

A Guide to the FENIX Storage Infrastructure

Offerings, Architectures, and Best Practices

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www.fenix-ri.eu

Agenda

- Introduction
- FENIX Storage Offerings
- Parallel Filesystems
 - Architecture
 - Usage (libraries)
 - Best Practices
- Object Stores
 - Architecture
 - Usage (Python, CLI)
 - Best Practices
- Comparison
- Conclusion, and Q&A





STORAGE OFFERINGS



Introduction

- Two environments plus associated storage
 - HPC environment: parallel filesystems Active Data Repositories (ACD)
 - Cloud environment: object stores
 Archival Data Repositories (ARD)
 - Goals of this talk
 - storage architectures
 - methods of access
 - short guide on performance
 - Disclaimer: vast generalisations



Active Data Repositories (ACD)

- Scalable Compute Services
- ACD usually built as Parallel Filesystems
 "Workhorse" of HPC storage
- Well-known interface: POSIX
 - emulates local storage
 - handles parallel access
 - strong consistency semantics
- Highly optimised for bandwidth
- Examples: SpectrumScale, Lustre, BeeGFS

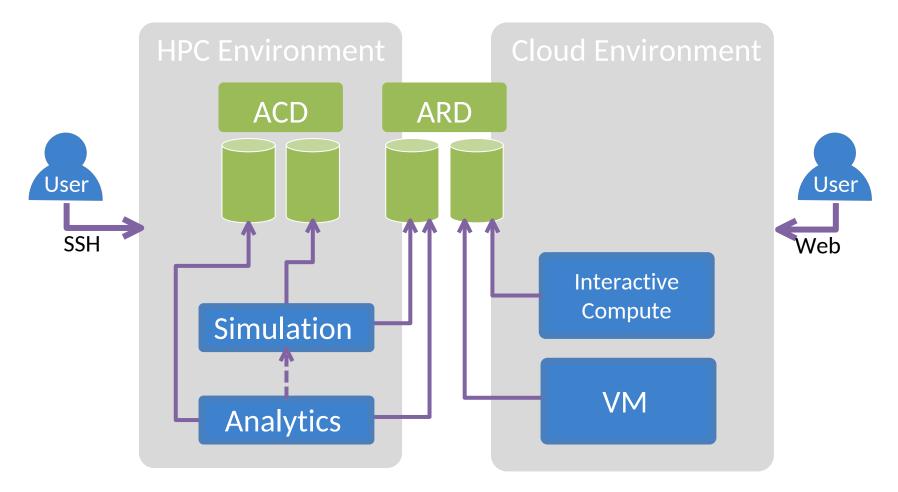


Archival Data Repositories (ARD)

- Interactive Compute/Virtual Machine Services
 Implemented as Object Stores
 - very popular in Cloud and Web contexts
 - less common in HPC
- Simpler than (parallel) filesystems
 - no guarantees on ordering, but: atomicity
 - put/get semantics
 - flat hierarchies
- Examples
 - Amazon S3
 - OpenStack SWIFT (used in FENIX)



