

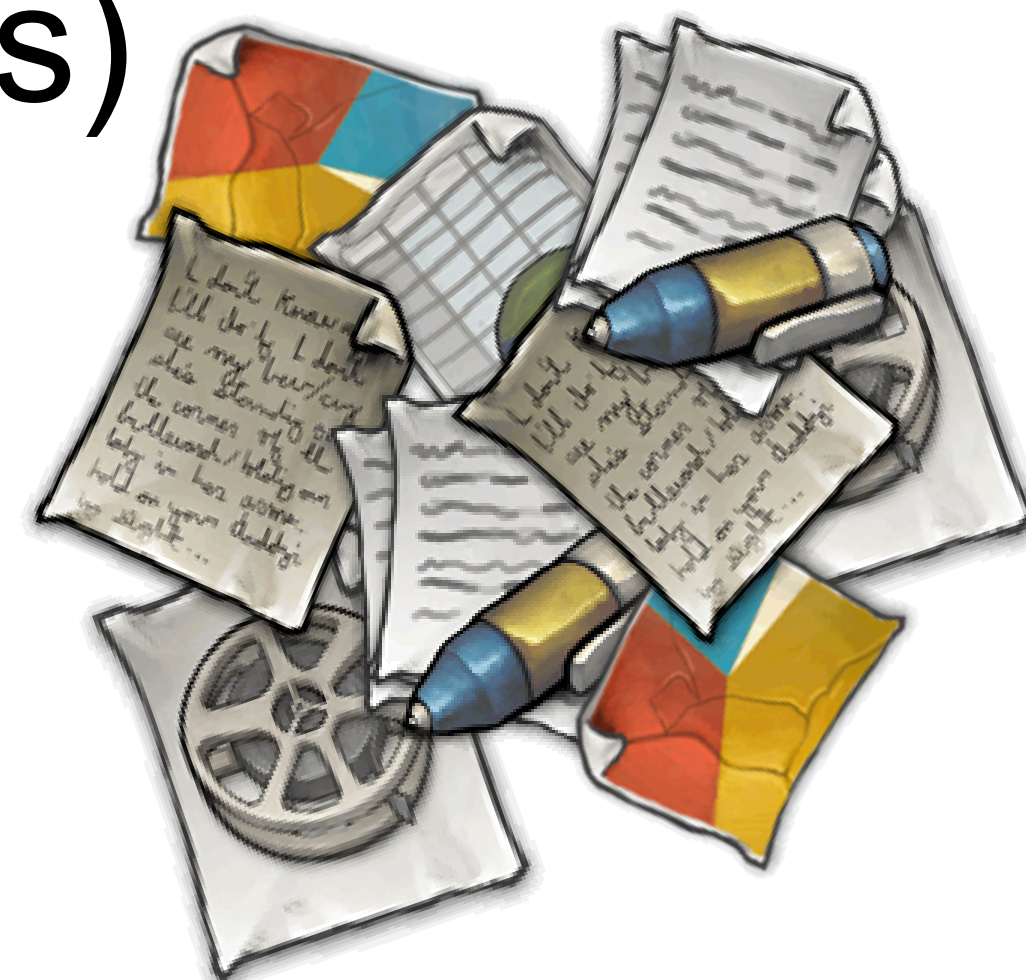
Organizing Chaos: HDF5 for Scientific Data

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The HDF Group

Why?

You're a scientist with massive, complex data.

- Multi-dimensional arrays (particle trajectories)
- Images
- Metadata (experiment parameters, timestamps)
- **Storing it in separate files is a mess.**
 - trajectories_run1.csv ... trajectories_run n^{th} .csv
 - metadata_run1.txt ... metadata_run n^{th} .txt
 - images_run1_001.png... images_run n^{th} _001.png
- **Accessing data is slow and complicated.**
 - To find images from a specific run, you have to open and parse multiple files



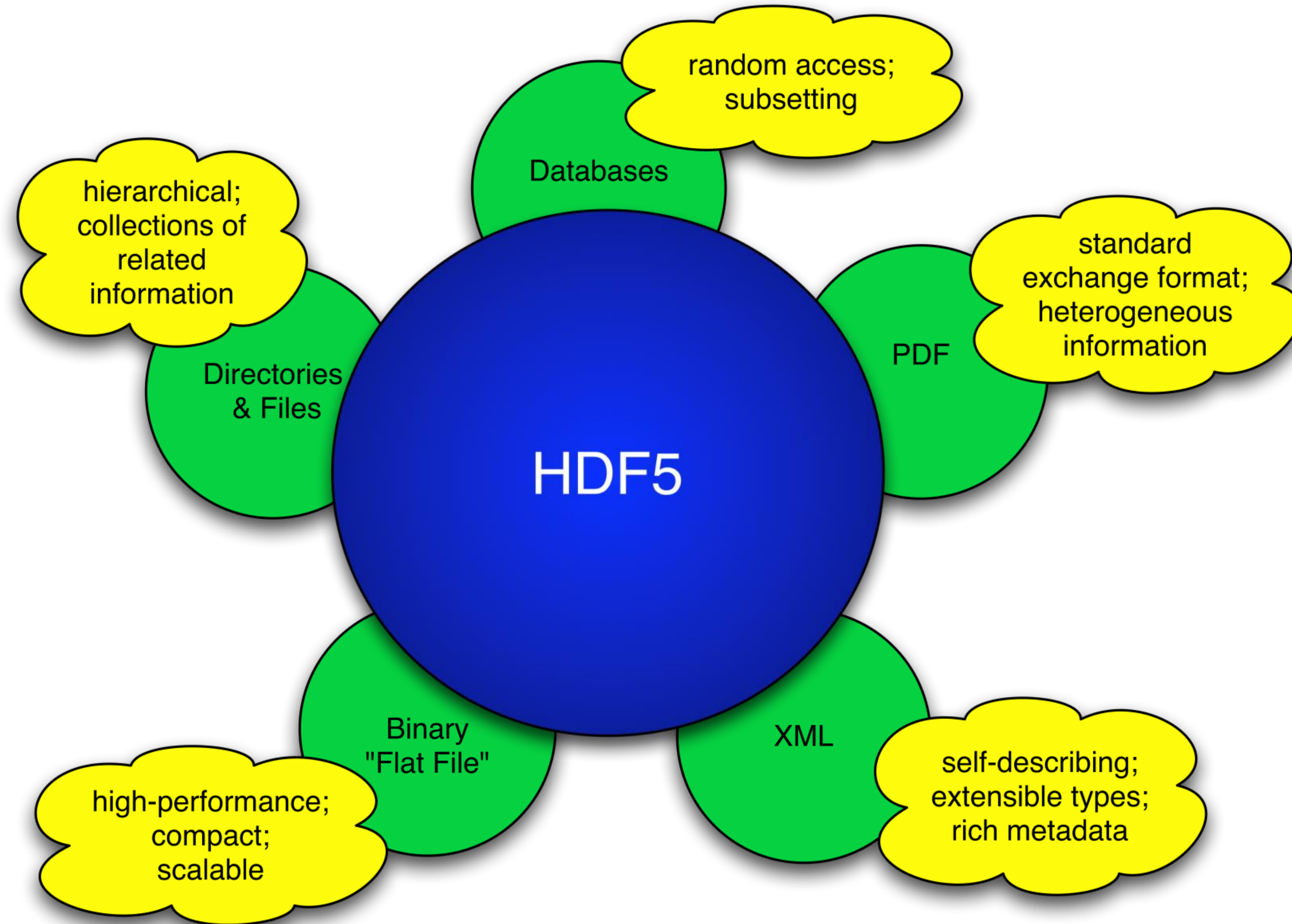
Talk Outline

- Understanding HDF5: The Essentials
- Introducing HDF5 into ML workflows
- Scaling Up: Parallel HDF5 and Advanced Features

What is HDF5?

- **Hierarchical Data Format version 5 (HDF5)**
 1. An extensible **data model**
 - Uses structures for data organization and specification
 2. Open source **software** (I/O library and tools)
 - Performs I/O on data organized according to the data model
 - Works with POSIX and other types of backing stores : Object Stores (DAOS, AWS S3, AZURE, Ceph, etc.), memory hierarchies and other storage devices
 3. Open **file format** (POSIX storage only)

HDF5 is like ...

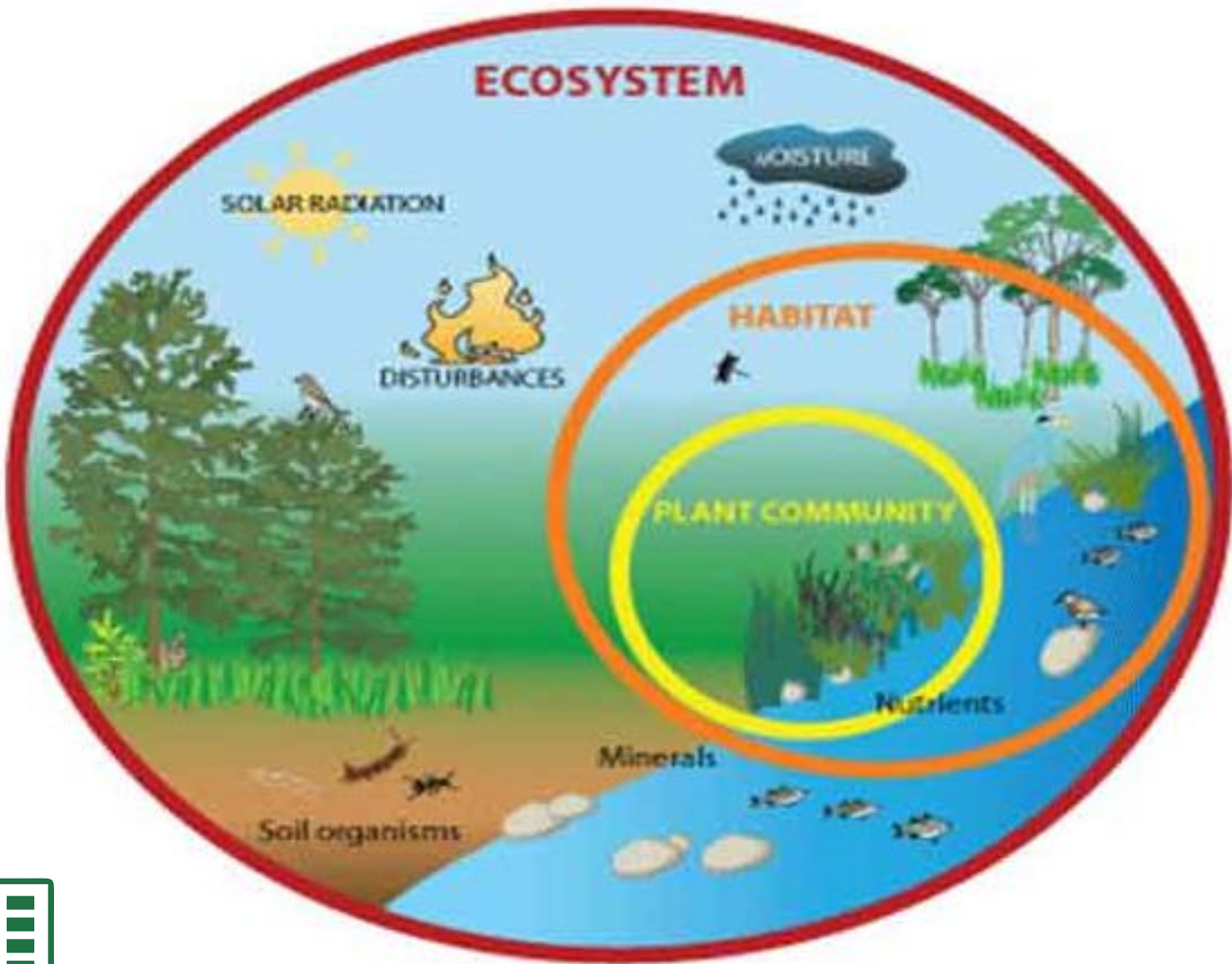


HDF5 is designed for...

