Data Management

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CHIME/FRB Telescopes

Data Governance

Industry Lingo

Architecture Storage & Operations Modelling & Design Integration Metadata Quality Documentation Security

Data Management

Academic Lingo

You have to do everything →
I will do everything!
I can do somethings...
What do I need to do to graduate?

Data

Overview

• Data

- Storage Mediums
- File Systems
- Archetypes

• Management

- Why?
- How?
- Best Practices

Case Study

• CHIME/FRB Telescopes

Storage Mediums











L1/L2/L3 Cache	Random Access Memory	Solid State Drives	Hard Disk Drives	Magnetic Tape
32→256 KB 256KB → 8MB	4GB → 128GB	$\textbf{120GB} \rightarrow \textbf{4TB}$	$500 ext{GB} ightarrow 16 ext{TB}$	~500TB \rightarrow 1PB
instant / ~ps ~100x / 25x > RAM \$\$\$\$\$	~1 → 100 ns ~10′s GB/s \$\$\$\$	$\sim 1 \rightarrow 10 \text{ us}$ $\sim 0.5 \rightarrow 3 \text{ GB/s}$ \$\$\$	1 → 10ms 100 → 500 MB/s \$\$	$1 \rightarrow 100 \text{ MB/s}$ secs $\rightarrow \text{minutes}$ \$
Volatile	Volatile	Non Volatile No Moving Part	Non Volatile Spinning Platter	Non Volatile Basically a Walkman!

Storage Systems











L1/L2/L3 Cache	Random Access Memory	Solid State Drives	Hard Disk Drives	Magnetic Tape
Lines of data!	Matrix Structure Rows x Columns Addressable Grid Random Access Bit/Byte Level Address	Hierarchical Structure Boot Sector File System Me File Allocation S Directories File Data		

Storage File Systems











L1/L2/L3 Cache	Random Access Memory	Solid State Drives	Hard Disk Drives	Magnetic Tape
Lines of data! 	Matrix Structure Rows × Columns Addressable Grid	Hierarchical Structure Boot Sector File System Met File Allocation S 	tadata Structure (File Systems)	
Hit or Miss	Random Access Bit/Byte Level Address	Directories Files Pirated Gaming Data etc.		

Storage File Systems

- Hierarchy
 - Fiction / Non-Fiction / Topic → Folders / Sub-Folders
- Metadata
 - Spine / Book Cover → Name, Size, Type of File
- Allocation
 - \circ Books on a Shelf Files on storage medium
- Access
 - Locate & Borrow → Open, Read & Write
- Permissions
 - Dark Art Sections!

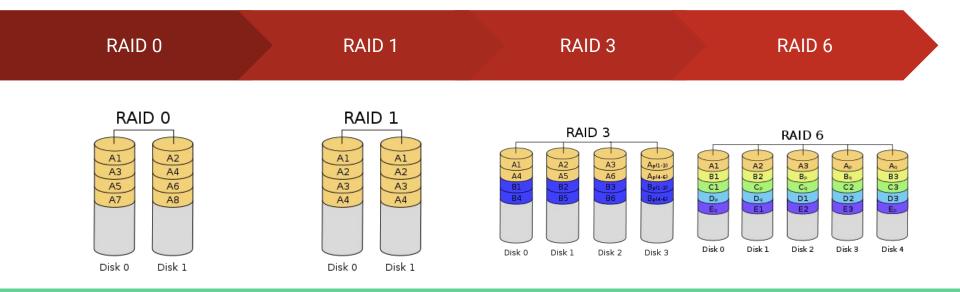


Storage File Systems

FAT	File Allocation Table – FAT32 / exFAT Windows, macOS, Linux	 Simple & Lightweight Max file size (4GB) & Max partition (2TB) Lacks permissions
NTFS	New Technology File System Mainly Windows	 Provides permissions / encryption / compression Improved reliability Read only access on other OS
APFS	Apple File System macOS & iOS	 Designed for modern storage, e.g. SSD's & Flash NTFS + Snapshotting & Shared Spaces Read only access on other OS
ext4	Fourth Extended File System Linux	 Performance & Scalability All the above features Limited Compatibility
ZFS	Zettabyte File System Advanced	 Data Integrity & RAID like capability Large Scale Deployments Native support for macOS & Windows

Storage Archetypes: RAID

- Redundant Array of Independent Disks
- Combines multiple physical storage systems into a single logical unit
- Improves redundancy (parity), performance (stripe) or both.



