**. They are static and unable to adapt to different workloads, making it basically impossible to optimize for a single job. This limitation affects all applications.

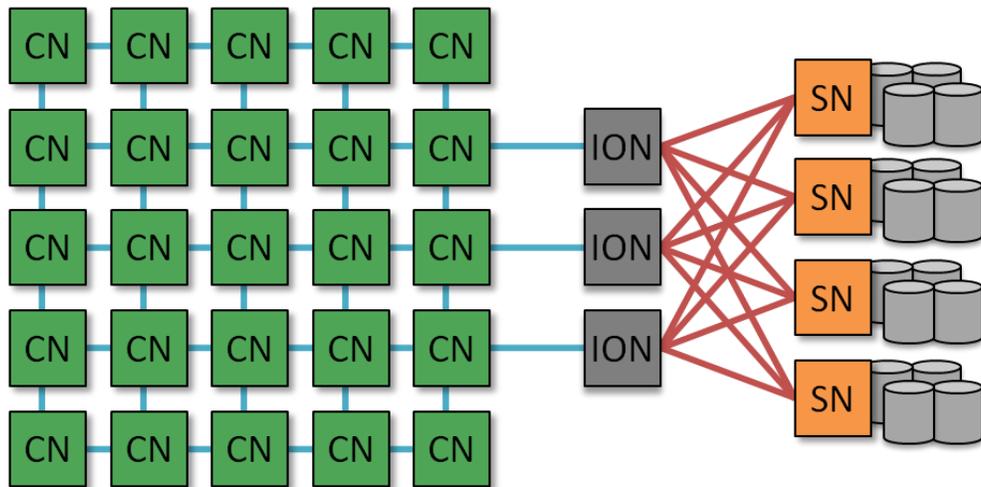


# Burst Buffers: Yet Another Storage Layer

*“Tape is Dead. Disk is Tape. Flash is Disk.” (at CIRG’07)*

*Jim Gray*

*(1998 Turing Award Winner)*

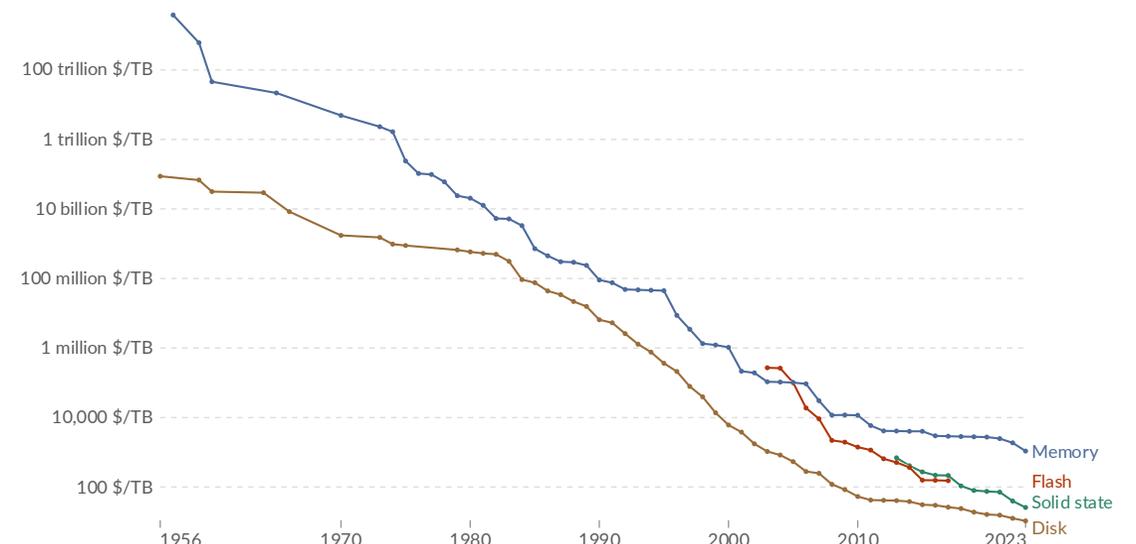


CN: Compute Node; ION: I/O Node; SN: Storage Node

## Historical price of computer memory and storage

This data is expressed in US dollars per terabyte (TB), adjusted for inflation. "Memory" refers to random access memory (RAM), "disk" to magnetic storage, "flash" to special memory used for rapid data access and rewriting, and "solid state" to solid-state drives (SSDs).

