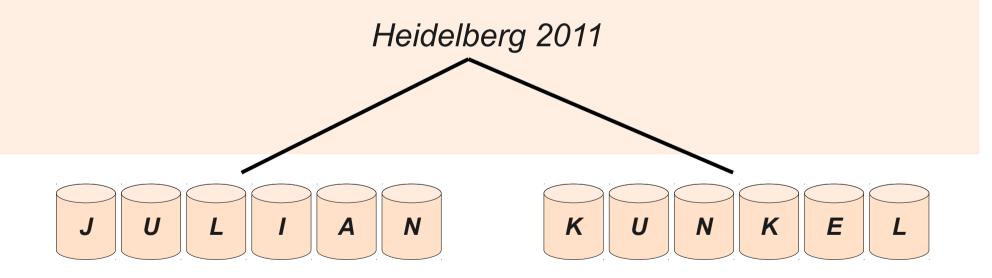
Parallel Programming with Respect to I/O



Agenda Overview

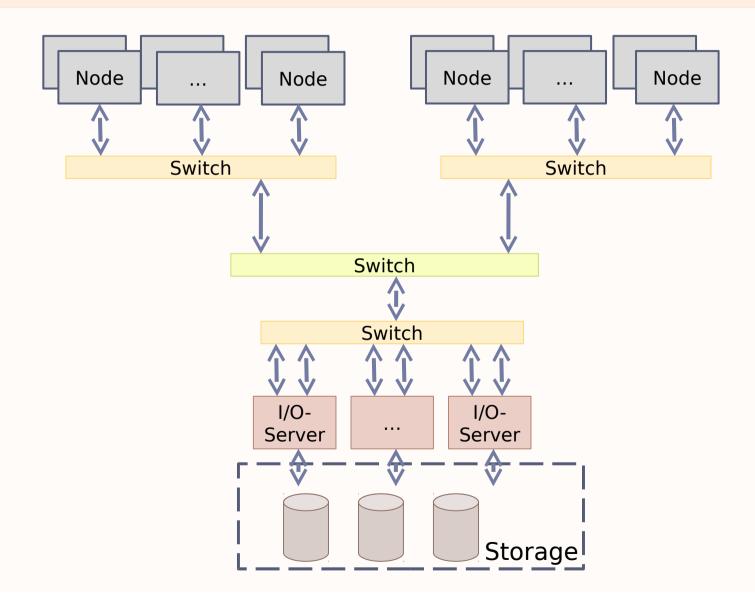
- 12.01 Morning
 - Introduction to Parallel Programming
 - Performance Estimation
- 12.01 Afternoon
 - Storage & File System concepts
- 13.01 Morning
 - Programming (Parallel) I/O

Agenda

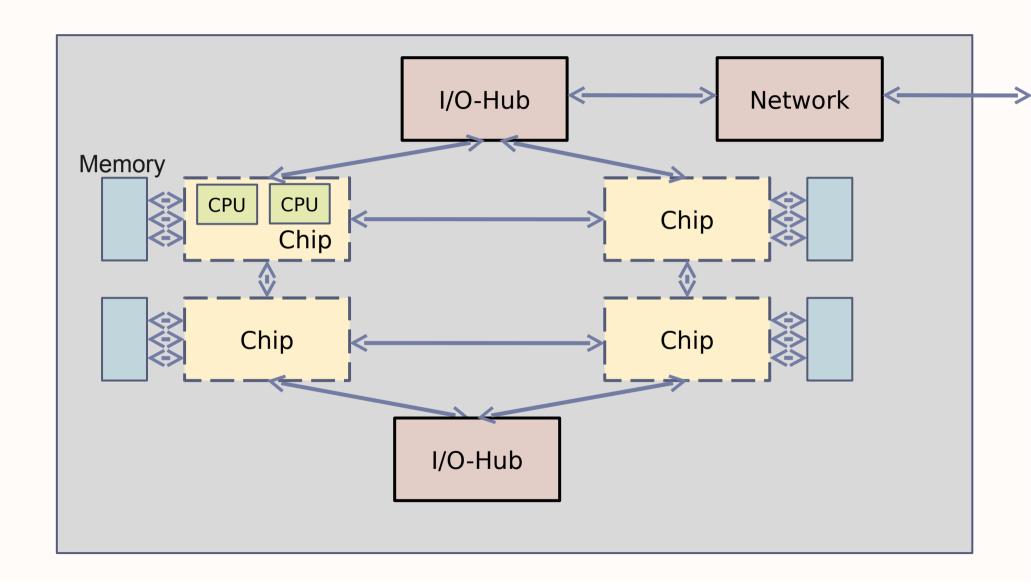
- Beowulf Clusters
- Parallel Processing in a Nutshell
 - State-of-the-art interfaces
 - Development cycle of parallel programs
 - Work partitioning
- Performance analysis of parallel programs
 - Theory
 - Hardware impact
 - Performance estimations

Introduction

Exemplary Beowulf Cluster



Cluster Node

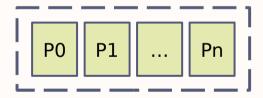


Parallel Processing

- Hardware provides resources
 - Computation
 - Storage
 - Network
- Program utilizes resources to solve a problem
- Communication enables collaboration
- Established standards are MPI and OpenMP

Message Passing Interface (MPI)

- Defines a standard API for inter-process communication
- Communication & I/O are programmed explicitly
 - A programmer is responsible for proper usage
 - Be careful with the semantics of the standard!
 - Processes are enumerated:



- An MPI implementation realizes the standard
 - By providing libraries and an execution environment
 - MPI maps the processes to the existing hardware

MPI Example

- Problem: Compute 200 independent values
 - For example compute f(x) for x in {1, ... 200}
- Two processes collaborate
 - Process 0 performs I/O and user interaction
 - Both compute 100 values
 - Process 1 sends results to process 0
 - Process 0 outputs results

MPI Schematic Example

```
#include "mpi.h"
Int main(int argc, char **argv) {
  int myrank; /* Used to store the process' unique number among all process */
  int nprocs; /* Number of processes */
  Int values[200]; /* Data which will be communicated between processes */
  MPI_Init(&argc, &argv); /* Initialize MPI to enable inter-process communication */
  /* Determine information about collaborating processes */
  MPI_Comm_rank(MPI_COMM_WORLD, & myrank); // My unique number
  MPI Comm size(MPI COMM WORLD, & nprocs); // Number of processes
  If (myrank == 0){
     <Compute first 100 values>
     MPI_Recv(values + 100, MPI_INT, 100, 1, 4711, MPI_COMM_WORLD, & status);
  }else if (myrank == 1){
    <Compute second 100 values>
    MPI Send(values, MPI INT, 100, 0, 4711, MPI COMM WORLD);
  MPI Finalize();
```