Storage Archetypes – RAID

Redundant Array of Independent Disks

- Combines multiple physical storage systems into a single logical unit
- Improves redundancy (parity), performance (stripe) or both.

RAID 0	RAID 1	RAID 3	RAID 6	
sudo apt-get ins [.]	tall mdadm			

sudo mdadm --create /dev/md0 --level=0 --raid-devices=2 /dev/sdX1 /dev/sdY1
sudo mkfs.ext4 /dev/md0

Storage Archetypes – Zettabyte File System

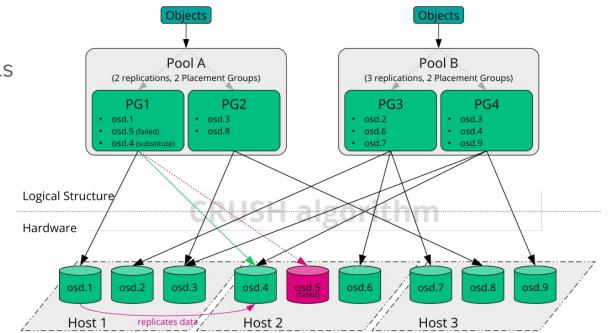
- The Librarian who manages with remarkable efficiency & intelligence.
- Comprehensive: File System + Volume Manager
- Operates at higher levels, sitting on top of storage media or RAID
- Integrity checks via checksums
- Snapshots, clones, deduplication, compression, scalable
- Copy-on-Write!
- Not simple but simpler than RAID

Storage Archetypes – Limitations of File Systems

- File Size & Count
- Total Storage Capacity
- Interoperability
- Metadata Overhead
- Limited to 1 node/server/computer
- Practical limit of ~1-2PB

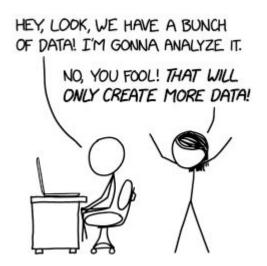
Storage Archetypes: Massively Scalable Storage

- Infinitely Scalable
- Store objects not files
- Objects served as URLs
- CRUSH Algorithm
 - Deterministic Mapping
 - Failure Aware
 - Load Balancing
- E.g. Google Drive.



Management

Why manage data?



Why manage data?

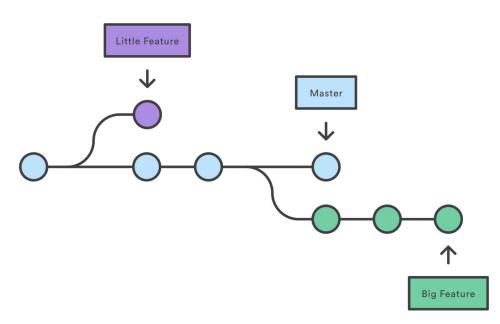
- Ensuring Accuracy
 - By organizing and documenting your data you can easily track changes
 - Ensures integrity of research
- Completeness
 - Standardized protocols of collecting and storage of data
 - Reduces risk of missing critical information
- Accessibility over time
 - Access and retrieval even after significant amount of time
 - Longevity of research!
- Sharing & Collaboration
 - Fosters transparency & reproducibility
 - Validation of research!
- Avoid Data Loss
 - \circ It's not important, unless it was.
 - All your research should not be on your laptop!

• Use Version Control to Tag data

```
import subprocess
import h5py
import pkg_resources
```

```
# Get the Git commit hash
git_hash = subprocess.check_output(['git', 'rev-parse', 'HEAD']).decode().strip()
# Get the version of your Python package
package_version = pkg_resources.get_distribution('mypackage').version
# Open the HDF5 file in write mode
with h5py.File('data.hdf5', 'a') as f:
    # Create or update a dataset in the header group to store the metadata
    header_group = f.require_group('/header')
    header_group.attrs['git_hash'] = git_hash
    header_group.attrs['package_version'] = package_version
```

- Use Version Control to tag data
 - Traceability
 - Enables Collaboration
 - Pinpoint introduction of bugs
 - Safe place to experiment



- Use Version Control to tag data
- Hierarchical Data Structure
 - Store data in folders & subfolders ...
 - <project>/<datasource>/<datatype>/<instrument>/...