Our World  
in Data



Data source: John C. McCallum (2023); U.S. Bureau of Labor Statistics (2024)

OurWorldinData.org/technological-change | CC BY

Note: For each year, the time series shows the cheapest historical price recorded until that year. This data is expressed in constant 2020 US\$.

Figures courtesy of Glenn K. Lockwood.

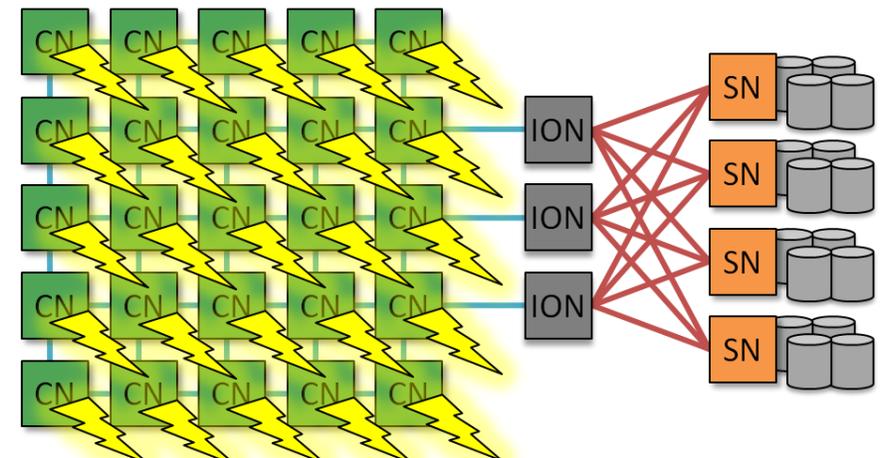
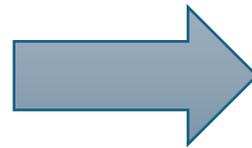
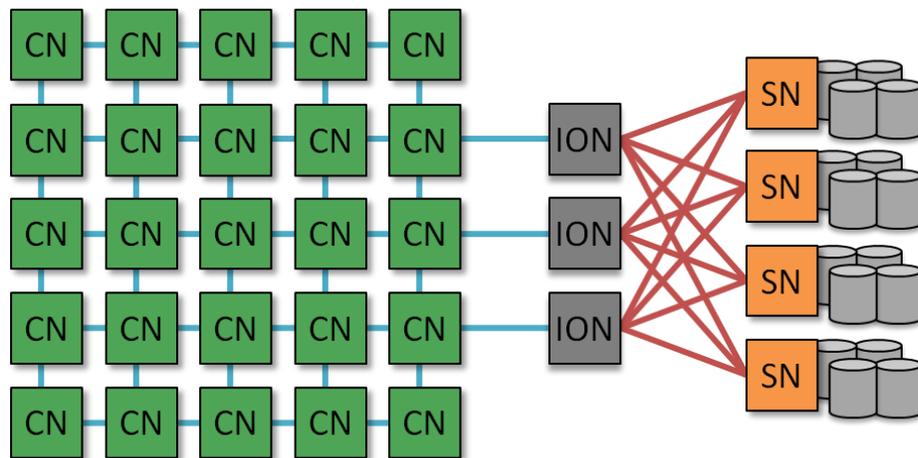
<https://blog.glennlockwood.com/2017/03/reviewing-state-of-art-of-burst-buffers.html>

Tutorial link: <https://wangchen.io/unifyfs-dyad-tutorial>

# Burst Buffers: Yet Another Storage Layer

Node-local burst buffer:

- Attach one SSD to each compute node.
  - Scales linearly.
- **Only accessible from the attached node.**
  - Users need to manage data transfers across layers and between nodes.



