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## A brief report on Parallel I/O

#### **ASSIGNMENT-2**

Name: Khan Md Shahedul Islam	Std id: 2018380130
Name: Amirbek Raimov	Std id: 2018380038

NORTHWESTERN POLYTECHNICAL UNIVERSITY

## 1. Introduction

In this report me and my teammate went through in depth of parallel I/O topics and elaborately explained each topic from a conceptual perspective. The motivation behind the report was the excellent classes conducted by the instructors and the materials provided to practice and learn hands on.

Parallel computing means that computers can execute code more efficiently, which can save time and money by sorting through "big data" faster than ever. Parallel programming can also solve more complex problems, bringing more resources to the table and as modern era are becoming more and more data driven, parallel programming has become an essential skillset to possess in 2021.

With that being kept on mind, we went ahead, did our extensive reading and prepared this report.

## 1.1. About this report

This report is based on High Performance I/O systems, parallel I/O APIs and I/O optimization techniques. The report presents a description of the storage architecture and the I/O software stack.

## 1.2. Report Structure

We have designed our report in the following structure:

- Introduction
- High Performance I/O Systems
- Parallel File Systems
- MPI-IO
- File per process
- High-level parallel I/O libraries
- Parallel I/O performance analysis

## 2. High Performance I/O Systems

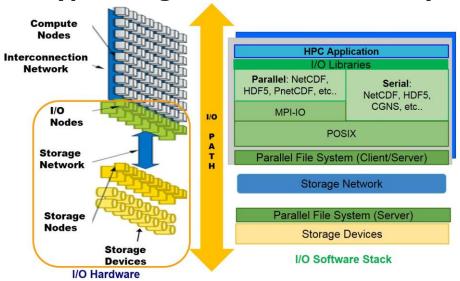
## 2.1. Introduction

Current HPC facilities are composed of thousands of compute nodes, high performance networking infrastructures, and parallel I/O environments capable of scaling to terabytes/second of I/O bandwidth while providing tens of petabytes of capacity. In HPC systems, the I/O is a subsystem composed by software and hardware as can be seen in Figure 1, "Typical

High Performance I/O System". The I/O software stack includes the I/O libraries, file system and some operating system utilities. The I/O hardware, also known as I/O infrastructure, is composed of the storage network, storage nodes, I/O nodes and the I/O devices. In most cases the available I/O resources for a single application scale together with the amount of reserved compute resources. However, their efficient usage depends on the individual application implementation.

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Figure 1. Typical High Performance I/O System

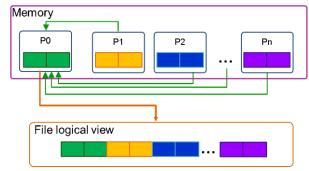


# 2.2. I/O Strategies in Parallel Applications

Parallel applications usually perform I/O in both serial and parallel. We briefly describe serial I/O below but the remainder of this report will concentrate on parallel I/O.

## 2.2.1. Serial I/O

Figure 2. Typical serial I/O in parallel applications

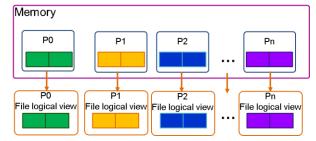


In serial I/O (see Figure 2, "Typical serial I/O in parallel applications"), a single process accesses a single file. As all of the data must flow to a single process, the bandwidth is generally limited by the amount of data that can be passed from the I/O system to a single client (usually a single compute node) leading to low I/O performance. Serial I/O operations should be limited to small data volume access performed infrequently.

#### **2.2.2. Parallel I/O**

In essence, a file is simply a stream of bytes so a user may have to substantially rearrange their program data before writing it to disk. For example, if a weather model has three main arrays storing air velocity, pressure and temperature then it might make sense for all values for a particular grid point to be stored together in memory within the application (e.g., for performance reasons). However, in the file it might be preferable for the velocities for all grid points to be stored in sequence, then all the pressure values then the temperatures (e.g. to help in postprocessing). The problem in parallel is that data rearrangement is almost always required if the parallel code should produce the same file as the serial one. For example, in a general domain decomposition parallel task do not own a single contiguous chunk of the global data set. Even in a simple 2D decomposition, the local data comes from many different locations in the file, with each local row coming from a different place.

## Figure 3. File per process approach to parallel I/O

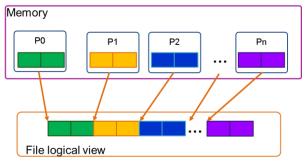


Single file, single writer

This is the other extreme, where a single master task coordinates the data rearrangement, e.g., receiving rows from many tasks and reconstructing the global data set prior to writing it out. This pattern is also called Master I/O. Normally a single process cannot benefit from the total available bandwidth and such an access scheme is also limited by the memory capabilities of the single master process. Larger chunks of data might be transferred step by step which serializes the write process even more.

Single file, multiple writers

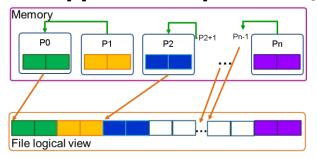
Figure 4. Single file, multiple writers' approach to parallel I/O



Single file, collective writers

This sits between the two approaches above, where either one or all of the parallel tasks perform I/O; here we identify a subset of tasks to perform the I/O operations. These I/O tasks must communicate with the computational tasks to receive and rearrange the data, and must coordinate with each other to avoid I/O clashes.

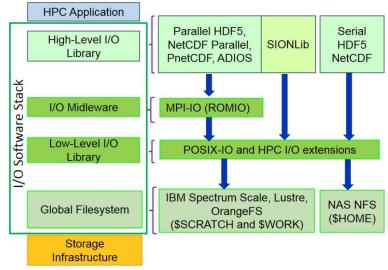
Figure 5. Single file, collective writers approach to parallel I/O



Note that, other than "multiple files, multiple writers", all these methods should produce identical output to each other on any number of processors. However, they may have very different performance characteristics. The "multiple files, multiple writers" scheme creates a local process related data view (in comparison of the global data view for all other schemes).

## 2.3. The I/O Software Stack

#### Figure 6. The I/O Software Stack



The I/O software stack provides users and administrators with application programming interfaces (APIs) that allow them to use the hardware in a common way without worrying about the specific hardware technology. Parallel I/O libraries provide APIs that enable parallel access to a single or several files. Unlike the parallel versions, serial I/O libraries (such as those that provide the basic file operations in high-level programming languages: C/C++, Fortran, Python) do not usually offer specific APIs for parallel access.

#### 2.3.1. Low-level: POSIX

The lowest level in the software stack we find is the POSIX interface, which refers to file operations such as open, close, read, write, stat and so on. POSIX HPC extensions were designed to improve performance of POSIX on large-scale HPC environments where the requirement for performance usually outweighs consistency considerations. For example, it often relaxes the rather strict POSIX consistency semantics [1]. In general, most other APIs are build on top of POSIX and POSIX approaches are often used in context of task local file access.

#### 2.3.2. Middle-level: MPI-IO

On HPC systems the middle level in the I/O software stack is dominated by MPI-IO. By using MPI-IO, it is possible to apply optimization techniques such as collective buffering and data sieving. ROMIO [12] is the most common implementation of the MPI-IO standard and it is used in MPI distributions such as MPICH (which also covers Cray MPT and HPE MTP), MVAPICH, IBM PE and Intel MPI.

#### 2.3.3. High level I/O software

The highest level in I/O software stack (See Figure 6, "The I/O Software Stack") presents the high-level libraries. These are APIs that help to express scientific simulation data in a more natural way such as multi-dimensional data, labels and tags, non-contiguous data and typed data. Parallel versions sit on top of the MPI-IO layer and can use MPI-IO optimizations. High-level libraries provide simplicity for visualization and analysis; and portable formats. HDF5 [7] and NetCDF [8] are the most popular high-level libraries. Over the last years Pentad [9] and ADIOS [10] have also been selected by HPC users to perform parallel I/O. Another library that is gaining popularity is SIONLib [11], a scalable I/O library for parallel access to task-local files. The library not only supports writing and reading binary data to or from several thousands of processors into a single or a small number of physical files but also provides global open and close functions to access SIONlib file formats in parallel.

## 2.4. General Pointers for Efficient I/O

A few "rules of thumb" are given below to consider when running or designing I/O-intensive applications on HPC systems.

- Avoid unnecessary I/O. For example, switch off debug output for production runs.
- Perform I/O in few and large chunks. In parallel file systems, the chunk size should be a multiple of the block size or stripe size.
- Prefer binary/unformatted I/O instead of formatted data.
- Avoid unnecessary/large-scale open/close statements. Remember that metadata operations are latency bound.
- Use an appropriate file system. Parallel file systems may not scale well for metadata operations, but provide high/scalable bandwidth. NFS-based file systems may show the reversed behavior.
- Avoid explicit flushes of data to disk, except when needed for consistency reasons.
- Use specialized I/O libraries based on the I/O requirements of your applications. These provide more portable way of writing data and may reduce metadata load when properly used.
- Convert to target / visualization format in memory if possible.

 For parallel programs, a file-per-process strategy can provide high throughput, but usually needs a further postprocessing stage to collate the outputs.

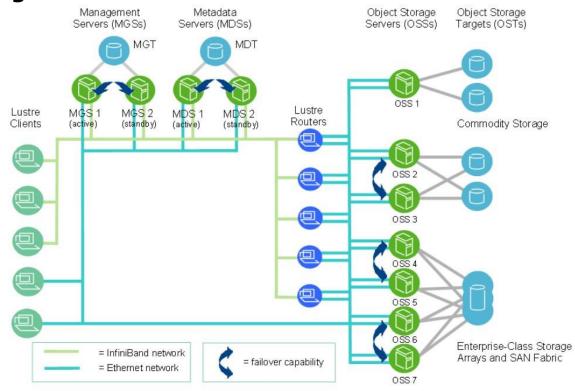
## 3. Parallel File Systems

## 3.1. Introduction

Parallel file systems provide high-performance I/O when multiple clients (a "client" usually meaning a compute node, in this instance) share the same file system. The ability to scale capacity and performance is an important characteristic of a parallel file system implementation. **Striping** is the basic mechanism used in parallel file systems for improving performance, where file data is split up and written across multiple I/O servers. Primarily, striping allows multiple servers, disks, network links to be leveraged during concurrent I/O operations, thus increasing available bandwidth. The most popular parallel file systems on HPC platforms are Lustre and IBM Spectrum Scale (formerly GPFS) [1] and form the focus of this chapter. A brief overview is given of BeeGFS, a file system gaining popularity, and object storage, a data storage method which is not based on files.

#### 3.2. Lustre

#### Figure 7. Lustre cluster at scale



Lustre is a Linux file system implemented entirely in the kernel and provides a POSIX standards-compliant file system interface. Its architecture is founded upon distributed object-based storage. This delegates block storage management to its back-end servers and eliminates significant scaling and performance issues associated with the consistent management of distributed block storage metadata.

