

Towards new storage interfaces – chance or curse?

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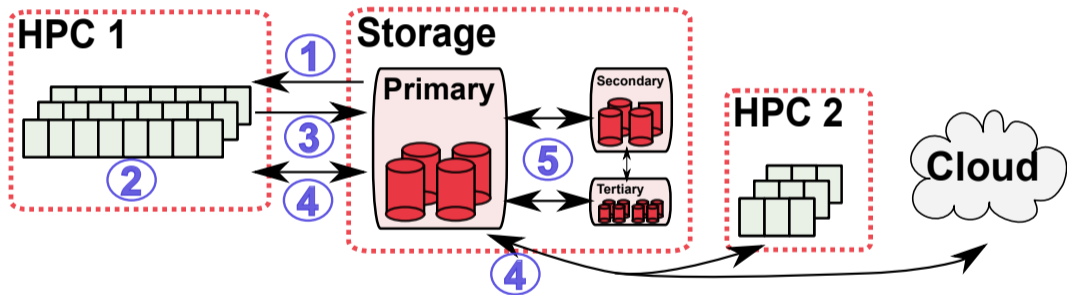
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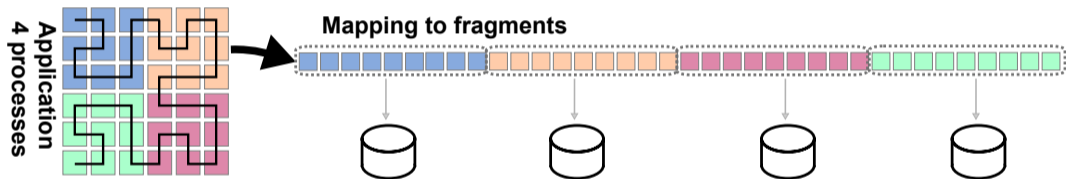
Outline

- 1 HPC Storage Landscape
- 2 Thoughts
- 3 Better Interfaces?
- 4 Community APIs

HPC Storage Usage: Workflows

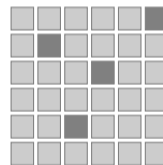
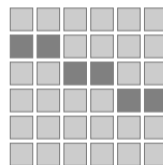
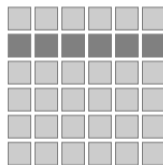
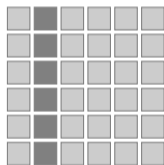
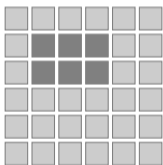


Mapping of a 2D field from a parallel application to storage



Mapping for Pre-Post

User-defined analysis of ND datasets leads to various patterns



User Perspective: Accessing Data

Multitude of data models

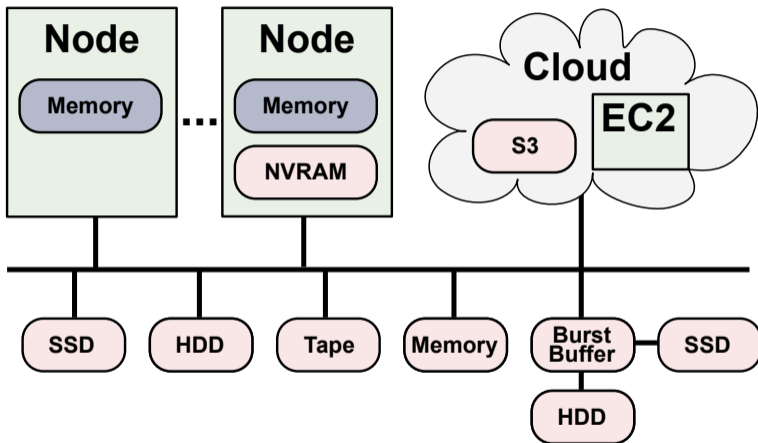
- POSIX File: Array of bytes
- HDF5: Container like a file system
 - Dataset: N-D array of a (derived) datatype
 - Rich metadata, different APIs (tables)
- Database: structured (+arrays)
- NoSQL: document, key-value, graph, tuple

Choosing the right interface is difficult – workflow may involve several

Properties / qualities

- Namespace: Hierarchical, flat, relational
- Access: Imperative, declarative, implicit (`mmap()`)
- Concurrency: Blocking vs. non-blocking
- Consistency semantics: Visibility and durability of modifications

Storage Landscape of Future Systems



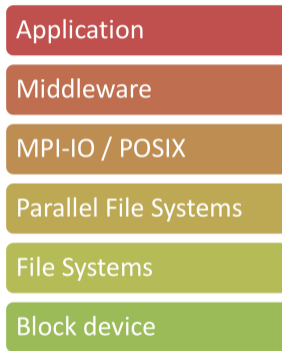
HPC system with compute nodes and storage