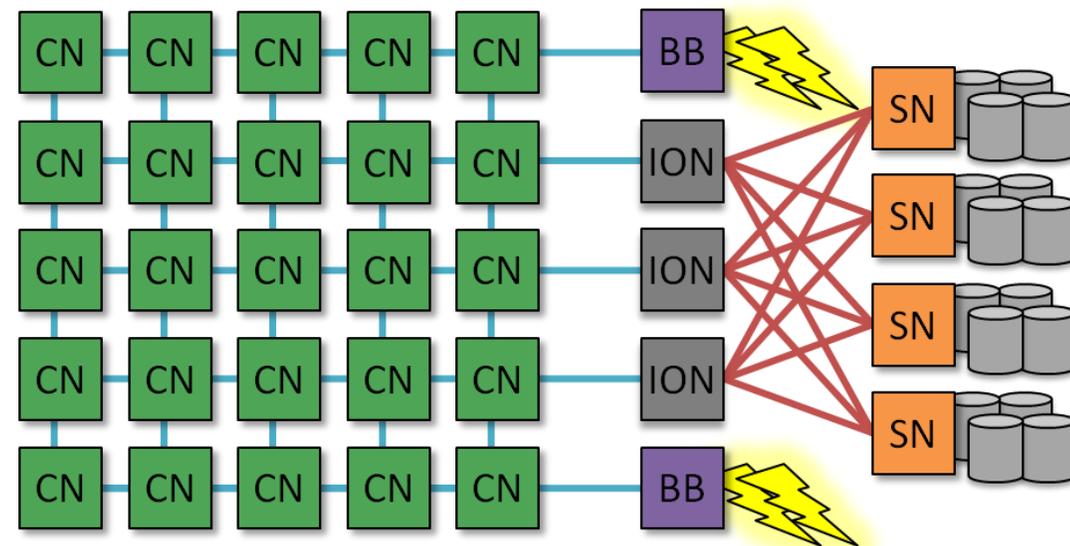
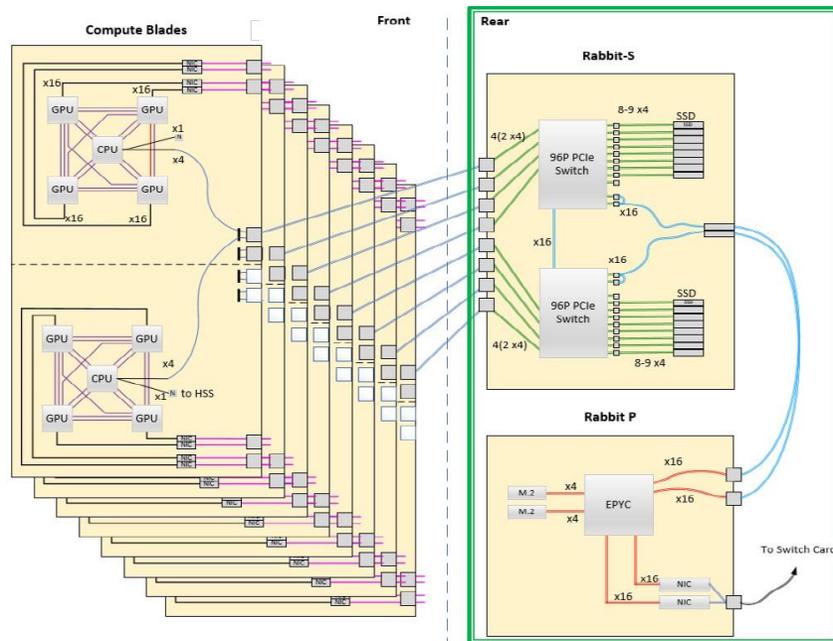
CN: Compute Node; ION: I/O Node; SN: Storage Node