- Metadata Servers (MDS): The MDS makes metadata stored in one or more Metadata Targets (MDTs) available to Lustre clients. Each MDS manages the names and directories in the Lustre file system(s) and provides network request handling for one or more local MDTs. Operations such as opening and closing a file can require dedicated access to the MDS and it can become a serial bottleneck in some circumstances.
- Metadata Targets (MDT): For Lustre software release 2.3 and earlier, each file system has one MDT; multiple MDTs are supported on later versions. The MDT stores metadata (such as filenames, directories, permissions and file layout) on storage attached to an MDS. An MDT on a shared storage target can be available to multiple MDSs, although only one can access it at a time. If an active MDS fails, a standby MDS can serve the MDT and make it available to clients. This is referred to as MDS failover.
- Object Storage Servers (OSS): The OSS provides file I/O service and network request handling for one or more local Object Storage Targets (OSTs). Typically, an OSS serves between two and eight OSTs, up to 16 TB each. A typical Lustre configuration consists of an MDT on a dedicated node, two or more OSTs on each OSS node, and a client on each of a large number of compute nodes.
- Object Storage Target (OST): User file data is stored in one or more objects, each object on a separate OST in a Lustre file system. Each OST can write data at around 500 MB/s. You can think of an OST as being equivalent to a disk, although in practice it may comprise multiple disks, e.g. in a RAID array. An individual file can be stored across multiple OSTs; this is called striping. The default is dependent on the particular system (1 is a common choice), although this can be changed by the user to optimize performance for a given workload.
- Lustre clients: Lustre clients are compute, visualisation or login nodes that are running Lustre client software, allowing them to mount the Lustre file system. Good performance is achieved when multiple clients (usually compute nodes for HPC calculations) simultaneously access the file system.

## 3.2.1. Lustre File Layout (Striping)

Lustre file systems have the ability to stripe data across multiple OSTs in a round-robin fashion. Users can optionally configure for each file the number of stripes, stripe size, and OSTs that are used. Although these parameters can be set on a per-file basis they are usually set on directory where your output files will be written so that all output files inherit the same settings. The *stripe\_size* indicates how much data to write to one OST before moving to the next OST. The *stripe\_count* indicates how many OSTs to use. The default values for *stripe\_count* and *stripe\_size* are system dependent but are often 1 and 1 MiB respectively.

You can use the Ifs getstripe command to see the layout of a specific file:

```
>lfs getstripe reads2.fastq
reads2.fastq
lmm_stripe_count: 1
lmm_stripe_size: 1048576
lmm_pattern: 1
lmm_layout_gen: 0
lmm_stripe_offset: 42
obdidx objid objid group 42
37138823 0x236b187
0
```

In this case, the file reads2.fastq has a single stripe of size 1 MiB (1048576 bytes). To change the layout of specific directory (or file) we can use the lfs setstripe command. (Note that this command will not repartition the data on existing files but will ensure that new files created within the directory use the updated settings.)

```
>lfs setstripe -c -1 results_dir/
```

In the example, we set the stripe count to -1 (maximal striping) to make the largest potential bandwidth available to new files in this directory.

## 3.2.2. Choosing a stripe count

The stripe count sets the number of OSTs (Object Storage Targets) that Lustre stripes the file across. In theory, the larger the number of stripes, the more parallel write performance is available. However, large stripe counts for small files can be detrimental to performance as there is an overhead in using more stripes.

## 3.2.3. Choosing a stripe\_size

The size of each stripe generally has less of an impact on performance than the stripe count but can become important as the size of the file being written increases. We outline some considerations below but bear in mind that the impact is linked to the I/O pattern used in the application (and that stripe count is usually a more important parameter anyway). Note:

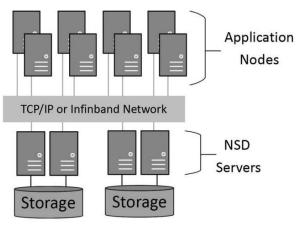
- The stripe size should be a multiple of the page size
- The smallest recommended stripe size is 512 KB
- A good stripe size for sequential I/O using high-speed networks is between 1 MB and 4 MB
- The maximum stripe size is 4 GB

Further information can be found on [32]

# 3.3. IBM Spectrum Scale (formerly GPFS)

#### Figure 8. NSD Server Model

Network Shared Disk (NSD) Server Model



Picture Source, High Performance Parallel I/O Book, Chapter 9. Editors Prabhat, Quincey Koziol. October 2014.

IBM Spectrum Scale is a cluster file system that provides concurrent access to a single file system or set of file systems from multiple nodes. The nodes can be SAN attached, network attached, a mixture of SAN attached and network attached, or in a shared nothing cluster configuration. This enables high performance access to this common set of data to support a scale-out solution or to provide a high availability platform.

#### 3.3.1. Block size

To display the amount of available disk space for each filesystem:

```
di98het@login05:~> df -Th
Filesystem Type Size Used Avail
Use% Mounted on
/dev/fs1 gpfs 12P 8.8P 3.0P 75% /gpfs
/dev/fs2 gpfs
5.2P 3.8P 1.4P 73% /gss/scratch
```

The I/O operation size should be a multiple of the block size. To display the properties of a GPFS, this command can be used:

```
di98het@login05:~> mmlsfs fs1 flag value
description ------
-f 262144
   Minimum fragment size in bytes
...
-B 8388608 Block size
...
```

Further information can be found on [31]

## 3.4. BeeGFS (formerly FhGFS)

BeeGFS, also known as FhGFS prior to 2014, is a file system which has gained popularity in recent years in the European HPC community. Attributed in part to its relative ease of deployment, it being free and open source (in contrast to proprietary solutions like IBM Spectrum Scale), and its support of a wide range of Linux kernel versions.

The beegfs-ctl utility enables clients to affect BeeGFS parameters. To query the existing stripe pattern for a file, beegfs-ctl --getentryinfo is used[34]:

```
> beegfs-ctl --getentryinfo /mnt/beegfs/testdir/test.txt
```

Path: /testdir/test Mount: /mnt/beegfs

EntryID: 0-5429B29A-AFA6
Metadata node: meta01 [ID: 1]

Stripe pattern details:

+ Type: RAID0 + Chunksize: 512K

+ Number of storage targets: desired: 4; actual: 4

+ Storage targets:

+ 102 @ storage01 [ID: 1] + 101 @ storage01 [ID: 1] + 201 @ storage02 [ID: 2] + 202 @ storage02 [ID: 2]

Indicating the test.txt file has four stripes, each with a size of 512KB. Similarly, setting the stripe pattern can be accomplished with beegfs-ctl -- setpattern as follows:

> beegfs-ctl --setpattern --chunksize=1m --numtargets=4 /data/test

which sets the stripe count to four, each with a size of 1MB, for all files created under the /data/test directory. Note that default installations of BeeGFS restrict the --setpattern mode to superuser/root access only. Non-root users may still set striping configurations but only if the site has explicitly allowed it by enabling the sysAllowUserSetPattern setting on the MDS[5].

More information on BeeGFS stripe settings and performance considerations is at [6]

## 3.5. Object Storage

A relatively new data storage technique being driven by advances in cloud and internet-based services is object storage. This eschews the directory hierarchy of a traditional file system in favour of a collection of data "objects", each containing their own unique identifier. A comparison can be drawn between object storage and key-value pairs, such as those implemented in Java Maps or Python dictionaries.

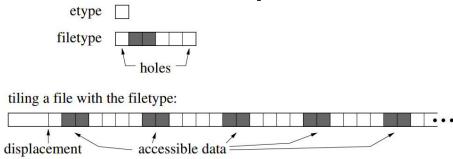
## **4. MPI-IO**

#### 4.1. Introduction

POSIX is a standard that maintains compatibility with the broadest base of applications while enabling high performance. Parallel file systems used across HPC provide parallel access while retaining POSIX semantics.

However, the portability and optimisation needed for parallel I/O cannot easily be achieved with the POSIX interface. In order to face this issue, the HPC community defined the MPI-IO interface. MPI-IO provides a high-level interface supporting partitioning of file data among processes and a collective interface supporting complete transfers of global data structures between process memories and files. An implementation of MPI-IO is typically layered on top of a parallel file system that supports the notion of a single and common file shared by multiple processes. (Both of the common parallel file systems currently in use: Lustre and IBM Spectrum Scale, support this concept.)

#### Figure 9. MPI-IO File concepts



Concept	Definition		
file	An ordered collection of typed data items		
etype	The unit of data access and positioning. It can be any MPI predefined or derived datatype		
filetype	The basis for partitioning a file among processes and defines a template for accessing the file. A filetype is either a single etype or a derived MPI datatype constructed from multiple instances of the same etype.		
view	Defines the current set of data visible and accessible from an open file as an ordered set of etypes. Each process has its own view of the file, defined by three quantities: a displacement, an etype, and a filetype. The pattern described by a filetype is repeated, beginning at the displacement, to define the view.		
Offset	It is a position in the file relative to the current view, expressed as a count of etypes. Holes in the view's filetype are skipped when calculating this position.		
Displacement	splacement It is an absolute byte position relative to the beginning of file. The displacement defines the location where a vibegins.		

file size and end of file	The size of an MPI file is measured in bytes from the beginning of the file.
file pointer	A file pointer is an implicit offset maintained by MPI. "Individual file pointers" are file pointers that are local to each process that opened the file. A "shared file pointer" is a file pointer that is shared by the group of processes that opened the file.
file handle	A file handle is an opaque object created by MPI_FILE_OPEN and freed by MPI_FILE_CLOSE.

## 4.1.1. MPI-IO data access operations

MPI-IO defines three orthogonal aspects to data access from processes to files: positioning, synchronism and coordination. Positioning can be an explicit offset or implicit through the file pointer. Synchronism provides three access modes blocking, nonblocking and split collective. Coordination allows to the MPI processes to perform noncollective or collective operations.

Figure 10. Data Access Operations

positioning	synchronism	coordination	
- A		noncollective	collective
explicit	blocking	MPI_FILE_READ_AT	MPI_FILE_READ_AT_ALL
offsets	8359	MPI_FILE_WRITE_AT	MPI_FILE_WRITE_AT_ALL
59.25	nonblocking &	MPI_FILE_IREAD_AT	MPI_FILE_READ_AT_ALL_BEGIN
	split collective		MPI_FILE_READ_AT_ALL_END
		MPI_FILE_IWRITE_AT	MPI_FILE_WRITE_AT_ALL_BEGIN
			MPI_FILE_WRITE_AT_ALL_END
individual	blocking	MPI_FILE_READ	MPI_FILE_READ_ALL
file pointers		MPI_FILE_WRITE	MPI_FILE_WRITE_ALL
	nonblocking &	MPI_FILE_IREAD	MPI_FILE_READ_ALL_BEGIN
	split collective		MPI_FILE_READ_ALL_END
		MPI_FILE_IWRITE	MPI_FILE_WRITE_ALL_BEGIN
			MPI_FILE_WRITE_ALL_END
shared	blocking	MPI_FILE_READ_SHARED	MPI_FILE_READ_ORDERED
file pointer	92	MPI_FILE_WRITE_SHARED	MPI_FILE_WRITE_ORDERED
	nonblocking &	MPI_FILE_IREAD_SHARED	MPI_FILE_READ_ORDERED_BEGIN
	split collective	Million Colors and American Colors	MPI_FILE_READ_ORDERED_END
	900	MPI_FILE_IWRITE_SHARED	MPI_FILE_WRITE_ORDERED_BEGIN
			MPI_FILE_WRITE_ORDERED_END

#### 4.1.2. General Hints

- Application developers should aim to use the highest level of abstraction possible to allow the MPI-IO library to implement I/O in the optimal way.
- Collective operations should be used as the non-collective operations limit
  the optimisation that the MPI-IO library can perform. Collective operations
  often perform orders of magnitude better than non-collective operations
  for large amounts of I/O. This is not always the case, but should be tested
  for all read and write commands where possible.

