

Parallel HDF5

The HDF Group

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Outline

- Overview of Parallel I/O
- Overview of Parallel HDF5 design
- PHDF5 Programming Model
- Performance Analysis and use cases

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Overview of Parallel IO

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Def. Parallel I/O

At the application level

 Concurrent writing to and reading from a single file from multiple processes

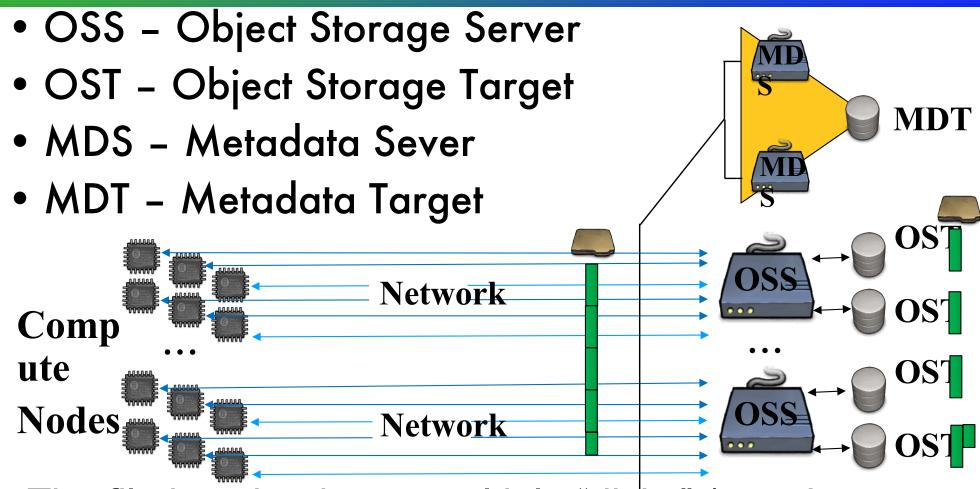
At the system level

Parallel file system which supports concurrent process access

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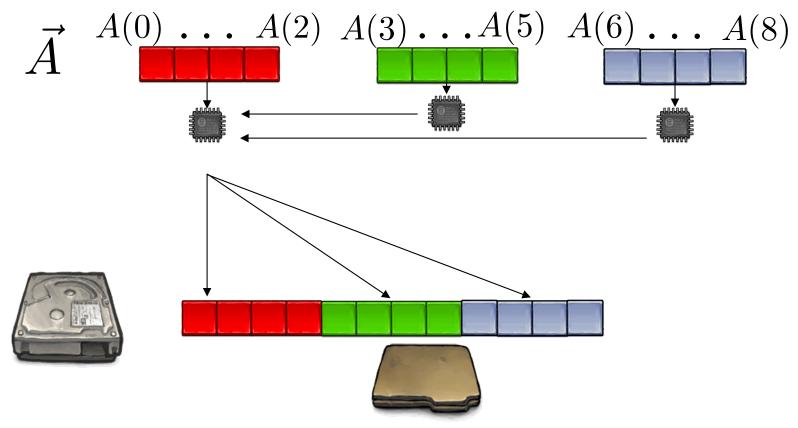
Parallel File System Nomenclature



- The file is striped over multiple "disks" (e.g., Lustre OSTs) depending on the stripe size and stripe count with which the file was created
- https://www.nics.tennessee.edu/computing-resources/file-systems/io-lustre-tips



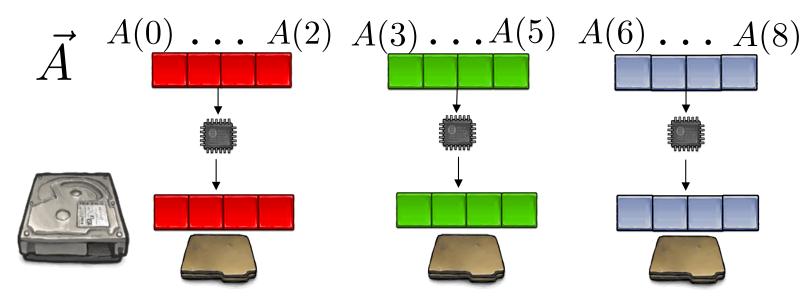
Non-parallel I/O¹



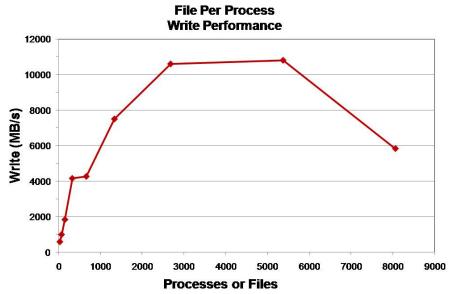
- Friends don't let friends to this
- Very bad performance
- Not scalable



Independent Parallel I/O¹



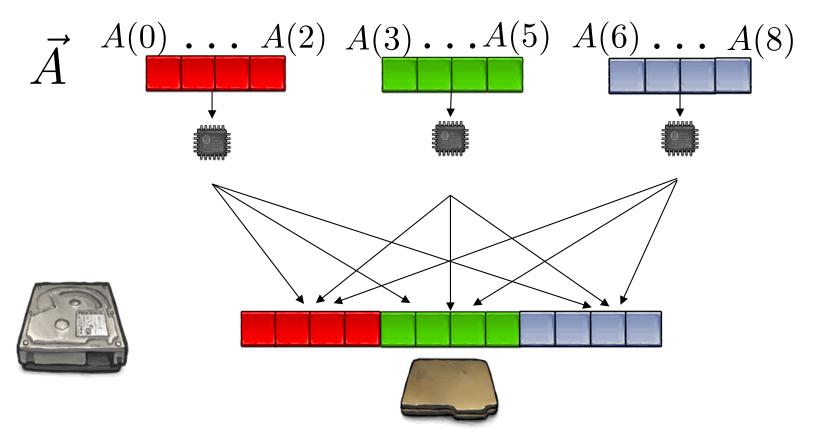
- Results in a large number of files
 - May run into file system limitation on the number of open files allowed
- Not usable from a different number of processes
- Usually, have to post-process the files
- Can achieve very good I/O performance



https://www.nics.tennessee.edu/computing-resources/file-systems/io-lustre-tips



Cooperative Parallel I/O¹



- Coordination between processes to a single file
- Must use MPI IO
- Can achieve excellent performance



MPI-IO vs. HDF5

- MPI-IO is an Input/Output API¹
 - Replacement functions for POSIX I/O
 - Handles multiple I/O schemes
 - Including schemes only available via MPI
 - It treats the data file as a "linear byte stream"
 - Each MPI process needs to provide its own file view and data representations to interpret those bytes.
 - Programmer handles
 - Collective coordination of operations
 - Creating user-defined datatypes for both in memory and file layout
 - Process layout of data in a file

¹W. Gropp, parallel@illinois



MPI-IO vs. PHDF5

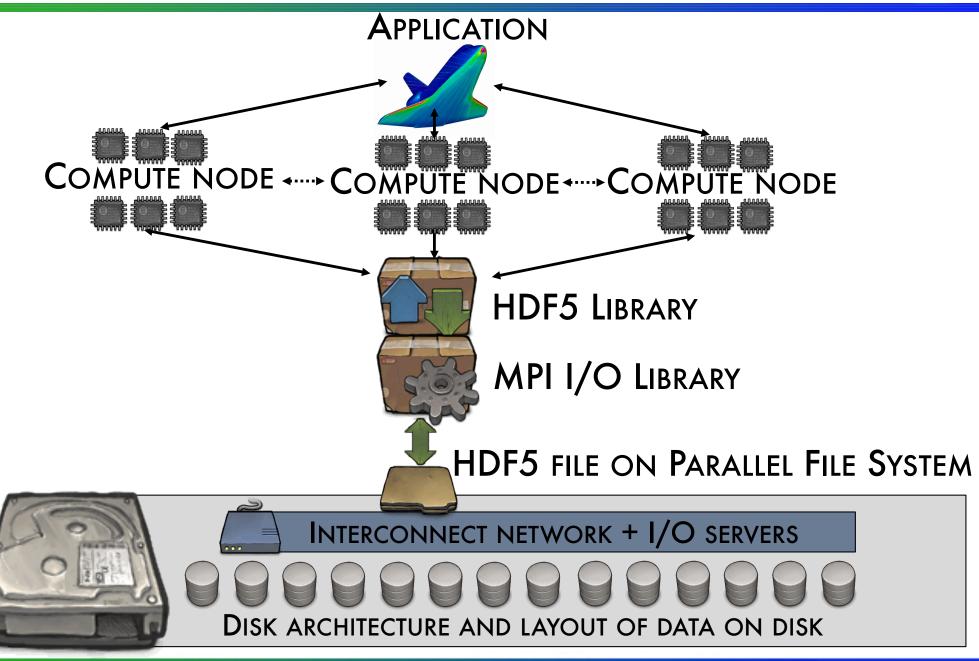
- PHDF5 is data management software
 - It stores data and metadata according to the HDF5 data format specification