Outline

- 1 HPC Storage Landscape
- 2 Thoughts**
 - Storage stack
 - Performance Optimization
- 3 Better Interfaces?
- 4 Community APIs

Peeking at the Current I/O Stack – System Perspective

- Coexistence of access paradigms
 - File (POSIX, ADIOS, HDF5), SQL, NoSQL
- Semantical information is lost through layers
 - Suboptimal performance
- Reimplementation of features across stack
 - Unpredictable interactions
 - Wasted resources
- Restricted (performance) portability
 - Optimizing each layer for each system?



Example I/O stack

Limitations of the current software stack

Platform

- 1 Zoo of interfaces
- 2 Low-level storage APIs
- 3 Loss of semantical information
- 4 Interference of applications / lack of coordination
- 5 All data treated identically

Software

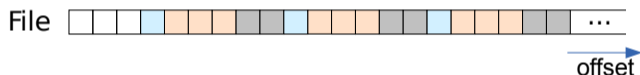
- 1 Explicit workflows
- 2 Unclear access patterns (users, sites)
- 3 No performance awareness
- 4 Lack of technological knowledge (from users, for tweaking)
- 5 Manual tiering (or with policies)

Semantical Gap of File Access (1)

Applications work with (semi)structured data

- Vectors, matrices, n-Dimensional data

A file is just a sequence of bytes!



Applications/Programmers must serialize data into a flat namespace

- Uneasy handling of complex data types
- Mapping is performance-critical (on HDDs)
- Vertical data access unpractical (e.g. to to pick a slice of multiple files)

Semantical Gap of File Access (2)

Information hidden from file systems

- Data types
- Data semantics
- Value of data
- Type: Checkpoint, computed, original, logfile
- Data lifecycle: production, usage, deletion

Characteristics can even vary within a file, e.g. for metadata

Storage systems could use this information for

- Improving performance: Automatic tiering, caching, replication
- Simplifying management: ILM, offering alternative data views
- Correctness: Ensuring data consistency

Performance Tweaks

- There are many options to tune the I/O-stack
 - API: `posix_fadvise()`, HDF5 properties, open flags, cache size
 - Via command line: `lfs setstripe`
 - Setup/initialization of a storage system
 - Mounting options and `procfs` settings
- Many options are of technical nature
 - Performance gain/loss depend on hardware, software
 - Specific to file system, API (MPI, POSIX, HDF5)
 - Many types of hints/tweaks are not portable
- Performance loss forces us to use these optimization

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Usually we are losing system performance!

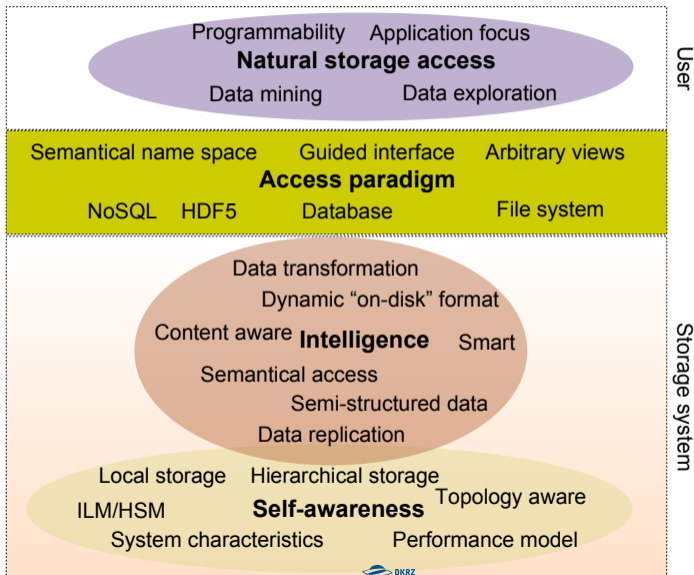
Critical Discussion

Questions from the users' perspective

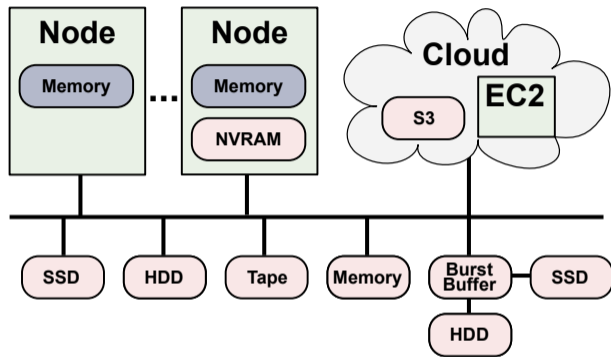
- Why do I have to organize the file format?
 - It's like taking care of the memory layout of C-structs
- Why do I have to convert data between storage paradigms?
- Why must I provide system specific performance hints?
 - It's like telling the compiler to unroll a loop exactly 4 times
- Why can't I rely on a correct implementation of the consistency model?
 - Parallel file systems have performance issues with most models
- Why is a file system not offering the consistency model I need?
 - My application knows the required level of synchronization

Would you rather like to code an actual application?