 Getting good performance out of MPI-IO also depends on configuring and using the parallel file system correctly, this is particularly important on Lustre file systems.

## 4.2. Manipulating Files in MPI

File manipulation in MPI-IO is similar to POSIX-IO:

- Open the file: MPI\_File\_open
- Write/Read to/from the file: MPI File write or MPI File read
- Close the file: MPI\_File\_close

#### 4.2.1. Opening a File

int MPI\_File\_open(MPI\_Comm comm, const char \*filename, int amode,
MPI Info info, MPI File \*fh)

**MPI\_File\_open** opens the file identified by the file name **filename** on all processes in the **comm** communicator group. It is a collective routine, all processes must provide the same value for **amode**, and all processes must provide filenames that reference the same file. A process can open a file independently of other processes by using the MPI\_COMM\_SELF communicator. The file handle returned, **fh**, can be subsequently used to access the file until the file is closed using **MPI\_File\_close(fh)**.

Initially, all processes view the file as a linear byte stream, and each process views data in its own native representation.

#### The supported **amode** are:

- MPI\_MODE\_RDONLY read only,
- MPI\_MODE\_RDWR reading and writing,
- MPI\_MODE\_WRONLY write only
- MPI MODE CREATE create the file if it does not exist,
- MPI\_MODE\_EXCL error if creating file that already exists,
- MPI MODE DELETE ON CLOSE delete file on close,
- MPI\_MODE\_UNIQUE\_OPEN file will not be concurrently opened elsewhere,
- MPI MODE SEQUENTIAL file will only be accessed sequentially,

• MPI MODE APPEND set initial position of all file pointers to end of file.

C users can use bit vector OR (|) and in Fortran addition "+" to combine these constants; or the bit vector **IOR** intrinsic in Fortran 90.

**info** is an opaque object with a handle of type **MPI\_Info** in C and Fortran with the mpi\_f08 module, and INTEGER in Fortran with the mpi module or the include file mpif.h. MPI\_INFO\_NULL can be used as a default.

Hints specified via **info** allow a user to provide information such as file access patterns and file system specifics to direct optimisation. Providing hints may enable an implementation to deliver increased I/O performance or minimise the use of system resources. **info** is a parameter in MPI\_FILE\_OPEN, MPI\_FILE\_SET\_VIEW,MPI\_FILE\_SET\_INFO.

To create a new **info** object MPI\_INFO\_CREATE(info,ierror) is used and MPI\_INFO\_SET(info, key, value, ierror) to add an entry to **info**.

#### 4.3. File View

In MPI-IO it is possible to define a view for MPI process using MPI\_File\_set\_view. A view is specified by a triplet (displacement, etype, and filetype). The default view is a linear byte stream, represented by the triple (0,

MPI BYTE, MPI BYTE).

MPI FILE SET VIEW (fh, disp, etype, filetype, datarep, info)

#### A file view:

- changes the process's view of the data in the file
- resets the individual file pointers and the shared file pointer to zero
- is a collective operation
- the values for datarep and the extents of etype in the file data representation must be identical on all processes in the group
- the datatypes passed in etype and filetype must be committed
- datarep argument is a string that specifies the format in which data is written to a file: "native", "internal", "external32", or user-defined
- "external32" is a data representation that is supposed to be portable across architectures. However not all MPI implementations support the

"external32" representation, so in general MPI-IO files are not portable between all combinations.

Below a simple example of a view is shown:

```
MPI Aint lb, extent;
MPI_Datatype etype, filetype, contig;
MPI Offset disp;
// Create the contig datatype composed by two
integers MPI Type contiguous(2, MPI INT, &contig);
lb = 0; extent = 6 * sizeof(int);
// Create the filetype with Ib as lower bound
// and extent from the contig datatype
MPI Type create resized(contig, lb, extent, &filetype);
MPI Type commit(&filetype);
// Define the displacement
disp = 5 * sizeof(int);
etype = MPI INT;
MPI_File_open(MPI_COMM_WORLD, "/pfs/datafile", \
MPI MODE CREATE | MPI MODE RDWR, MPI INFO NULL, &fh);
MPI_File_set_view(fh, disp, etype, filetype, "native",
MPI INFO NULL);
MPI File write(fh, buf, 1000, MPI INT, MPI STATUS IGNORE);
```

## 4.4. ROMIO optimisation

One of the most common MPI-IO implementation is ROMIO, which is used in the major MPI distributions such as MPICH, MVAPICH, PE IBM and Intel MPI.

Hints for Data Sieving:

- ind\_rd\_buffer\_size controls the size (in bytes) of the intermediate buffer used when performing data sieving during read operations.
- ind\_wr\_buffer\_size Controls the size (in bytes) of the intermediate buffer when performing data sieving during write operations.
- romio\_ds\_read Determines when ROMIO will choose to perform data sieving for read. Valid values are enable, disable, or automatic.

 romio\_ds\_write Determines when ROMIO will choose to perform data sieving for write. Valid values are enable, disable, or automatic.

Hints for Collective Buffering (Two-Phase I/O):

- cb\_buffer\_size Controls the size (in bytes) of the intermediate buffer used in two-phase collective I/O.
- cb\_nodes Controls the maximum number of aggregators to be used.
- romio\_cb\_read Controls when collective buffering is applied to collective read operations. Valid values are enable, disable, and automatic.
- romio\_cb\_write Controls when collective buffering is applied to collective write operations.
- romio\_no\_indep\_rw This hint controls when "deferred open" is used.
- cb\_config\_list Provides explicit control over aggregators (i.e., one process per node).

## 4.5. Setting Hints for MPI-IO

Hints for MPI-IO can be set by using:

• an "info" object, as in this example:

```
integer info, ierr call MPI_Info_create(info, ierror) call MPI_Info_set(info, 'romio_cb_read', 'disable', ierr) call MPI_Info_set(info, 'romio_cb_write', 'disable', ierr) ... call MPI_File_open(comm, filename, amode, info, fh, ierror)
```

 the ROMIO\_HINTS environment variable. Here the user must create a file with a list of hints to be set at execution time:

```
>cat $HOME/romio-
hints romio_cb_read
enable
romio cb write enable
```

Before running the application, the ROMIO\_HINTS variable is set as follows: export ROMIO\_HINTS= \$HOME/romio-hints

#### 4.6. MPI-IO General Considerations

MPI-IO has many features that can help users achieve high performance.
 The different MPI-IO routines provide flexibility as well as portability.

- The most important features are the ability to specify non-contiguous accesses, the collective I/O functions, and the ability to pass hints to the MPI implementation.
- When accesses are non-contiguous, users must create derived datatypes, define file views, and should use the collective I/O functions.
- Use of MPI I/O is often limited to parallel file systems; do not expect to obtain good performance using it with NFS.

## 5. File Per Process

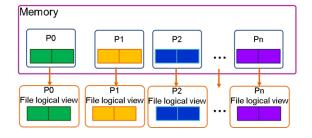
#### 5.1. Introduction

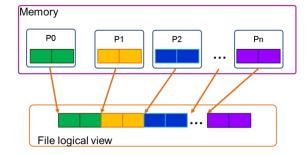
Often maligned in HPC circles, the file-per-process model can be a useful approach to parallel I/O for many applications if they can live within the constraints it introduces.

#### 5.2. File Per Process vs. Shared File

When deciding on a parallel I/O strategy, it is important to be informed of the pros and cons of available approaches in order to select one that best meets your requirements. File-per-process is often not given appropriate consideration, with a single shared file assumed to be superior in all situations. However, this is not the case. The following section details the relative advantages and drawbacks to the two opposite extremes patterns of file-per-process and a single shared file, and seeks to justify why the lesser considered strategy should not be discarded so readily.

## Figure 11. Recap of file per process (left) and single shared file (right)





The shared file strategy has the advantage in terms of disk storage footprint. Each file comes with its own overhead, such as header information, which is only recorded once for a single file but must be recorded for each file in the file-per-process model. However, in practice, users are rarely limited by disk space so this is seldom a concern.

To summarise, file-per-process **pros**:

- Easier to implement
- More straightforward to maintain
- Can achieve performance on par with single shared file and file-perprocess cons:

- Adds complexity to data analysis
- Metadata operations limit scalability
- Has a larger disk storage footprint

## 5.3. Optimizing File Per Process

Several of the drawbacks of file-per-process can be mitigated by the user. Use of data management approaches can alleviate much of the strain from metadata operations in the approach. Breaking up the files into a set of directories, for instance, can reduce the number of metadata queries made when copying, moving, reading or otherwise accessing the files.

File systems that support user tuning can be reconfigured to better support the file-per-process pattern. For example, a Lustre space is best configured to a single stripe per file on the corresponding directory, to minimise the overhead of communication with the OSTs:

lfs setstripe -c 1 files\_dir/

## 5.4. Sample Performance Data

Experimental data is provided by the National Supercomputing Service of the UK, ARCHER[13] [14]. The service is based around a Cray XC30 supercomputer with a Seagate Sonexion implementation of Lustre and results published primarily focus on its fs3 file system consisting of 48 available OSTs and 12 OSSes. The file-perprocess (FPP) and single-shared-file (SSF) schemes are directly compared in terms of the maximum achievable bandwidth of standard unformatted Fortran writes. Results are gathered from 1-512 fully-populated ARCHER nodes, i.e. 24-12288 CPU cores at 24 processes per node, and the plot in Figure 12, "ARCHER sample FPP vs. SSP results" produced.

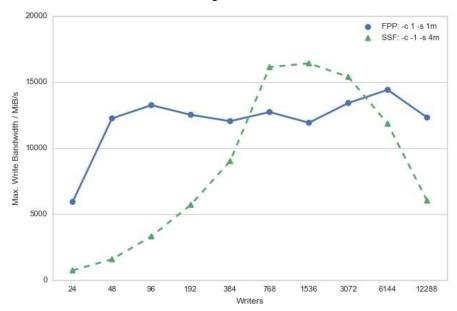


Figure 12. ARCHER sample FPP vs. SSP results

While SSF reaches a higher maximum bandwidth than FPP between 768-3072 processes, FPP is still comparable at these core counts and even noticeably outperforms the shared file approach at a lower number of cores, 24-192. Furthermore, FPP outperforms SSF at the maximum of 12288 cores where SSF performance drops off, suggesting

## 6. High-Level I/O Libraries

#### 6.1. Introduction

Scientific applications work with structured data and desire more self-describing file formats. A high-level I/O library is an API which helps to express scientific simulation data in a more natural way. Using high-level I/O libraries it is possible represent multi-dimensional data, labels and tags, noncontiguous data and typed data. These kind of libraries offers simplicity for visualization and analysis, portable formats can run on one machine and take output to another; and longevity where output will last and be accessible with library tools with no need to remember a specific version number of the code. Many HPC applications make use of higher-level I/O libraries such as the Hierarchical Data Format (HDF) and the Network Common Data Format (NetCDF). Parallel versions of high-level libraries are built on top of MPI-IO and they can use MPI-IO optimizations.