3.048km View





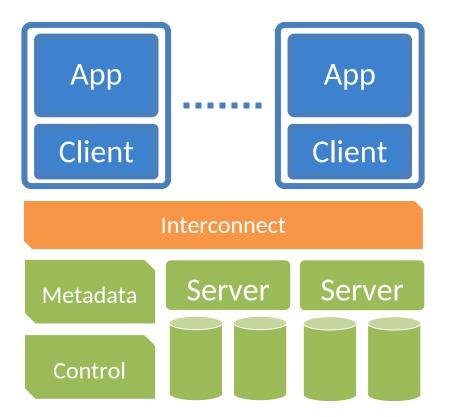




Part I ACTIVE DATA REPOSITORIES



Typical Organisation



Files are distributed in blocks

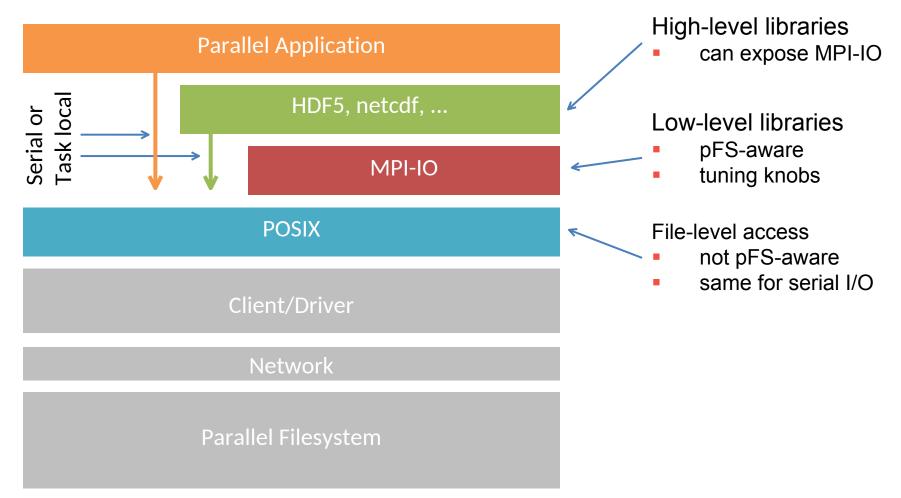
- striping for performance
- RAID for resilience

Global metadata

- directory contents
- Size, RWM times, ...
- can be a source of contention
- Consistency model options
 - relaxed: not POSIX
 - using locks: scalability issues
 - various trade-offs in between



Methods of Access





User Interface

- POSIX: open, write, read, close, ...
 - Files: byte-level access
 - Directories: group files and other directories
- Strong guarantees on Ordering and Visibility
 - eg Read after Write
 - hold in multi-process environment
 - especially: exclusive ownership
 - hard to implement performantly
- Reachable inside a site
 - Access control: owner/group, permissions



Do's and Don'ts

Common Pitfalls

- Many small files
 - metadata
 - false sharing
- Concurrent use of files
 - false sharing
 - locks
- Random access
 - caches
 - pre-fetching
- Redundant accesses
 - bandwidth

Optimised Use

- A few, **not** one, files
 - one per node
 - at least block sized
- Contiguous access
 - aligned to FS blocks
 - block granularity
 - exclusive to a process
- Reduce operation count
 - read once, broadcast
 - coalesce writes
 - especially within nodes



```
import numpy as np
import h5py as h5
from mpi4py import MPI
# Setup parameters
N = 4
comm = MPI.COMM WORLD
size, rank = comm.size, comm.rank
# Generate a block of data containing our rank
data = np.zeros((N,), dtype=np.float32) + rank
# Create a pHDF5 file and a chunked dataset inside
fd = h5.File("ranks.h5", "w", driver="mpio", comm=comm)
fd.create dataset("R", dtype=np.float32,
                  shape=(N*size,), chunks=(N,),)
# All tasks coordinate, each writes one block
with fd["R"].collective:
```

fd["R"][N*rank : N*(rank + 1)] = data[:]



```
$ mpirun -n 4 python3 ranks.py
$ h5dump ranks.h5
HDF5 "ranks.h5" {
  GROUP "/" {
   DATASET "R" {
       DATATYPE H5T IEEE F32LE
       DATASPACE SIMPLE { ( 16 ) / ( 16 ) }
       DATA {
       (\bigcirc): \bigcirc, \bigcirc, \bigcirc, \bigcirc, \bigcirc,
             1, 1, 1, 1,
             2, 2, 2, 2,
             3, 3, 3, 3
```

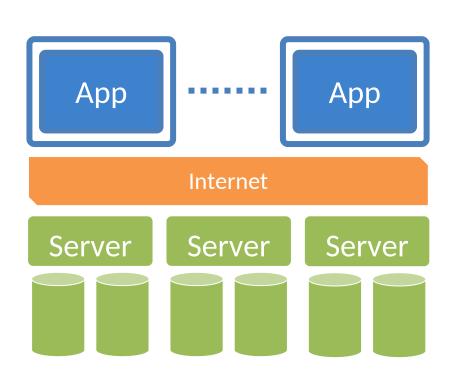




Part II ARCHIVAL DATA REPOSITORIES



Typical Organisation



Data model: abstract objects

- arbitrary size
- metadata inline
- resilient storage
- (optionally) content addressable