- Use Version Control to tag data
- Hierarchical Data Structure
 - Store data in folders & subfolders ...
 - Use meaningful naming conventions
 - Consistent naming conventions
 - → Include metadata in the filename e.g. analysis_success.h5
 - Parametrize and automated file paths
 - → Where ever possible, automate file path generation

import os
import datetime
import subprocess

```
# Parameters
dataset_name = "steady"
subset_name = "data_management"
file_extension = ".csv"
```

```
# Get the current date in YYMMDD format
current_date = datetime.datetime.now().strftime("%y%m%d")
# Get the Git commit hash
git_sha = subprocess.check_output(['git', 'rev-parse', 'HEAD']).decode().strip()
# Define the base directory where the files will be stored
base_directory = "/path/to/data"
# Generate the file path using placeholders or variables
file_path = os.path.join(
    base_directory,
    dataset_name,
    current_date,
    f"{subset_name}_{current_date}_{git_sha[:7]}{file_extension}"
```

- Use Version Control to tag data
- Hierarchical Data Structure
 - Store data in folders & subfolders ...
 - Use meaningful naming conventions
 - Parametrize and automated file paths
 - Consider file system limitations
 - → Character Restrictions: Windows does not allow : <, >, :, ", /, \, |, ?, and *.
 - → Case Sensitivity & Path Separator: / vs. \
 - → Max File Length: Windows 256 // Linux 4096
 - Python Note: Create virtualenv in folders

- Use Version Control to tag data
- Hierarchical Data Structure
 - Store data in folders & subfolders ...
 - Use meaningful naming conventions
 - Parametrize and automated file paths
 - Consider file system limitations
 - Create the highest bisecting hierarchy
 - → You should be able to go from the biggest to smallest data product intuitively

Data Completeness

- Use data models where possible
 - → For analysis parameters
 - When saving data
 - → When loading data
 - Especially when transforming!

from pydantic import BaseModel, Field, FilePath
import yaml

```
# Define the data model using pydantic
class DataModel(BaseModel):
    datapath: FilePath(exists=True)
    iterations: int = Field(..., gt=0, lt=10)
    threshold: float = Field(..., ge=0, le=1)
```

```
# Load YAML data into the data model
def load_parameters(file_path):
    with open(file_path, 'r') as f:
        config_data = yaml.safe_load(f)
    parameters = DataModel(**config_data)
    return parameters
```

```
# Example usage
file_path = 'data.yaml'
data_model = load_data_model_from_yaml(file_path)
print(data_model)
```

Data Accessibility / Sharing & Collaboration

- Document common practices
 - → Include a README.md with the code that analyzes data
 - Define the expected input / output data structures
- Use common and stable data formats from your field
 - → csv, json, par, hdf5, fits etC.
 - → e.g. even though commonly used, would advise against npy format.
- Use packages and formats with the largest community support

Avoid Data Loss

- Redundant Storage Archetypes
- Regular Data Backups
 - → Local Backups > Cloud Backups
 - → Cloud syncs can cause 10-100x slow down in access time
- Uninterrupted Power Supply (UPS)
- Robust Security Options
 - → When in doubt do not give write permissions
 - → Generally advise against password access, instead always use ssh keys
- Training & Awareness
 - → Mean Time Before Failures (MTBF)

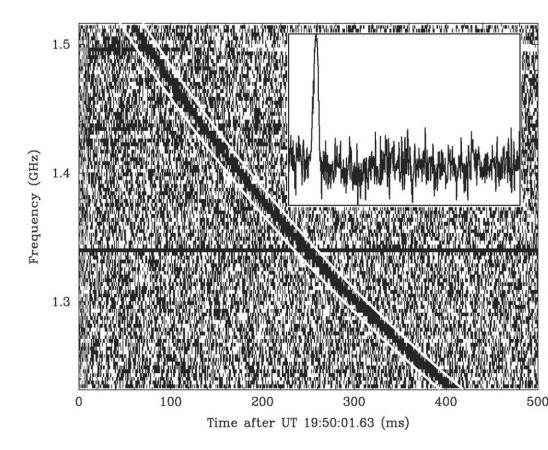
Best Practices

- Time Rule: I/O should take no more than 10% of compute time
 - → Faster storage medium
 - → Faster storage archetype
 - → Split your saved data products
- Size Rule: Depends on your budget, but...
 - → ~10GB → SSD
 - → ~10TB → HDD
 - → ~10PB → Tape
- More I/O Speed
 - → Read the Time Rule
- Always be redundant or backed up!
- Documentations

Case Study: CHIME/FRB Telescopes

Fast Radio Bursts

- Transient, ms-scale
- Astrophysical phenomenon
- Intense pulse of radio waves
- Extragalactic in origin.
- Enormous energy.