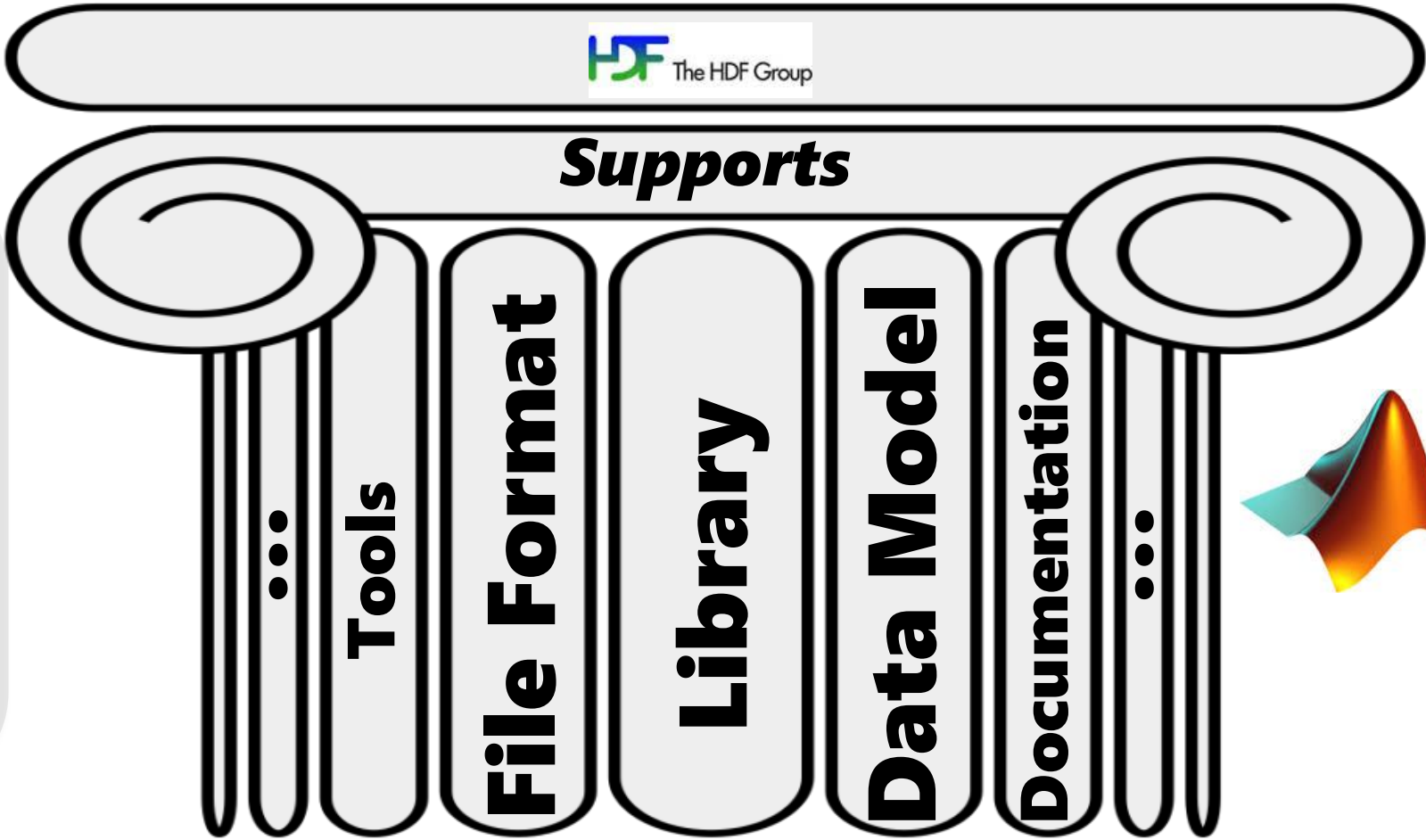
- High-volume and complex data
 - HDF5 files of TBs+ sizes are common
- Every size and type of system (portable)
 - Works on embedded systems, desktops and laptops, to exascale systems
- Flexible, efficient storage and I/O
 - Works for a variety of backing storage
- Enabling applications to evolve in their use of HDF5 and to accommodate new models
 - Data can be added, removed and reorganized in the file
- Supporting long-term data preservation
 - Petabytes of remote sensing data including data for long-term climate research in NASA archives now

HDF5 Ecosystem









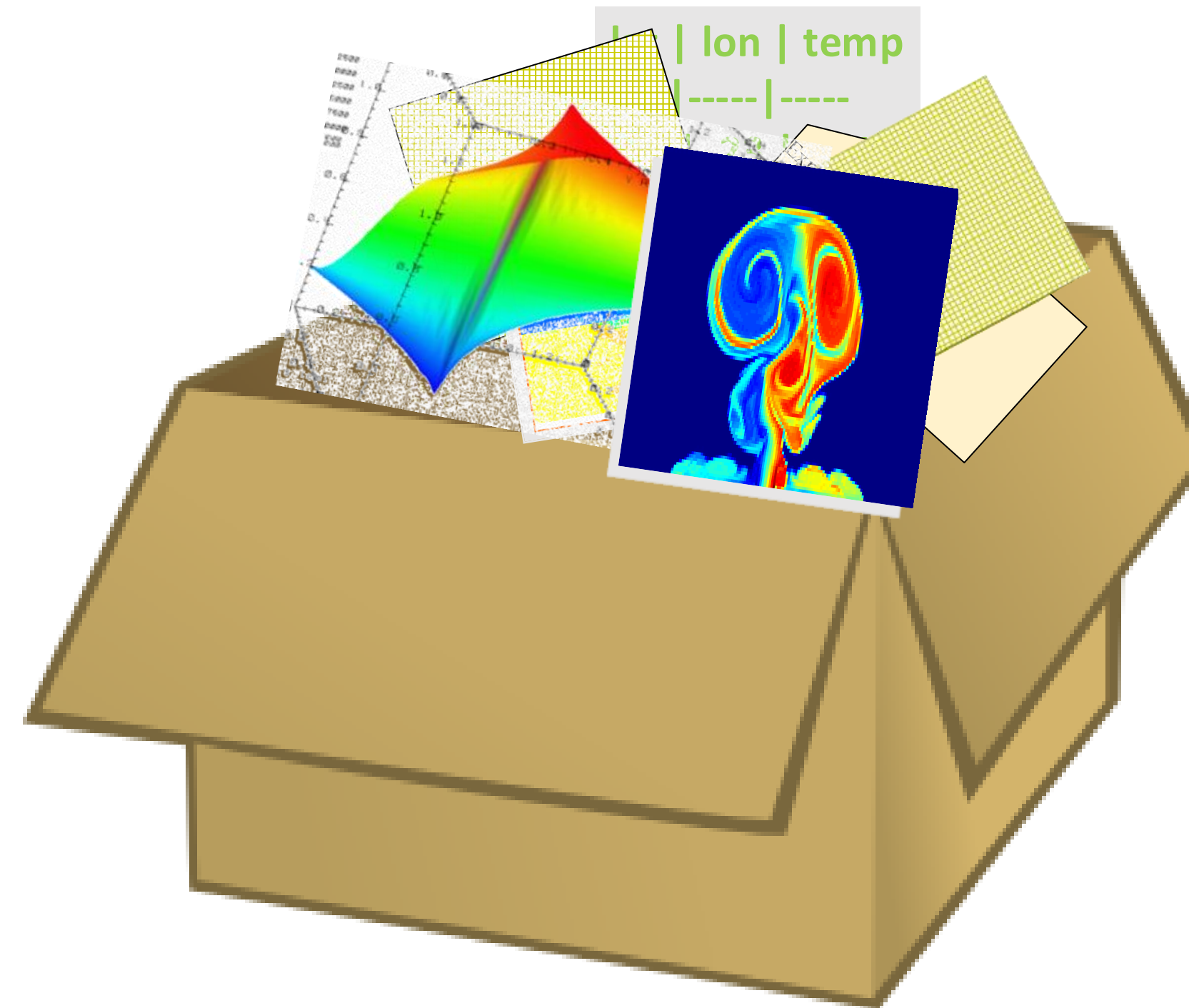




HDF5 Data model

HDF5 File

An HDF5 file is a **container** that holds data objects.



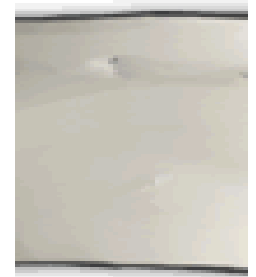
HDF5 Data Model



Dataset –
Organize and contain data elements



Dataspace –
Describes logical layout of the data elements



Attribute –
User-defined metadata

HDF5 Objects



File



Datatype –
Describes individual data elements

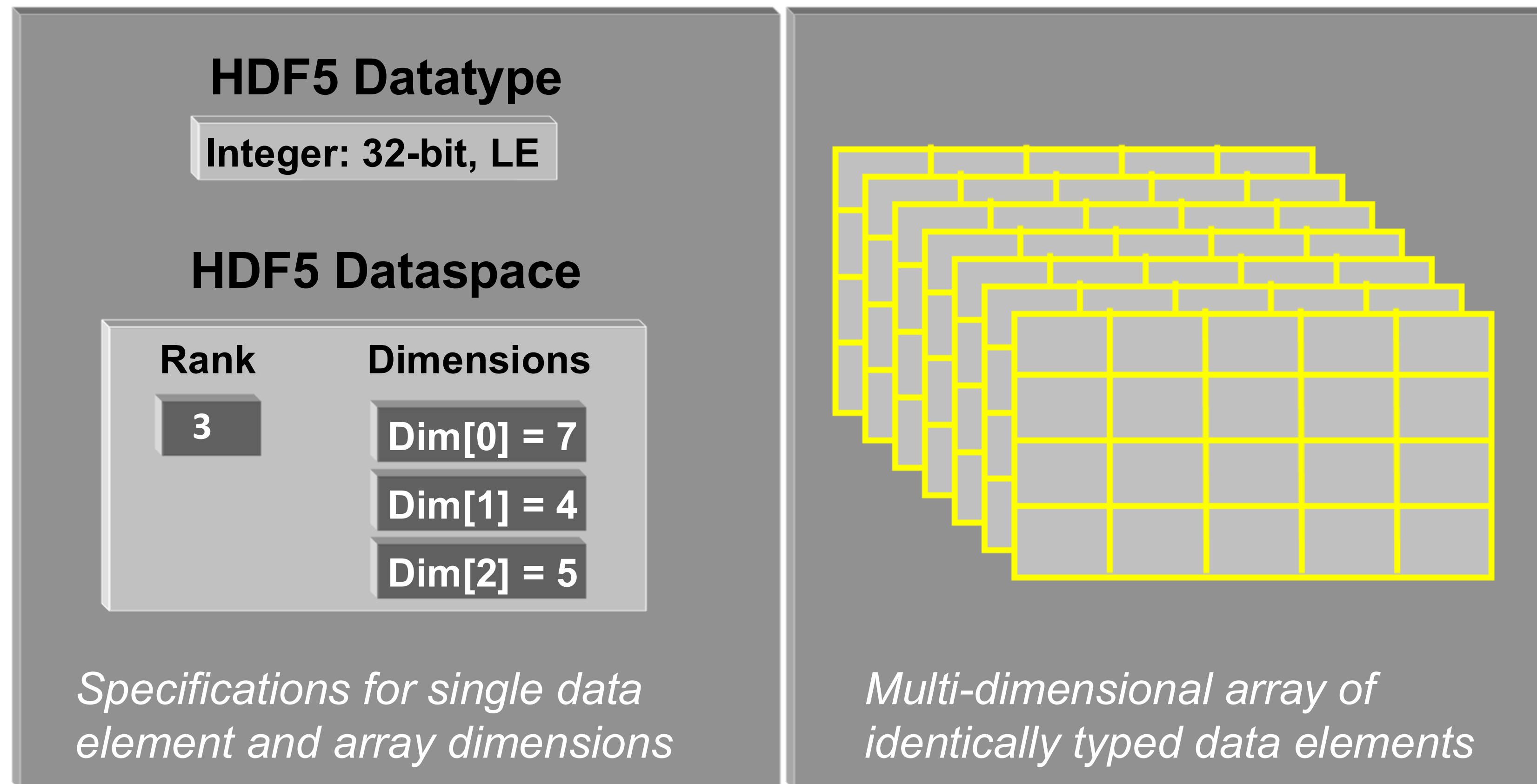


Link –
Organize data objects



Group –
Organize data objects

HDF5 Dataset



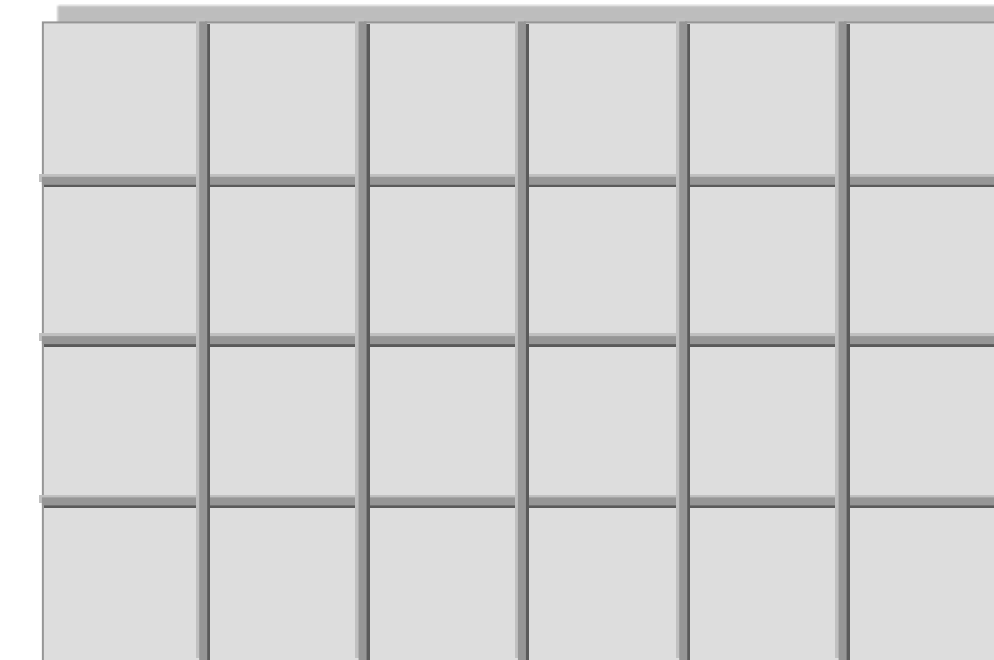
- HDF5 datasets **organize and contain** data elements
 - HDF5 datatype describes individual data elements
 - HDF5 dataspace describes the logical layout of the data elements

HDF5 Dataspace

Two roles:

(1) Spatial information for Datasets and Attributes

- Empty sets and scalar values
- Multidimensional arrays
 - Rank and dimensions
- A permanent part of object definition



Rank = 2

Dimensions = 4 x 6

(2) Partial I/O: Dataspace and subset describe the application's data buffer and data elements participating in I/O