OpenMP

- Semi-automatic parallelization
 - Programmer specifies parallelism in compiler directives
 - OpenMP aware compiler generates parallel code
 - Provides lightweight library to manage parallelism
- Uses threads to utilize available CPUs
 - Works on shared memory machines
- Example code computes vector addition
 - ◆ C = A + B

OpenMP Example Code

```
#include <omp.h>
int main (int argc, char *argv[]) {
#pragma omp parallel
 Int tid = omp_get_thread_num();
 if (tid == 0)
  Int nthreads = omp_get_num_threads();
  printf("Number of threads = %d\n", nthreads);
 #pragma omp for
 for (i=0; i<N; i++) { /* work in this loop is automatically distributed */
  c[i] = a[i] + b[i];
 } /* end of parallel section */
```

Development of Parallel Programs

- Loop of development:
 - 1. Parallelization of the algorithm
 - 2. Implementation
 - 3. Debugging
 - 4. Performance analysis & optimization

Parallelizing an algorithm requires to consider I/O

Partitioning of Work

- Utilize computation resource by assigning work
- Paradigms:
 - Functional decomposition

Processes perform different tasks

Examples: Pipeline, coupled climate codes

Domain decomposition

Processes work on different data

Examples: Segmenting images, feature extraction

- Flexible, fine grained partitioning is preferable
 - Allows to utilize any number of (different kinds of) CPUs

Performance of Parallel Programs

Performance of Parallel Programs

- Relevant aspects
 - Hardware characteristics
 - Determine how fast resources and communication are
 - Operating system and system configuration
 - How well are local resources utilized
 - Communication and I/O libraries
 - How well could a parallel program utilize the parallel computer
 - Code
 - Should perform only computation relevant for the solution
 - Everything else including communication: Overhead

Theoretic Performance Considerations

Speedup:

$$S_p = \frac{T_1}{T_p}$$

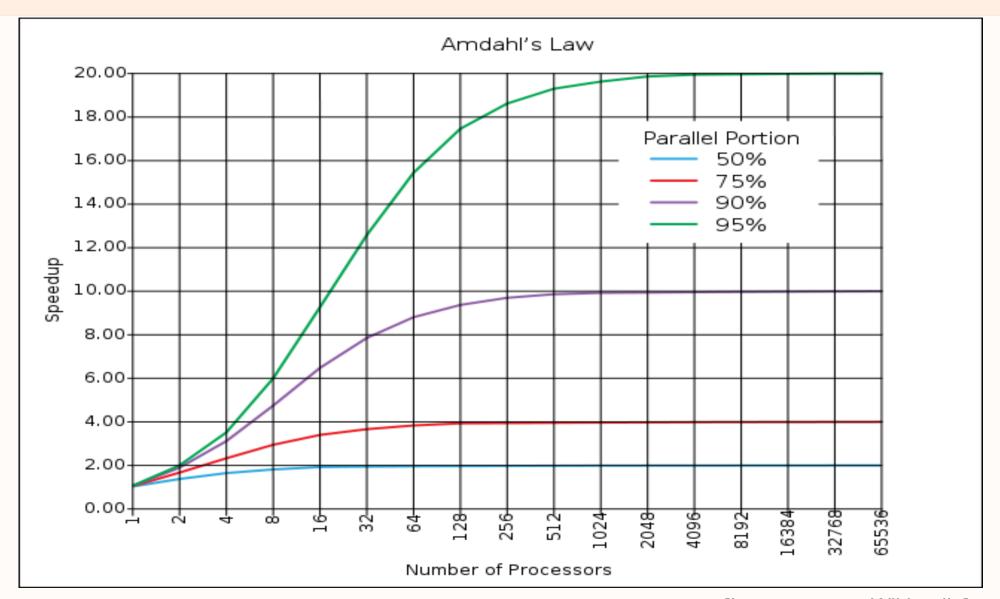
- Assume a fixed problem
- Runtime of the best sequential program: T₁
- Runtime of the parallel program with p CPUs: T_p
- Efficiency:

$$E_p = \frac{S_p}{p}$$

- Amdahl's law:
 - P: fraction of parallelizable code
 - N: number of processors

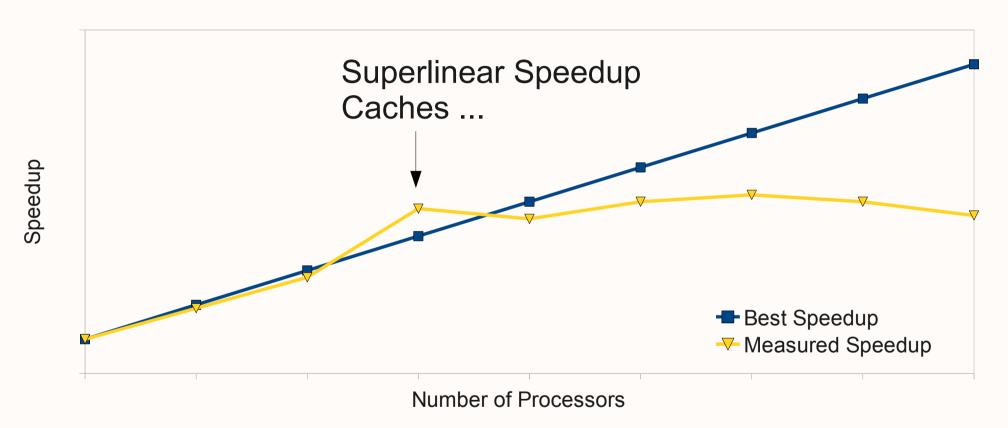
$$S_N \leqslant \frac{1}{(1-P) + \frac{P}{N}}$$

Achievable Speedup



[Image source: Wikipedia]

Realistic Speedup

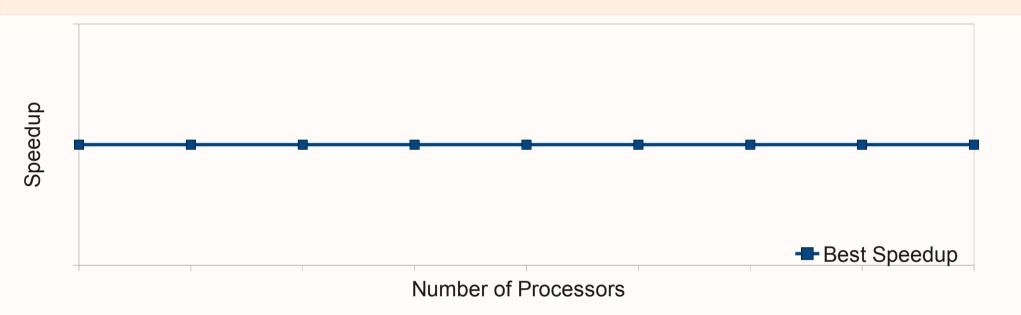


- Scalability of parallel programs is limited!
- At some point overhead > improvement