### 6.2. NetCDF

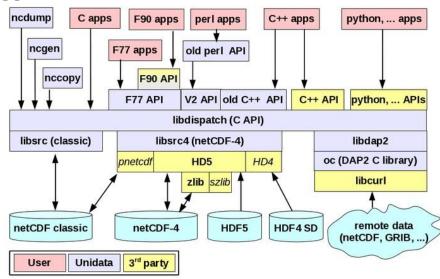
NetCDF (Network Common Data Form) is a set of software libraries and self-describing, machine-independent data formats that support the

creation, access, and sharing of array-oriented scientific data sets (vectors, matrices and higher dimensional arrays). The file format defined by netCDF allows scientists to exchange data and makes the I/O independent of any particular machine architecture. The portability of its files is the most important feature in netCDF. It is commonly used in climatology, meteorology and oceanography applications (e.g., weather forecasting, climate change) and GIS applications.

## **6.2.1.** Architecture of NetCDF APIs and Libraries

Figure 14, "Enhanced NetCDF Data Model" shows the layering architecture of netCDF C/C++/Fortran libraries and applications.

## Figure 13. Architecture of NetCDF APIs and Libraries

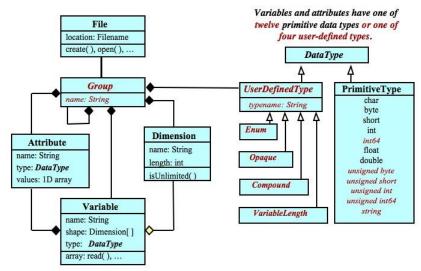


The software libraries provide read-write access to the netCDF format, encoding and decoding the necessary arrays and metadata. The core library is written in C, and provides an API for C, C++, Fortran 77 and Fortran 90. Additionally, a complete independent implementation is written in Java, which extends the core data model and adds additional functionality. Interfaces to netCDF based on the C library are also available in other languages including R, Perl, Python, Ruby, MATLAB, IDL and Octave.

#### 6.2.1.1. The netCDF data model

A netCDF dataset contains dimensions, variables, and attributes, which all have both a name and an ID number by which they are identified. These components can be used together to capture the meaning of data and relations among data fields in an array-oriented dataset.





A file has a top-level unnamed group. Each group may contain one or more named subgroups, user-defined types, variables, dimensions, and attributes. Variables also have attributes. Variables may share dimensions, indicating a common grid. One or more dimensions may be of unlimited length.

- Dimension: An entity that can either describe a physical dimension of a dataset, such as time, latitude, etc., as well as an index to sets of stations or model-runs.
- Variable: An entity that stores the bulk of the data. It represents an ndimensional array of values of the same type.
- Attribute: An entity to store data on the datasets contained in the file or the file itself. The latter are called global attributes.

#### **6.2.2.** Parallel I/O with netCDF

The netCDF library supports parallel I/O based on MPI-IO+pnetcdf or MPI-IO+HDF5. Parallel I/O support must be enabled at configure time when building these libraries:

- To build netCDF-4 with HDF5 parallel support: \$ CC=mpicc CPPFLAGS=-I\${H5DIR}/include LD-FLAGS=-L\${H5DIR}/lib ./configure --disable-shared --enable-parallel-tests --prefix=\${NCDIR} , where H5DIR must be a parallel HDF5 installation (--enable-parallel in HDF5 configure).
- To build netCDF-4 with PnetCDF parallel support: \$ CC=mpicc CPPFLAGS="-I\${H5DIR}/include
  - -I\${PNDIR}/include" LDFLAGS="-L\${H5DIR}/lib -L\${PNDIR}/lib" ./configure -enable-pnetcdf --enable-parallel-tests --prefix=\${NCDIR}

Parallel I/O support establishes a dependency on the MPI implementation. The pnetcdf library enables parallel I/ O operations on files in classic formats (CDF-1 and 2), and CDF-5 format since the release of NetCDF 4.4.0.

To support parallel I/O there are additional parameters and new functions:

- Fortran nf90\_create() and nf90\_open() have two additional optional arguments: an MPI communicator comm, and an MPI\_Info object info (may be MPI\_INFO\_NULL)
- For switching between collective and independent access: nf90\_var\_par\_access(ncid, varid, access), where access may be NF90\_INDEPENDENT or NF90\_COLLECTIVE. The default value is independent access. This applies for writes of variables while the file is open.

Figure 15, "netCDF Parallel Example: 2D array" gives a Fortran program for a 2D array with dimensions =  $5 \times 8$  where each process writes a subset of  $5 \times (dimsf(2)/mpi\_size)$ . The main parameters are: count = (5, 2) and start =  $(1, mpi\_rank*count(2)+1)$ 

#### Figure 15. netCDF Parallel Example: 2D array

```
Create the netCDF file:
call check(nf90_create(FILE_NAME, &
                                                     NC_PNETCD
IOR(NF90_NETCDF4, NF90_MPIIO), ncid, &
                                                     NC_NETCDF4,NC_MPIIO
comm = MPI_COMM_WORLD, info = MPI_INFO_NULL))
call check(nf90_{def_dim(ncid,"x", row_x,x_{dimid})) Define the dimensions
call check(nf90_def_dim(ncid, "y", col_y, &
                                                     The dimids array is used to
y_dimid))
                                                     pass the IDs of the dimen-
dimids = (/ y_dimid,x_dimid /)
                                                     sions of the variables
call check(nf90_def_var(ncid, "data", &
NF90_INT, dimids, varid))
                                                     Define the variable
call check(nf90_enddef(ncid))
                                                     End define mode.
                                                     defining metadata.
count = (/ row_x, col_y/mpi_size /)
                                                     cesses will write their meta-
start = (/1, (my_rank * count(2))+1/)
                                                     data to disk.
call check(nf90_put_var(ncid, varid, &
                                                     Write data to the file.
data_out, start = start, count=count))
                                                     Close the file
call check( nf90_close(ncid) )
```

#### 6.3. HDF5

HDF (Hierarchical Data Format) is an I/O library that serves the same purposes as NetCDF and more. As NetCDF, HDF Version 4 (HDF4) data models include annotated multidimensional arrays (called also scientific data sets), as well as raster files and lists of records. HDF4 does not support parallel I/O and files are limited to 2GB. To address these limitations, a new

version was designed: HDF Version 5 (HDF5). HDF5 has no file size limitation and is able to manage files as big as the largest allowed by the operating system. Unlike classic NetCDF, HDF5 supports more than one unlimited dimension in a data type. HDF5 provides support for C, Fortran, Java and C ++ programming languages.

.

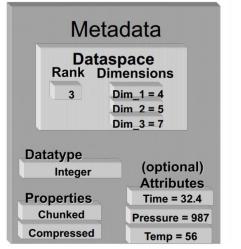
#### 6.3.1. HDF5 Design

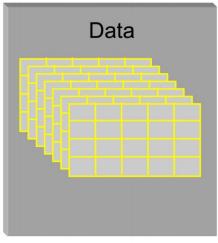
HDF5 is designed at three levels:

- A data model: consists of abstract classes, such as files, datasets, groups, datatypes and dataspaces, that developers use to construct a model of their higher-level concepts.
- A software library: to provide applications with an object-oriented programming interface that is powerful, flexible and high performance.
- A file format: provides portable, backward and forward compatible, and extensible instantiation of the HDF5 data model.

## 6.3.2. HDF5 File Organization and Data Model

Figure 16. HDF5 Dataset Model





Source: http://press3.mcs.anl.gov/computingschool/files/2014/01/QKHDF5-Intro-v2.pdf

HDF5 files are organized in a hierarchical structure, with two primary structures: groups and datasets.

• HDF5 group: a grouping structure containing instances of zero or more groups or datasets, together with supporting metadata.

 HDF5 dataset: a multidimensional array of data elements, together with supporting metadata.

As can be seen in Figure 16, "HDF5 Dataset Model", a dataset relies on two parts: the data itself and the metadata. The metadata covers the datatype, optional attributes, data properties and the dataspace which represents the data dimensionality and the data layout. The file dataspace can be handled completely independently from the data layout within the memory.

#### 6.3.3. Selecting a Portion of a Dataspace

HDF5 allows reading or writing to a portion of a dataset by use of hyperslab selection. A hyperslab selection can be a logically contiguous collection of points in a dataspace, or it can be a regular pattern of points or blocks in a dataspace. A hyperslab can be selected with the function: H5Sselect\_hyperslab/ h5sselect\_hyperslab\_f.

Hyperslabs are described by four parameters:

start: (or offset): starting location

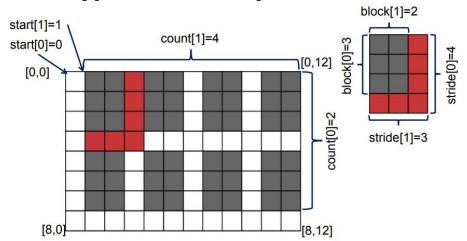
stride: separation blocks to be selected

count: number of blocks to be selected

block: size of block to be selected from dataspace

The dimensions of these four parameters correspond to dimensions of the underlying dataspace. Figure 17, "Hyperslab example" shows a hyperslab example and values of four parameters to select a portion of a dataset.

Figure 17. Hyperslab example



Creating a Hyperslab:

- In C: herr\_t H5Sselect\_hyperslab(hid\_t space\_id, H5S\_seloper\_t op, const hsize\_t \*start, const hsize\_t \*stride, const hsize\_t \*count, const hsize\_t \*block)
- In Fortran: H5SSELECT\_HYPERSLAB\_F(SPACE\_ID, OPERATOR, START, COUNT, HDFERR, STRIDE, BLOCK) INTEGER(HID\_T), INTENT(IN) :: SPACE\_ID INTEGER, INTENT(IN) :: OP INTEGER(HSIZE\_T), DIMENSION(\*), INTENT(IN) :: START, COUNT INTEGER, INTENT(OUT) :: HDFERR INTEGER(HSIZE\_T), DIMENSION(\*), OPTIONAL, INTENT(IN) :: STRIDE, BLOCK

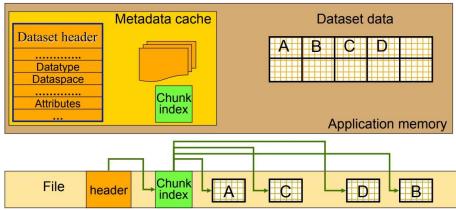
The following operators are used to combine old and new selections:

- H5S\_SELECT\_SET[\_F]: Replaces the existing selection with the parameters from this call. Overlapping blocks are not supported with this operator.
- H5S\_SELECT\_OR[\_F]: Adds the new selection to the existing selection.
- H5S\_SELECT\_AND[\_F]: Retains only the overlapping portions of the new selection and the existing selection.
- H5S\_SELECT\_XOR[\_F]: Retains only the elements that are members of the new selection or the existing selection, excluding elements that are members of both selections.
- H5S\_SELECT\_NOTB[\_F]: Retains only elements of the existing selection that are not in the new selection.
- H5S\_SELECT\_NOTA[\_F]: Retains only elements of the new selection that are not in the existing selection.