http://support.hdfgroup.org/HDF5/doc/H5.format.html

- PHDF5 files are HDF5 files
 - Portable to different platforms
- PHDF5 handles the MPI I/O details at a higher level
 - It requires MPI I/O for parallel I/O[‡]
- The PHDF5 API consists of:
 - The standard HDF5 API
 - A few extra knobs and calls
 - A parallel "etiquette"



PHDF5 implementation layers





OVERVIEW OF PARALLEL HDF5 DESIGN

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How to Compile PHDF5 Applications

- Cmake
- Autotools (configure, make, etc..)
- Wrappers contain compiler/linker flags
 - h5pcc HDF5 C compiler command
 - Similar to mpicc
 - h5pfc HDF5 Fortran compiler command
 - Similar to mpif90
- To compile:
 - h5pcc h5prog.c
 - h5pfc h5prog.f90



h5pcc/h5pfc -show option

 show displays the compiler commands and options without executing them, i.e., dry run

```
h5pcc -show Sample_mpio.c

mpicc -I/home/packages/phdf5/include \
-D_LARGEFILE_SOURCE -D_LARGEFILE64_SOURCE \
-D_FILE_OFFSET_BITS=64 -D_POSIX_SOURCE \
-D_BSD_SOURCE -std=c99 -c Sample_mpio.c

mpicc -std=c99 Sample_mpio.o \
-L/home/packages/phdf5/lib \
home/packages/phdf5/lib/libhdf5_hl.a \
/home/packages/phdf5/lib/libhdf5.a -lz -lm -Wl,-rpath \
-Wl,/home/packages/phdf5/lib
```

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Collective vs. Independent Calls

- MPI definition of collective calls
 - All processes of the communicator must participate in the right order. E.g.,

```
Process 1

call A(); call B();

call A(); call B();

call A(); call B();

call A(); call B();

call A(); call A(); ...WRONG
```

• Neither mode is preferable a priori

Collective I/O: Attempts to combine multiple smaller independent I/O ops into fewer larger ops



Programming Restrictions

- Most PHDF5 APIs are collective
- PHDF5 opens a parallel file with a communicator
 - Returns a file-handle
 - Future access to the file via the file-handle
- All processes must participate in collective PHDF5 APIs
- Different files can be opened via different communicators

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Collective HDF5 calls

- All HDF5 APIs that modify structural metadata are collective!
 - File operations
 - H5Fcreate, H5Fopen, H5Fclose, etc
 - Object creation
 - H5Dcreate, H5Dclose, etc
 - Object structure modification (e.g., dataset extent modification)
 - H5Dset_extent, etc
- http://www.hdfgroup.org/HDF5/doc/RM/CollectiveCalls.html



Other HDF5 calls

- Array data transfer can be collective or independent
 - Dataset operations: H5Dwrite, H5Dread
- Collectiveness is indicated by function parameters, not by function names as in MPI API

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What does PHDF5 support?

- After a file is opened by the processes of a communicator
 - All parts of file are accessible by all processes
 - All objects in the file are accessible by all processes
 - Multiple processes may write to the same data array
 - Each process may write to individual data array

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PHDF5 API languages

- C and Fortran language interfaces
- Most platforms with MPI-IO supported. e.g.,
 - IBM AIX
 - Linux clusters
 - Cray XT

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Programming model

- HDF5 uses access template object (property list) to control the file access mechanism
- General model to access an HDF5 file in parallel:
 - Set-up MPI I/O access template (file access property list)

```
H5Fcreate (H5Fopen)

H5Screate_simple/H5Screate

H5Dcreate (H5Dopen)

H5Dread, H5Dwrite

H5Dclose

H5Sclose

H5Fclose

Create (open) File

create dataSpace

create (open) Dataset

create (open) Dataset

close Dataset

close dataSpace

close File
```

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Writing patterns

EXAMPLE

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Parallel HDF5 tutorial examples

 For simple examples how to write different data patterns see

http://www.hdfgroup.org/HDF5/Tutor/parallel.html

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General Programming model

- Each process defines memory and file hyperslabs using H5Sselect_hyperslab
- Each process executes a write/read call using hyperslabs defined, which can be either collective or independent
- The hyperslab parameters define the portion of the dataset to write to
 - Contiguous hyperslab
 - Regularly spaced data (column or row)
 - Pattern
 - Blocks

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Setup MPI-IO access template

Each process of the MPI communicator creates an access template and sets it up with MPI parallel access information

C:

```
herr t H5Pset fapl mpio(hid t plist id,
      MPI Comm comm, MPI Info info);
```

Fortran:

```
h5pset fapl mpio f(plist id, comm, info)
integer(hid t) :: plist id
integer
           :: comm, info
plist id is a file access property list identifier
```

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C Example Parallel File Create

```
23
         comm = MPI COMM WORLD;
  24
         info = MPI INFO NULL;
  26
         /*
  27
          * Initialize MPI
  28
          */
  29
         MPI Init(&argc, &argv);
         /*
  30
  34
          * Set up file access property list for MPI-IO access
  35
          */
->36
         plist id = H5Pcreate(H5P FILE ACCESS);
         H5Pset fapl mpio(plist id, comm, info);
->37
  38
->42
         file id = H5Fcreate(H5FILE NAME, H5F ACC TRUNC,
            H\overline{5}P DEFAULT, plist id);
         /*
  49
  50
          * Close the file.
  51
          */
  52
         H5Fclose(file id);
  54
         MPI Finalize();
```

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Fortran Example Parallel File Create

```
23 comm = MPI COMM WORLD
  24 info = MPI INFO NULL
  26 CALL MPI INIT (mpierror)
 29!
  30 !Initialize FORTRAN predefined datatypes
  32 CALL h5open f(error)
 34!
  35 !Setup file access property list for MPI-IO access.
->37 CALL h5pcreate f(H5P FILE ACCESS F, plist id, error)
->38 CALL h5pset fapl mpio f(plist id, comm, info, error)
 40 !
 41 !Create the file collectively.
->43 CALL h5fcreate f(filename, H5F ACC TRUNC F, file id,
         error, access prp = plist id)
 45!
  46 !Close the file.
  49 CALL h5fclose f(file id, error)
  51 1
  52 !Close FORTRAN interface
  54 CALL h5close f(error)
  56 CALL MPI FINALIZE (mpierror)
```

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Creating and Opening Dataset

- All processes of the communicator open/close a dataset by a collective call
 - ✓ C: H5Dcreate or H5Dopen; H5Dclose
 - √ Fortran: h5dcreate_f or h5dopen_f; h5dclose f
- All processes of the communicator must extend an unlimited dimension dataset before writing to it
 - ✓ C: H5Dextend
 - √ Fortran: h5dextend_f



C Example: Create Dataset

```
56
      file id = H5Fcreate(...);
  57
      /*
  58
       * Create the dataspace for the dataset.
  59
       */
  60
      dimsf[0] = NX;
  61
      dimsf[1] = NY;
  62
      filespace = H5Screate simple(RANK, dimsf, NULL);
  63
  64
      /*
  65
       * Create the dataset with default properties collective.
 66
       */
      dset id = H5Dcreate(file_id, "dataset1", H5T_NATIVE_INT,
->67
                           filespace, H5P DEFAULT);
  68
  70
      H5Dclose(dset id);
 71
     /*
  72
      * Close the file.
  73
     */
  74
     H5Fclose(file id);
```

