

# Optimizing Storage Utilization in the HL-LHC Era

*Qiulan Huang, Vincent Garonne*  
Scientific Data and Computing Center (SDCC)  
Brookhaven National Laboratory

US ATLAS Computing Facilities Meeting  
11/30/2022

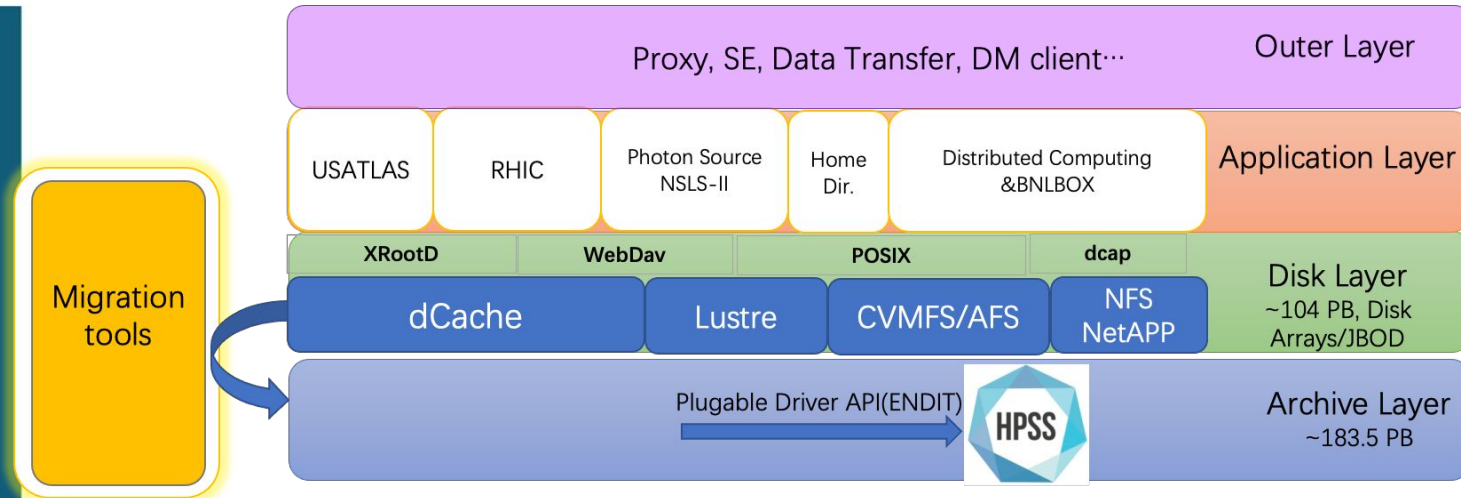


@BrookhavenLab

# LDRD project

- Project title: Data popularity, placement optimization and storage usage effectiveness at Data Center
- Investigators: Qiulan Huang(PI, SDCC), Vincent Garonne(SDCC), Ai Kagawa (CSI, Computational Science Initiative), Xin Dai (CSI)
  - AI/ML team(CSI): Develop the state-of-the-art data usage prediction model
  - Storage team(SDCC): Quality Assurance & Control of training data & performance monitoring
- Term: 2022- 2024

# Current Storage at SDCC



**Storage "classes"** (DISK,TAPE, FLASH,etc) have various cost and quality of services, E.g.,

- slow and cheaper storage for archiving vs. fast and expensive storage for quick and efficient data access

**Two storage classes : Disk and Tape**

- Disk storage: ~104 PB (fast storage)
- Tape storage: ~183.5PB(slow storage)