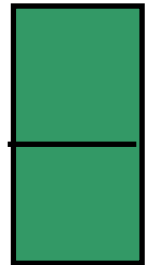
Rank = 1

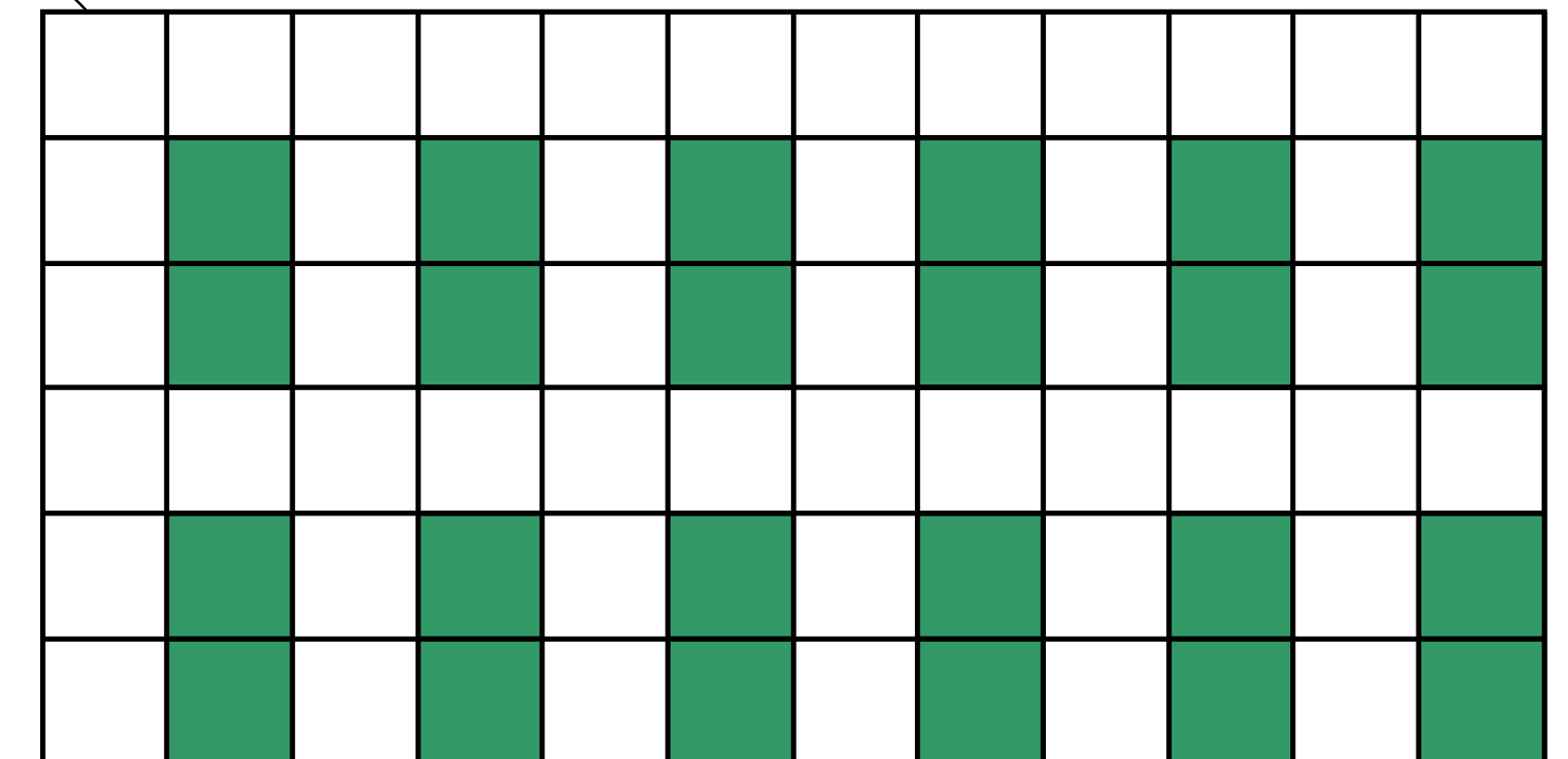
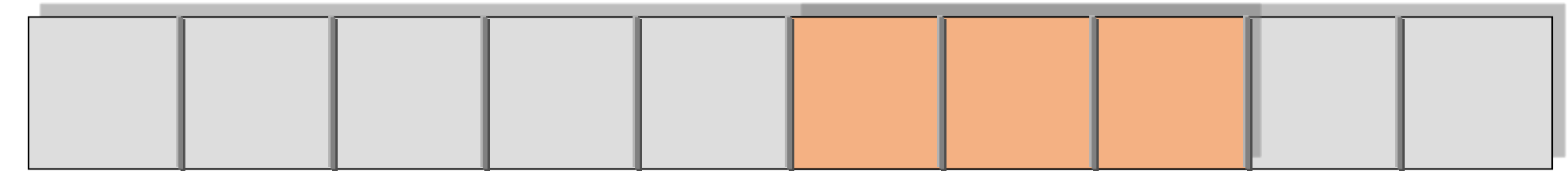
Dimension = 10

How to describe a subset in HDF5?

- Before writing and reading a subset of data, one must describe it to the HDF5 Library.
- The HDF5 APIs and documentation refer to a subset as a “***selection***,” for example “*hyperslab* selection.”
- If specified, HDF5 performs I/O on a selection *only* and not on all dataset elements.

Describing elements for I/O: HDF5 Hyperslab

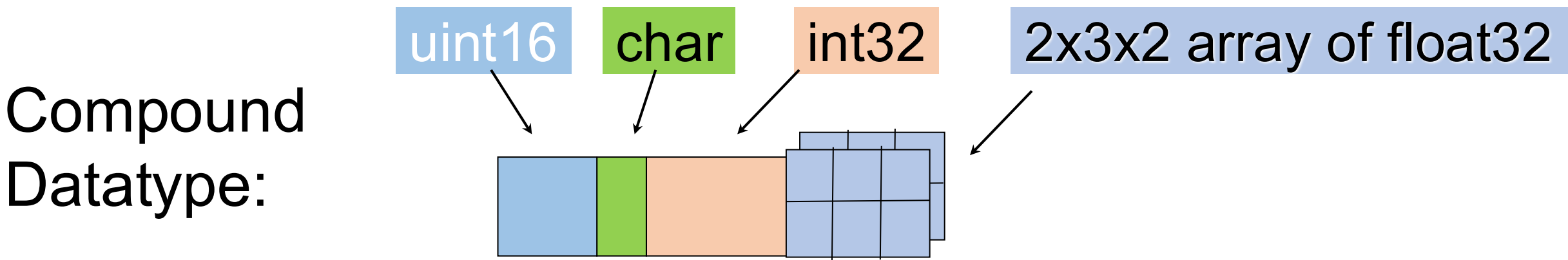
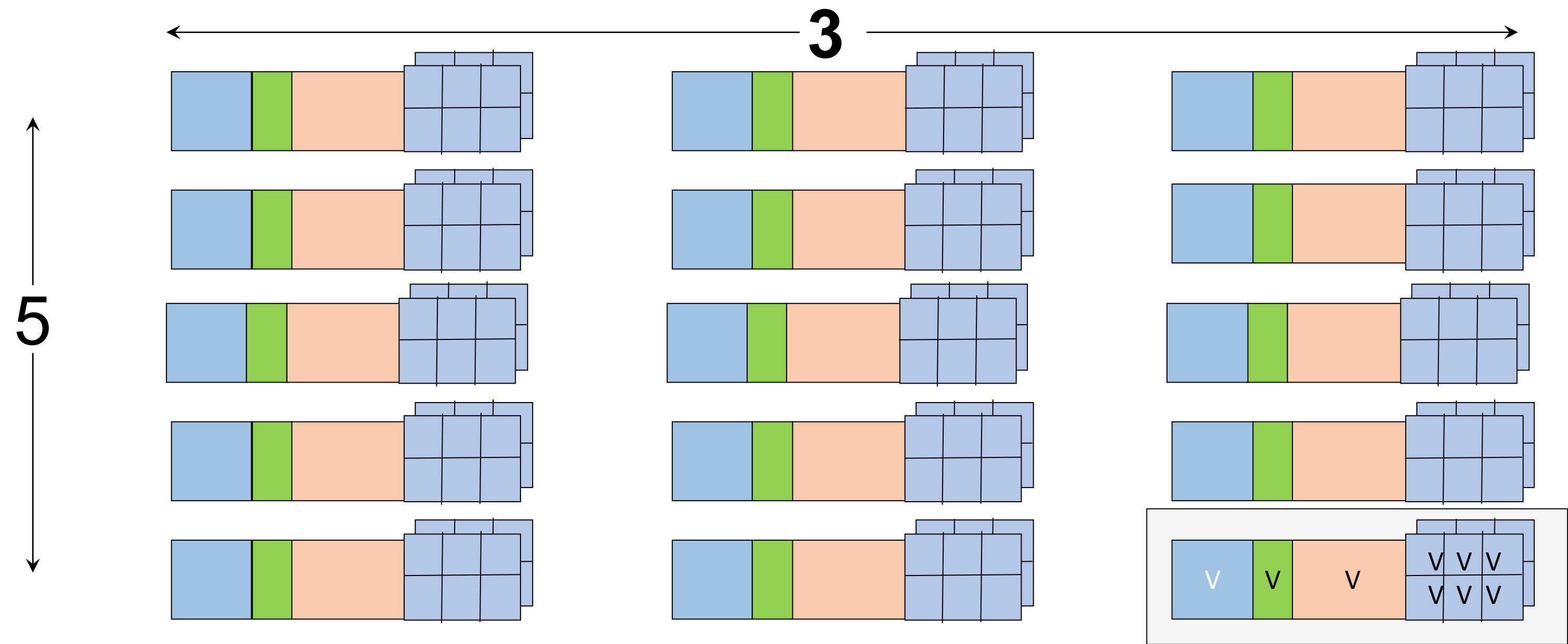
- *Everything is “measured” in the number of elements; 0-based*
- Example 1-dim:
 - Start - starting location of a hyperslab (5)
 - Block - block size (3)
- Example 2-dim:
 - Start - starting location of a hyperslab (1,1)
 - Stride - number of elements that separate each block (3,2)
 - Block - block size (2,1) 
 - Count - number of blocks (2,6)
- All other selections are built using set operations



HDF5 Datatypes

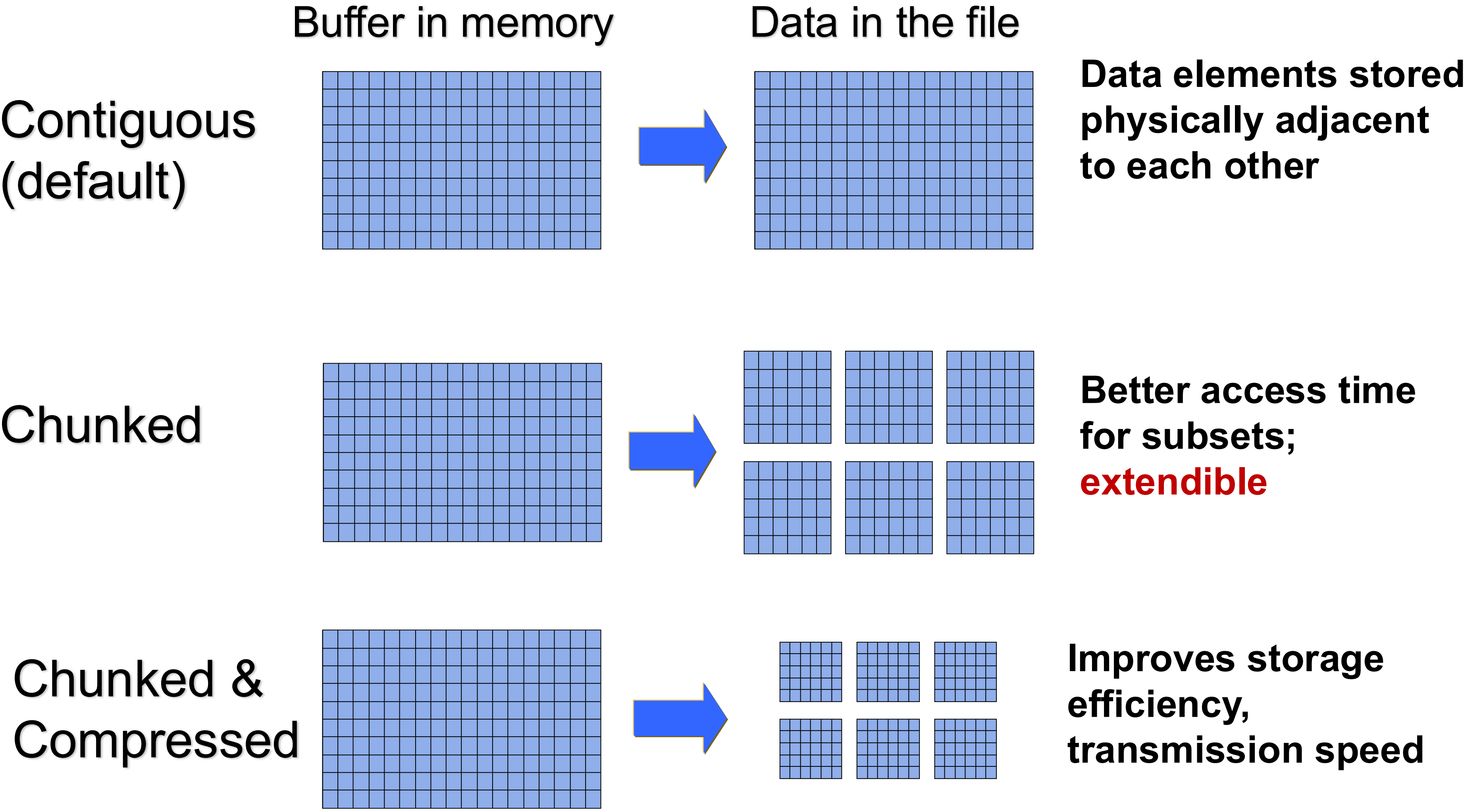
- Describe individual data elements in an HDF5 dataset
- A wide range of datatypes is supported
 - Atomic types: integer, floats
 - User-defined (e.g., 12-bit integer, 16-bit float)
 - Enum
 - References to HDF5 objects and selected elements of datasets
 - Variable-length types (e.g., strings, vectors)
 - Compound (similar to C's structures or Fortran's derived types)
 - Array (similar to matrix)
- HDF5 library provides predefined symbols to describe atomic datatypes