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Fortran Example: Create Dataset

```
43 CALL h5fcreate f(filename, H5F ACC TRUNC F, file id,
         error, access prp = plist id)
  73 CALL h5screate simple_f(rank, dimsf, filespace, error)
  76!
 77 ! Create the dataset with default properties.
  78!
->79 CALL h5dcreate f(file id, "dataset1", H5T NATIVE INTEGER,
                      filespace, dset id, error)
  90!
  91! Close the dataset.
  92 CALL h5dclose f(dset id, error)
  93!
  94! Close the file.
  95 CALL h5fclose f(file id, error)
```

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Accessing a Dataset

- All processes that have opened dataset may do collective I/O
- Each process may do independent and arbitrary number of data I/O access calls
 - C: H5Dwrite and H5Dread
 - Fortran: h5dwrite_f and h5dread_f

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Programming model for dataset access

- Create and set dataset transfer property
 - C: H5Pset_dxpl_mpio
 - H5FD_MPIO_COLLECTIVE
 - H5FD_MPIO_INDEPENDENT (default)
 - Fortran: h5pset_dxpl_mpio_f
 - H5FD_MPIO_COLLECTIVE_F
 - H5FD_MPIO_INDEPENDENT_F (default)
- Access dataset with the defined transfer property

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C Example: Collective write

```
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);
```

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Fortran Example: Collective write

```
88
      ! Create property list for collective dataset write
 89
      CALL h5pcreate f(H5P DATASET XFER F, plist id, error)
  90
      CALL h5pset dxpl mpio f(plist id, &
->91
                              H5FD MPIO COLLECTIVE F, error)
  92
  93
  94
      ! Write the dataset collectively.
  95
  96
      CALL h5dwrite f(dset id, H5T NATIVE INTEGER, data, &
                      error, &
                      file space id = filespace, &
                      mem space id = memspace, &
                      xfer prp = plist id)
```

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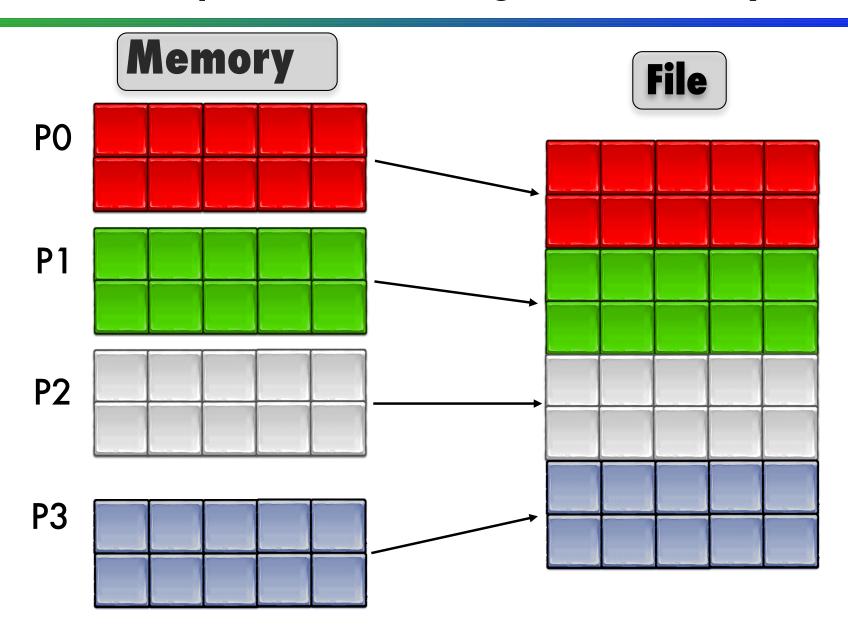
Writing and Reading Hyperslabs

- Distributed memory model: data is split among processes
- PHDF5 uses HDF5 hyperslab model
- Each process defines memory and file hyperslabs

```
H5Sselect hyperslab(
   filespace, H5S SELECT SET,
   offset, stride, count, block
)
```

- Each process executes partial write/read call
 - Collective calls
 - Independent calls

Example 1: Writing dataset by rows



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Example 1: Writing dataset by rows

Memory Process P1 offset[1] count[1] offset[0] count[0] count[0] = dimsf[0]/mpi size count[1] = dimsf[1]; offset[0] = mpi rank * count[0]; /* = 2 */ offset[1] = 0;

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Example 1: Writing dataset by rows

```
/*
71
72
     * Each process defines dataset in memory and
     * writes it to the hyperslab
73
     * in the file.
74
     */
75
     count[0] = dimsf[0]/mpi size;
76 count[1] = dimsf[1];
77
     offset[0] = mpi rank * count[0];
     offset[1] = 0;
78
79
     memspace = H5Screate simple(RANK,count,NULL);
80
81
    /*
82
     * Select hyperslab in the file.
83
     */
84
     filespace = H5Dget space(dset id);
     H5Sselect hyperslab (filespace,
85
         H5S SELECT SET, offset, NULL, count, NULL);
```

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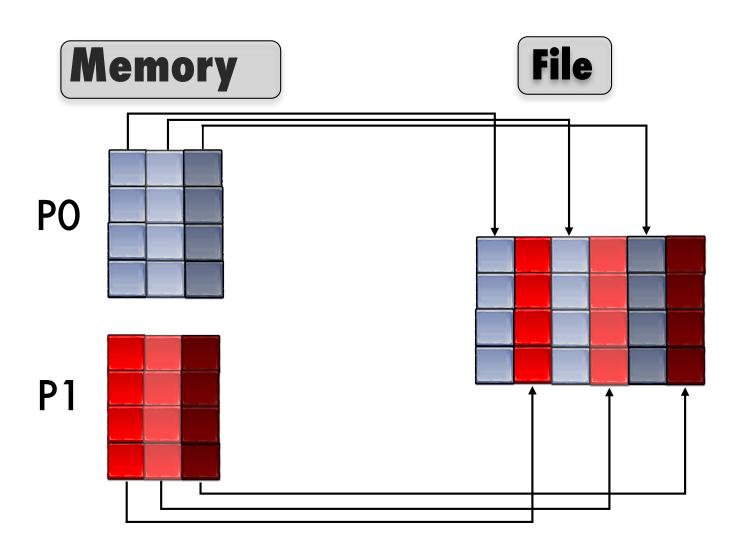
Writing by rows: Output of h5dump

```
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
```

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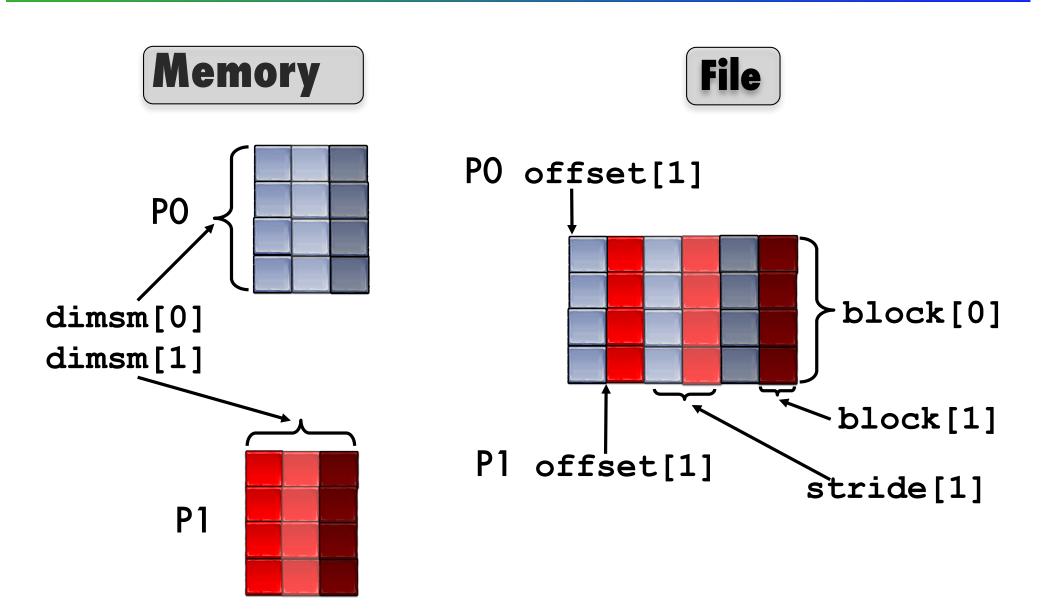