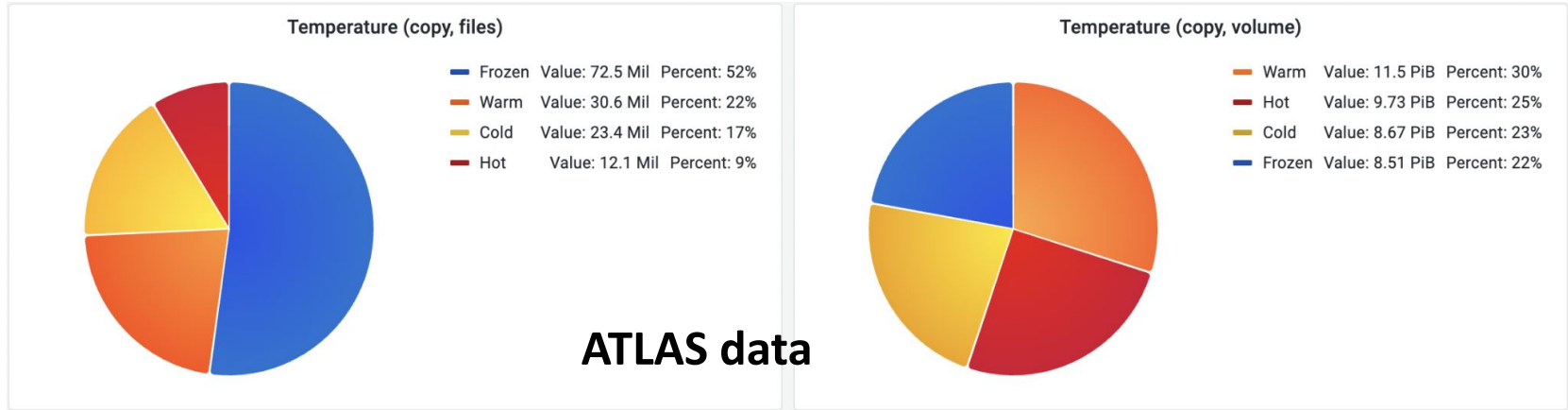
- Several millions of files created/deleted/transferred per day
- Peak traffic at **100GB/s**
- Data accessed by millions of jobs per day
- Hundred of thousands of storage devices

# Data temperature of ATLAS

~50% files are frozen, ~22% of total size (~8PB) is frozen

Cannot be solved by adding more storage but by better use of storage resources

- ★ Plot done more than one year ago on all dCache replica categories (cache, precious, etc) without curation



**Hot:** Last access in the last month

**Cold:** Last access between 6 months and one year

**Warm:** Last access in the last 6 months

**Frozen:** Not accessed in the last year

⇒ How to map data category with activities and storages classes?

⇒ How to provide a better accurate view for more experiments and technology: dcache, lustre (robinhood), etc

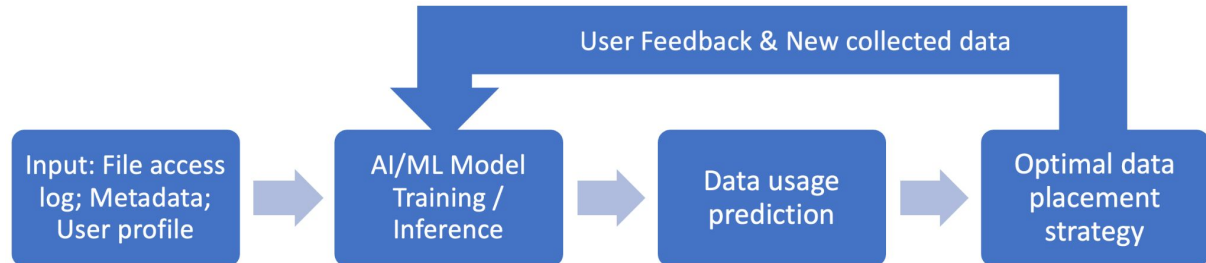
# AI/ML For Storage Optimization

## Motivation

- In the current multi-tier storage "class" system at the Data Center
  - Unused data is stored on expensive storage
  - Fast IO storage is not currently used effectively