Scalability

- Scalability:
 - Variation of solution time with the number of processors
- Good scalability:
 - Efficient algorithm & implementation
- Two common terms in HPC:
 - Strong scaling
 - Fixed problem size
 - What we analyzed with Amdahl (Speedup + Efficiency)
 - Scaling of a fixed problem is always limited
 - Weak scaling
 - Fixed problem size per processor
 - Easier to achieve than strong scaling

"Speedup" with Weak Scaling



- Speedup definition not appropriate
- Each processor has the same amount of work
- Efficiency more useful:

$$E_p = \frac{T_p}{T_1}$$

Theoretic Performance Considerations

- In Amdahl etc. Hardware was not considered
- Without knowledge of the system we achieve suboptimal performance
 - Gear the algorithm / system / code towards hardware
- Knowledge allows to assess observed performance
- But, hardware components are complex
- Model cluster performance with important characteristics:
 - Simple to understand
 - Sufficient detailed to allow to identify bottlenecks
 - Provides a starting point for further analysis

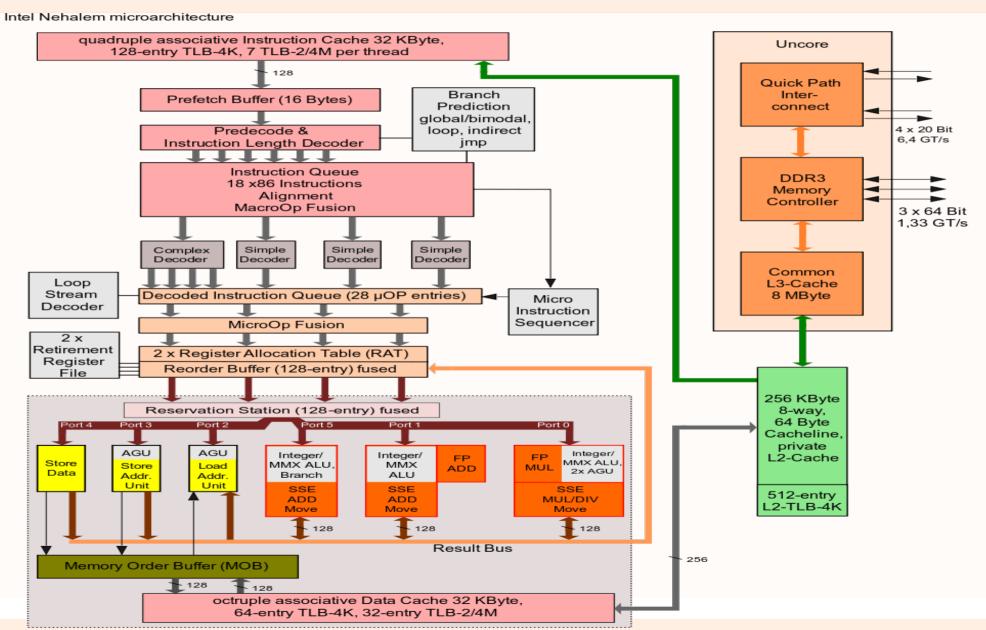
Hardware Characteristics

- How do we obtain hardware characteristics?
 - Vendor information
 - Often optimistic, best case performance
 - Benchmark components
 - Benchmark results are tailored to a given workload
- Assessment of observed performance:
 - How well is a "program" capable to utilize hardware?
 - Software and hardware tries to hide undesired characteristics
 - Measured <= Benchmark <= Theoretic peak</p>

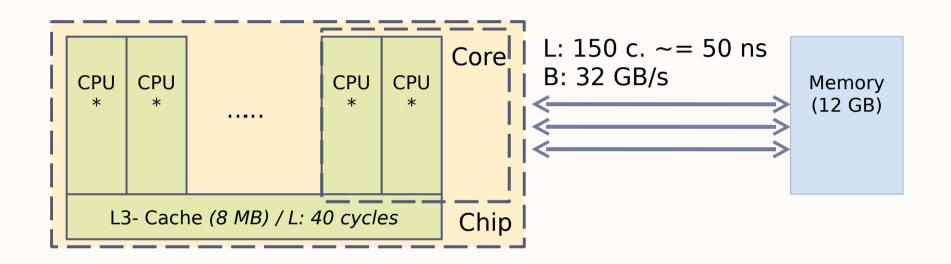
Relevant Characteristics

- Processor
 - Instructions per second
 - Size of L1, L2 and L3 caches
- Memory subsystem
 - Interconnect topology: single bus, bus per chip (e.g. Nehalem)
 - Latency and bandwidth
- I/O-Subsystem
 - Bandwidth (per client node and per server)
 - IOPS Number of I/O operations per second (for metadata)
- Network
 - Interconnect topology
 - Latency and bandwidth
 - Remember: communication is considered as overhead!

Processor Microarchitecture



Logical View of a Processor



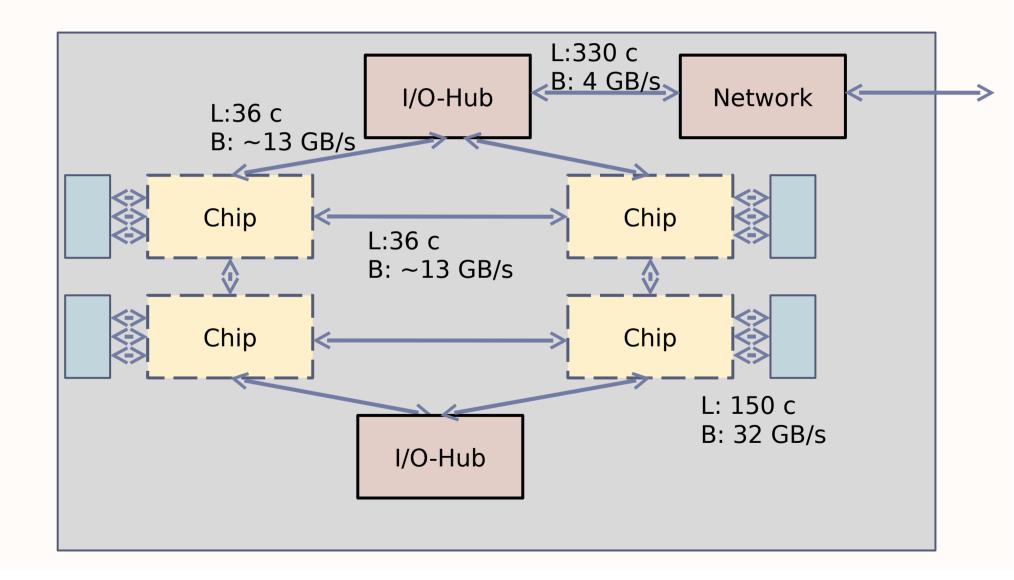
CPU might execute 2 Flop per cycle (Add+Mult)