HDF5 Dataset with Compound Datatype

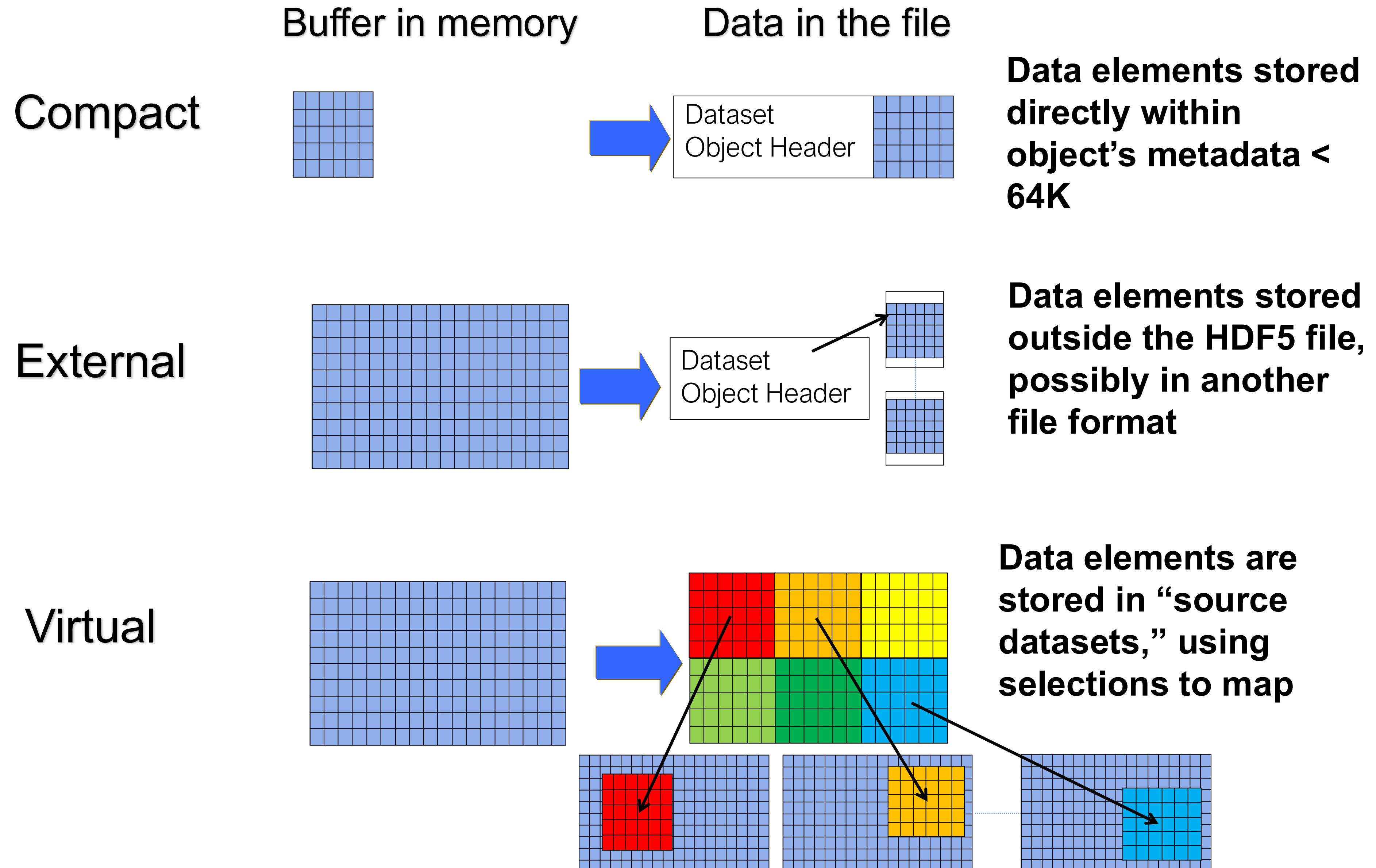


Dataspace: Rank = 2
Dimensions = 5 x 3

How are data elements stored? (1/2)



How are data elements stored? (2/2)

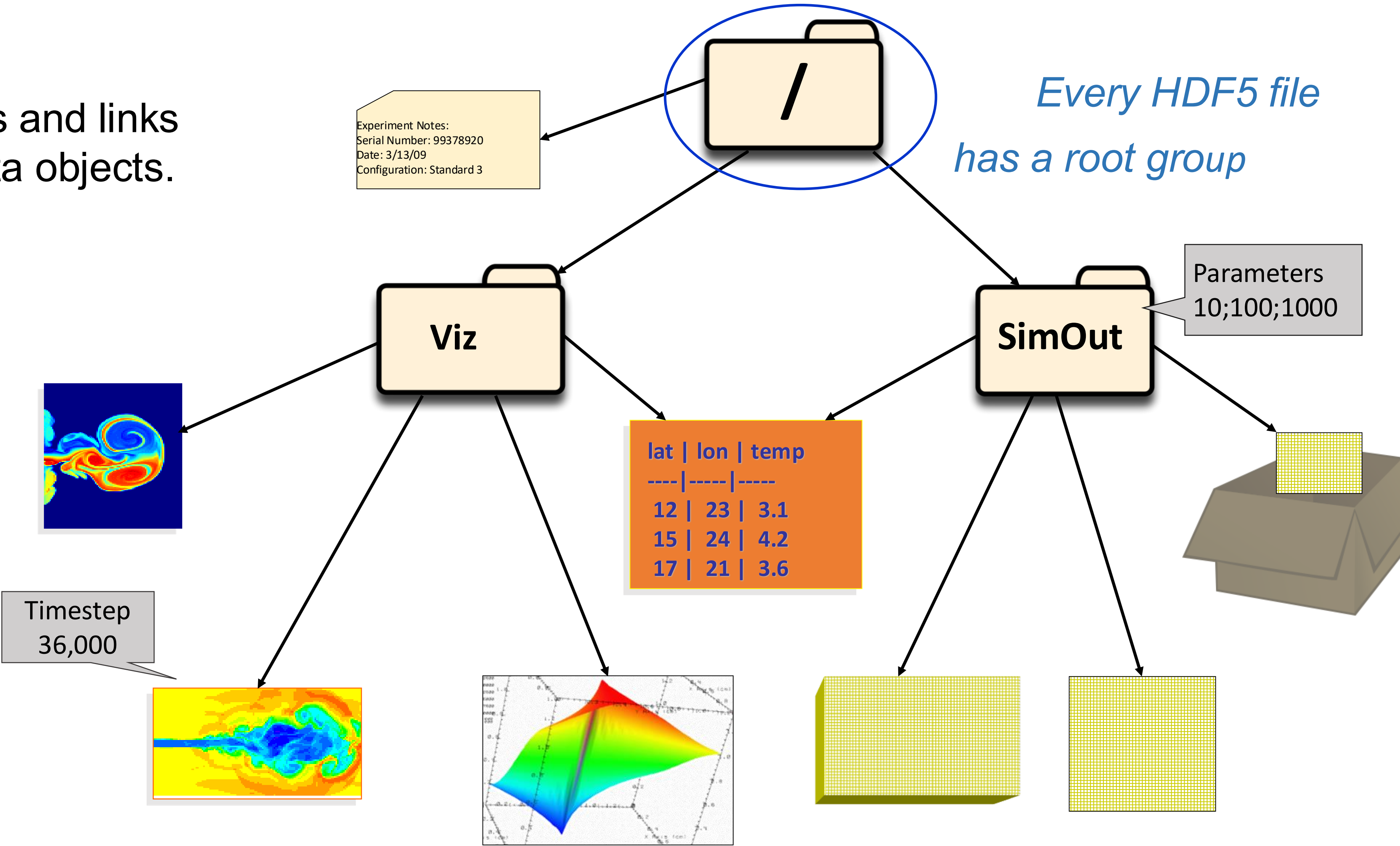


HDF5 Attributes

- Attributes “decorate” HDF5 objects
- Contain *user-defined* metadata
- Similar to Key-Values:
 - Have a unique name (for that object) and a value
- Analogous to a dataset
 - “Value” is described by a datatype and a dataspace
 - **Do not** support partial I/O operations; nor can they be compressed or extended

HDF5 Groups and Links

HDF5 groups and links **organize** data objects.



HDF5 software and architecture

HDF5 Software

→ HDF5 home page: <http://hdfgroup.org/HDF5/>

- Latest releases: 2.2.0 (Retired versions ~~1.8~~, ~~1.10~~, ~~1.12~~, ~~1.14~~)

HDF5 source code:

- Available on GitHub: <https://github.com/HDFGroup/hdf5>
- Written in C and includes optional C++, Fortran, Java APIs, and High-Level APIs
- Contains command-line utilities (h5dump, h5repack, h5diff, ..) and compile scripts

HDF5 pre-built binaries:

- Include C, C++, Fortran, Java, and High-Level libraries when possible. Check `./lib/libhdf5.settings` file.
- Built with the SZIP and ZLIB external libraries

3rd party software:

- Python → h5py
- Rust → <https://github.com/redclaw-quantum/clawhdf5>
- Contemporary C++, including support for MPI I/O
 - <https://github.com/ess-dmsc/h5cpp>, <https://github.com/steven-varga/h5cpp>
 - <https://github.com/highfive-devs/highfive>

Useful Tools For New Users

h5dump

Tool to “dump” or display contents of HDF5 files

Scripts to compile applications:

h5cc, h5c++, h5fc (*h5pcc, h5pfc – parallel variants*)

HDFView:

Java browser to view HDF5 file

<https://www.hdfgroup.org/downloads/hdfview/>

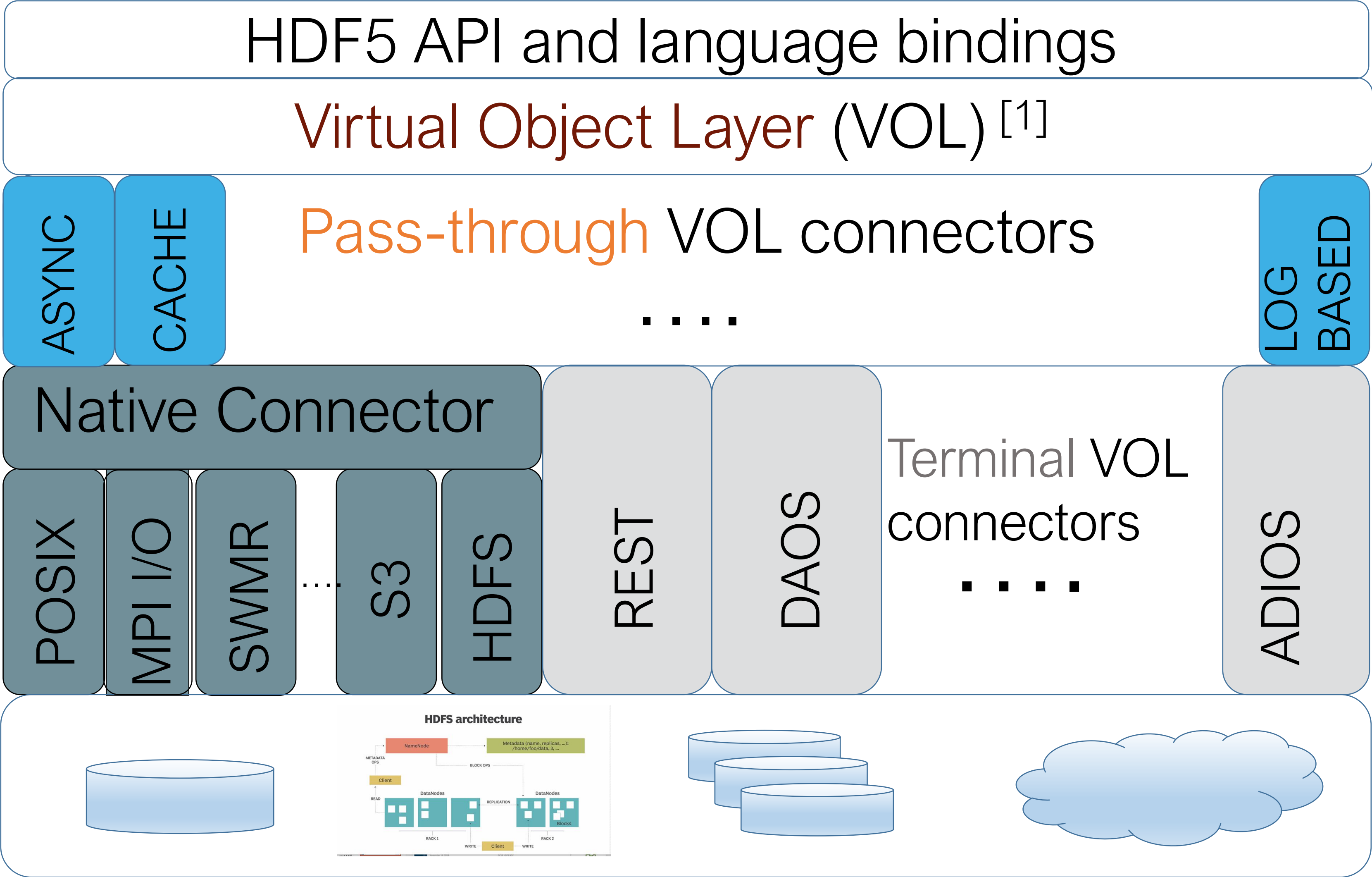
HDF5 Examples (C, Fortran, Java, Python, Matlab, ...)

https://support.hdfgroup.org/documentation/hdf5/latest/_h_d_f5_examples.html

HDF5 Library Architecture (1.12.0 +)

HDF5 Core Library

VFDs



[1] https://portal.hdfgroup.org/documentation/hdf5-docs/registered_vol_connectors.html

HDF5 Programming model and API

The General HDF5 API

- C, FORTRAN, Java, and C++
- The APIs begin with the prefix: H5🔑
🔑 corresponds to the type of object the function acts on

Example Functions:

H5D : Dataset interface e.g., **H5Dread**

H5F : File interface e.g., **H5Fopen**

H5S : data**S**pace interface e.g., **H5Sclose**

- The language wrappers follow the same trend
- There are more than 300 APIs – but one can start with less than 50

General Programming Paradigm

- Object is opened or created
 - Creation properties applied
 - Access properties applied
 - Supporting objects are defined (datatype, dataspace)
- Object is accessed possibly many times
 - Access property can be changed
- Object is closed
- Properties (H5P) of an object are optionally defined
 - Creation properties (e.g., use chunking storage)
 - Access properties (e.g., using MPI I/O driver to access file)

H5Fcreate (H5Fopen)

create (open) File

H5Screate_simple/H5Screate

create dataSpace

H5Dcreate (H5Dopen)

create (open) Dataset

H5Dread, H5Dwrite

access Dataset

H5Dclose

close Dataset

H5Sclose

close dataSpace

H5Fclose

close File

General best practices

RULE #1: You Don't Always Need A DIY HDF5 Implementation

If your scientific field already has a community library built on HDF5, it has likely tuned the chunking, compression, and metadata conventions (i.e., standards) your data needs—better than a from-scratch implementation would.