### 6.3.4. Chunking in HDF5

Datasets in HDF5 not only provide a convenient, structured, and self-describing way to store data, but are also designed to do so with good performance. In order to maximise performance, the HDF5 library provides ways to specify how the data is stored on disk, how it is accessed, and how it should be held in memory. The way in which the multidimensional dataset is mapped to the serial file is called the layout. The most obvious way to accomplish this is to simply flatten the dataset in a way similar to how arrays are stored in memory, serialising the entire dataset into a monolithic block on disk, which maps directly to a memory buffer the size of the dataset. This is called a contiguous layout.

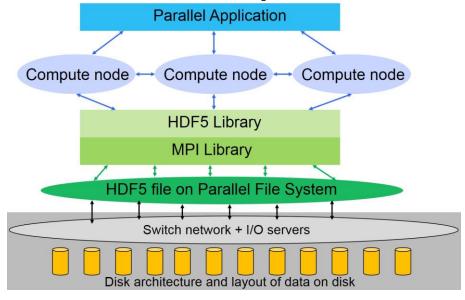
Figure 18. Chunking Scheme



#### 6.3.5. Parallel HDF5

The parallel HDF5 (PHDF5) library is implemented upon the MPI-IO layer, meaning users can directly benefit from MPI-IO optimisation techniques such collective buffering and data sieving. Figure 19, "Parallel HDF5 Layers" shows the different layers to implement parallel HDF5.

Figure 19. Parallel HDF5 Layers



PHDF5 has the following programming restrictions:

- PHDF5 opens a parallel file with a MPI communicator,
- · Returns a file ID and future access to the file is done via the file ID,
- · All processes must participate in collective PHDF5 APIs and
- Different files can be opened via different communicators.

Considerations for collective calls [26]:

- 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 • 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.

PHDF5 presents the following characteristics:

- After a file is opened by all the processes of a communicator:
- All parts of the 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 (i.e. collective I/O).
- Each process may write to individual data array (i.e. independent I/O).
- API languages: C and F90, 2003 language interfaces.
- Programming model: HDF5 uses an access property list to control the file access mechanism. Therefore, the general model to access HDF5 file in parallel must follow these steps: 1) set up MPI-IO file access property list, 2) open file, 3) access data and 4) close file.

#### 6.3.6. Parallel HDF5 Example

Figure 20, "PHDF5 Example: 2D dataset" corresponds to an 5 x 8 array dataset dimsf = (/5,8/), where each process will write a subset of the data 5 x (dimsf(2) / mpi\_size). In this example, the main parameters for the hyperslab are set as follows: offset =  $(/0, mpi_rank * count(2) /)$ ; count =  $(/5, dimsf(2)/mpi_size /)$ ; stride = (/1, 1 /)

#### Figure 20. PHDF5 Example: 2D dataset

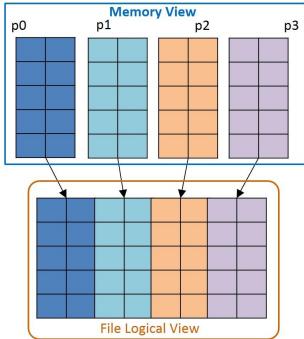


Figure 21, "Fortran program: file, dataspace, dataset and hyperslab" is the code for the PHDF5 example that shows the file open, dataspace and dataset definition and hyperslab selection.

## Figure 21. Fortran program: file, dataspace, dataset and hyperslab

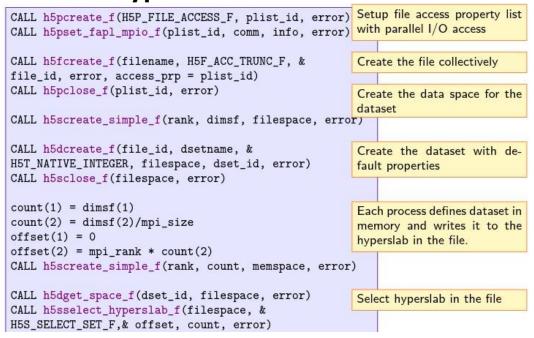
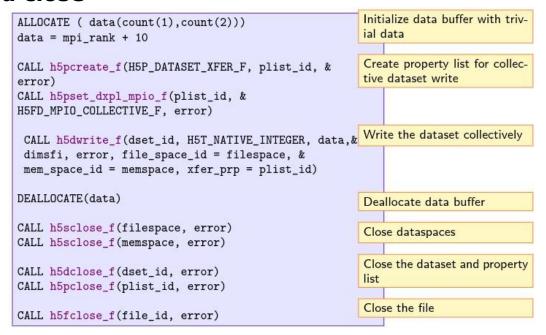


Figure 22, "Fortran program: property list, write and close" continues from the previous figure, showing collective operations, the writing of the data and finally the close step for all HDF5 elements.

## Figure 22. Fortran program: property list, write and close



## 6.4. pNetCDF

Parallel netCDF (officially abbreviated PnetCDF) is a library for parallel I/O providing higher-level data structures (e.g. multi-dimensional arrays of typed data). PnetCDF creates, writes, and reads the same file format as the serial netCDF library, meaning PnetCDF can operate on existing datasets, and existing serial analysis tools can process PnetCDF-generated files. PnetCDF is built on top of MPI-IO, which guarantees portability across various platforms and high performance. Originally PnetCDF was created as a first approach to provide parallel netCDF capabilities on top of netCDF-3. However with netCDF-4 a separate approach was implemented on top of HDF5. That's why two approaches are available today. However only PnetCDF is able to read netCDF-3 files in parallel (netCDF-4 uses PnetCDF underneath if available).

#### 6.5. SIONLIB

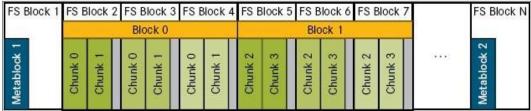
SIONlib is a scalable I/O library for parallel access to task-local files. The library allows mixing the good performance of a task local approach together with a small amount of created files. Data stored within a SIONlib fileset is a direct byte stream, no additional metadata information is stored by SIONlib. Due to the close connection to task-local files a separate post-processing step might be necessary to create a global data view of multiple local data views. The library provides different interfaces: parallel access using MPI, OpenMP, or their combination and sequential access for post-processing utilities.

#### 6.5.1. SIONLib file format

One of the strategies for SIONlib to increase I/O performance is preventing file system block contention, i.e. different tasks trying to modify the same file system block at the same time. To avoid this, SIONlib needs additional information from the user when opening a file. The chunksize supplied during the open call is communicated globally and lets SIONlib calculate the ranges inside the file which belongs to each task. In case where there is not enough space left for a write request in the current block, SIONlib skips all the file ranges that belong to other tasks and has new chunksize bytes of space available[27].

Figure 23, "SIONLib file layout"[27] shows file format and the different concepts used to its implementation.

## Figure 23. SIONLib file layout



#### 6.5.2. SIONLib API and utilities

The general API includes:

- Parallel interface: sion\_paropen\_mpi, sion\_parclose\_mpi.
- Serial interface: sion\_open, sion\_open\_rank, sion\_close, sion\_get\_locations.
- Common interface: sion\_ensure\_free\_space, sion\_feof, sion\_bytes\_avail\_in\_block, sion\_seek, sion\_seek\_fp, sion\_fwrite, sion\_fread.

Detailed API description can be found in [28].

SIONLib also provides utilities to file manipulation[29]:

- siondump: Dump meta data information of a SIONlib file.
- sionsplit: Split one SIONlib file into separate files.
- siondefrag: De-fragment a SIONlib file.
- sioncat: Extract all data or data of one tasks.

- sionconfig: Print compiler and linker flags.
- · partest: Benchmark using SIONlib.

# 6.6. Touching on Low-level I/O: POSIX, C and Fortran File Manipulation

The POSIX OS standard, C standard library and Fortran provide different methods for manipulating files in parallel applications. Although they are broadly discouraged for HPC applications due to the availability of the higher-level libraries detailed in this chapter, they do see use regardless. In this section, a brief overview of POSIX, C and Fortran file operations is given, as well as optimisation techniques that should be considered when I/O is performed with these interfaces. Note that threading or other concurrency models for low-level I/O are not covered in this section.

### 6.6.1. POSIX and the C Standard Library

In the C library, a stream is represented by a FILE\* pointer. File closing is done with the fclose function. Functions such as fputc, fputs, and fwrite can be used to write data to the stream, while fscanf, fgetc, fgets, and fread read data. For lower-level I/O operations in POSIX operating systems such as Linux, a handle called a file descriptor is used instead of a FILE\* pointer. A file descriptor is an integer value that refers to a particular instance of an open file in a single process. The function open creates a file descriptor and subsequent read, write and close functions take this as an argument.

#### Opening a file:

- FILE \*fopen(const char \*path, const char \*mode);
- FILE \*fdopen(int fildes, const char \*mode);
- int open(const char \*pathname, int flags [, mode t mode]);

The function fdopen() converts an already open file descriptor (fildes) to a stream. The flags arguments to open refers to the access mode, for example:

- O\_DIRECTORY, O\_NOFOLLOW (enforce directory / not-a-link)
- O\_SYNC, O\_ASYNC (synchronous vs. asynchronous)
- O\_LARGEFILE, O\_DIRECT (large files / bypass cache)
- O\_APPEND, O\_TRUNCO

The mode argument for open sets file access permissions and for fopen refers to read/write/read-write and so on.

It is recommended to perform unformatted fwrite()/fread() calls rather than formatted I/O fprintf()/ fscanf(). For repositioning within the file use fseek().

## 6.6.1.1. **O\_DIRECT** flag

Using the O\_DIRECT flag when opening a file will bypass the buffer cache and send data directly to the storage system. This can be useful in some special cases to avoid memory copies and improve multi-client consistency/parallel access. When performing direct I/O, the request length, buffer alignment, and file offsets generally must all be integer multiples of the underlying device's sector size. There is no block size requirement for O\_DIRECT on NFS, although this is not the case for local filesystems and IBM Spectrum Scale.

For example, Spectrum Scale may provide some performance benefits with direct I/O if:

- The file is accessed at random locations.
- · There is no access locality.

In which case, direct transfer between the user buffer and the disk can only happen if all of the following conditions are also true:

- The number of bytes transferred is a multiple of 512 bytes.
- The file offset is a multiple of 512 bytes.
- The user memory buffer address is aligned on a 512-byte boundary.

#### 6.6.1.2. **Buffering**

C-standard I/O implements three types of user buffering, and provides developers with an interface for controlling the type and size of the buffer. There are three types:

- Unbuffered: No user buffering is performed. Data is submitted directly to the kernel.
- Line-buffered: Buffering is performed on a per-line basis (default for e.g., stdout). With each newline character, the buffer is submitted to the kernel.

• Block-buffered: Buffering is performed on a per-block basis. This is ideal for files. By default, all streams associated with files are block-buffered. Standard I/O uses the term full buffering for block buffering.