Node-local Burst Buffer

# Burst Buffers: Yet Another Storage Layer

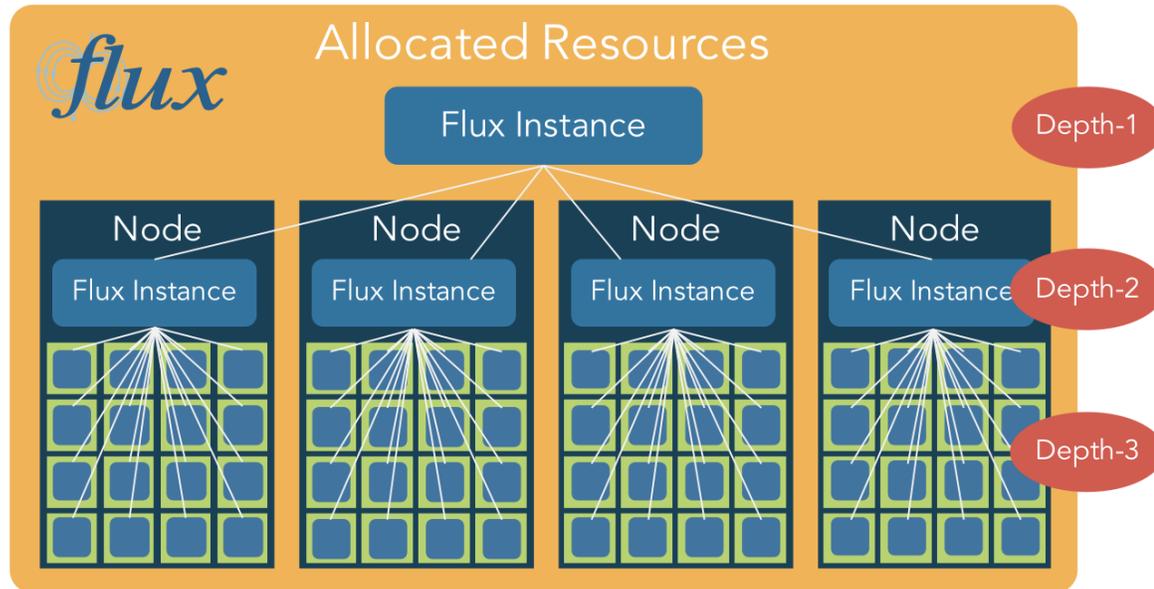
The “Rabbit” way:

- Each Rabbit node consists of  $N$  SSDs and one processor.
- Two Rabbit nodes sit in each rack; Each is directly connected to all compute nodes within the same rack.
- Provide a shared address space to all compute nodes.