HDF5

the general-purpose engine



Domain library layer

Encodes your field's schemas, coordinate systems, and access patterns on top of HDF5 — so you inherit years of tuning instead of rebuilding it.

Use HDF5 directly only when your data doesn't fit an existing convention.

1

EARTH & CLIMATE SCIENCE NetCDF-4

CF metadata conventions, coordinate/grid semantics, and dimension scales layered over HDF5.

2

NEUROSCIENCE NWB / HDMF (PyNWB)

Standardized schemas for neurophysiology recordings, built and maintained with HDF5 as the storage backend.

3

COMPUTATIONAL FLUID DYNAMICS CGNS

Mesh and solution-field conventions for CFD codes, so files are portable across solvers.

4

SINGLE-CELL GENOMICS AnnData

Annotated data matrices for scRNA-seq, with HDF5 (.h5ad) as the on-disk format.

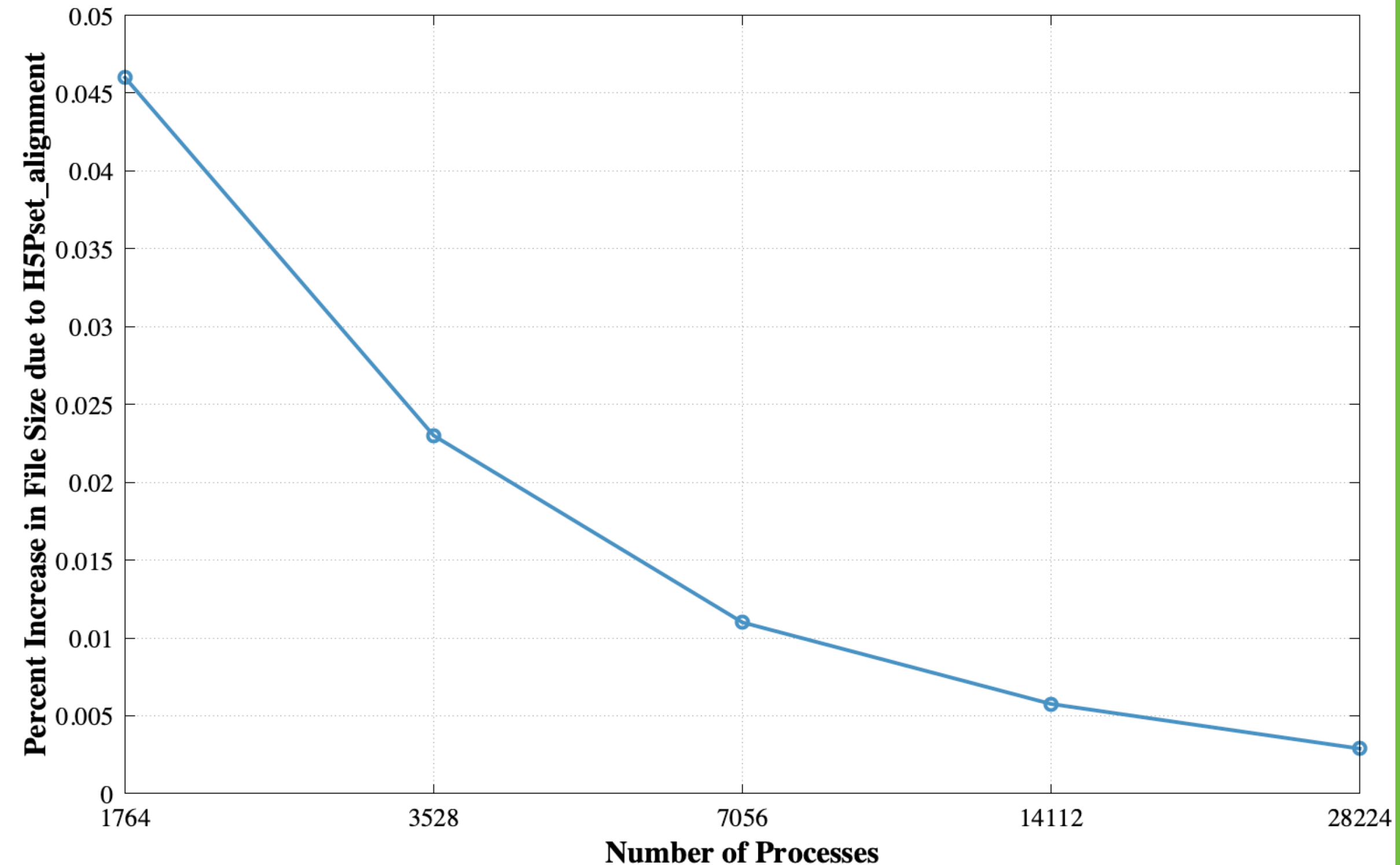
MISC RULES: HDF5 Dataset I/O

- Issue large I/O requests
 - At least as large as the file system block size
- Avoid **dataspace conversion** ⓘ
 - One-dimensional buffer in memory to a two-dimensional array in the file
- Avoid **datatype conversion**
 - Use the same data type in the file as in memory
 - If conversion is necessary, increase datatype conversion buffer size (default 1MB) with *H5Pset_buffer()*

ⓘ Can break collective operations; check what mode was used
[H5Pget_mpio_actual_io_mode](#), and why
[H5Pget_mpio_no_collective_cause](#)

General HDF5 Efficiency

- Faster HDF5 Performance: **Metadata**
 - Use the “latest” file format features
 - *H5Pset_libver_bounds()*
 - Increase the size of metadata data structures
 - *H5Pset_istore_k()*, *H5Pset_sym_k()*, etc.
 - Aggregate metadata into larger blocks
 - *H5Pset_meta_block_size()*
 - Align objects in the file
 - *H5Pset_alignment()*
 - Control metadata cache
 - Paged allocation and page buffering (serial only)
 - Aggregate and align metadata and small data, perform I/O in aligned pages
 - See File Space Management Documentation
https://support.hdfgroup.org/documentation/hdf5-docs/advanced_topics/FileSpaceManagement.html



Fast Reduced-Precision Conversions in HDF5 for ML workflows

Hardware-friendly **bfloat16** / **FP8** / **FP6** / **FP4** conversions via
HDF5's public **H5Tregister** API — up to **280×** faster*

**Benchmarked on Ryzen 9 9950X3D + RTX 5060 Ti (Blackwell)*

Why the default path is slow

HDF5's general converter checks byte order, format, and exception callbacks **per element** — too generic for the compiler to auto-vectorize.

FLOAT32 — 32 BITS

KEPT

DISCARDED

BFLOAT16 — 16 BITS

= BFLOAT16

bfloat16 keeps the sign, exponent, and top 7 mantissa bits of a float32 — conversion is just a cast and a 16-bit shift.

Same trick applies to FP8 / FP6 / FP4

Register a fast path

```
H5Tregister(H5T_PERS_HARD,  
"flt_bf16", H5T_NATIVE_FLOAT,  
H5T_FLOAT_BFLOAT16LE,  
conv_float_bfloat16);
```

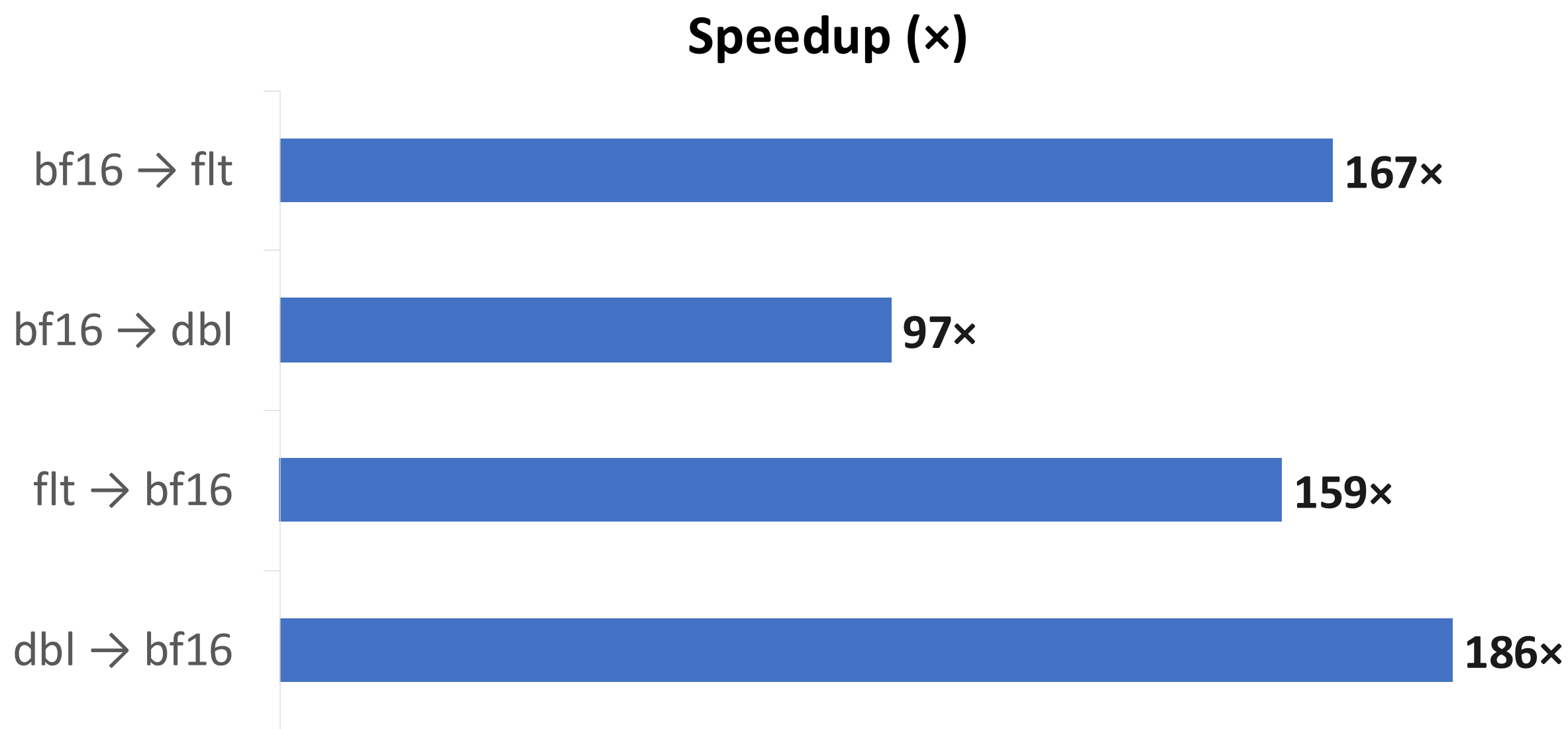
(register the reverse direction – bfloat16 → float – the same way)

- Call it once at startup and every subsequent H5Tconvert() call uses it
- A plain C loop is auto-vectorized by the compiler (confirmed via objdump).
- Falls back to HDF5's general loop for strided or misaligned buffers.
- Performance is not the issue; correctness is....
- Testing random values, tie cases, low-payload NaNs, and zero-length calls
- ✓ **0 mismatches for float-to-bfloat16 / 1,000,000 vs. software path (used same rounding convention)**

RESULTS

30 × – 280 × faster, every precision type

BFLOAT16



134 × – 180 × on GCC 15 & Clang 21 alike — bound by memory bandwidth, not compute (shift-and-cast is too cheap to be the bottleneck).

FP8 / FP6 / FP4 (stored in single byte)

130–165 ×
widening (256-entry lookup table)
Trivially vectorizable

37–57 ×
narrowing (exponent rebias + clamp)
Branchy -> vectorizes less cleanly

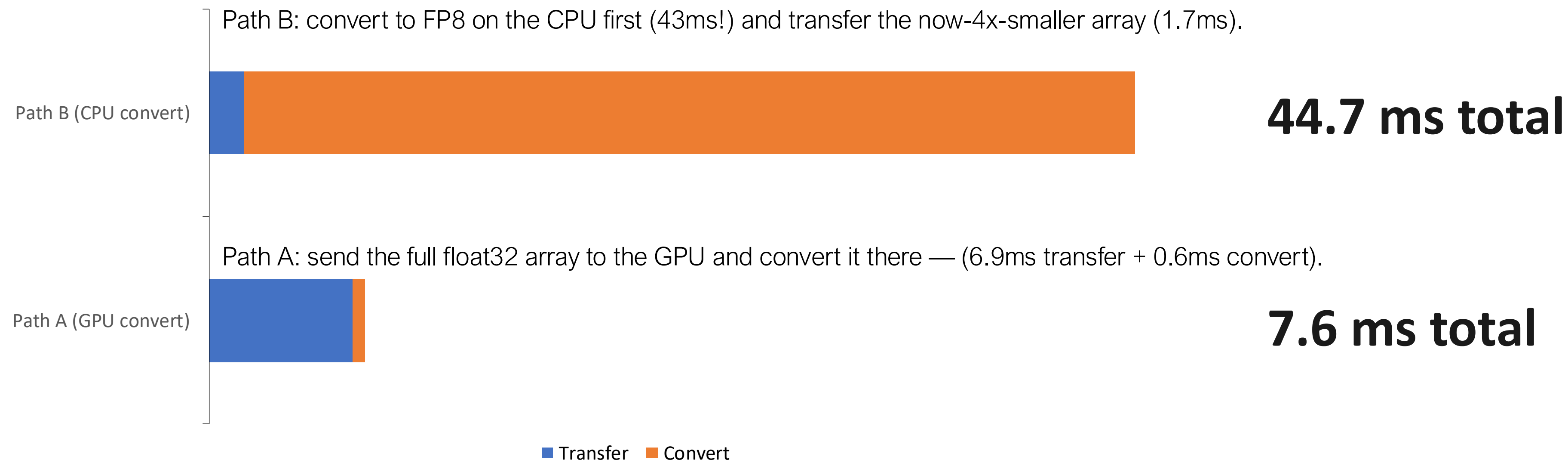
FP6 / FP4 have no x86 hardware path at all — a plain C loop still wins big.