Additionally, C-standard I/O provides an interface for controlling the type of buffering:

```
int setvbuf(FILE *stream, char *buf, int mode, size_t size);
```

This must be done after opening a file but before any I/O operations. The buffering mode is controlled by macros: \_IONBF (Unbuffered), \_IOLBF (Line-Buffered) or \_IOFBF (Block-buffered). The default buffer size for block buffering is BUFSIZ, defined in stdio.h. Function fflush() can be used to force block out early.

Example code snippet demonstrating changing the buffering type:

```
#define BUF 10000000 double
data[SIZE]; char* myvbuf; FILE* fp;
fp=fopen(FILENAME, "w"); myvbuf =
(char *) malloc(BUF) setvbuf(fp,
myvbuf,_IOFBF,BUF); fseek(fp, 0,
SEEK_SET); // start of file
fwrite(data, sizeof(double),SIZE, fp);
// close file, then deallocate buffer
```

#### **6.6.1.3. POSIX Hints**

The POSIX standard allows certain hints to be given to the OS. This specifies the future intentions for how a file will be manipulated to influence buffer cache behaviour and read-ahead. The actual effect of these hints is implementation specific - even different versions of the Linux kernel may react dissimilarly.

```
#include <fcntl.h> int posix_fadvise (int fd, off_t offset, off_t
len, int advice);
```

A call to posix\_fadvise() provides the OS with the hints for file descriptor fd in the interval

[offset,offset+len). If len is 0, the advice will apply to the range [offset,length of file]. Common usage is to specify 0 for len and offset, applying the advice to the entire file.

#### 6.6.2. Fortran Files

In fortran, a file is associated with a logical device which is in turn associated with a file by a unit specifier (UNIT=). A unit is connected or linked to a file through the OPEN statement in standard Fortran. Filenames are given with the FILE= specifier in the OPEN statement:

```
OPEN(UNIT=11, FILE="filename", options)
```

Options and considerations for making Fortran I/O more efficient:

- Specify the operation you intend with the ACTION keyword: read, write or both for ACTION='READ' / 'WRITE' / 'READWRITE'
- Perform direct access (ACCESS='DIRECT') with a large maximum record length (RECL=rl). If possible, making rl a multiple of the disk block size. Note that data in direct-access files can be read or written to in any order and that records are numbered sequentially, starting with record number 1. The units used for specifying record length depend on the form of the data:
- Formatted files (FORM= 'FORMATTED'): Specify the record length in bytes.
- Unformatted files (FORM= 'UNFORMATTED'): Specify the record length in 4-byte units.

#### 6.6.2.1. I/O Formatting

Formatted I/O can be the list-directed (i.e. using the Fortran default formats):

```
write(unit,fmt=*)
```

or fmt can be given a format string representing a statically compiled or dynamically generated format:

```
write(unit,fmt='(es20.13)')
write(unit,fmt=iof)
For unformatted I/O, it is possible to use sequential:
write(unit)
or direct access
write(unit, rec=i)
```

As stated, unformatted I/O is recommended over formatted I/O due to the performance benefits.

#### 6.6.2.2. Conversion of Fortran Unformatted Files

Although formatted data files are more easily ported to other systems, Intel Fortran can convert unformatted data in several formats.

By default, the Intel compiler uses little-endian format but it is possible to write unformatted sequential files in big-endian format, as well as read files produced in big-endian format by using the little-endian-to-big-endian conversion feature:

F\_UFMTENDIAN=MODE | [MODE;] EXCEPTION

with the following options:

MODE = big | little EXCEPTION =

big:ULIST | little:ULIST | ULIST

ULIST = U | ULIST,U U =

decimal | decimal -decimal

Examples:

F\_UFMTENDIAN=big; File format is big-endian for all units.

F\_UFMTENDIAN=big:9,12 ;big-endian for units 9 and 12, little-endian for others.

F\_UFMTENDIAN="big;little:8"; big-endian for all except unit 8.

#### 6.6.2.3. I/O Pattern Issues

To eliminate unnecessary overhead, write whole arrays or strings at once rather than individual elements in different operations. Each item in an I/O list generates its own calling sequence. This processing overhead becomes most significant in an implicit loop. When accessing whole arrays, use the array name (Fortran array syntax) instead of using an implicit loop. The following snippets give the code for each of these I/O approaches. Only the last option is recommended with the others given as examples of what to avoid. Implicit loop

```
write(...) ((a(i,j),i=1,m),j=1,n)
```

Array section

write(...) a(1:m,1:n)

#### Complete Array

write(...) a

Use the natural ascending storage order whenever possible. This is column-major order, with the leftmost subscript varying fastest and striding by 1. If the whole array is not being written, natural storage order is the best order possible.

# 6.6.2.4. Tuning Fortran Using the Intel Compiler **Buffering**

In an Intel-compiled Fortran application, any records, read or written, are transferred between the user's program buffers and one or more disk block I/O buffers. These buffers are established when the file is opened by the Intel Fortran Run-Time Library. Unless very large records are being read or written, multiple logical records can reside in the disk block I/O buffer when it is written to disk or read from disk, minimising physical disk I/O.

The OPEN statement BUFFERCOUNT keyword specifies the number of I/O buffers. The default for BUFFERCOUNT is 1. Any experiments to improve I/O performance should increase the BUFFERCOUNT value and not the BLOCKSIZE value, to increase the amount of data read by each disk I/O.

If the OPEN statement has BLOCKSIZE and BUFFERCOUNT specifiers, then the internal buffer size in bytes is the product of these specifiers. If the open statement does not have these specifiers, then the default internal buffer size is 8192 bytes.

- BLOCKSIZE=<bytes> specifier (rounded up to multiples of 512).
- BUFFERCOUNT=<count> specifier (default 1, at most 127 possible).

To enable buffered writes; that is, to allow the disk device to fill the internal buffer before the buffer is written to disk, use one of the following:

- The OPEN statement BUFFERED specifier.
- The FORT\_BUFFERED run-time environment variable.

#### Intercepting the Run-time's libc calls (Linux)

The Intel Fortran Run-time Library allows uses of the USEROPEN specifier in an OPEN statement to pass control to a routine that directly opens a file. The called routine can use system calls or library routines to open the file and establish special context that changes the effect of subsequent Intel Fortran I/O statements.

The USEROPEN specifier takes the following form:

USEROPEN = function-name

function-name is the name of an external function; it must be of type INTEGER(4) or (INTEGER\*4).

The external function can be written in Fortran, C, or other languages.

For example, the following Intel Fortran code might be used to call the USEROPEN procedure UOPEN (associated with linker uppen ):

# 6.7. I/O Libraries Summary

- The netCDF classic data model is simple and flat composed by Dimensions, Variables and Attributes. The netCDF enhanced data model adds primitive types, multiple unlimited dimensions, hierarchical groups and user-defined data types.
- The HDF5 data model has even more features such as non-hierarchical groups, user-defined primitive data types, References (pointers to objects and data regions in a file) and Attributes attached to user-defined types.
- SIONLib optimises binary one-file-per-processes approach by usage of a shared container file using a custom file format, as is more convenient where portability is not the main priority. Furthermore, SIONLib provides a transparent mechanism to avoid file system block contention.

- Specialised I/O libraries may provide more a portable way of writing data and may reduce metadata load when properly used.
- For parallel programs the output to separate files for each process can provide high throughput, but usually needs post-processing.
- Binary files may need to use library/compiler support for conversion. If binary files are transferred between different architectures (little vs.bigendian byte order) then the limitations may apply on file sizes and data types.

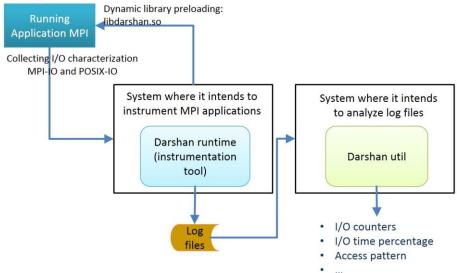
# 7. I/O Performance Analysis

#### 7.1. Introduction

Parallel I/O performance evaluation for HPC applications is a nontrivial task because it depends on the I/O software stack, variety of application I/O patterns, hardware configuration and heterogeneity of the I/O system infrastructure.

#### 7.2. Darshan Tool

#### Figure 24. Darshan Overview



Darshan is a lightweight, scalable I/O characterisation tool that transparently captures I/O access pattern information from production applications. It was developed by the Argonne Leadership Computing Facility (ANL). Darshan provides I/O profile for C and Fortran calls including: POSIX and MPI-IO (and limited to HDF5 and PnetCDF). Darshan does not provide information about the I/O activity along the runtime. It uses a LD\_PRELOAD mechanism to wrap the I/O calls.

#### 7.3. Darshan Runtime

Starting with version 3.x, the Darshan runtime environment and log file format have been redesigned such that new "instrumentation modules" can be added without breaking existing tools. Darshan can then manage these modules at runtime and create a valid Darshan log regardless of how many or what types of modules are used [19]. Figure 25, "Darshan Runtime Environment" depicts the components of Darshan Runtime.

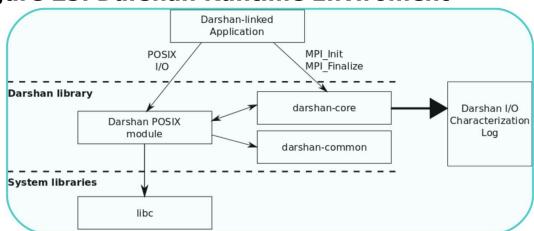


Figure 25. Darshan Runtime Enviroment

### 7.3.1. Using the Darshan Runtime

Darshan can trace MPI applications linked statically or dynamically. The instrumentation method to use depends on whether the executables produced by the MPI compiler are statically or dynamically linked (for more detail refer to [16]).

Currently, Darshan is a tool commonly provided in HPC centers through a module system. Site-specific documentation for facilities that deploy Darshan in production can be found in [17]. Although, each center provides some specific steps to enable Darshan, there are some commonalities and relevant commands are often similar. For example, at Leibniz Supercomputing Centre (LRZ), Darshan [18] is available on all its HPC systems and is enabled by the following commands in the user submission:

- 1. module load darshan: prepare environment for both dynamically and statically linked applications. Other HPC systems often provide two modules for Darshan, e.g. module load darshan-runtime for I/O profiling and module load darshan-util for analyzing Darshan logs.
- 2. export LD\_PRELOAD=`darshan-user.sh \$FORTRAN\_PROG`: load the appropriate library depending on the programming lenguage of the parallel application. For dynamically-linked executables, Darshan relies on the

LD\_PRELOAD environment variable to insert instrumentation at run time. For Fortran applications compiled with MPICH, users may have to take the additional step of adding libfmpich.so to the LD\_PRELOAD environment variable. At LRZ, this is checked by the \$FORTRAN\_PROG variable, so that the appropriate library is loaded.