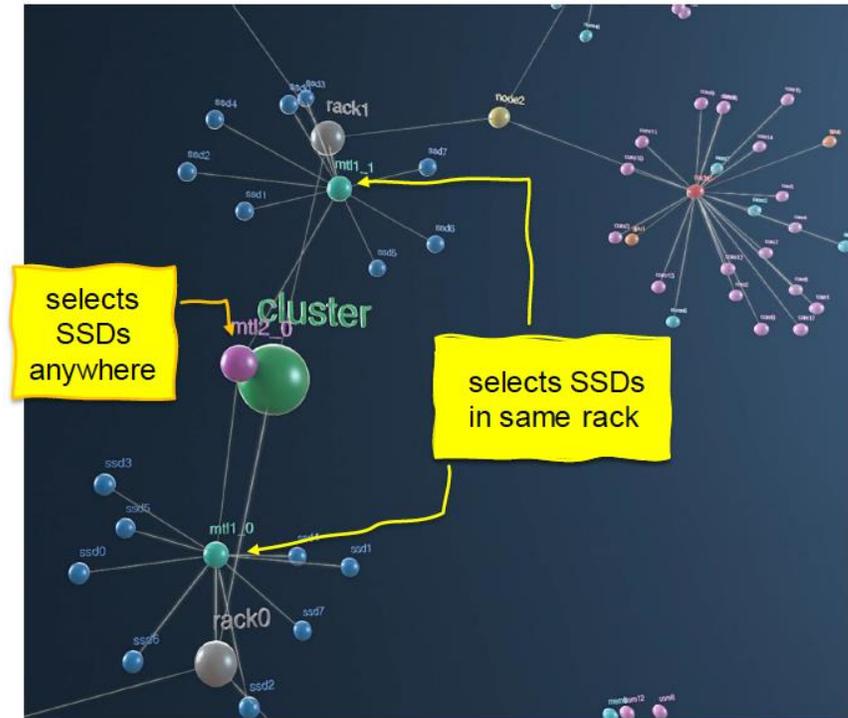
BB: Burst Buffer Node

# Flux hierarchically manages of resources and jobs for scalability



- Hierarchical and modular management
- Mitigates the centralized scheduler bottleneck
- Deployed at El Capitan
  - Ranked 1st in top500 supercomputers

# Flux's graph-based resource representation addresses challenges with Rabbit storage system



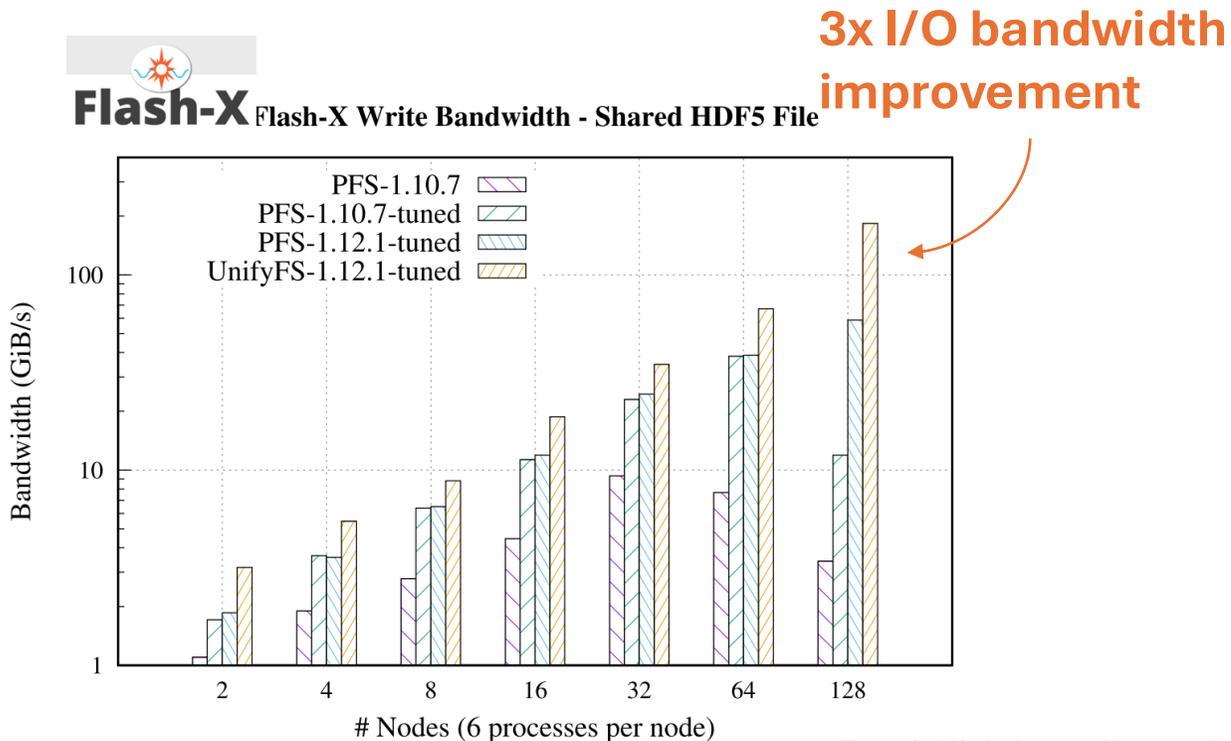
- Better suited for dynamic and heterogeneous resources
- Rabbit is a near-compute disaggregate storage system that can be dynamically allocated per job as either shared or node-local.