Baseline: HDF5's general loop, same data — all registered via H5Tregister

Blackwell GPUs push it further

float → FP8 on an RTX 5060 Ti: $67\times$ faster than the CPU fast path (0.64 ms vs. 42.9 ms).

- Branchy exponent-rebias code that limits CPU vectorization isn't a bottleneck for GPU SIMT execution



Path A wins $5.9\times$ — GPU conversion is negligible next to the transfer. Pipelined: 7.92 ms, near the 6.94 ms PCIe ceiling.

- Slightly counterintuitive to “transfer less data”, but GPU conversion is essentially free next to the PCIe transfer, while CPU narrowing is expensive relative to everything else

RECOMMENDATIONS

What should scientists use?

WHICH TYPE

- **bfloat16**: safe default, ~0.4% error (mantissa precision) -> file size and I/O time ~halved
- **FP8**: ML inference and training only, needs scaled data (or range-analyzed)
- **FP6 / FP4**: GPU-only ML formats, no CPU hardware path

HOW TO CONVERT

- **CPU**: register once at startup (H5Tregister)
- **GPU**: transfer float32, convert on device, never shrink on the CPU first
- **Round trips**: pipeline multi-chunk round trips across CUDA streams

BOTTOM LINE

30–280 ×

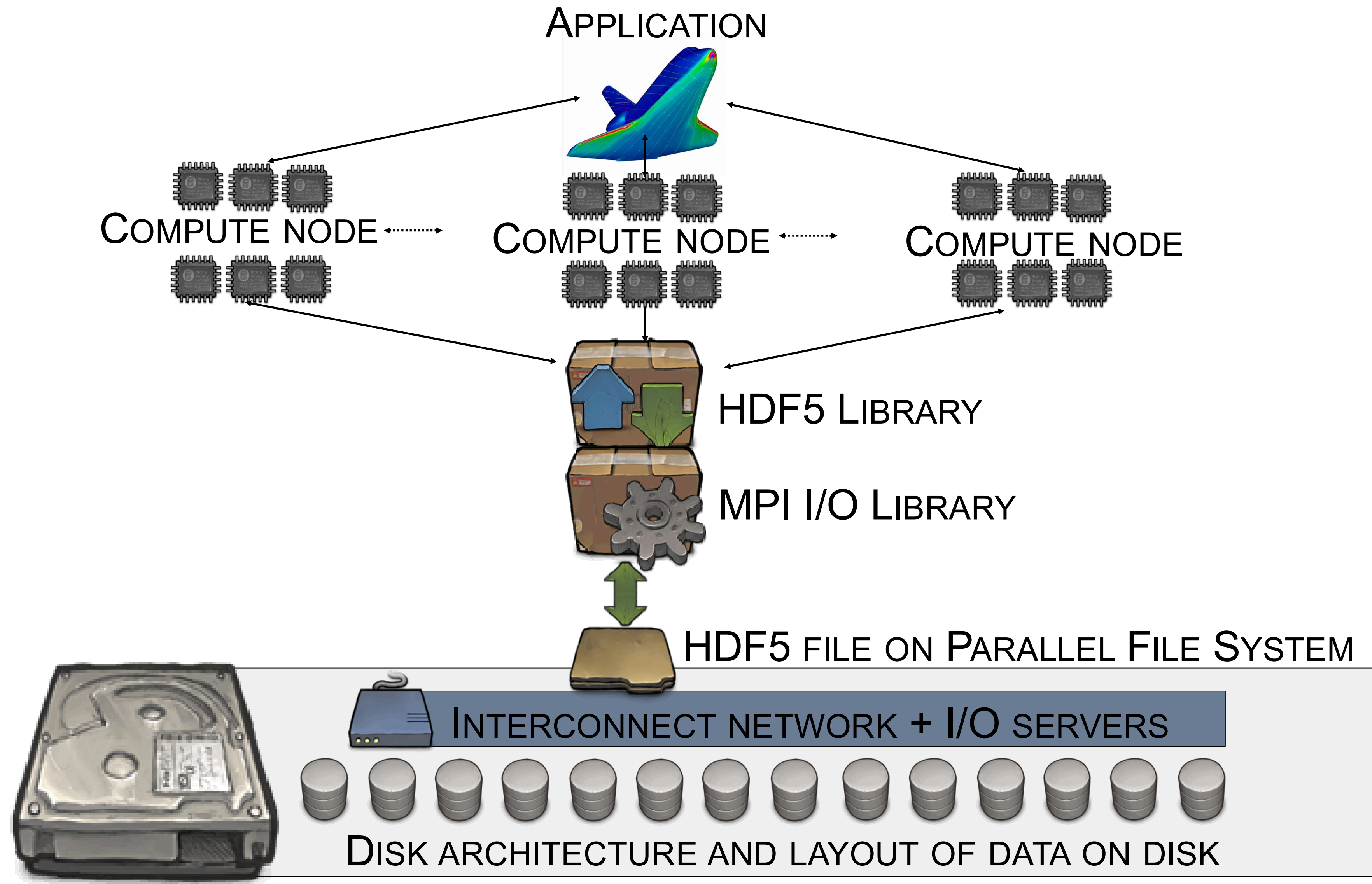
CPU, public API

67 ×

Further improvement on GPU

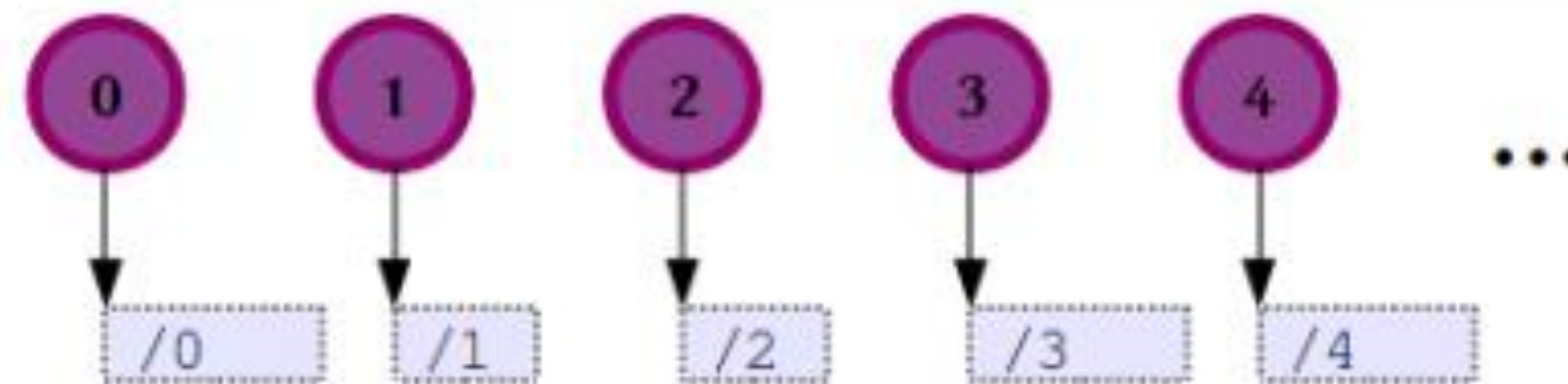
Parallel I/O with HDF5

PHDF5 implementation layers

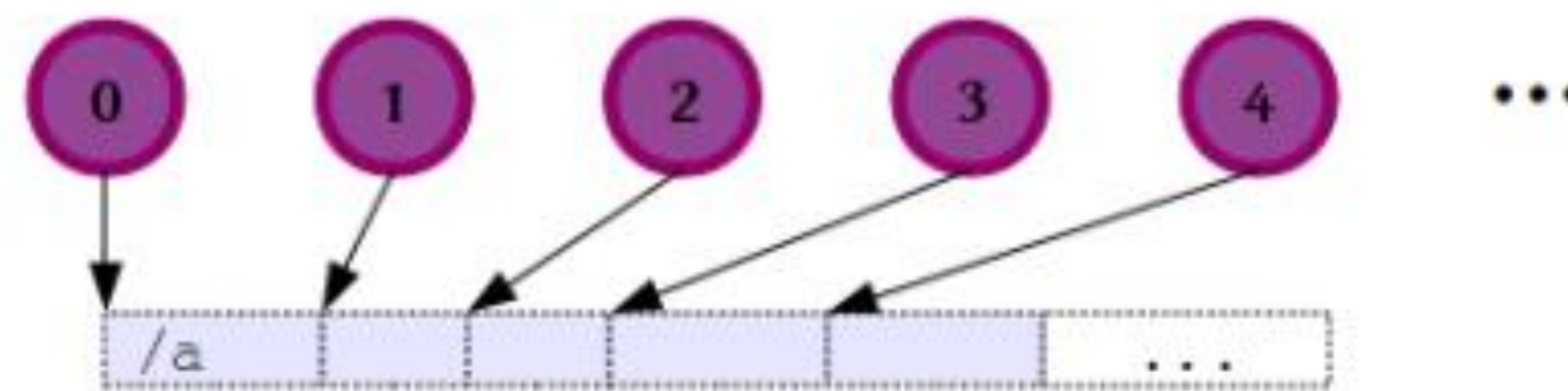


Types of Application I/O to Parallel File Systems

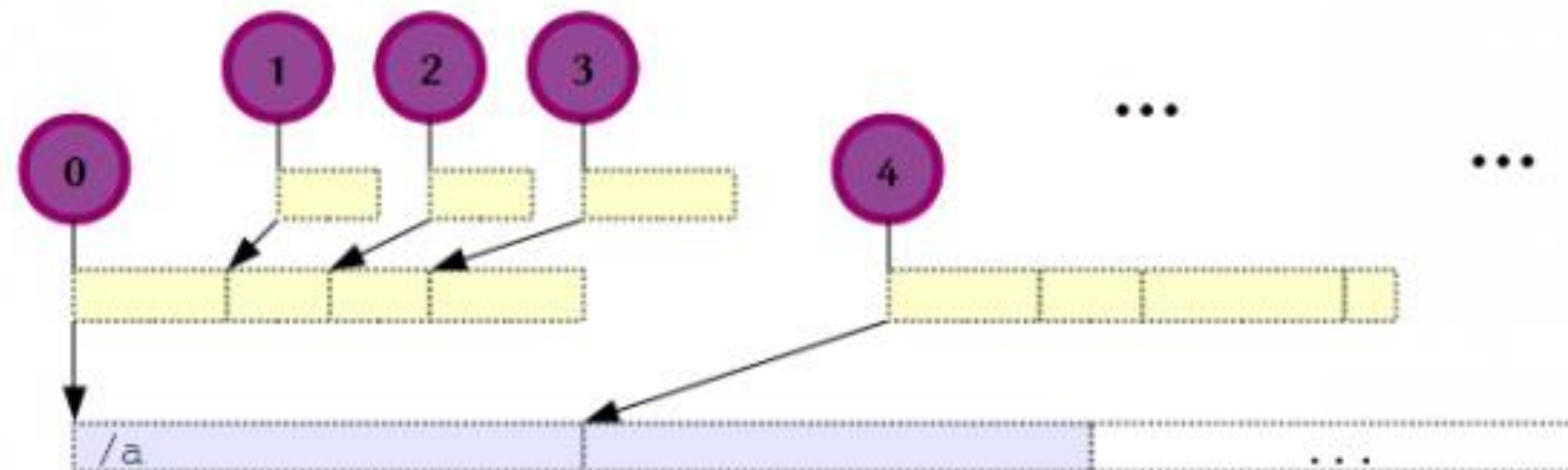
File-per-processor



Shared file (independent)



Shared file (collective buffering)



Why Parallel HDF5?

- Take advantage of high-performance parallel I/O while reducing complexity
 - Use a well-defined high-level I/O layer instead of POSIX or MPI-IO
 - Use only a single or a few shared files
- Maintained code base, performance and data portability
 - Rely on HDF5 to optimize for the underlying storage system

Parallel HDF5 (PHDF5) vs. Serial HDF5

- PHDF5 allows multiple MPI processes in an MPI application to perform I/O to a single HDF5 file
- PHDF5 uses a standard parallel I/O interface (MPI-IO)
- Portable to different platforms
- PHDF5 files ARE HDF5 files conforming to the [HDF5 file format specification](#)
- The PHDF5 API consists of:
 - The standard HDF5 API
 - A few extra knobs and calls
 - A parallel “schema”