Example 2: Writing dataset by columns



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Example 2: Writing dataset by column



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column

```
/*
85
86
       Each process defines a hyperslab in
        * the file
88
     */
89
       count[0]
90
       count[1]
                  = dimsm[1];
91
       offset[0]
                  =
                    0;
92
                  = mpi rank;
       offset[1]
93
       stride[0]
                  =
       stride[1]
94
                  =
95
       block[0] = dimsf[0];
96
       block[1]
97
98
    /*
99
       Each process selects a hyperslab.
100
        */
101
       filespace = H5Dget space(dset id);
102
       H5Sselect hyperslab(filespace,
              H5S SELECT SET, offset, stride,
              count, block);
```

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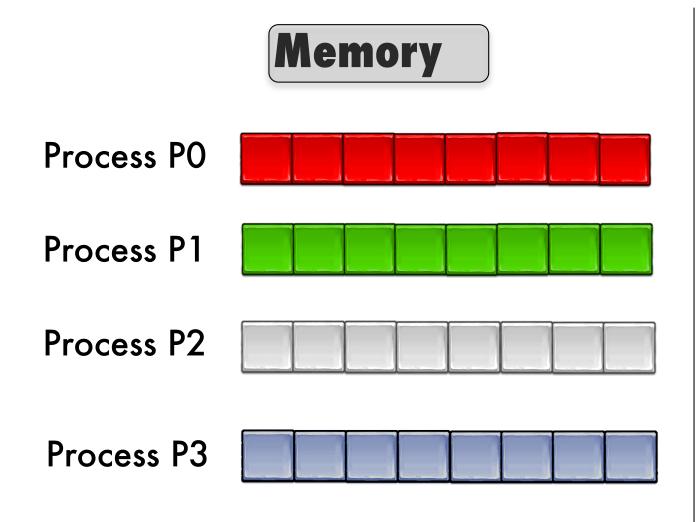
Writing by columns: Output of h5dump

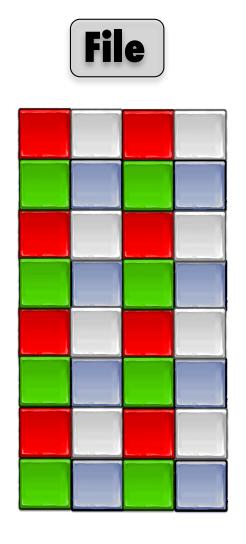
```
HDF5 "SDS col.h5" {
GROUP "/" {
   DATASET "IntArray" {
     DATATYPE H5T STD I32BE
     DATASPACE SIMPLE { (8, 6) / (8, 6) }
     DATA {
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200,
         1, 2, 10, 20, 100, 200
```

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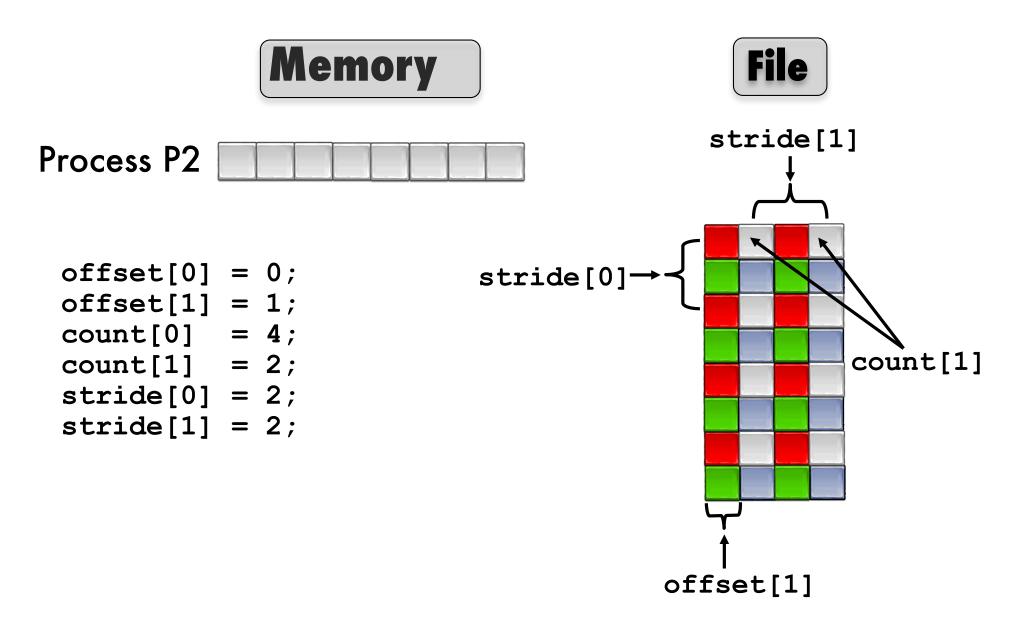
Example 3: Writing dataset by pattern







Example 3: Writing dataset by pattern



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Example 3: Writing by pattern

```
90
        /* Each process defines dataset in memory and
 91
         * writes it to the hyperslab in the file.
 92
         */
 93
        count[0] = 4;
 94
        count[1] = 2;
        stride[0] = 2;
 95
 96
        stride[1] = 2;
 97
        if(mpi rank == 0) {
 98
           offset[0] = 0;
 99
           offset[1] = 0;
100
101
        if(mpi rank == 1) {
102
           offset[0] = 1;
103
           offset[1] = 0;
104
105
        if(mpi rank == 2) {
106
           offset[0] = 0;
           offset[1] = 1;
107
108
        if(mpi rank == 3) {
109
110
           offset[0] = 1;
111
           offset[1] = 1;
112
```

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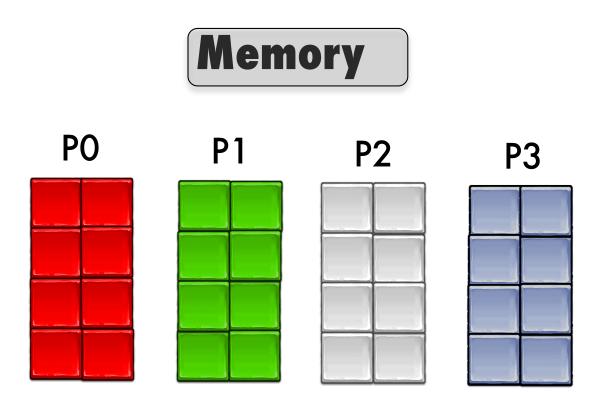
Writing by Pattern: Output of h5dump

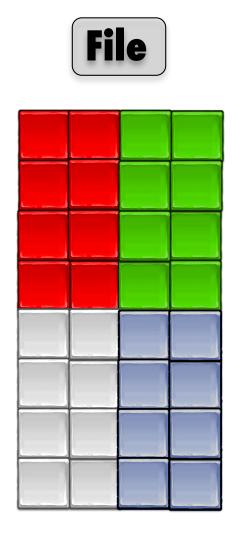
```
HDF5 "SDS pat.h5" {
GROUP "/" {
   DATASET "IntArray" {
      DATATYPE H5T STD I32BE
      DATASPACE SIMPLE { ( 8, 4 ) / ( 8, 4 ) }
      DATA {
         1, 3, 1, 3,
         2, 4, 2, 4,
         1, 3, 1, 3,
         2, 4, 2, 4,
         1, 3, 1, 3,
         2, 4, 2, 4,
         1, 3, 1, 3,
        2, 4, 2, 4
```

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Example 4: Writing dataset by chunks