# UnifyFS and DYAD

**UnifyFS:** A specialized burst buffer parallel file system for supercomputers. Designed for **write-heavy HPC applications**.

- <https://github.com/LLNL/UnifyFS>

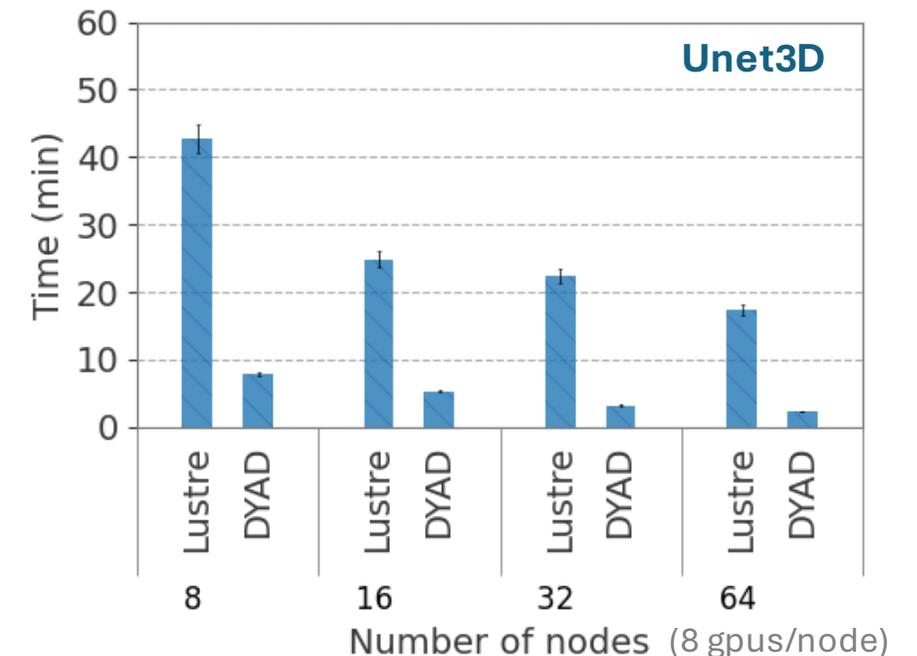


Tutorial link: <https://wangchen.io/unifyfs-dyad-tutorial>

**DYAD:** is a data streamer optimized for deep learning (DL) training.

- <https://github.com/flux-framework/dyad>

Up to 8x training time improvement



# Thank You!

**UnifyFS:** A specialized burst buffer parallel file system for supercomputers. Our goal is to provide easy, portable, and fast support for I/O-intensive applications.

- <https://github.com/LLNL/UnifyFS>
- *Michael Brim, Adam Moody, Seung-Hwan Lim, Ross Miller, Swen Boehm, Cameron Stanavige, Kathryn Mohror, Sarp Oral, “UnifyFS: A User-level Shared File System for Unified Access to Distributed Local Storage,” 37th IEEE International Parallel & Distributed Processing Symposium (IPDPS), May 2023.*