Parallel HDF5 Schema

- PHDF5 opens a shared file with an MPI communicator

- Returns a file ID (as usual)
 - All future access to the file via that file ID

 Different files can be opened via different communicators

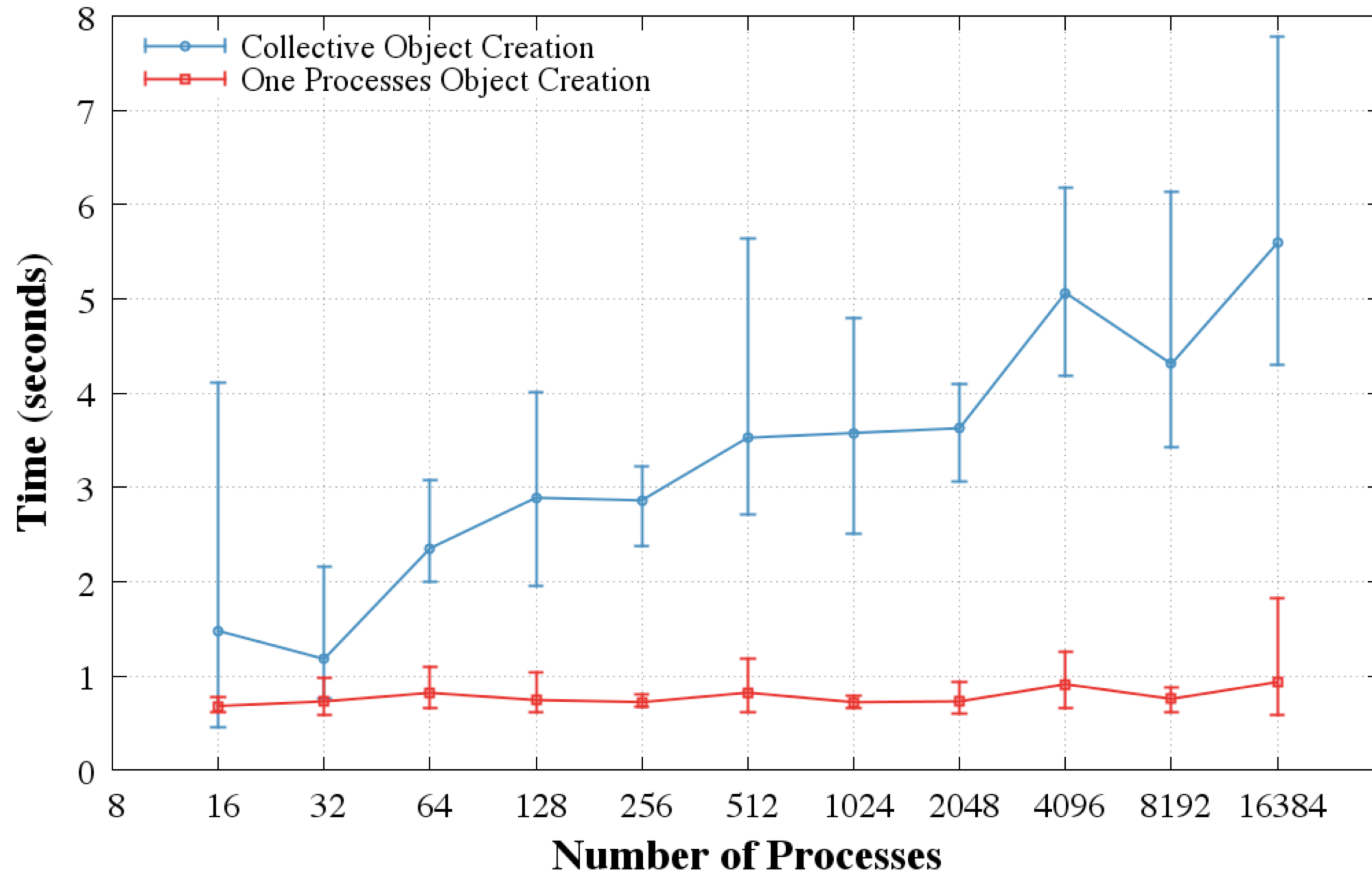
 All processes must participate in collective PHDF5 APIs

- All HDF5 APIs that modify the HDF5 namespace and structural metadata are collective!

https://support.hdfgroup.org/documentation/hdf5/latest/collective_calls.html#sec_collective_calls_func

- File ops., group structure, dataset dimensions, object life-cycle, etc.
 - Raw data operations can either be collective or independent
 - For collective, all processes must participate, but they don't need to read/write data.

Object Creation (Collective vs. Single Process)



Collective vs. Independent Operations

- MPI Collective Operations:

- All processes of the communicator must participate in the right order. E.g.,

Process1

Process2



call A(); call B();

call A(); call B();

...CORRECT



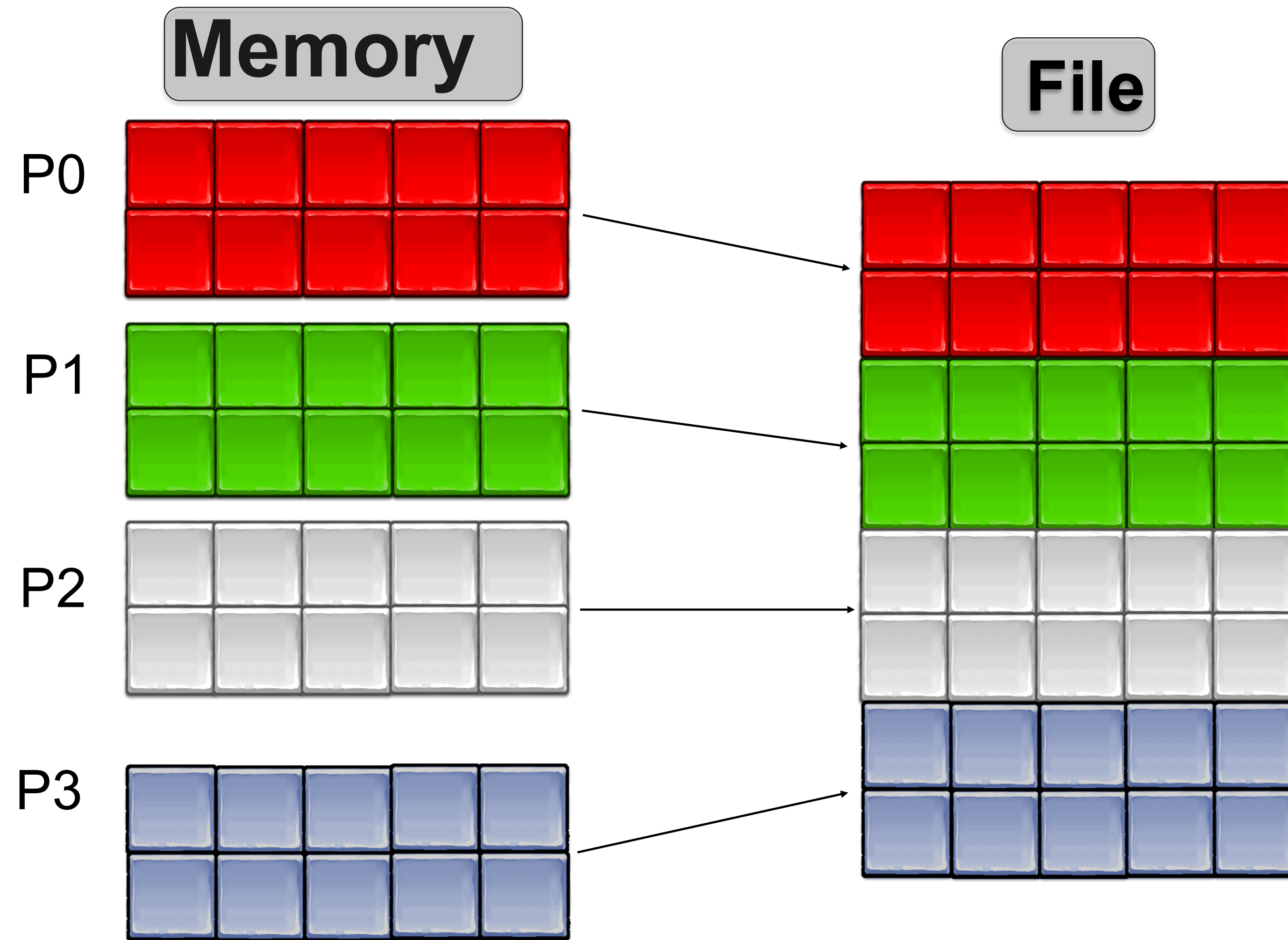
call A(); call B();

call B(); call A();

...WRONG

- Collective I/O attempts to combine multiple smaller independent I/O ops into fewer larger ops; neither mode is preferable *a priori*

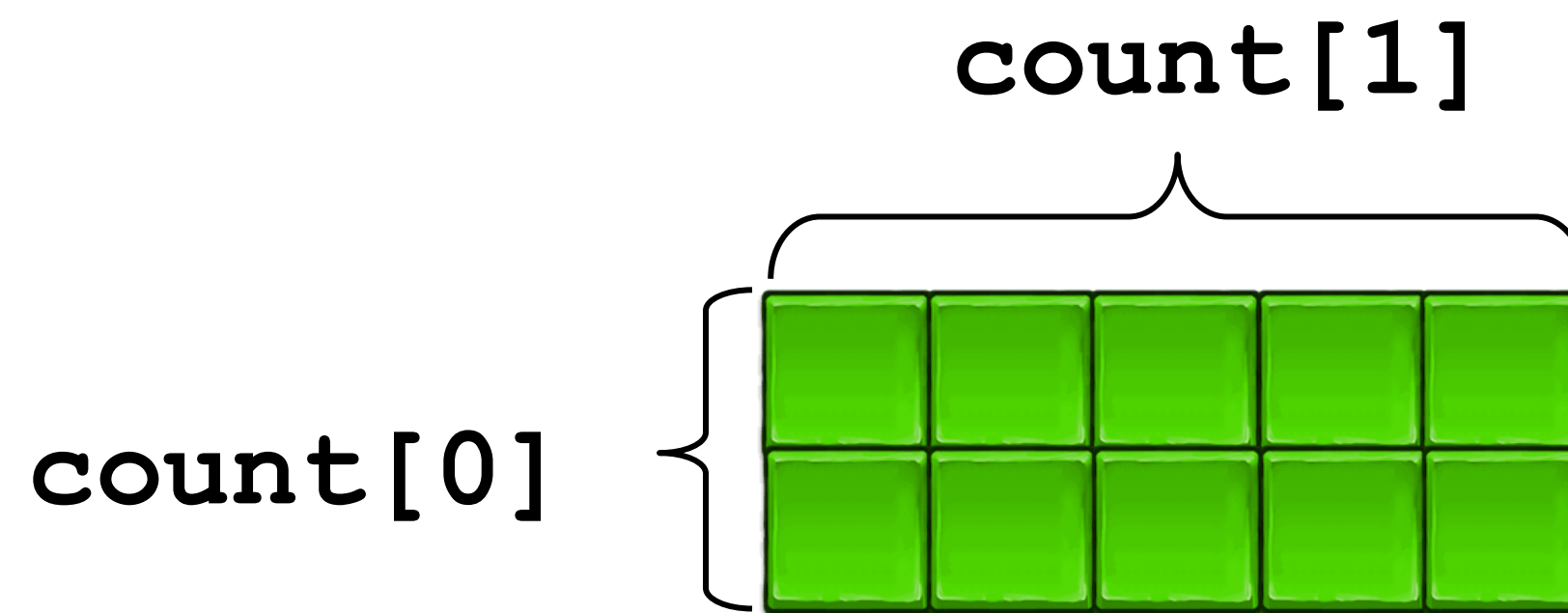
Example 1: Writing dataset by rows



Example 1: Writing dataset by rows

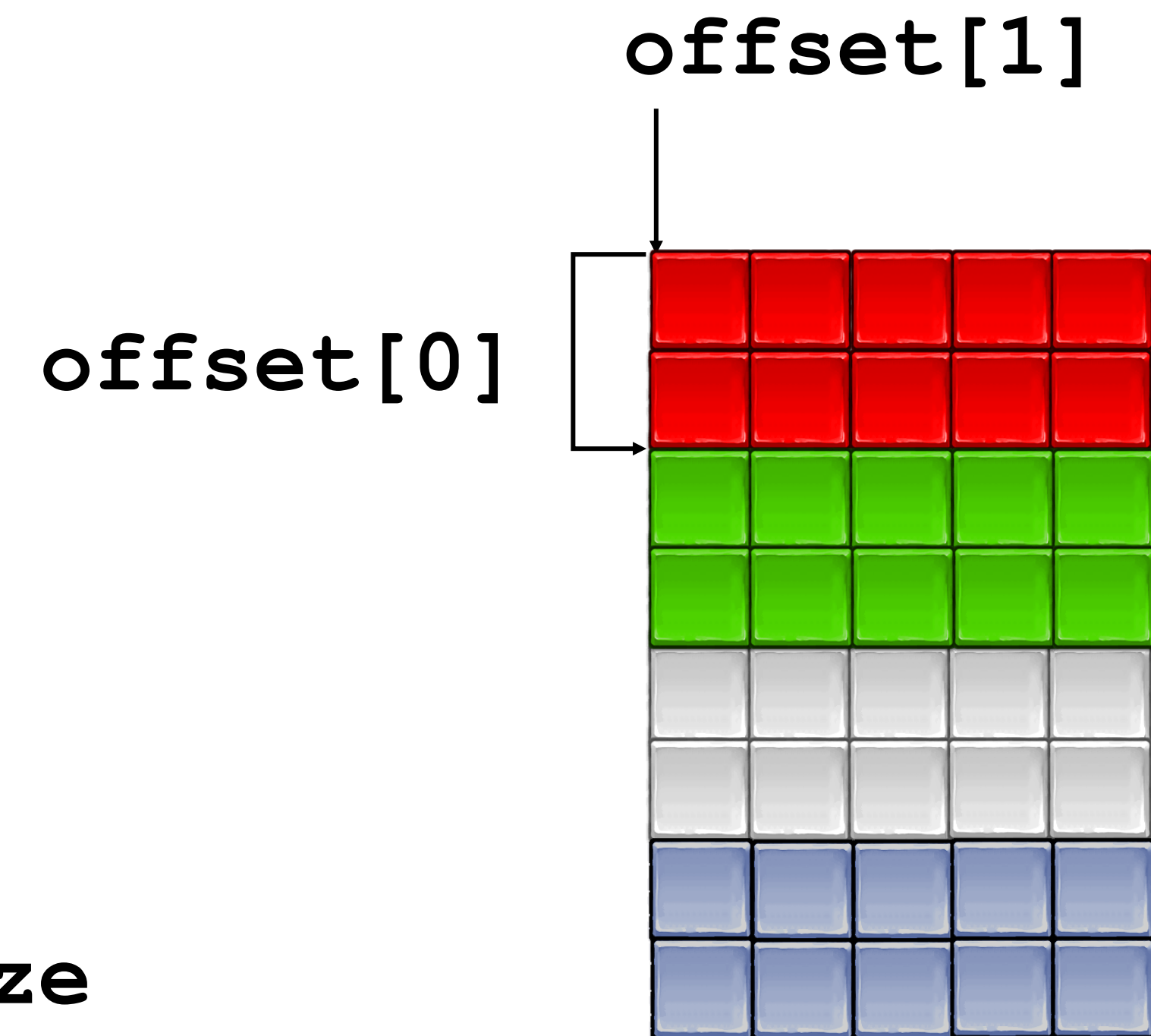
Memory

Process P1



```
count[0] = dims[0]/mpi_size  
count[1] = dims[1];  
offset[0] = mpi_rank * count[0]; /* = 2 */  
offset[1] = 0;
```