Atomic operations

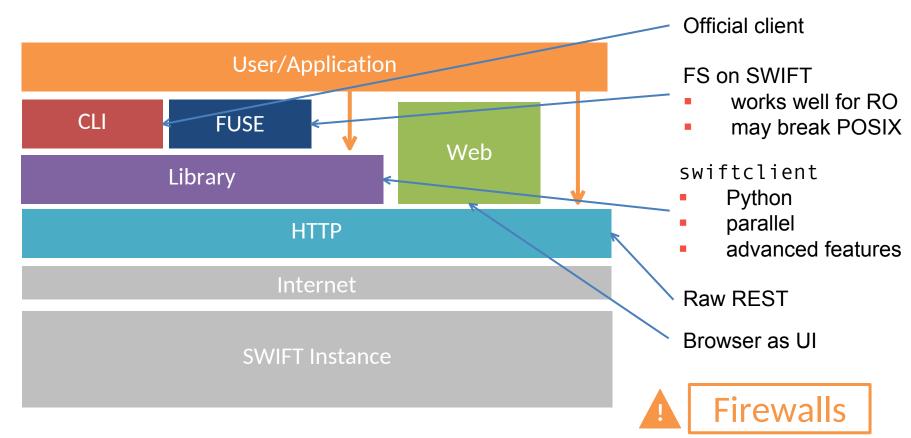
- all-or-nothing
- no ordering guarantees

Performance

- (usually) slower than pFS
- high latency/low IOp/s
- HTTP overheads
- not a HPC network



Methods of Access





User Interface

HTTP REST

- Objects: atomic units of data
- Containers: group objects
- Flat hierarchy
- Atomic operations on objects
 - no guarantees on order
 - no byte-granular access
- Exposed on a public endpoint
 - access protected by authentication
 - traffic encrypted
- Access Control Lists



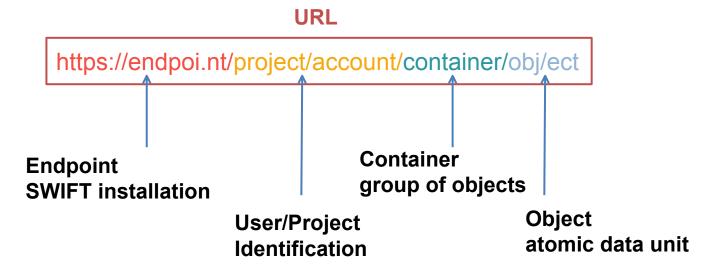
Introduction to REST

REpresentational State Transfer

- Pattern for Web API design
- Client-Server models
- Stateless, queries contain all context
- HTTP verbs express operations
- Resources identified by URL
- Atomic transitions between states

Verbs

- GET: Read content/List items
- HEAD: Get partial information, headers
- PUT/POST: Write to location
- DELETE: Remove item
- Less common: PATCH, TRACE, ...





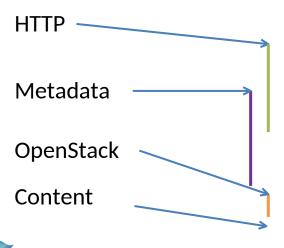
SWIFT Objects

Object: Any byte stream

- POST: Write metadata
- GET: Content + metadata
 - Can ask for most recent version (expensive)
 - Can use eTag for caching
- PUT: Write data
- HEAD: Retrieve Metadata
- DELETE: Remove object

Metadata

- Key-Value pairs of form X-Object-Meta-<name>
- Mime-Type
- Encoding
- Expire objects at/after
- Checksums



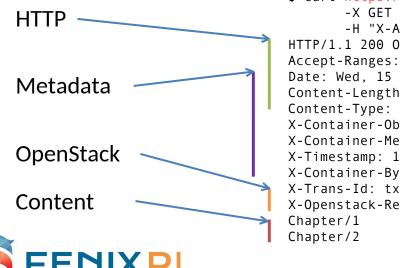
SWIFT Containers

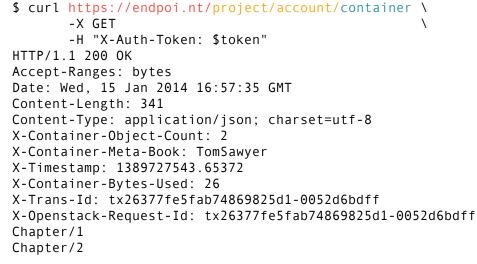
Containers

- POST: Write metadata
- GET: Metadata + List of objects
 - Paging/Sorting
 - Pseudo-directory operations
- PUT: Create container
- HEAD: Retrieve Metadata
- DELETE: Remove containers