CHIME Telescope



CHIME/FRB Telescopes: Data

- 4 Telescope Sites
- 2.2 TB/sec → 190 PB/day → 69 EB/year
- ~ Every computer / phone / video stream for entirety of Canada

- 69 EB/year reduced to ~500PB/year
- Search this data to find FRB's
- All FRB data from previous 5 years → 750 TB
- $^{\circ}0.002\%$ of all data \rightarrow $^{\circ}5$ minutes of data.
- Everything else is discarded.

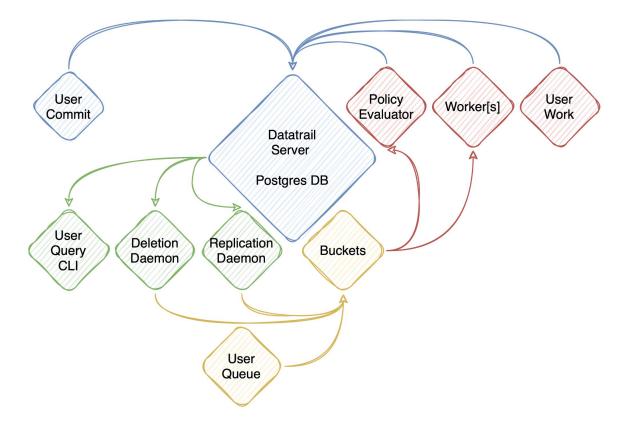
CHIME/FRB Telescopes: Storage Medium / File System / Archetype

- ~23 TB RAM spread over ~400 nodes
- ~40TB High Speed Flash Storage
- (60+40) x 10TB HDDs = 1 PB Raw HDD Storage
- Configured ZFS File System
- RAID z5 with 10% Hot Spares
- Defaults User Permissions: Read Only
- Parameterized Path Structure /data/<telescope-site>/<data-product>/[raw|processed]/YYYY/MM/DD/astro_[ev ent]/*.h5
- Data backed up in "realtime to CANFAR (Compute Canada)

What is Datatrail?

- Data management platform built for CHIME
- Scalable to run at multiple sites CHIME, outriggers, future outriggers?
- Registers data into database
- Handles deletion and replication between sites
- Policy driven approach to products
 - Allows different rules to be applied to different types of data
 - Eg. Classified FRBs backed up and kept forever
 - RFI stays local until deleted a few weeks later

Datatrail: Overview



Interacting with Datatrail

• Interaction with Datatrail via CLI

• Users can:

- List all datasets in Datatrail
- Download datasets
- View dataset policies
- List where data is stored
- \circ ~ See number of files and file size

•••

>> datatrail --help
Usage: datatrail [OPTIONS] COMMAND [ARGS]...

Datatrail Command Line Interface.

Options:

--help Show this message and exit.

Commands:

config	Datatrail CLI Configuration.
list (ls)	List scopes & datasets
ps	Details of a dataset.
pull	Download a dataset.
version	Show versions.

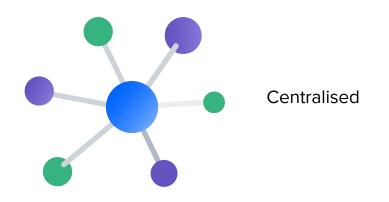
Version Control - Types

Centralised:

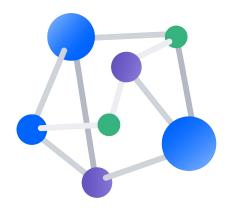
- Single complete copy
- Users write to main branch
- Unavailable while checked out

Pros/Cons:

- Works well with large files
- Easier to understand
- One point of failure
- Lack of stability
- Online



Distributed



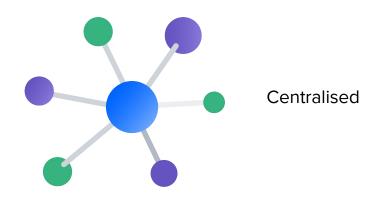
Version Control - Types

Distributed:

- Full copy of repository checked out
- Commit, branch, and merge locally
- Requires more storage space

Pros/Cons:

- Backups
- Faster workflow
- Offline
- Less intuitive
- More prone to conflicts



Distributed