Per Core: Latency:

L1: 32K Instruction/32K data 4 cycles L2: 256K 10 cycles

Data is taken from a Nehalem @3GHz, depends on Memory

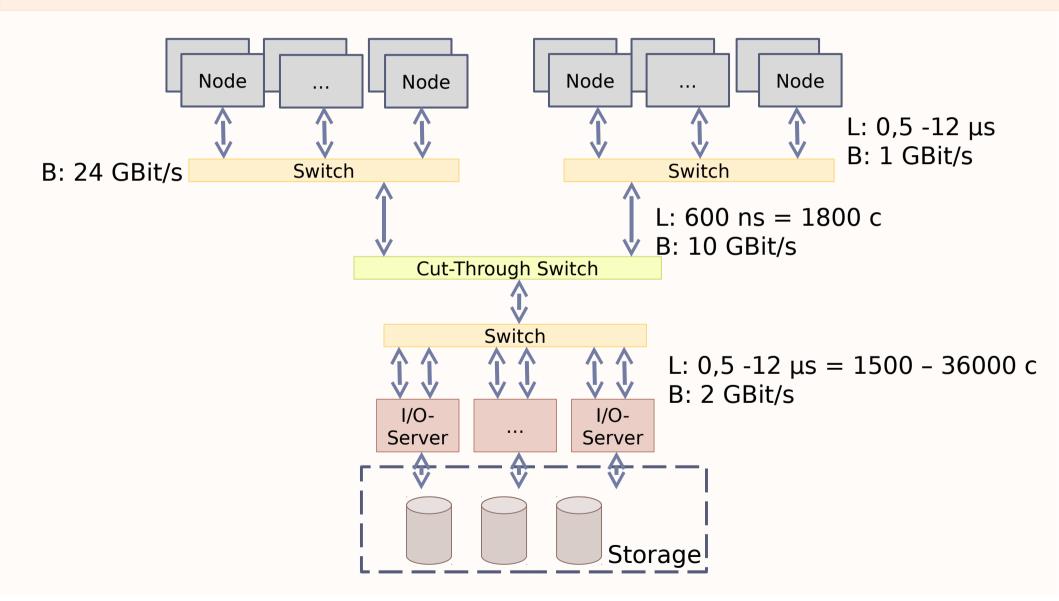
Cluster Node



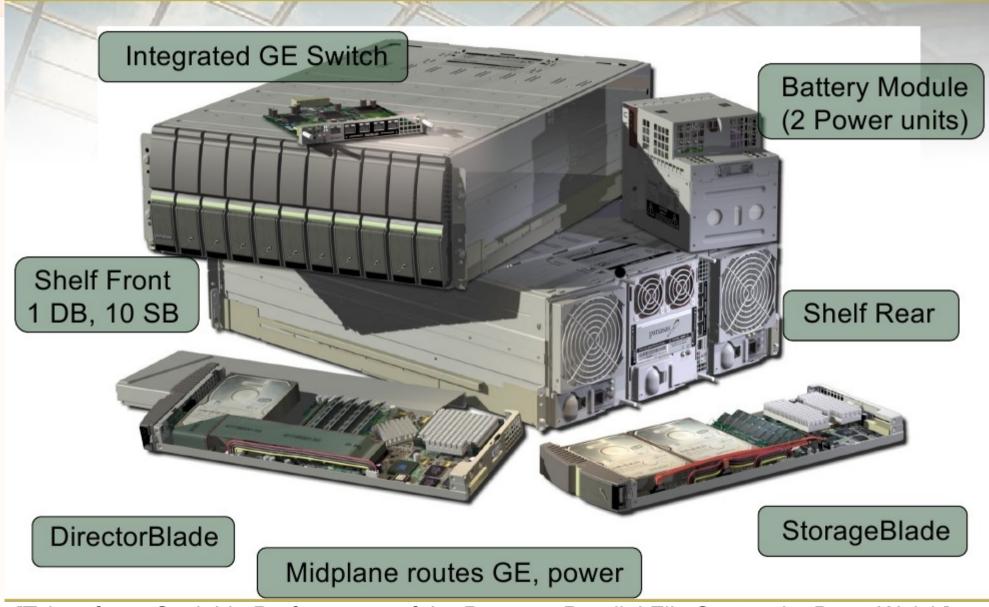
Observations

- Cache sizes vary
- Cache miss is expensive
- NUMA characteristics
- Upper CPUs have faster access to the network
- On some machines: Network may be saturated by simultaneous accesses from multiple CPUs

Exemplary Beowulf Cluster

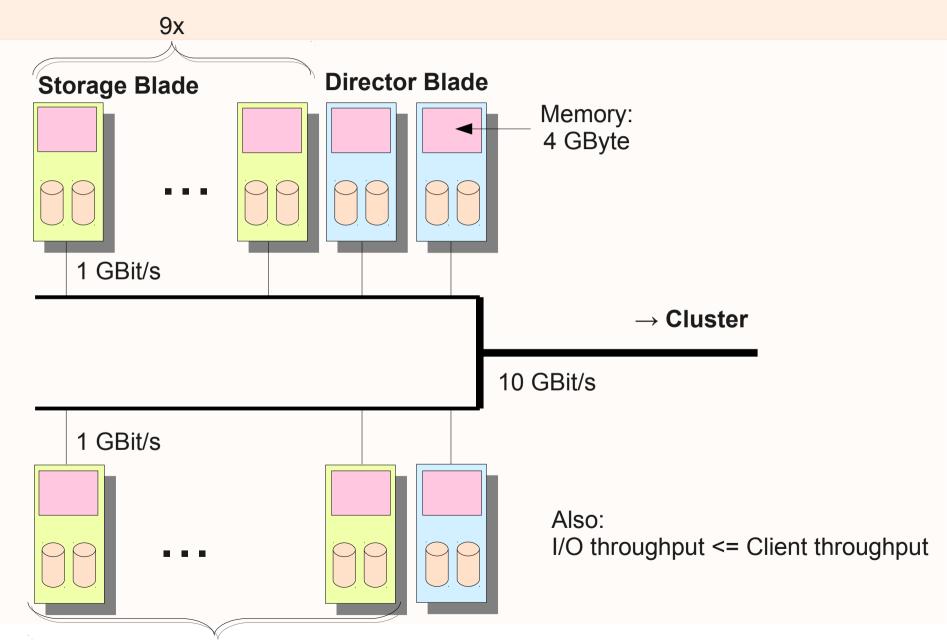


Panasas Hardware



[Taken from: Scalable Performance of the Panasas Parallel File System by Brent Welch]