File



Example 1: *Writing dataset by rows*

```
71  /*
72  * Each process defines dataset in memory and
73  * writes it to the hyperslab
74  * in the file.
75  */
76  count[0] = dims[0]/mpi_size;
77  count[1] = dims[1];
78  offset[0] = mpi_rank * count[0];
79  offset[1] = 0;
80  memspace = H5Screate_simple(RANK, count, NULL);
81  /*
82  * Select hyperslab in the file.
83  */
84  file_space = H5Dget_space(dset_id);
85  H5Sselect_hyperslab(file_space,
86                     H5S_SELECT_SET, offset, NULL, count, NULL);
```

C Example: Collective write and read

```
95  /*
96   * Create property list for collective dataset write.
97   */
98  plist_id = H5Pcreate(H5P_DATASET_XFER);
->99  H5Pset_dxpl_mpio(plist_id, H5FD_MPIO_COLLECTIVE);
100
101  status = H5Dwrite(dset_id, H5T_NATIVE_INT,
102                  memspace, filespace, plist_id, data);
103  /*
104   * Collective dataset read.
105   */
106
->107 status = H5Dread(dset_id, H5T_NATIVE_INT,
108                 memspace, filespace, plist_id, data);
109
```

Writing by rows: *Output of h5dump*

```
HDF5 "SDS_row.h5" {
  GROUP "/" {
    DATASET "IntArray" {
      DATATYPE  H5T_STD_I32BE
      DATASPACE  SIMPLE { ( 8, 5 ) / ( 8, 5
    ) }

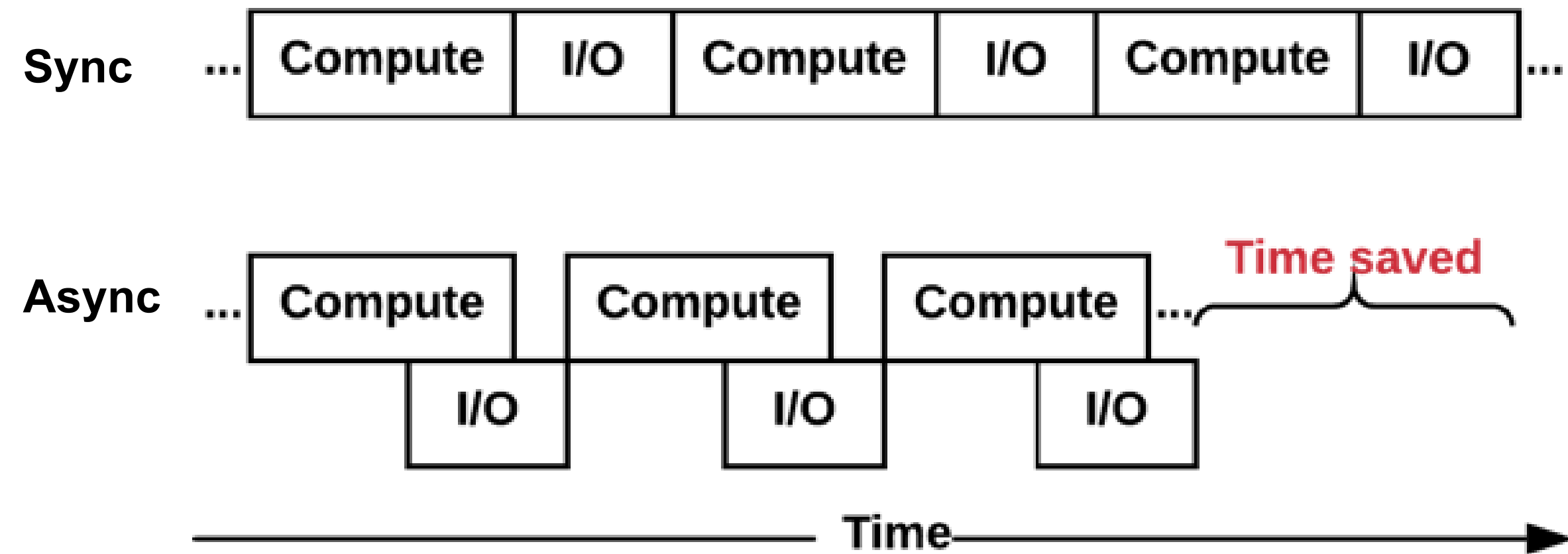
    DATA {
      10, 10, 10, 10, 10,
      10, 10, 10, 10, 10,
      11, 11, 11, 11, 11,
      11, 11, 11, 11, 11,
      12, 12, 12, 12, 12,
      12, 12, 12, 12, 12,
      13, 13, 13, 13, 13,
      13, 13, 13, 13, 13
    }
  }
}
```

General HDF5 Best Practices and Case Studies for Parallel Performance

Strongly Recommended Options

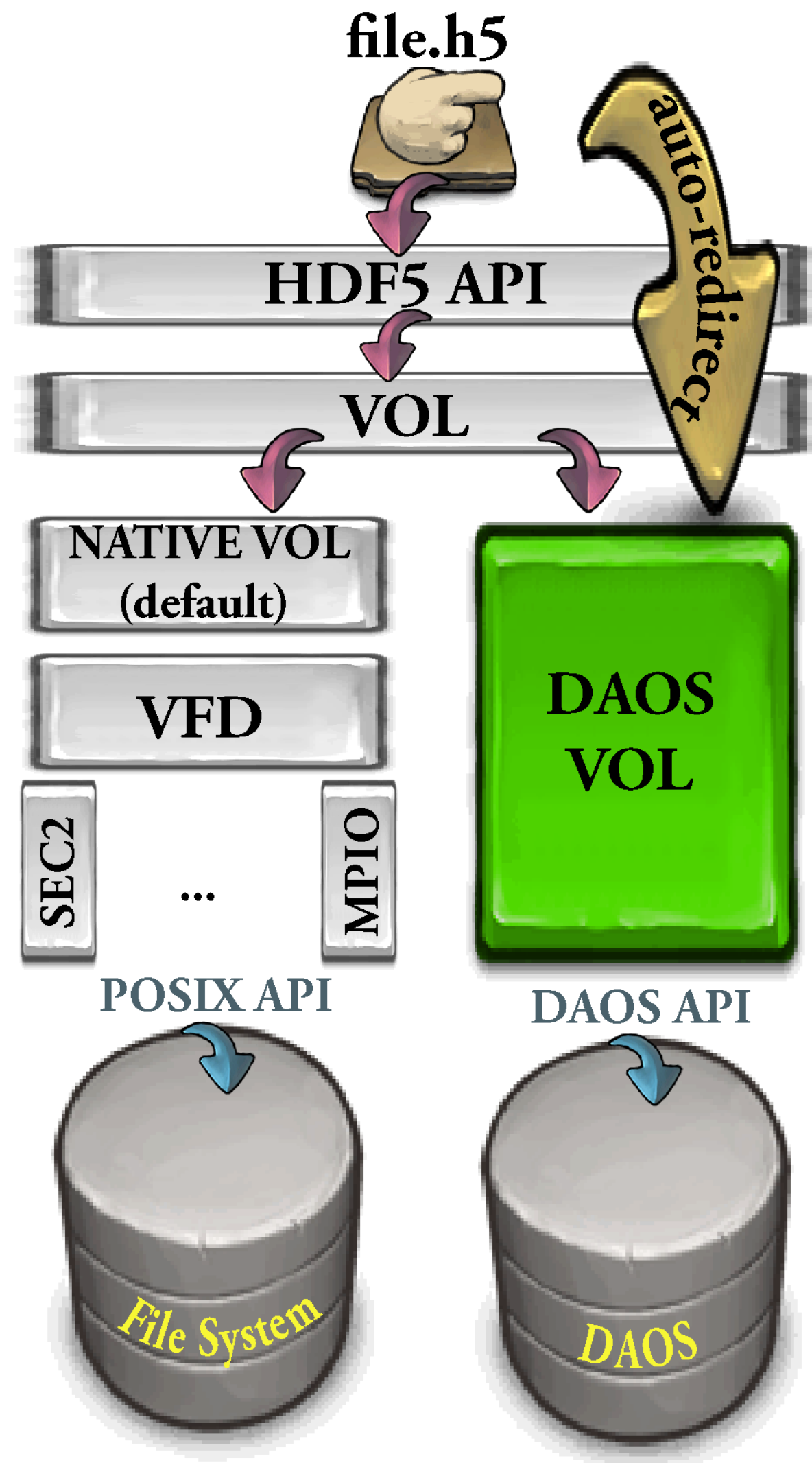
- ✓ • Hint that metadata access is done collectively
 - `H5Pset_coll_metadata_write`, `H5Pset_all_coll_metadata_ops`
 - A property on an access property list
 - If set on the file access property list, then all metadata read operations will be required to be collective
 - Can be set on individual object property list
 - When set, MPI rank 0 will issue the read for a metadata entry to the file system and broadcast to all other ranks
- ✓ • Set HDF5 to never fill chunks (`H5Pset_fill_time` with `H5D_FILL_TIME_NEVER`)

Features: Asynchronous I/O



- Allows asynchronous operations for HDF5 applications:
 - Applications use the `_async` versions for the **H5** APIs
 - Return “request tokens” to applications to track I/O tasks.
- Requires a VOL (async or DAOS) which supports asynchronous I/O, otherwise defaults to synchronous I/O.

Feature: DAOS VOL Connector

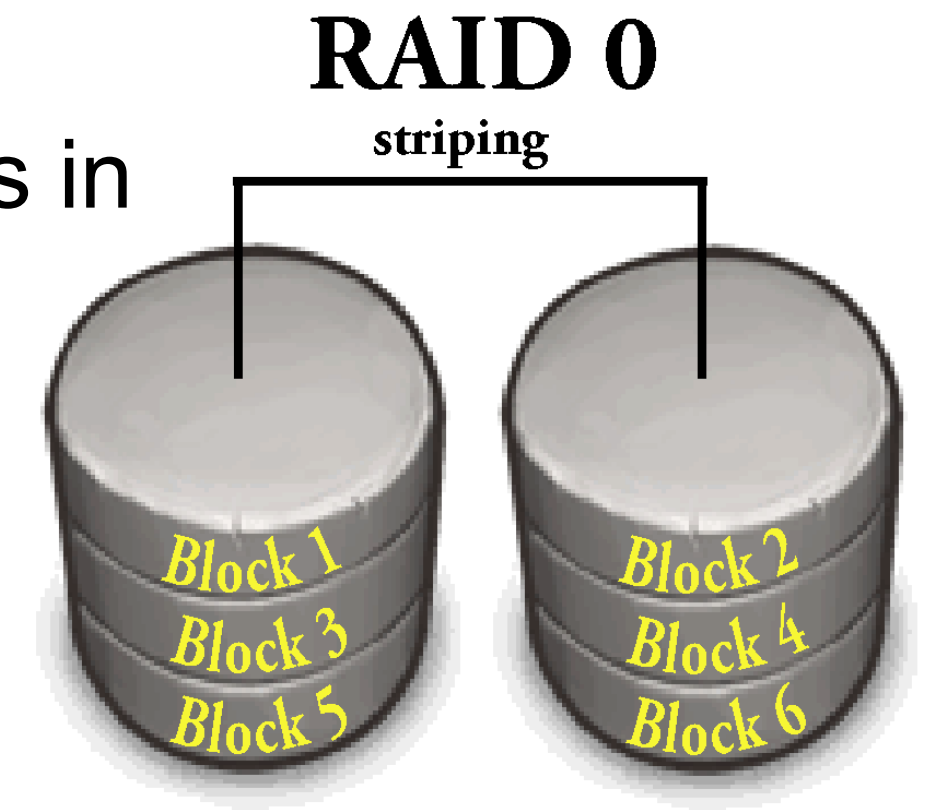


- HDF5 VOL connector for I/O to Distributed Asynchronous Object Storage (DAOS)
- Deployed at ANL.
- Minimal code changes needed to use, enable via environment variables or through HDF5 APIs.
- HDF5 tools are supported
 - h5dump, h5ls, h5diff, h5repack, h5copy, etc.
- Supports async I/O

<https://github.com/HDFGroup/vol-daos>

Subfiling

- An MPI-based parallel file driver is used to split an HDF5 file across a collection of subfiles in equally sized data segment stripes.
 - Data stripe size is the amount of data (in bytes) that can be written to a subfile before data is placed in the next subfile in a round-robin (default) fashion
 - Defaults to 1 subfile **per machine node** with 32MiB data stripes
- Subfiling is a compromise between file-per-process (*fpp*) and a single shared file (*ssf*)
 - Minimize the locking issues of *ssf* approach
 - Avoid some complexity and reduce total number of files compared to *fpp* approach
 - Designed to be flexible and configurable for different machines



- HDF5 stub file
 - Appears as a normal HDF5 file; only contains HDF5 superblock information and subfiling parameter information

```
bash-5.1$ ls
outFile.h5
outFile.h5.subfile_12190989.config
outFile.h5.subfile_12190989_1_of_4
outFile.h5.subfile_12190989_2_of_4
outFile.h5.subfile_12190989_3_of_4
outFile.h5.subfile_12190989_4_of_4
```

```
stripe_size=1048576
aggregator_count=4
subfile_count=4
hdf5_file=/home/jhenderson/subfiling/outFile.h5
subfile_dir=/home/jhenderson/subfiling
outFile.h5.subfile_12190989_1_of_4
outFile.h5.subfile_12190989_2_of_4
outFile.h5.subfile_12190989_3_of_4
outFile.h5.subfile_12190989_4_of_4
```

THANK YOU!

Questions & Comments?

Need Help?

HDF-FORUM – <https://forum.hdfgroup.org/>

HDF Helpdesk – help@hdfgroup.org

ARGONNE TRAINING PROGRAM ON EXTREME-SCALE COMPUTING

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