Metadata

- Key-Value pairs of form X-Container-Meta-<name>
- Access control lists (ACLs)
- Quota
- Versioning
- Synchronisation





Do's and Don'ts

Common Pitfalls

- Incremental updates
 - redundant
 - consistency with other writers
- Small objects
 - Performance suffers
- Redundant operations
 - individual operations are slow
- Concurrent modifications
 - no locking, one writer wins

Optimised Use

- Use a few MB per object
 - see benchmarks later on
- Consider bundling data
 - HDF5, "tar", …
- Investigate compression
 - HDF5, "zip", ...
- Cache RO objects
- Update via RMW cycle
 - GET, update, PUT
 - condense updates
- Consider versioning objects
- Leverage atomicity



```
# setup a virtual environment
```

```
$ python3 -mvenv swift
```

\$ cd swift

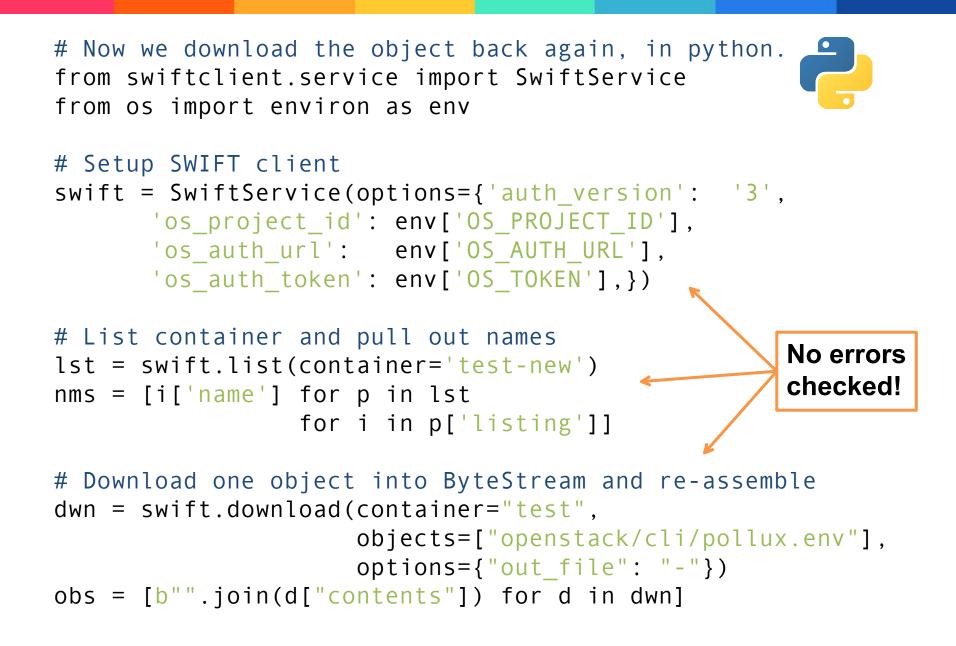
- \$ source bin/activate
- # install required software
- \$ git clone https://github.com/eth-cscs/openstack.git
- # authenticate against SWIFT
- \$ source openstack/cli/pollux.env
- > User: *****
- > Password: *****

```
# upload to a new container `test`
```

\$ swift upload test openstack

```
# check container
$ swift list test
... list of all files in `openstack` folder ...
```