- 3. export JOBID\_LL=`darshan-JOBID.sh \$LOADL\_STEP\_ID`; export DARSHAN\_JOBID=JOBID\_LL: Darshan gives the log file a name based on the job identifier assigned by the job management system, to facilitate correlating logs with a specific job. At LRZ, the DARSHAN\_JOBID environment variable is set with the Loadleveler identifier in SuperMUC. In a Linux Cluster with Slurm as the job management system DARSHAN\_JOBID is set to SLURM\_JOB\_ID.
- 4. export LOGPATH\_DARSHAN\_LRZ=`darshan-logpath.sh`: set up the folder to save the Darshan logs. This option is available if darshan-runtine was built with --with-log-path-by-env. Otherwise, the darshan-mk-log-dirs.pl utility is applied to specify the path at configure time to include log file subdirectories organised by year, month, and day. At LRZ by default log files are placed in \$SCRATCH/.darshan-log, but users can change the log path by setting LOGPATH DARSHAN LRZ to a more convenient folder.

Once the application executes, a log file is generated in log path, e.g. LOGPATH\_DARSHAN\_LRZ. If the execution finishes without errors, the user can analyze the \*.darshan file by using darshan-util tools. The log file can also be analysed on another system.

#### 7.4. Darshan Util

Darshan provides command line tools that enable the analysis of I/O performance metrics. Keys tools:

- darshan-job-summary.pl: creates a pdf with graphs useful for initial analysis.
- darshan-summary-per-file.sh: creates a separate pdf for each file opened by the application
- darshan-parser: dumps all information into ASCII (text) format.

>darshan-parser --help

Usage: darshan-parser [options] <filename>

--all : all sub-options are enabled

--base : darshan log field data [default]

--file: total file counts

--file-list : per-file summaries

--file-list-detailed : per-file summaries with additional detail

--perf: derived perf data

--total : aggregated darshan field data

#### 7.4.1. Darshan Plots: FLASH-IO Benchmark

FLASH-IO [36] is a block-structured adaptive mesh hydrodynamics code. The computational domain is divided into blocks which are distributed across the processors. Typically a block contains 8 zones in each coordinate direction (x,y,z) and a perimeter of guardcells (presently 4 zones deep) to hold information from the neighbors. FLASH-IO will produce a checkpoint file (containing all variables in 8-byte precision) and two plotfiles (4 variables, 4-byte precision, one containing corner data, the other containing cell-centered data). The plotfiles are smaller than the checkpoint file.

#### Figure 26. Job Information and Performance

jobid: 1689071	uid: 3366230	nprocs: 1024	runtime: 206 seconds
J. Commission of the Commissio	Treatment and transmission and treatment		1 miles 1 mile

I/O performance estimate (at the MPI-IO layer): transferred 584657.2 MiB at 3570.53 MiB/s

Figure 26, "Job Information and Performance" shows the ID, number of MPI processes and runtime of a FLASHIO job. The Darshan pdf file associated with this job will include executable name and date as a header, and the binary location and commands used as footer.

Figure 27. Average I/O Cost and Operation Count

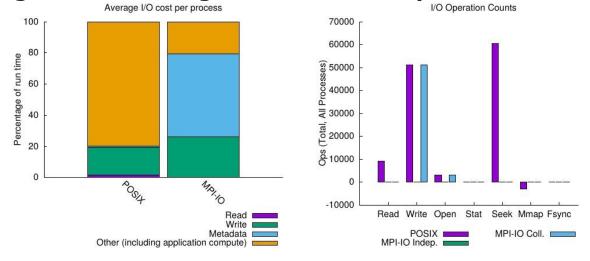


Figure 28. Access Sizes at POSIX and MPI-IO level

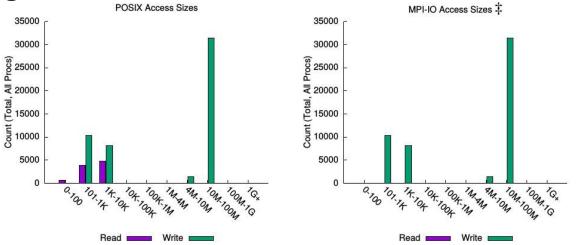


Figure 28, "Access Sizes at POSIX and MPI-IO level" shows a histogram of MPI-IO and POSIX reads/write and their associated sizes. This information is useful for identifying small I/O operations. Reasons for small I/O include poorly implemented I/O or, occasionally, incorrect optimisation attempts by an I/O library which changes operation sizes and negatively affects the I/O pattern.

Figure 29. Common Access Sizes and File Count

Most Common Access Sizes
(POSIX or MPI-IO)

	access size	count		
	20971520	8208		
POSIX	21495808	8184		
	21233664	8184		
	984	2639		
MPI-IO ‡	20971520	8208		
	21495808	8184		
	21233664	8184		
	10485760	1368		

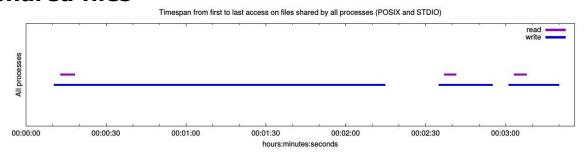
NOTE: MPI-IO accesses are given in terms of aggregate datatype size.

File Count Summary (estimated by POSIX I/O access offsets)

type	number of files	avg. size	max size
total opened	3	191G	487G
read-only files	0	0	0
write-only files	3	191G	487G
read/write files	0	0	0
created files	3	191G	487G

Figure 29, "Common Access Sizes and File Count" presents the most common sizes for MPI-IO and POSIX. This information helps confirm whether collective operations are being performed or whether this has been disabled by the I/O library. Usually, the MPI-IO implementation attempts to apply an appropriate optimisation technique based on the I/O pattern. Further information in this figure mainly relates to access mode and file size.

# Figure 30. Timespan from first to last access on shared files



Darshan is also capable of producing a timespan of I/O activity, for example Figure 30, "Timespan from first to last access on shared files". These plots are useful for observing the temporal I/O pattern. Information such as I/O sequentiality or overlapping can help identify possible inefficiency in the data access operations.

Figure 31. Average I/O and Data Transfer

Average I/O per process (POSIX and STDIO)

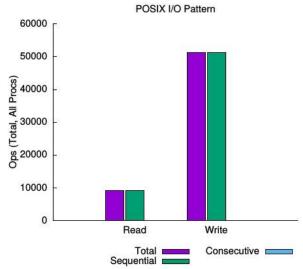
	Cumulative time spent in	Amount of I/O (MB)
	I/O functions (seconds)	
Independent reads	0	0
Independent writes	0	0
Independent metadata	0	N/A
Shared reads	3.32754123144531	0.013781301677227
Shared writes	36.5967756923828	570.940493404865
Shared metadata	1.63780214160156	N/A

Data Transfer Per Filesystem (POSIX and STDIO)

File System	Write		Read		
	MiB	Ratio	MiB	Ratio	
/gss/scratch	584643.06525	1.00000	14.11205	1.00000	

In Figure 31, "Average I/O and Data Transfer", the average I/O per process is given, categorised by I/O strategy (shared or independent files). This information is important for perceiving possible problems at a large scale. In this figure, it is possible to observe the impact of read and metadata operations on run time and I/O size. Read operations within the I/O library could be an I/O bottleneck for a larger number of processes, optimisation at the MPI-IO level could be applied to mitigate this impact.

Figure 32. I/O Pattern



*sequential*: An I/O op issued at an offset greater than where the previous I/O op ended. *consecutive*: An I/O op issued at the offset immediately following the end of the previous I/O op.

Figure 32, "I/O Pattern" shows the I/O pattern for each file opened by the application. These plots should be analysed per file, as the noise of other files is removed and the sequential or consecutive pattern is clearer and easier to identify. Usually, a consecutive pattern is higher performance than an sequential pattern. In this case, the three files present a sequential pattern.

Figure 33. Variance in shared files

variance in shared riles (POSIX and STDIO)									
File	Processes	Fastest		Slowest			σ		
Suffix	· ·	Rank	Time	Bytes	Rank	Time	Bytes	Time	Bytes
df5_chk_0000	1024	145	3.595872	487M	175	64.058370	487M	9.19	5.14e+06
plt_cnt_0000	1024	73	0.342752	41M	827	13.076260	42M	2.56	4.28e + 05
plt_crn_0000	1024	4	0.583471	45M	1006	11.148950	45M	1.77	4.7e + 05

Figure 33, "Variance in shared files" presents the variance for I/O time and bytes for shared files. This is useful for identifying possible mapping problems (I/O time) or I/O imbalance (bytes).

## 7.5. Vampir

The Vampir performance visualisation tool consists of a performance monitor (e.g., Score-P or VampirTrace) that records performance data and a performance GUI, which is responsible for the graphical representation of the data. Figure 34, "Vampir Architecture" depicts the components of the Vampir Tool.

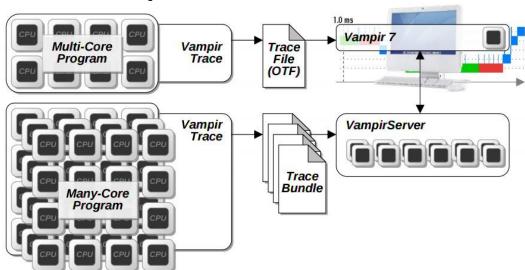


Figure 34. Vampir Architecture

To trace I/O events, VampirTrace is recommended as, if this is built with I/O tracing support, it intercepts calls to I/O functions of the standard C library. Therefore, it is possible to capture the serial and parallel I/O done by a parallel application.

The following functions are intercepted by VampirTrace:

close creat creat64 dup dup2 fclose fcntl fdopen fgetc fgets flockfile fopen fopen64 fprintf fputc fputs fread fscanf fseek fseeko fseeko64 fsetpos fsetpos64 ftrylockfile funlockfile fwrite getc gets lockf lseek lseek64 open open64 pread pread64 putc puts pwrite pwrite64 read readv rewind unlink write writev

Tracing I/O events has to be activated for each tracing run by setting the environment variable VT\_IOTRACE to "yes". Setting the environment variable VT\_IOTRACE\_EXTENDED to "yes" enables the collection of additional function arguments for some of the I/O function mentioned above. For example, this option additionally stores offsets for pwrite and pread to the I/O event record. Enabling VT\_IOTRACE\_EXTENDED automatically enables VT\_IOTRACE[35].