Personal Vision: Towards Intelligent Storage Systems and Interfaces



Coexistence of Storage Systems vs. Tiering



- We shall be able to use all storage technologies concurrently
 - Without explicit migration etc. put data where it fits
 - Administrators just add a new technology (e.g., SSD pool) and users benefit
 - Should be steered by a standard and open interface
 - Open ecosystem for any vendor...

Additional Responsibilities of Storage System

- Mapping of data structures
- Flexible semantics
- Compute offloading, see success of big data tools
- Tight integration of workflows
- Advanced performance assessment
- Namespace based on metadata
- Management tools
- ...

Outline

- 1 HPC Storage Landscape
- 2 Thoughts
- 3 Better Interfaces?**
 - Guided Interfaces
 - Compression Example
 - SCIL
 - ESDM
- 4 Community APIs

Exascale10 Initiative Term: Guided Interfaces

Guiding vs. automatism vs. technical hints

Users provide additional information to guide an intelligent system.

The I/O stack may exploit this information or not.

Systems could improve over time by using the information better.

Information which could be provided by users

- Data types
- Semantics
- Relations between data
- Lifecycle (especially usage)

Several issues have been addressed in different access paradigms.

Also some behavioral hints exist: `open()` flags, `fadvise()`, ...

Compression Research: Involvement

- **Scientific Compression Library (SCIL)**
 - Separates concern of **data accuracy** and **choice of algorithms**
 - Users specify necessary accuracy and performance parameters
 - Metacompression library makes the choice of algorithms
 - Supports also new algorithms
 - Ongoing: standardization of useful compression quantities
- Development of algorithms for lossless compression
 - MAFISC: suite of preconditioners for HDF5, pack data optimally, reduces climate data by additional 10-20%, simple filters are sufficient
- Cost-benefit analysis: e.g., for long-term storage MAFISC pays off
- Analysis of compression characteristics for earth-science related data sets
 - Lossless LZMA yields best ratio but is very slow, LZ4fast outperforms BLOSC
 - Lossy: GRIB+JPEG2000 vs. MAFSISC and proprietary software
- Method for system-wide determination of ratio/performance
 - Script suite to scan data centers...

SCIL: Supported User-Space Quantities

Quantities defining the residual (error):

absolute tolerance: compressed can become true value \pm absolute tolerance

relative tolerance: percentage the compressed value can deviate from true value

relative error finest tolerance: value defining the abs tol error for rel compression for values around 0

significant digits: number of significant decimal digits

significant bits: number of significant decimals in bits

field conservation: limits the sum (mean) of field's change

Quantities defining the performance behavior:

compression throughput

decompression throughput

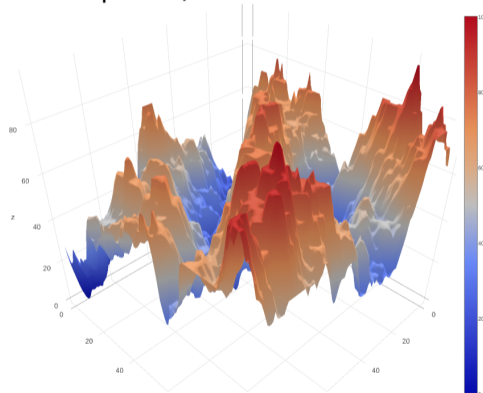
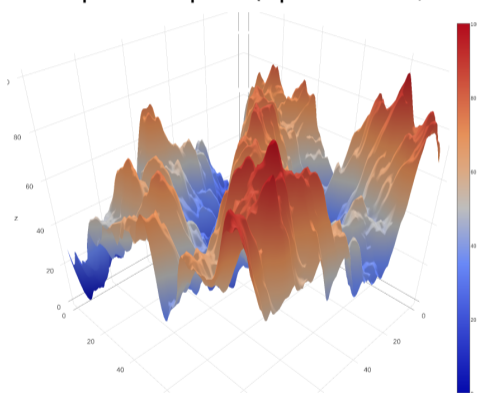
- in MiB or GiB, or relative to network or storage speed

Aim to standardize user-space quantities across compressors!