-Administration -- Administration of

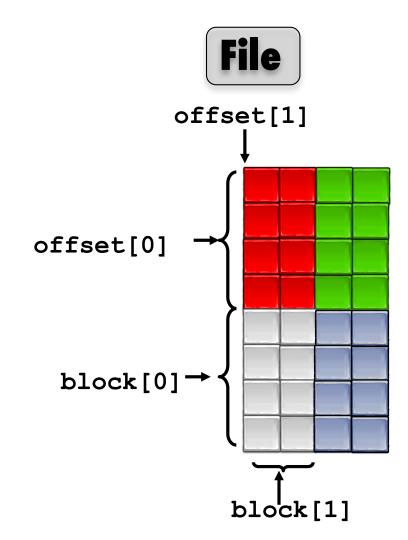
chunks

Memory

Process P2

chunk_dims[1]
chunk_dims[0]

```
block[0] = chunk_dims[0];
block[1] = chunk_dims[1];
offset[0] = chunk_dims[0];
offset[1] = 0;
```





Example 4: Writing by chunks

```
97
       count[0] = 1;
 98
              count[1] = 1 ;
              stride[0] = 1;
 99
100
              stride[1] = 1;
             block[0] = chunk dims[0];
101
              block[1] = chunk dims[1];
102
              if (mpi rank == 0) {
103
                 offset[0] = 0;
104
105
                 offset[1] = 0;
106
107
              if(mpi rank == 1) {
                 offset[0] = 0;
108
                 offset[1] = chunk dims[1];
109
110
111
              if(mpi rank == 2) {
                 offset[0] = chunk dims[0];
112
                 offset[1] = 0;
113
114
115
              if(mpi rank == 3) {
                 offset[0] = chunk dims[0];
116
                 offset[1] = chunk dims[1];
117
118
```



Writing by Chunks: Output of h5dump

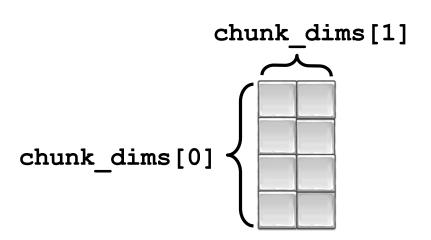
```
HDF5 "SDS chnk.h5" {
GROUP "/" {
   DATASET "IntArray" {
      DATATYPE H5T STD I32BE
      DATASPACE SIMPLE { ( 8, 4 ) / ( 8, 4 ) }
      DATA {
         1, 1, 2, 2,
         1, 1, 2, 2,
         1, 1, 2, 2,
         1, 1, 2, 2,
         3, 3, 4, 4,
         3, 3, 4, 4,
         3, 3, 4, 4,
         3, 3, 4, 4
```

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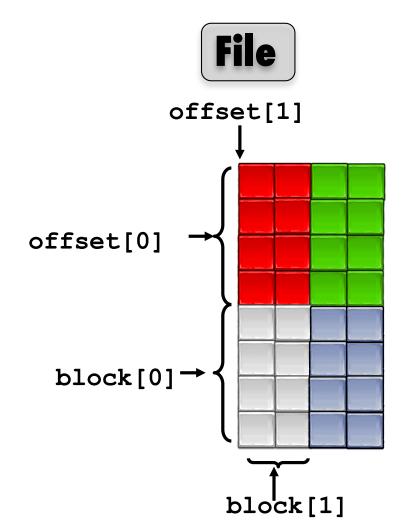


Demo

Memory



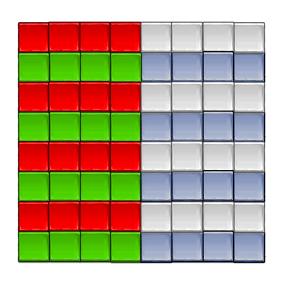
```
block[0] = chunk_dims[0];
block[1] = chunk_dims[1];
offset[0] = chunk_dims[0];
offset[1] = 0;
```

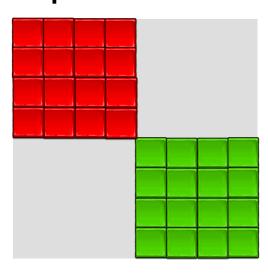


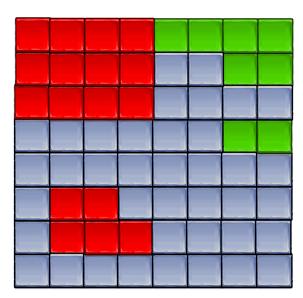
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Complex data patterns

HDF5 doesn't have restrictions on data patterns and balance







 Irregular hyperslabs created by union operators

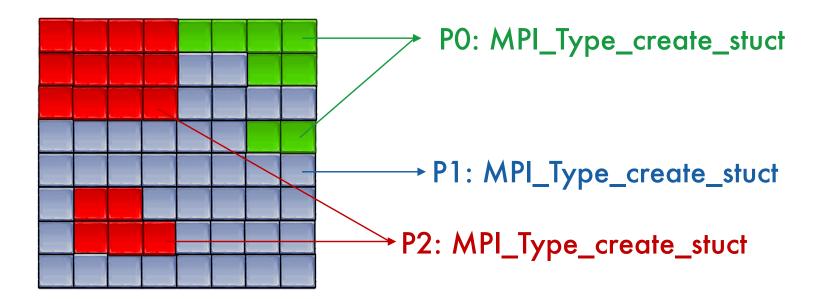
H5Sselect_hyperslab(space_id, op,
start, stride, count, block)

Complex data patterns – Selection





Examples of irregular selection



Internally...

- 1. The HDF5 library creates an MPI datatype for each lower dimension in the selection
- 2. It then combines those types into one giant structured MPI datatype

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PERFORMANCE ANALYSIS

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My PHDF5 Application I/O is slow

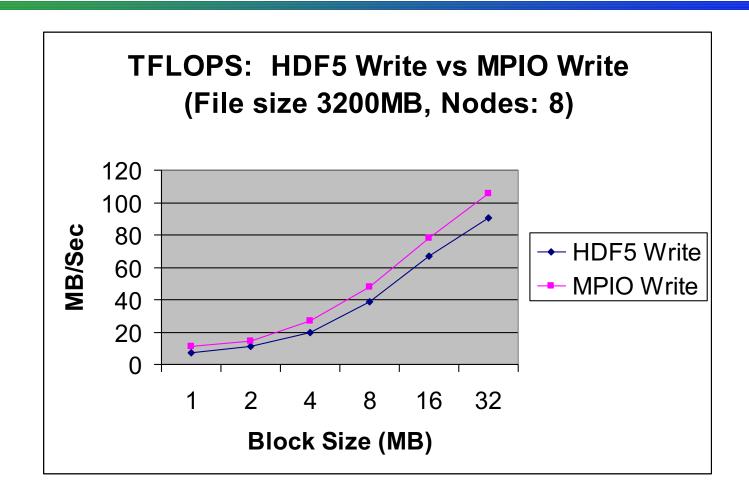
Common Solutions to poor performance

- Use larger I/O data sizes
- Independent vs. Collective I/O
- Specific I/O system hints

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Write Speed vs. Block Size



Tip: Minimize I/O calls by performing large data I/O

My PHDF5 Application I/O is slow

- Use larger I/O data sizes
- Independent vs. Collective I/O
- Specific I/O system hints

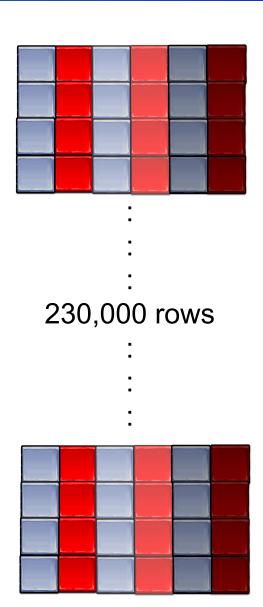
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Independent vs. Collective Access

User Report

- Independent data transfer mode is much slower than the Collective data transfer mode
- Data array is tall and thin:
 230,000 rows by 6 columns





Debug Slow Parallel I/O Speed(1)

- Writing to one dataset
 - Using 4 processes == 4 columns
 - Datatype is 8-byte float (doubles)
 - 4 processes $\times 1000$ rows $\times 8$ bytes = 32,000 bytes
 - mpirun -np 4 ./a.out 1000
 - Execution time: 1.783798 s
 - mpirun -np 4 ./a.out 2000
 - Execution time: 3.838858 s
- 2 seconds for 1000 more rows = 32,000 bytes.

Speed of 16KB/sec!!! Way too slow.