## Goals

- Design an efficient monitoring platform to collect the relevant information from various distributed data sources
- Develop an optimal data management system for the data center to maximize usable space while minimizing access latency, within budget, hardware, and compliance constraints
  - Heavy use of storage, metadata and data popularity information
  - Detect early failures and pathological usage pattern
  - Develop a precise AI/ML prediction model to possibly forecast the future usage of the data
  - Orchestration of data for optimal movement and placement

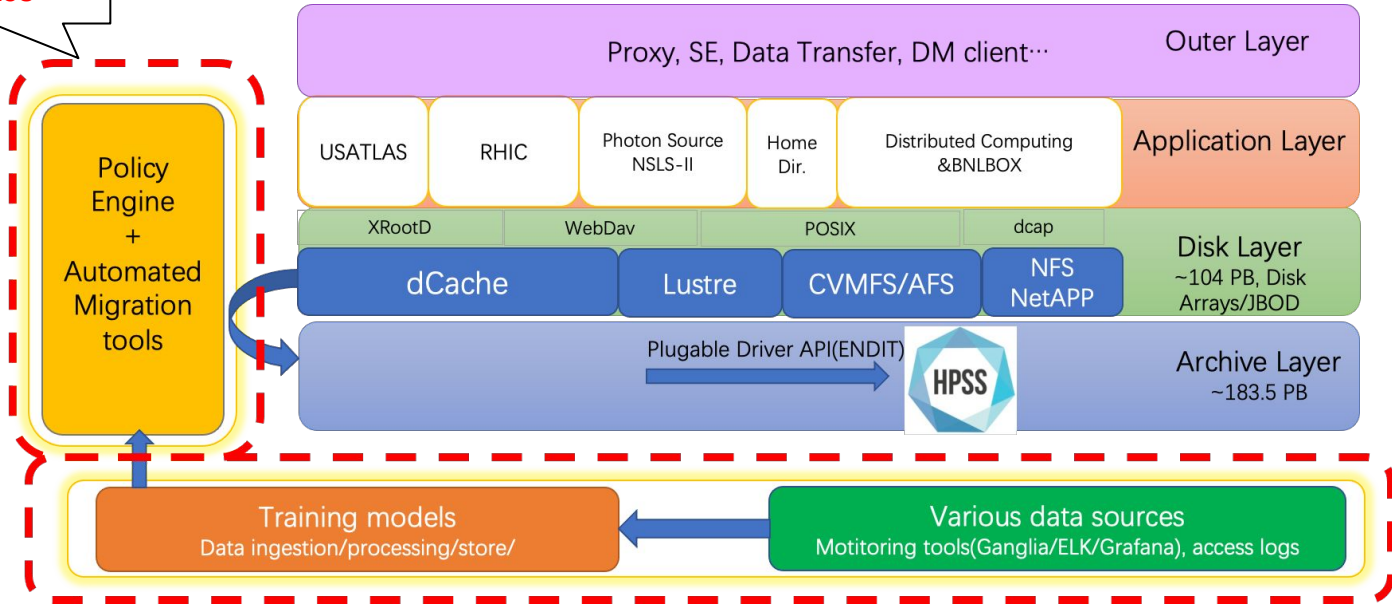


# What benefits of the LDRD

- Help in making important business decisions, policies and lead to more cost effective operations
  - Implement the automatic migration of data between different levels of storage
  - **Measurable objectives**, e.g.,  $\min(\text{access time})$  and  $\max(\text{total usable space})$  on a same budget
- Define optimal data placement decisions
  - Data distribution in a reasonable tiered storage and provide good IO performance for scientists

# Novel Storage at SDCC

The LDRD contributions are shown in the red dashed boxes



- Introduce AI/ML combined with our big data, conduct extensive training, and build data prediction model
- Define the policy engine and perform automated migration actions transparently

# Status

- Collect and analyze data samples is undergoing
  - dCache billing, chimeradb, access logs,etc
  - [File id],[File size],[access\_time],[disk usage],etc
- Machine learning analyzes all of the collected data samples
  - Start from a simple definition to predict near future(like 30 days) data usage
- Biweekly meetings are scheduled