**DYAD:** is a data streamer optimized for deep learning (DL) training and scientific workflows.

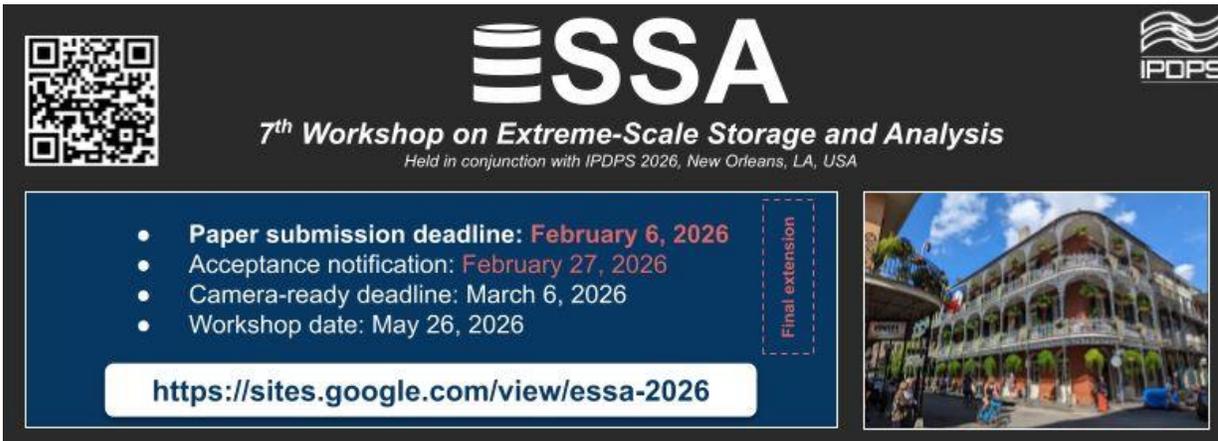
- <https://github.com/flux-framework/dyad>
- DYAD data movement coordination strategy reduces network contention and thus improving data movement
- DYAD employs a hierarchical sample discovery technique to isolate metadata accesses, and that improves lookup throughput
- DYAD’s novel streaming RPC over RDMA protocol boosts inter-node access
- DYAD optimizes large-scale DL workloads by 8.2x as compared to state-of-the-art solutions.



# Call for Papers

**7th International Workshop on Extreme-Scale Storage and Analysis (ESSA 2026)**, held in conjunction with IPDPS (New Orleans, May 2026)!

Deadline: Feb. 6, 2026.



The banner features a QR code on the left, the ESSA logo in the center, and the IPDPS logo on the right. Below the ESSA logo, it reads "7th Workshop on Extreme-Scale Storage and Analysis" and "Held in conjunction with IPDPS 2026, New Orleans, LA, USA". A list of key dates is provided in a blue box, and a photograph of a building is shown on the right side of the banner.

**ESSA**  
7<sup>th</sup> Workshop on Extreme-Scale Storage and Analysis  
Held in conjunction with IPDPS 2026, New Orleans, LA, USA

- Paper submission deadline: **February 6, 2026**
- Acceptance notification: **February 27, 2026**
- Camera-ready deadline: March 6, 2026
- Workshop date: May 26, 2026

<https://sites.google.com/view/essa-2026>

Final extension



## Special Issue

### New Advances in Parallel and Distributed Computing

#### Message from the Guest Editors

This Special Issue focuses on emerging advances that enhance the performance, scalability, and intelligence of parallel and distributed computing systems. It emphasizes new technologies and paradigms in high-performance computing (HPC) in the era of AI, particularly in the context of the convergence of HPC, cloud, and edge computing. Topics of interest include, but are not limited to, the following:

- Advanced architectures for the HPC–cloud–edge computing continuum.
- Parallel and distributed algorithms for large-scale AI and large language models (LLMs).
- Data-driven optimization techniques for complex scientific and AI workflows.
- Cross-facility resource management and scheduling.
- Workflow management frameworks for distributed and federated infrastructures.
- Performance modeling, optimization, and benchmarking of distributed systems.
- Disaggregated storage and high-throughput data flow optimization.
- Green and sustainable computing strategies for data-intensive workloads.
- Integration of HPC, quantum, and AI accelerators into distributed platforms.
- Emerging applications, including digital twins, autonomous systems, IoT, and 6G networking.



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