## 7.5.1. Using Vampir

Users can instrument a parallel application by using VampirTrace:

- compile-time instrumentation:
- Fortran 77: vtf77 [-g] -c <further options> myprog.f
- Fortran 90 and higher: vtf90 [-g] -vt:f90 mpif90 -c <further options> myprog.f90
- C: vtcc [-g] -vt:cc mpicc -c <further options> myprog.c
- C++: vtcxx [-g] -c -vt:cxx mpiCC <further options> myprog.cpp
- run-time instrumentation (vtrun): mpirun -np 16 vtrun ./a.out

The following environment variables must set before running the application with VampirTrace:

export

VT\_PFORM\_LDIR=/gpfs/scratch/project/user/vampir-tmp export VT\_FILE\_UNIQUE='yes' export VT\_IOTRACE='yes'

After execution a .otf file as well as a number of \*.events.z files are generated. The Vampir tool is then used for visualisation and trace analysis:

- For small traces: vampir <filename>.otf
- For large traces: vampirserver start -n <tasks>

#### 7.5.1.1. Vampir Examples: FLASH-IO and BT-IO

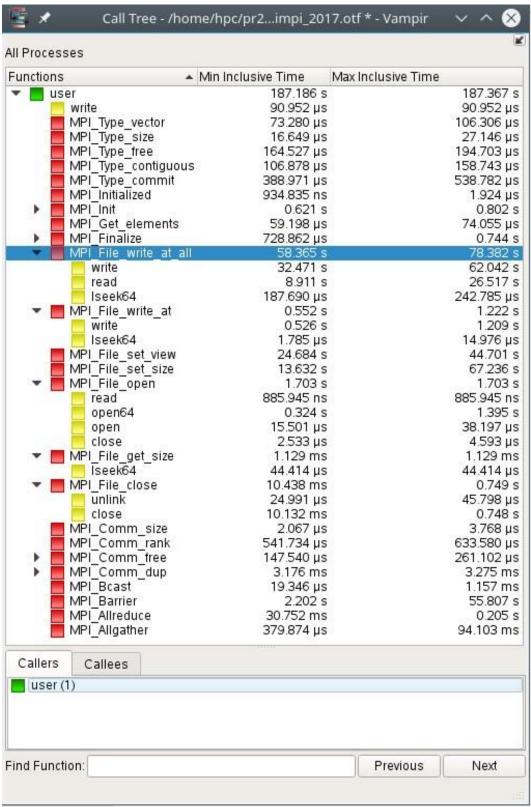
To demonstrate the I/O profiling information produced by Vampir, two benchmarks are detailed here. One (FLASH-IO) was previously described in the Darshan section above.



Figure 35. FLASH-IO Processes Timeline

Figure 35, "FLASH-IO Processes Timeline" depicts the execution of FLASH-IO using Intel MPI over 32 processes and HDF5 for I/O. The dark green triangles represent I/O events, yellow bars give the time of I/O operations, green bars give time of user operations, and the red bars show the time for MPI events. Here, all process are doing I/O, meaning the collective buffering optimisation technique is not being performed.

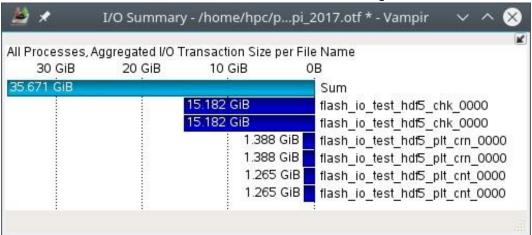
Figure 36. FLASH-IO Call Tree



Vampir provides several views for performance analysis. A call tree can be produced showing the MPI-IO operations and the corresponding POSIX operations, as demonstrated in Figure 36, "FLASH-IO Call Tree". Here, the collective operation MPI File\_write\_at\_all is shown to be composed of

POSIX write, read and Iseek64 operations. This provides an explanation for the read and seek operations observed with Darshan. A similar analysis can be done for the other MPI-IO operations.

Figure 37. FLASH-IO I/O Summary



A further example of Vampir profiling capability is shown in Figure 37, "FLASH-IO I/O Summary". Here, the aggregated I/O transaction size per file is given with Vampir providing a further set of metrics: Number of I/O Operations, Aggregated I/O Transaction Size, Aggregated I/O Transaction Time, and values of I/O Transaction Size, I/O Transaction Time, or I/O Bandwidth with respect to their selected value type. More detail about Vampir I/O summaries can be find in [22]

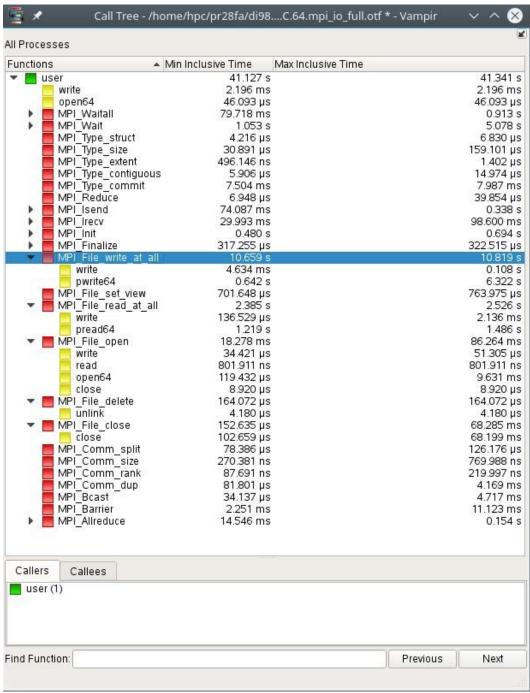


Figure 38. BT-IO Processes Timeline

Process 28 Process 29 Process 30 Process 31

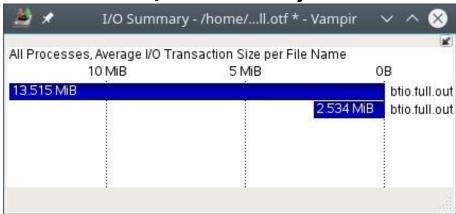
Here, BT-IO was executed under IBM MPI (IBM's Parallel Environment (PE)) on 64 processes, over 4 compute nodes and using MPI for I/O. Figure 38, "BT-IO Processes Timeline" shows the timeline for the first 32 processes. It can be seen that only process 0 and 16 perform data access operations (yellow bars) as collective operations are performed and collective buffering is enabled for BT-IO explicitly. The system where this trace was performed has collective buffering configured for one I/O aggregator per compute node.

#### Figure 39. BT-IO Call Tree



The BT-IO call tree is shown in Figure 39, "BT-IO Call Tree". Here, the MPI-IO operations use different underlying POSIX operations than Intel MPI. As can be seen in the figure, the collective operation MPI\_File\_write\_at\_all is composed of write and pwrite64 operations at the POSIX level.

Figure 40. BT-IO I/O Summary



Finally, in Figure 40, "BT-IO I/O Summary" the I/O summary is shown. BT-IO only writes and reads a single 6 GiB file but the summary presents two files. This is due to Vampir depicting the file at both the POSIX and MPIIO levels for the metric selected. In the case, the Average I/O Transaction Size is selected, showing the effect of collective buffering techniques on operation size at the POSIX level (13.515 MiB).

#### 7.5.1.2. Comparing traces using Vampir

Another useful view of the Vampir Tool is the comparison of traces. To demonstrate this, the IOR benchmark is executed using MPI-IO for independent and collective operations for a strided pattern. IOR [23] is a synthetic benchmark for testing the performance of parallel filesystems. The benchmark supports a variety of different APIs to simulate I/O load. IOR can be used for testing performance of parallel file systems using various I/O libraries: MPI-IO, POSIX-IO, HDF5 and PnetCDF.

The IOR parameters for this test were: MPI processes = 64, request size = 1 MiB, 16 MPI processes per compute node, a MPI process per core, 1GiB of data per process. and the strided pattern was configured as follows:

- Independent I/O: IOR-MPIIO-ibmmpi -a MPIIO -s 1024 -b 1m -t 1m
- Collective I/O by default: IOR-MPIIO-ibmmpi -a MPIIO -c -s 1024 -b 1m -t
   1m
- Collective I/O with collective buffering enabled: IOR-MPIIO-ibmmpi -a MPIIO -s 1024 -b 1m t 1m, but setting export ROMIO\_HINTS=romio-hints where romio\_hints contains romio\_cb\_read enable; romio\_cb\_write enable.

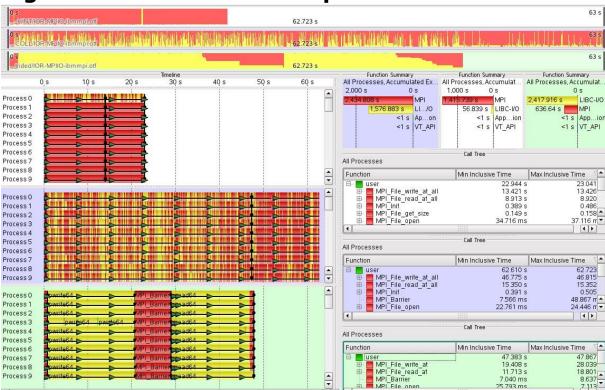


Figure 41. IOR traces comparison

Figure 41, "IOR traces comparison" shows the traces for the three cases. The first timeline corresponds to collective operations with collective buffering enabled, the second to collective operations by default and the final timeline to the independent I/O. Collective buffering is generally recommended for giving the best performance non-contiguous patterns. However, performance depends on the configuration of the underlying system as well. Here, the transfer size (1MiB) is less than the blocksize of the filesystem (8MiB), and the buffer size for collective operations is 16MiB. Therefore, aggregated I/O performs well for this system. When selecting an optimisation technique for your I/O pattern, the I/O configuration of the system itself must also be considered.

# 7.6. MPI-IO Reporting with Cray MPICH

The custom version of MPI provided by Cray (cray-mpich) available on XC-series clusters, as well as others based around Cray hardware, has an additional feature that reports a variety of I/O statistics to aid with profiling efforts. Setting the environment variable MPICH\_MPIIO\_STATS=1 before running a parallel application enables the report:

export MPICH MPIIO STATS=1

which gives output such as:

```
+-----+
  | MPIIO write access patterns for benchmark files/mpiio.dat
    independent writes
                       = 0
                     = 24
    collective writes
    independent writers
                    = 24
    aggregators
                   = 48
    stripe count
    stripe size
                    = 1048576
                    = 3072
    system writes
    aggregators active = 0,0,0,24 (1, <= 12, > 12, 24)
    total bytes for writes = 3221225472 = 3072 \text{ MiB} = 3 \text{ GiB}
    ave system write size = 1048576
    read-modify-write count = 0
    read-modify-write bytes = 0
    number of write gaps = 0
    ave write gap size = NA
                             | See "Optimizing MPI I/O on Cray XE
Systems" S-0013-20 for explanations.
  +----+
```

for each MPI-IO operation performed in the job. This allows for confirmation that the application and MPI library have been tuned effectively by, for example, exclusively using collective operations to optimise I/O.

Additionally, the environment variable MPICH\_MPIIO\_HINTS\_DISPLAY=1 is available on Cray systems to print the MPI-IO hints used by open operations on each file accessed by an application. Setting:

```
export
MPICH MPIIO HINTS DISPLAY=1
```

before running the same job as above produces:

```
PE 0: MPIIO hints for benchmark_files/mpiio.dat:
                                        = 16777216
            cb buffer size
romio_cb_read
                                       = automatic
romio cb write
                    = automatic
                                            cb nodes
= 24
                        cb_align
romio_no_indep_rw
                      = false
                                        romio cb pfr
= disable
                        romio_cb_fr_types
                                               = aar
romio cb fr alignment
                                                   1
romio cb ds threshold
romio cb alltoall
                                           automatic
```

ind rd buffer size 4194304 ind wr buffer size 524288 romio\_ds\_read disable romio ds write striping factor = disable = 48striping\_unit = 1048576romio lustre start iodevice = 0 direct io = false aggregator placement stride = -1 abort on rw error \*:\* cb config list romio filesystem type = CRAY ADIO:

#### Reference:

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#### **Contribution:**

Md Shahedul Islam Khan – extensive reading and resource gathering and implement it in the report.

Amirbek Raimov - extensive reading and resource gathering and implement it in the report.

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