Linear

scaling



Debug slow parallel I/O speed(2)

- Build a version of PHDF5 with
 - ./configure --enable-build-mode=debug --enable-parallel ...
 - This allows the tracing of MPIO I/O calls in the HDF5 library such as MPI_File_read_xx and MPI_File_write_xx
- To trace

setenv H5FD_mpio_Debug "rw"

The report will look something like this...

Debug slow parallel I/O speed(3)

```
setenv H5FD_mpio_Debug 'rw'
mpirun -np 4 ./a.out 1000 # Indep.; contiguous.
 in H5FD_mpio_write mpi_off=0 size_i=96
 in H5FD_mpio_write mpi_off=0 size_i=96
 in H5FD_mpio_write mpi_off=0 size_i=96
 in H5FD_mpio_write mpi_off=0 size_i=96
 in H5FD_mpio_write mpi_off=2056 size_i=8
 in H5FD_mpio_write mpi_off=2048 size_i=8
 in H5FD_mpio_write mpi_off=2072 size_i=8
 in H5FD_mpio_write mpi_off=2064 size_i=8
 in H5FD_mpio_write mpi_off=2088 size_i=8
 in H5FD_mpio_write mpi_off=2080 size_i=8
```

 Total of 4000 of these little 8 bytes writes == 32,000 bytes.

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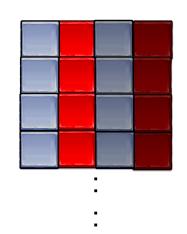


Independent calls are many and small

Diagnosis

 Each process writes one element of one row, skips to next row, writes one element, and so on...

Each process issues 230,000 writes of 8 bytes each.



230,000 rows



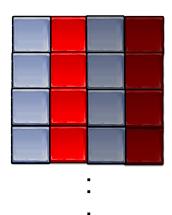


Use Collective Mode or Chunked Storage

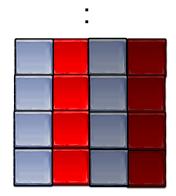
Remedy

 Collective I/O mode will combine many small independent calls into few but bigger calls

Chunks of columns speeds up too



230,000 rows





Debug slow parallel I/O speed (4)

```
setenv H5FD mpio Debug 'rw'
 mpirun -np 4 ./a.out 1000
                                   # Indep., Chunked by column.
                                   size_i=96
in H5FD_mpio_write mpi_off=0
in H5FD_mpio_write mpi_off=0
                                   size_i=96
                                   size_i=96
in H5FD_mpio_write mpi_off=0
in H5FD_mpio_write mpi_off=0
                                   size_i=96
in H5FD_mpio_write mpi_off=3688
                                   size_i=8000
in H5FD_mpio_write mpi_off=11688
                                   size i=8000
in H5FD_mpio_write mpi_off=27688
                                   size_i=8000
                   mpi_off=19688
in H5FD_mpio_write
                                   size i=8000
in H5FD_mpio_write mpi_off=96
                                   size i=40
in H5FD_mpio_write mpi_off=136
                                   size_i=544
in H5FD_mpio_write mpi_off=680
                                   size_i=120
in H5FD_mpio_write mpi_off=800
                                   size_i=272
```

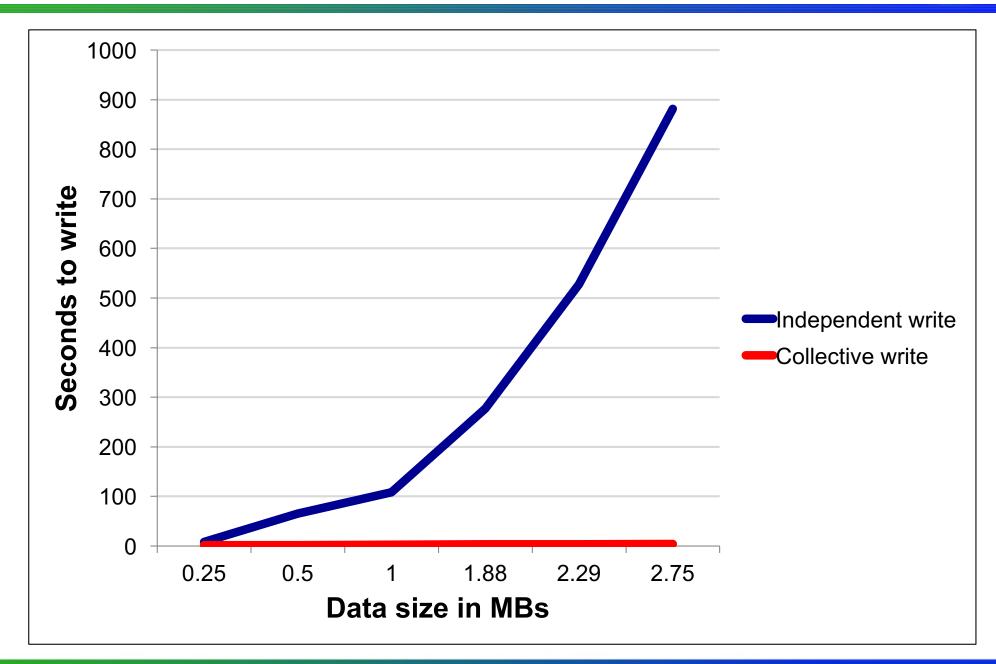


Execution time: 0.011599 s.





Collective vs. independent write



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Collective I/O in HDF5

- Set-up using a Data Transfer Property List (DXPL)
- All processes must participate in the I/O call (H5Dread/write) with a selection (which could be a NULL selection)
- Some cases where collective I/O is not used even when the use asks for it:
 - Data conversion
 - Compressed Storage
 - Chunking Storage:
 - When the chunk is not selected by a certain number of processes



Collective I/O in HDF5

- Can query Data Transfer Property List (DXPL) after I/O for collective I/O status:
 - H5Pget_mpio_actual_io_mode
 - Retrieves the type of I/O that HDF5 actually performed on the last parallel I/O call
 - H5Pget_mpio_no_collective_cause
 - Retrieves local and global causes that broke collective I/O on the last parallel I/O call
 - H5Pget_mpio_actual_chunk_opt_mode
 - Retrieves the type of chunk optimization that HDF5 actually performed on the last parallel I/O call. This is not necessarily the type of optimization requested



Enabling Collective Parallel I/O

```
/* Set up file access property list w/parallel I/O access */
fa_plist_id = H5Pcreate(H5P_FILE_ACCESS);
H5Pset_fapl_mpio(fa_plist_id, comm, info);
/* Create a new file collectively */
file_id = H5Fcreate(filename, H5F_ACC_TRUNC,
      H5P_DEFAULT, fa_plist_id);
/* <omitted data decomposition for brevity> */
/* Set up data transfer property list w/collective MPI-IO */
dx_plist_id = H5Pcreate(H5P_DATASET_XFER);
H5Pset_dxpl_mpio(dx_plist_id, H5FD_MPIO_COLLECTIVE);
/* Write data elements to the dataset */
status = H5Dwrite(dset_id, H5T_NATIVE_INT,
      memspace, filespace, dx_plist_id, data);
```

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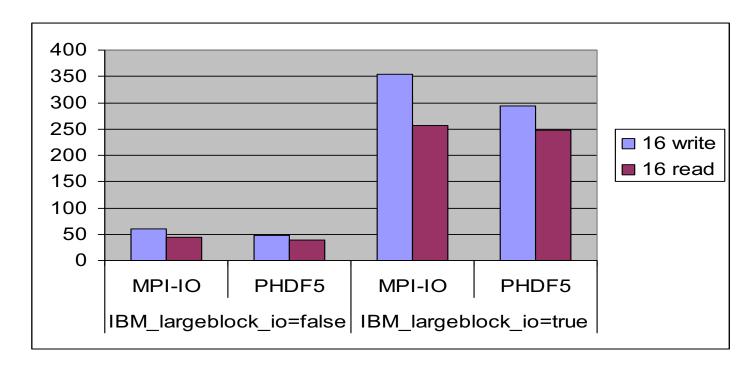
My PHDF5 Application I/O is slow

- Use larger I/O data sizes
- Independent vs. Collective I/O
- Specific I/O system hints