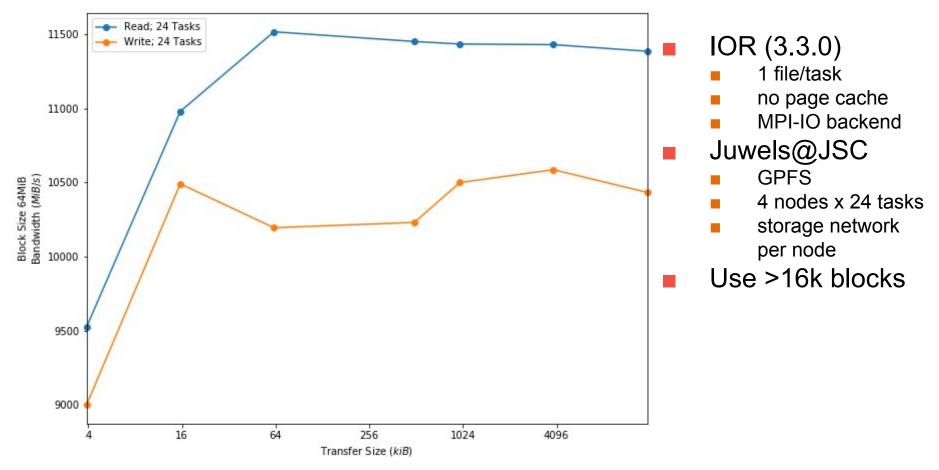






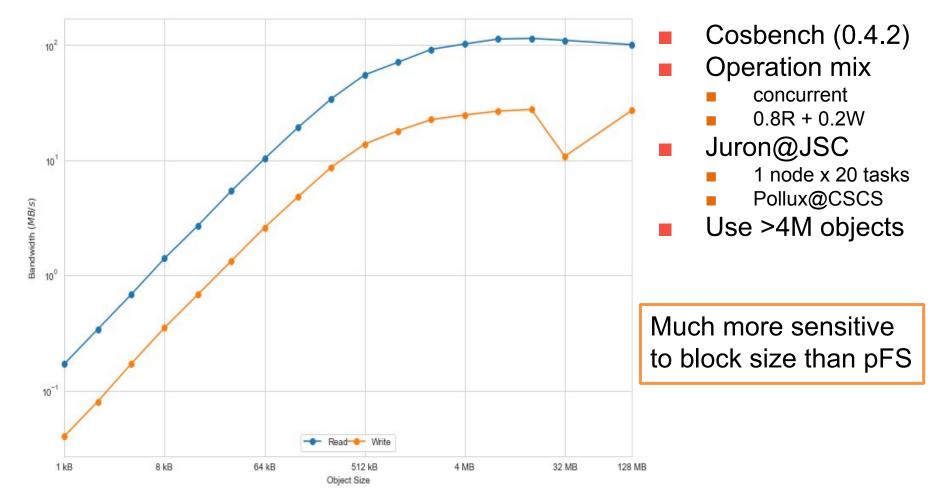


Performance Example: ACD





Performance Example: ARD











Take-home Messages

- Two breeds of storage with unique strengths
 - HPC + Active Data Repositories (ACD)
 - Cloud + Archival Data Repositories (ARD)
- ACD for high-performance I/O on-site
 - Fast, byte-oriented
 - Strong guarantees for coordinated access
- ARD for long term, federated storage
 - Slower, object-based
 - Atomic put/get
 - Data can be made accessible as a plain http link



Learn More

General information on SWIFT and I/O

- CSCS Object Storage user.cscs.ch/storage/object_storage
- HBP T7.2.3 I/O Guides
 - wiki.humanbrainproject.eu/bin/view/Collabs/how-to-dataaccess-and-efficient-io
- OpenStack SWIFT docs.openstack.org/swift/latest
- Talk to us
 - HBP Support support@humanbrainproject.eu
 - Fenix User Forum fenix-ri.eu/infrastructure/fenix-user-forum
- Get Access to FENIX fenix-ri.eu/access









References

- ior github.com/LLNL/ior
- cosbench github.com/open-io/cosbench
- https://www.nextplatform.com/2017/09/11/whats-bad-posix-io/
- High Performance Parallel IO (2019; Prabhat, Koziol et al)



POSIX Specification (2008)

After a **write** to a regular file has successfully returned

Any successful read from each byte position in the file that was modified by that write shall return the data specified by the write for that position until such byte positions are again modified.

Meaning: reads sync writes and are ordered relative to writes

Any subsequent successful write to the same byte position in the file shall overwrite that file data.

Meaning: Writes are ordered relative to each other.

Consequence

Parallel FS must either break POSIX compliance or very carefully perform synchronisation, which is expensive.



High Performance Storage Tier at JSC

- NVMe-based storage cluster at JSC
 - Tied into HPC interconnect
 - Small capacity
 - High bandwidth
- Part of FENIX effort
- SLURM integration for Co-Scheduling

Usecases

- Burst Buffer
- Check pointing
- Pre-staging
- Data processing campaigns
- Live visualization
- Complex workflows