See <https://www.vi4io.org/std/compression>

SCIL Provides Typical Synthetic Data

Example: Simplex (options 206, 2D: 100x100 points)



Right picture compressed with Sigbits 3bits (ratio 11.3:1)

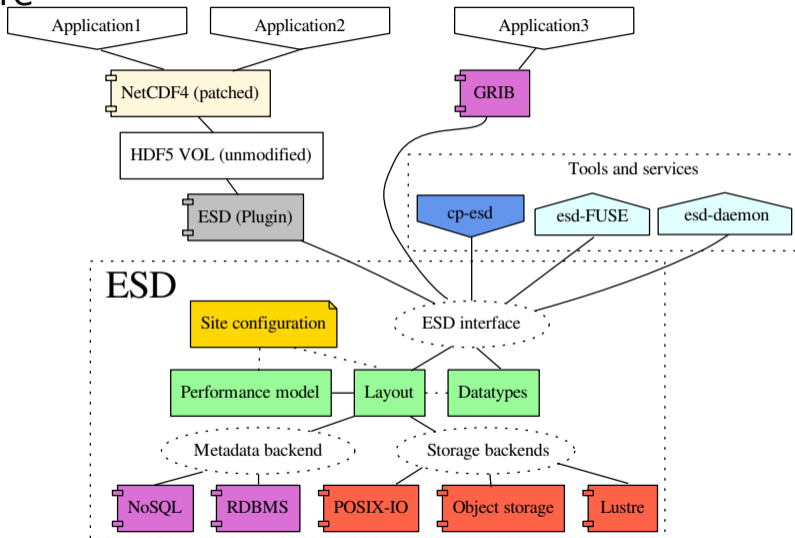
Ongoing Activity: Earth-Science Data Middleware

Part of the ESiWACE Center of Excellence in H2020

Design Goals of the Earth System Data Middleware

- 1 Understand application data structures and scientific metadata
- 2 Flexible mapping of data to multiple storage backends
- 3 Placement based on site-configuration + performance model
- 4 Site-specific optimized data layout schemes
- 5 Relaxed access semantics, tailored to scientific data generation
- 6 A configurable namespace based on scientific metadata

Architecture

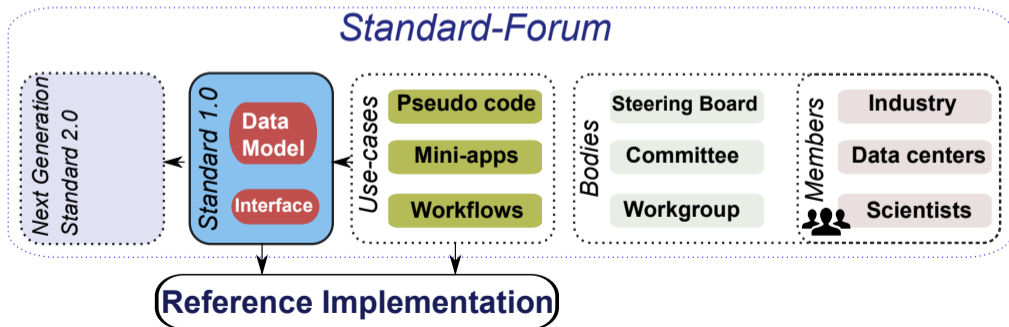


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 - Community

A Potential Approach in the Community: Following MPI

- The **standardization** of a high-level *data model & interface*
 - Targeting data intensive and HPC workloads
 - Lifting semantic access to a new level
- Development of a reference implementation of a **smart runtime system**
 - Implementing key features
- Demonstration of benefits on socially relevant data-intensive apps



API Key Features

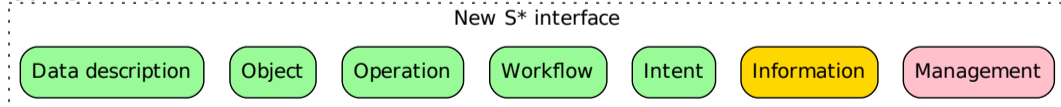
- High-level data model for HPC
 - Storage understands data structures vs. byte array
 - Relaxed consistency
- Semantic namespace
 - Organize based on domain-specific metadata (instead of file system)
 - Support domain-specific operations and addressing schemes
- Integrated processing capabilities
 - Offload data-intensive compute to storage system
 - In-situ/In-transit workflows
- Workflow management
 - Managed data-driven workflow
- Performance-portability
 - Guided interfaces: Intents vs. technical hints
- Enhanced data management features
 - Embedded performance analysis
 - Resilience, import/export, ...