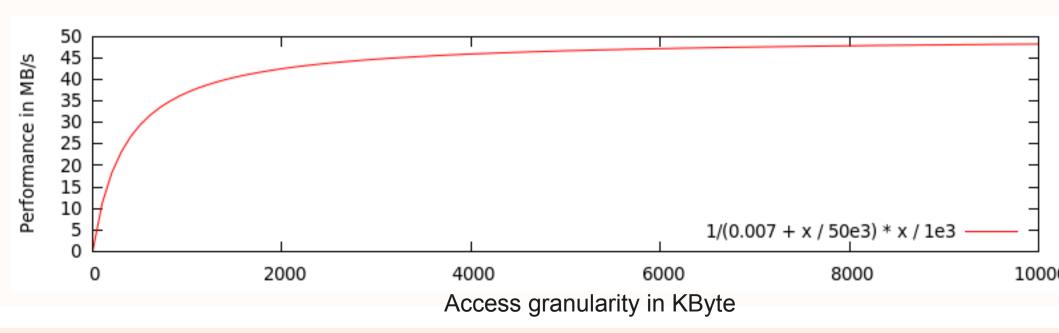
Exemplary Panasas Storage



13.01.2011 10x *Julian M. Kunkel* 31

Theoretical I/O-Performance

- Persistent storage devices
 - Hard disk
 - Latency: ~7 ms = 21.000.000 cycles!
 - Bandwidth: 50 MB/s
 - ★ With access granularity of 100 KB only 11.1 MB/s!



Improving I/O-Performance

- Software and hardware tries to hide I/O penalty
- Caching of data
 - Allows application to continue while I/O completes in the background
 => Write-behind
 - Allow to aggregate multiple (small) operations into larger operations
 - Read data from disk before it is needed (Read-ahead)
 - Requires memory! Hiding vs. increased problem size
- Programming:
 - Overlap I/O (or communication) with computation
 - I/O and communication comes almost for free
 - In the best case a speedup of 2 can be achieved vs. non-overlapped
 - Optimize file format and access pattern (later)

Storage & File Systems

Agenda 2

- Abstraction layers
- Low-level to high-level
- (Parallel) file systems

Abstraction Layers

	Abstraction	Examples
Parallel application	Task	Checkpoint
High-level I/O libraries	Domain specific	HDF5
Low-level I/O interface	Inter-Application/FS	POSIX, MPI
File System	Logical objects	Ext3, FAT32
Block storage	Block	iSCSI, SATA
Storage device	Technology	Controller

Block Storage

- Sequence of blocks which can be accessed (randomly)
 - Nowadays, block size is often 512 Byte
 - Only full blocks can be accessed
- Simple interface to access data
 - Read (offset, size)
 - Write (offset, size, <data>)
 - Additional management commands
 - Size must be multiple of a block
 - Modification might require Read-Modify-Write
- Examples: SATA, SCSI

File System Overview

- Provides an interface to logical (persistent) objects
 - Interface to access and manipulate objects
 - Semantics
- Store information about data (metadata)
 - Permissions, timestamps, ...
- Structure objects by providing a namespace
 - Usually the namespace is hierarchical
- On-disk-format defines physical representation
 - How are physical blocks mapped into logical objects

Low-level I/O interface

- Abstracts from file system
 - Common interface and semantics
 - Linux "Virtual Filesystem Switch" (VFS)
 - Windows "Installable File System"
- Operations:
 - Open(), Read(), Write(), Close()
- Drawbacks:
 - Individual file system features not usable
 - Low-level tools must be used
 - Common semantics does not exploit local features
 - Eventually fast operations must be done "slowly"

Storage Devices



Solid state drive



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HDD vs. SSD

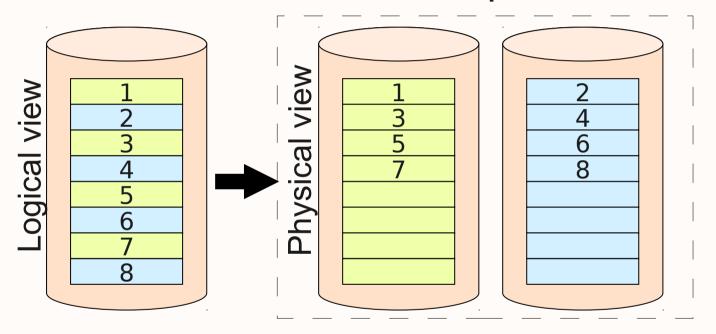
- SSDs have no mechanical parts
 - Technology: Flash-EEPROM
 - Time for random access == sequential access
 - Write-Access in block granularity (e.g. 256 KByte)
 - More expensive for capacity, but faster (200 MB/s)
 - Controller defines performance
- Requires new optimizations in OS and middleware
 - Sequential optimizations (might) harm performance

RAID

- Multiple storage devices are combined into
 - Redundant Array of Independent Disks
- Hardware or Software distributes blocks of data
 - Among multiple devices
 - Transparent for the user or OS
- Depending on data distribution function increase
 - Capacity
 - Throughput
 - Availability and Durability

Data Distribution

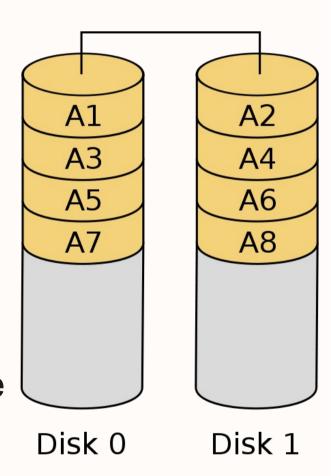
- Simple functions to allow hardware implementation
- Example:



- Block size can be configured towards application
 - Small block size increase parallelism of a single I/O access
 - Large block size increase concurrency of I/Os

RAID-0

- Round-robin block distribution
- Improves performance
 - All devices participate in I/O
- Improves capacity
 - Sum of all capacities
- Error-prone
 - Not protected against device failure

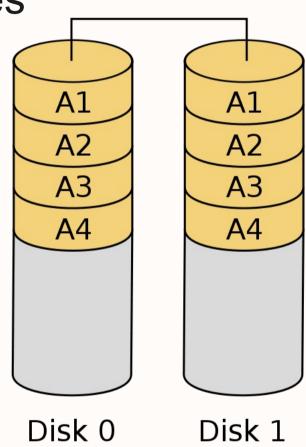


[Source: Wikipedia]

RAID-1

Mirror data blocks between devices

- Data is written to all devices
- Improves read performance
 - One device required to read data
- No capacity improvement
 - RAID == smallest device
- Protects against errors
 - Only one device required