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Effects of I/O Hints: IBM_largeblock_io

- GPFS at LLNL ASCI Blue machine
 - 4 nodes, 16 tasks
 - Total data size 1024MB
 - I/O buffer size 1MB





Parallel I/O Profiling Tools

- Two kinds of tools
 - I/O benchmarks for measuring a system's I/O capabilities
 - I/O profilers for characterizing applications' I/O behavior
- Couple of examples
 - H5perf (in the HDF5 source code distro)
 - Darshan (from Argonne National Laboratory)
- Profilers have to compromise between
 - A lot of detail → large trace files and overhead
 - Aggregation → loss of detail, but low overhead



h5perf

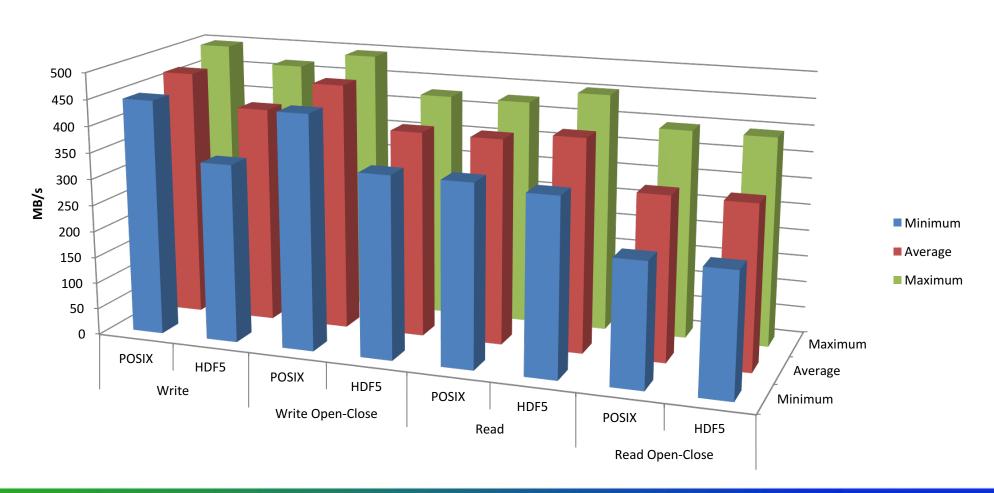
- Measures filesystem performance for different I/O patterns and APIs
- Three File I/O types in one convent package
 - 1. POSIX I/O (open/write/read/close...)
 - 2. MPI I/O (MPI_File_{open,write,read,close})
 - 3. PHDF5
- An indication of I/O speed ranges and HDF5 overhead



h5perf

A Serial Run

h5perf_serial, 3 iterations, 1 GB dataset, 1 MB transfer buffer, HDF5 dataset contiguous storage, HDF5 SVN trunk, NCSA BW



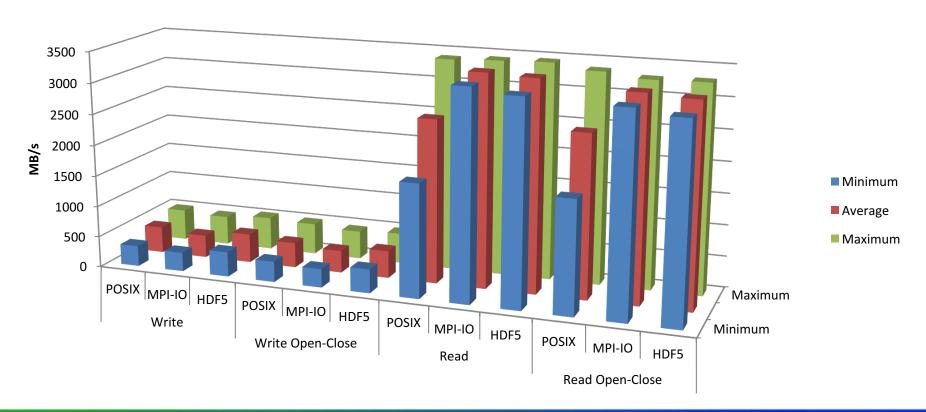
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h5perf

A Parallel Run

h5perf, 3 MPI processes, 3 iterations, 3 GB dataset (total), 1 GB per process, 1 GB transfer buffer, HDF5 dataset contiguous storage, HDF5 SVN trunk, NCSA BW





h5perf: example output 1/3

```
mpirun -np 4 h5perf
                           # Ran in a Linux system
Number of processors = 4
    Transfer Buffer Size: 131072 bytes, File size: 1.00 MBs
      # of files: 1, # of datasets: 1, dataset size: 1.00 MBs
        IO API = POSIX
            Write (1 iteration(s)):
                Maximum Throughput: 18.75 MB/s
                Average Throughput: 18.75 MB/s
                Minimum Throughput: 18.75 MB/s
            Write Open-Close (1 iteration(s)):
                Maximum Throughput: 10.79 MB/s
                Average Throughput: 10.79 MB/s
                Minimum Throughput: 10.79 MB/s
            Read (1 iteration(s)):
                Maximum Throughput: 2241.74 MB/s
                Average Throughput: 2241.74 MB/s
                Minimum Throughput: 2241.74 MB/s
            Read Open-Close (1 iteration(s)):
                Maximum Throughput: 756.41 MB/s
                Average Throughput: 756.41 MB/s
                Minimum Throughput: 756.41 MB/s
```

h5perf: example output 2/3

```
mpirun -np 4 h5perf
         IO API = MPIO
             Write (1 iteration(s)):
                 Maximum Throughput: 611.95 MB/s
                 Average Throughput: 611.95 MB/s
                 Minimum Throughput: 611.95 MB/s
             Write Open-Close (1 iteration(s)):
                 Maximum Throughput: 16.89 MB/s
                 Average Throughput: 16.89 MB/s
                 Minimum Throughput: 16.89 MB/s
             Read (1 iteration(s)):
                 Maximum Throughput: 421.75 MB/s
                 Average Throughput: 421.75 MB/s
                 Minimum Throughput: 421.75 MB/s
             Read Open-Close (1 iteration(s)):
                 Maximum Throughput: 109.22 MB/s
                 Average Throughput: 109.22 MB/s
                 Minimum Throughput: 109.22 MB/s
```



h5perf: example output 3/3

```
mpirun -np 4 h5perf
      IO API = PHDF5 (w/MPI-I/O driver)
          Write (1 iteration(s)):
              Maximum Throughput: 304.40 MB/s
              Average Throughput: 304.40 MB/s
              Minimum Throughput: 304.40 MB/s
          Write Open-Close (1 iteration(s)):
              Maximum Throughput: 15.14 MB/s
              Average Throughput: 15.14 MB/s
              Minimum Throughput: 15.14 MB/s
          Read (1 iteration(s)):
              Maximum Throughput: 1718.27 MB/s
              Average Throughput: 1718.27 MB/s
              Minimum Throughput: 1718.27 MB/s
          Read Open-Close (1 iteration(s)):
              Maximum Throughput: 78.06 MB/s
              Average Throughput: 78.06 MB/s
              Minimum Throughput: 78.06 MB/s
  Transfer Buffer Size: 262144 bytes, File size: 1.00 MBs
    # of files: 1, # of datasets: 1, dataset size: 1.00 MBs
```