API development

Development of the data model

- Establishing a Forum similarly to MPI
- Define data model for HPC
 - Must be beneficial for Big Data + Desktop, too
- Open board: encourage community collaboration

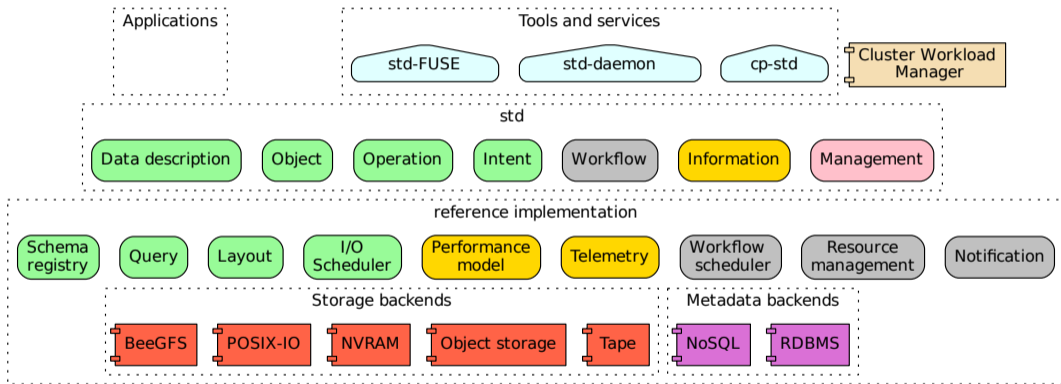
Current Draft



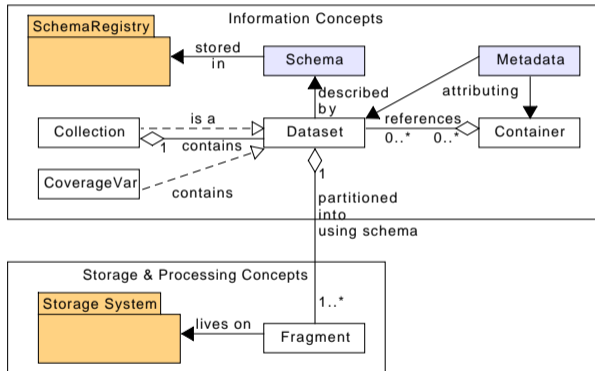
Reference Implementation: Goals

- Semantic access
 - Search and access based on metadata
- Self-aware
 - Understand performance characteristics
- Automatic layouting + smart data replication
 - Adapt data layout during runtime
- Managed workflows
 - Scheduler considers compute and I/O requirements
- Compatibility

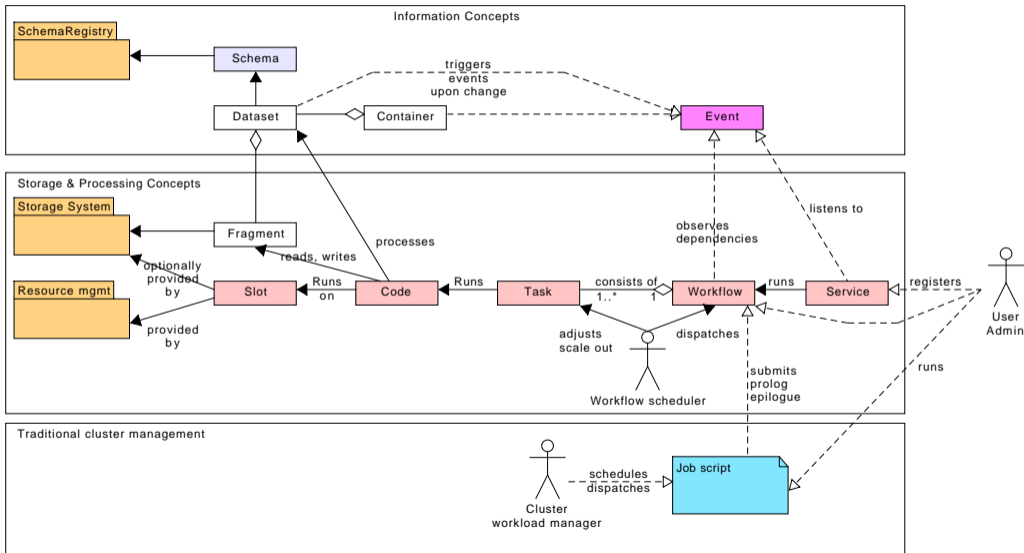
Architecture Draft



Data Model



Processing Model



The ESiWACE project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No **675191**



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