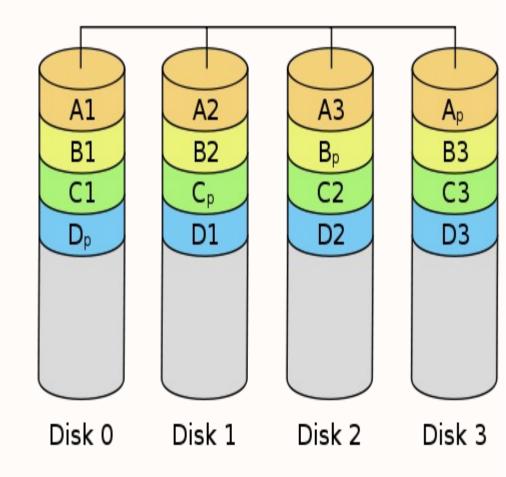


[Source: Wikipedia]

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RAID-5

- Data blocks are distributed in RR
- Additional parity block per row
 - One device may fail
 - XOR function
 - Requires to maintain parity block
 - Wanders between devices
- Compromise of performance, high availability and capacity



[Source: Wikipedia]

Tolerating Failures

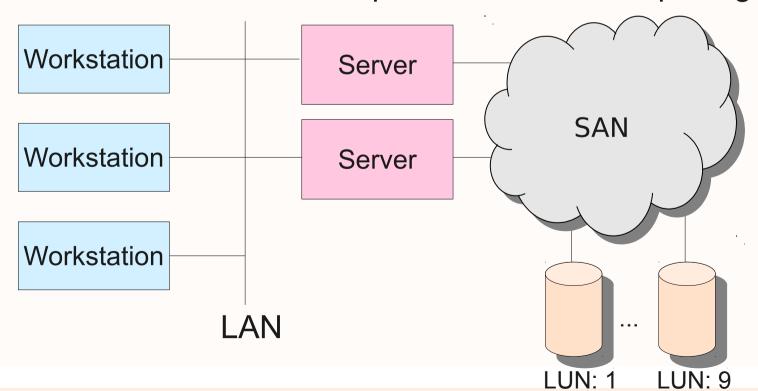
- Rebuild of degraded RAID(-5) systems
 - Requires to recreate blocks on broken device
 - Use XOR on all blocks in a row to compute missing data
 - Transparent to the user (but lower performance)
- Time intensive as device capacity improves
 - During rebuild procedure data protection is lower
 - A second disk failure in RAID-5 and data gets lost!

RAID Conclusions

- RAID distributes data blocks among devices
- Improve availability and persistence
- RAID-levels 2 to 4 are not important
- RAID-10 == RAID-0 over multiple RAID-1
- RAID-6 (2-dimensional parity)
 - Similar to RAID-5
 - Tolerates two hardware failures
 - State-of-the-art in industry
- RAID concept is used in higher abstraction levels

Storage Area Network

- SAN is a network with attached block storage
 - Multiple servers can access the same device
 - Only one server has dedicated access to one unit
 - But, devices can be partitioned into multiple logical units

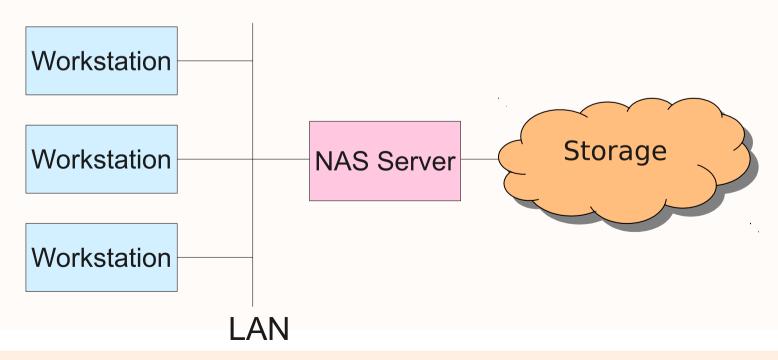


SAN

- Access via low level access in blocks
 - SCSI protocol
 - Servers must provide a file system for workstations
- Can be configured to provide high availability
 - Redundant servers, switches...
 - If one server/switch fails another takes over
- Technology
 - iSCSI (IP network, can use the LAN technology)
 - Fibre Channel

Network Attached Storage

- NAS provides network storage in a box
 - Buy a NAS, plug in, use it
 - High-level access: FTP, CIFS, NFS
 - Sometimes block storage via iSCSI



File Systems Aspects

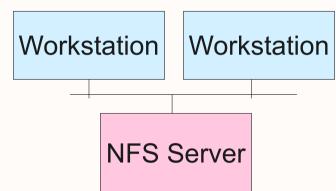
- Data integrity
- Fault-tolerance
- (High)-availability
- Access semantics
 - Concurrent access to file / directories
 - (Meta)Data caching
- Performance
 - Metadata ops per second and I/O-throughput
- Management features
 - Snapshots to ease backup
 - Performance monitoring

File Systems

- Disk
 - Manage/Access (local) direct attached storage
 - Examples: Ext3, FAT32, NTFS
- Shared disk / Cluster
 - Multiple nodes can access one block storage
 - SAN + software to allow concurrent access
 - Examples: GFS, OCFS2

Distributed (Network) File Systems

- Provide storage to many workstations (clients)
 - Like a NAS system
- Examples: NFS, CIFS
- NFS v4 (Network File System)
 - Client is connected to exactly one server
 - Operations are performed via Remote Procedure calls
 - (Meta)-data caching on the client
 - Close-to-open cache consistency
 - Clients can read stale (Meta)-data
 - File delegation
 - File or byte-range locks supported