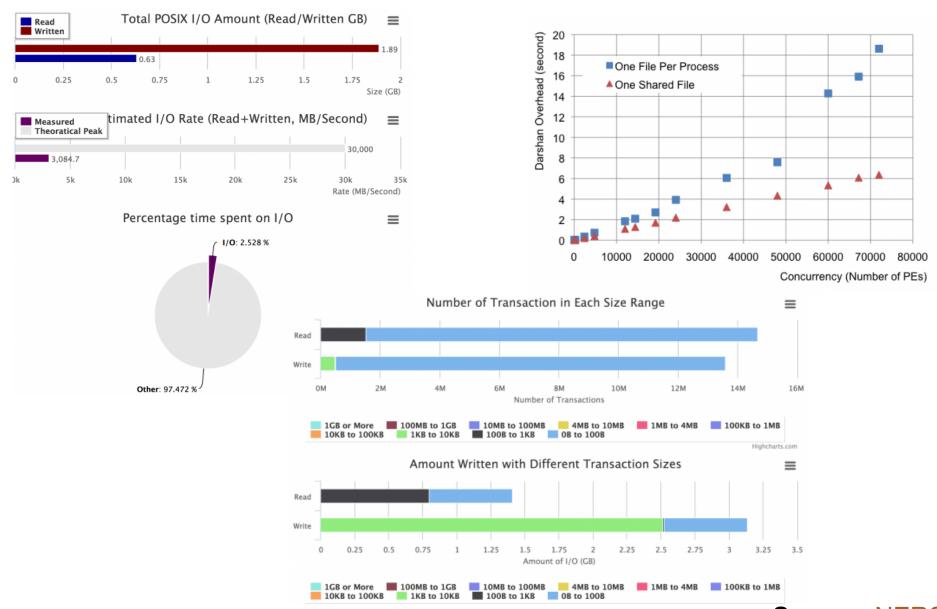


Darshan (ANL)

- Design goals:
 - Transparent integration with user environment
 - Negligible impact on application performance
- Provides aggregate figures for:
 - Operation counts (POSIX, MPI-IO, HDF5, PnetCDF, ...)
 - Datatypes and hint usage
 - Access patterns: alignments, sequentially, access size
 - Cumulative I/O time, intervals of I/O activity
- Does not provide I/O behavior over time
- Excellent starting point, maybe not your final stop



Darshan Sample Output



Source: NERSC



Lastly, Don't reinvent the wheel...

- Make use of libraries which utilize HDF, but represent the scientific data using a set of conventions, i.e., standard way for
 - Representing meshes
 - Variable definitions
 - Multiple datasets
 - Component definitions
- Some high-level parallel formats by field using HDF
 - Computation Fluid Dynamics: CGNS
 - Meshless/Particle Methods: H5Part
 - Finite element method: MOAB
 - Earth science: NetCDF
- Hides the complexity of PHDF5, but still must know concepts of parallel I/O and the underlying file systems





Examples

CGNS

Reference:

Parallel and Large-scale Simulation Enhancements to CGNS, By Scot Breitenfeld, The HDF Group, 2015.



CFD Standard

- CGNS = Computational Fluid Dynamics (CFD) General Notation System
- An effort to standardize CFD input and output data including:
 - Grid (both structured and unstructured), flow solution
 - Connectivity, boundary conditions, auxiliary information.
- Two parts:
 - A standard format for recording the data
 - Software that reads, writes, and modifies data in that format.
- An American Institute of Aeronautics and Astronautics Recommended Practice





CGNS Storage Evolution

- CGNS data was originally stored in ADF ('Advanced Data Format')
 - ADF lacks parallel I/O or data compression capabilities
 - Doesn't have HDF5's support base and tools
- HDF5 superseded ADF as the official storage mechanism for CGNS
- CGNS introduced parallel I/O APIs w/ parallel HDF5 in 2013
 - However, poor performance of the new parallel APIs in most circumstances
 - In 2014, NASA provided funding for The HDF Group with the goal to improve the under-performing parallel capabilities of the CGNS library

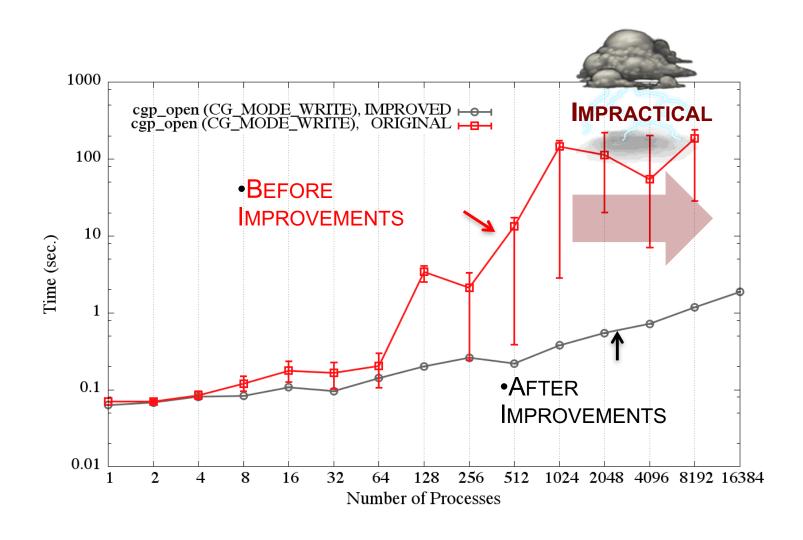


CGNS Performance Problems

- Opening an existing file
 - CGNS reads the entire HDF5 file structure, loading a lot of (HDF5) metadata
 - Reads occur independently on ALL ranks competing for the same metadata
 - →Termed "Read Storm"
- Closing a CGNS file
 - Triggers HDF5 flush of a large amount of small metadata entries
 - Implemented as iterative, independent writes, an unsuitable workload for parallel file systems



Opening CGNS File ...



HOF Metadata Read Storm Problem (I)

All metadata "write" operations are required to be collective:

```
if(0 == rank)
    H5Dcreate("dataset1");
else if(1 == rank)
    H5Dcreate("dataset2");
```

```
/* All ranks have to call */
H5Dcreate("dataset1");
H5Dcreate("dataset2");
```

Metadata read operations are not required to be collective:

```
•if(0 == rank)
• H5Dopen("dataset1");
•else if(1 == rank)
• H5Dopen("dataset2");
```

```
/* All ranks have to call */H5Dopen("dataset1");H5Dopen("dataset2");
```



Metadata Read Storm Problem (II)

- Metadata read operations are treated by the library as independent read operations.
- Consider a very large MPI job size where all processes want to open a dataset that already exists in the file.

All processes...

- Call H5Dopen("/G1/G2/D1");
- Read the same metadata to get to the dataset and the metadata of the dataset itself
- IF metadata not in cache, THEN read it from disk.
- Might issue read requests to the file system for the same small metadata.





Avoiding a Read Storm

- Application sets hint that metadata access is done collectively
 - 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
- If set, MPI rank 0 will issue the read for a metadata entry to the file system and broadcast to all other ranks

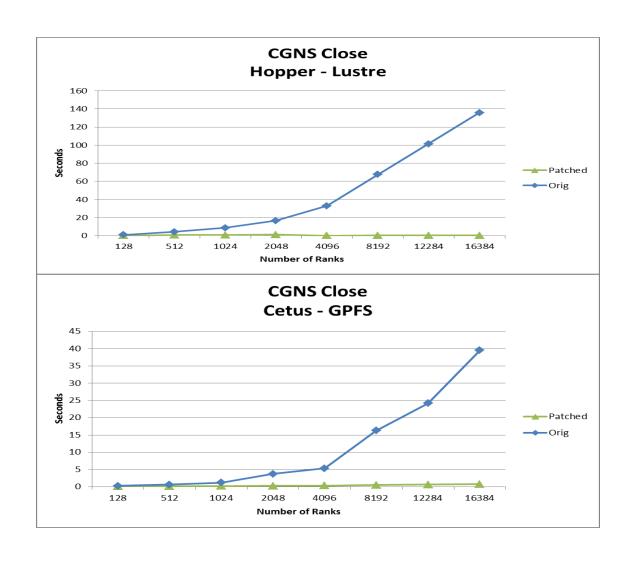


Write Metadata Collectively!

- **Symptoms:** Many users reported that H5Fclose() is very slow and doesn't scale well on parallel file systems.
- Diagnosis: HDF5 metadata cache issues very small accesses (one write per entry). We know that parallel file systems don't do well with small I/O accesses.
- Solution: Gather up all the entries of an epoch, create an MPI derived datatype, and issue a single collective MPI write.



Closing a CGNS File ...





Useful parallel HDF5 links

- Parallel HDF information site
 http://www.hdfgroup.org/HDF5/PHDF5/
- Parallel HDF5 tutorial available at http://www.hdfgroup.org/HDF5/Tutor/
- HDF Help email address help@hdfgroup.org



Questions?

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Support: Lawrence Berkeley Nation Lab - SCIDAC

Graphics: Buuf, CC BY-NC-SA 2.5