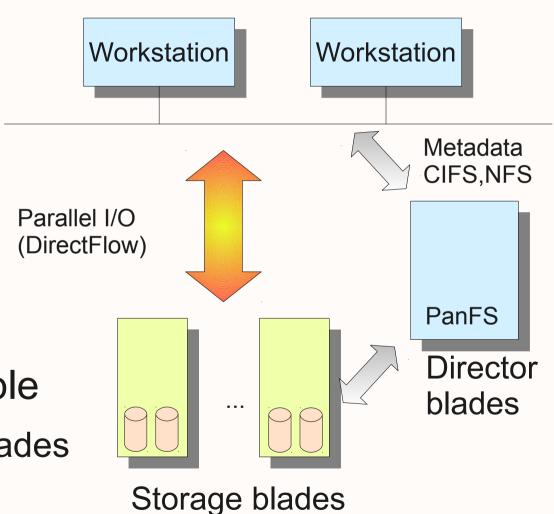


(Distributed) Parallel File Systems

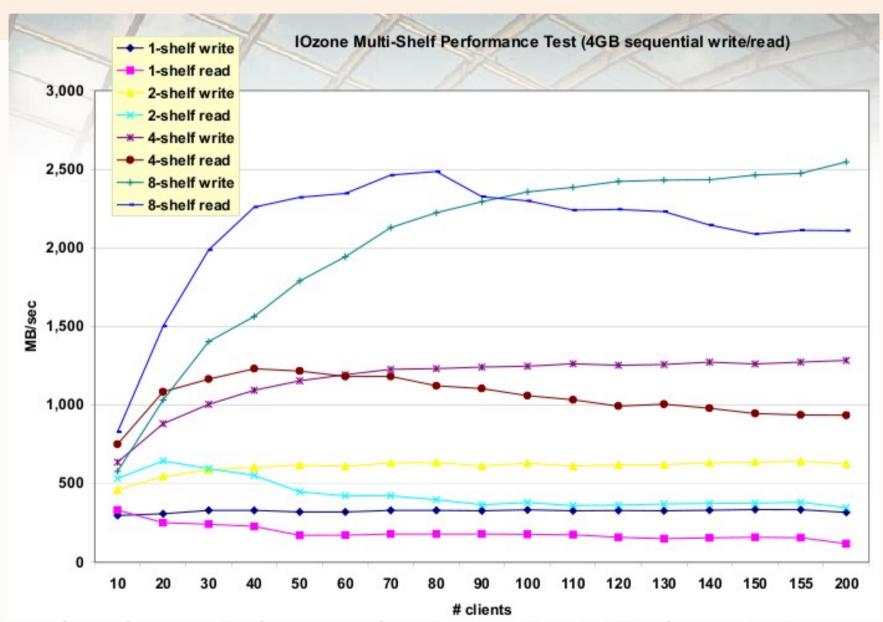
- A file can be accessed concurrently by all clients
- Data of a file is distributed among servers
 - Data is distributed by RAID concepts
 - Scalable performance
- Metadata may be distributed among servers
- Examples: GPFS, PanFS, PVFS2

Panasas Active Scale OS (PanFS)

- Metadata
 - Director blades
- Object storage
 - Storage blades
- Kernel module
 - Enables parallel I/O
- CIFS and NFS possible
 - Routed via director blades
- Dynamic RAID-level
 - Depending on file size

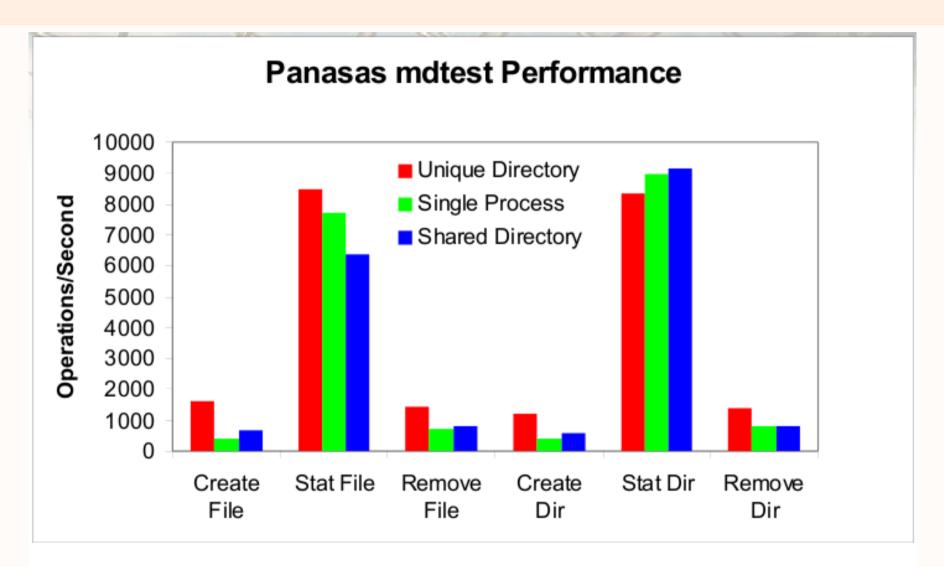


Panasas I/O Performance



[Taken from: Scalable Performance of the Panasas Parallel File System by Brent Welch]

Metadata Benchmark Example



mpirun -n 64 mdtest -d \$dir -n 100 -i 3 -N 1 -v -u

[Taken from: Scalable Performance of the Panasas Parallel File System by Brent Welch]

Programming (Parallel) I/O

Agenda

- Low-level I/O concept
- On-Disk-Format
- POSIX interface in brief
- MPI-I/O
- High-level I/O libraries

Low-level I/O Concept

A file is a sequence of bytes



- File pointer shows last (access) position
- A number of bytes can be read/written
- File pointer can be repositioned

Semantic Gap in I/O

- Applications work with structured data
 - Vectors, Matrixes, 3-dimensional climate data
- A file is just a sequence of bytes

- Applications must serialize data into the file
 - Complex data types?
 - Mapping defines performance

On-Disk-Format

- Depends on application behavior
 - Sequential, large accesses favorable
- Example: Storing a 2-D matrix

$$A = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix}$$

d

File

a

- Variants:
 - Serialize by row
 - A row can be written directly from memory
 - Serialize by column

a d g b e h c f i

e

Offset

POSIX Interface Aspects

- POSIX requires serial processing of I/O operations
 - A read() following a write() must return new data
 - Cheap within one node
 - Expensive in a distributed environment
 - Cache-coherence between nodes
 - Lock mechanisms required
- Exception handling of calls expensive
- In Linux: large file support compile options

POSIX Interface Recommendations

- Use: open(), close(), pwrite(), pread() functions
 - pread() and pwrite() are thread safe
- Don't use asynchronous I/O (aio)
 - Performance varies between GLIBC implementation
- Don't use fwrite() etc.
 - Sometimes 64 Bit issues
 - Buffering unclear / might vary (even with setvbuf())
- Build a wrapper which ensures proper writing
 - Read man pages

MPI-I/O Concepts

- MPI-I/O: (coordinated) parallel I/O for MPI
 - Similar to POSIX-I/O
- File pointer:
 - Individual vs. Shared File Pointer
- File view
 - Filters uninteresting file areas for the processes
 - Allows noncontiguous accesses
- Collective I/O
 - Multiple clients participate in a collaborative I/O
 - Enables collective optimizations
- Non-blocking I/O
 - Overlap computation with I/O

Accessing a File with MPI-I/O

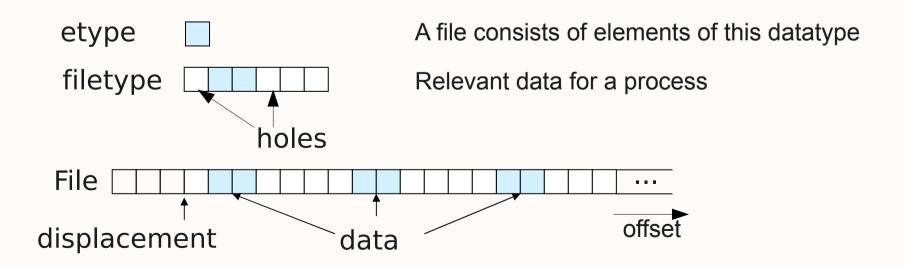
```
MPI Offset my offset, my current offset;
MPI File fh;
MPI Status status;
int buffer[100];
MPI File open(MPI COMM WORLD, filename, MPI MODE RDWR
   | MPI MODE CREATE, MPI INFO NULL, & fh);
MPI File write(fh, buffer, 100, MPI INT, &status);
MPI Get count(&status, MPI INT, &count);
printf("process %3d wrote %d ints\n", my rank, count);
MPI File close(&fh);
```

Shared File Pointer

- Processes share only one file pointer
- Read/Write from global file offset
- Useful for log-files, appends to the file
- Drawbacks:
 - Serialization of operation => Bottleneck
 - Not always implemented
- Advice: Avoid shared file pointers

File Views

- Each process can set an individual file view
 - MPI_File_set_view(MPI_File fh, MPI_Offset displacement, MPI_Datatype etype, MPI_Datatype filetype, char *datarep, MPI_Info info)
 - MPI_File_set_view() is a collective operation!



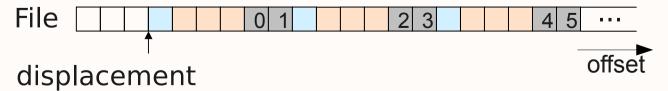
File Views (2)

- I/O operations return only accessible data
- Positioning happens in multiple of etype
 - Example: MPI_File_seek(fh, 10, MPI_SEEK_SET)
 - File pointer is set to the 10 element which can be accessed
- Displacement is used to skip the file header
- For portability a data representation can be set
 - "native": data is stored in the file as it is in memory
 - Not portable
 - "internal": MPI implementation chooses format
 - Portable as long as the same MPI implementation is used
 - "external32": I/O is converted to a defined format

File Views Example

- Example: assume three processes
 - File data is distributed among processes

- Now they use MPI_File_set_view(filetype...)
- File data is distributed in the following way:



Offsets are given for process 2 in the boxes

File View Code Example

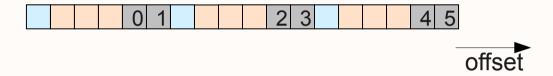
```
etype = MPI INT
 filetype =
/* Create a datatype consisting of two integers */
MPI Type contiguous(2,MPI INT,&contig);
/* add four holes */
lower boundary=0;
extent=6*sizeof(int); /* extend size is a total of 6 int */
MPI Type create resized(contig, lower boundary, extent,
  &filetype);
/* finalize the new datatype */
MPI Type commit(&filetyp);
/* set the file view according to the new datatype */
MPI File set view(filehandle, displacement, etype, filetype,
 "native", MPI INFO NULL);
```

Collective I/O

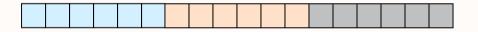
- MPI_File_[read|write]_all()
 - All clients of a communicator perform I/O
- MPI_File_open() sets participating communicator
- Usually file is partitioned by setting a file view
- Clients try to aggregate small accesses
 - Large requests to I/O subsystem
 - Clients exchange required data by communication
 - Additional communication might slow down I/O!

Collective I/O Example

Assume three processes access their data



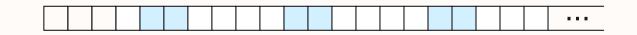
Processes read individual potions of the file:



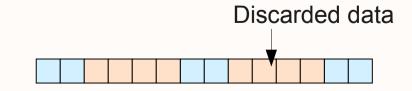
- Then communicate data to target processes
 - Process continues until all data is processed
 - I/O granularity depends on the size of the collective buffer
 - At most each client fills its collective buffer (e.g. 8 MByte)

Data-Sieving

- Applicable to non-contigous access
- Read/Write a contiguous block of data
 - Throw away uninteresting data
 - Reduces number of I/O calls
- Application requests data:
- File view:

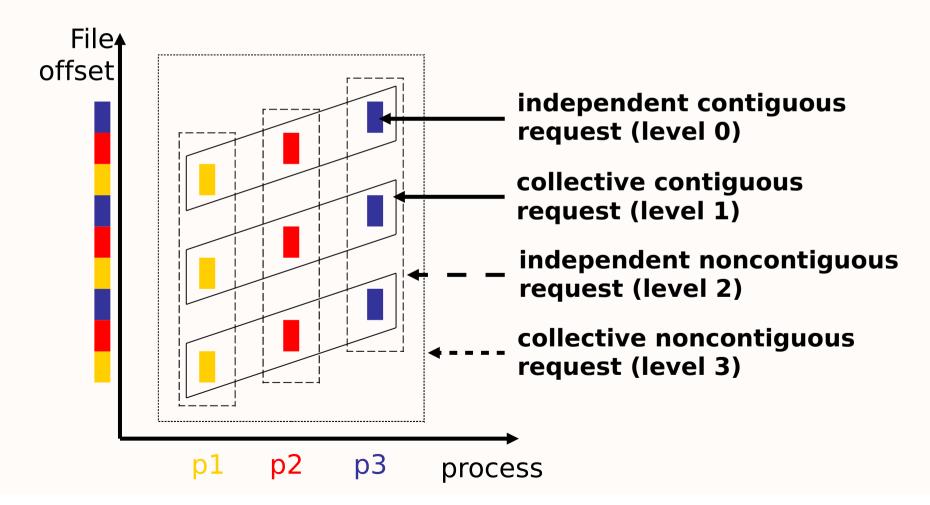


• Accessed file data:



Level of Access

Non-contiguous and collective can be combined



Non-blocking I/O

- All blocking I/O operations have a non-blocking variant
 - Name convention MPI_File_I... [read|write|...]
- Allows to overlap computation & I/O
- In the best case I/O access time is hidden

Blocking	Compute	I/O	Compute	I/O
Non-	Compute	Compute	Compute	
Blocking	Compute	Compute	Compute	
Blocking		I/O	I/O	

time

MPI Consistency Semantics

 Concurrent modifications of the same block may destroy data:

Write (A)

Write (B)

Result or ... all combinations

- Atomic mode (Sequentially consistency)
 - MPI_File_set_atomicity(MPI_File fh, int flag)

Write (A)

Write (B)

Result or

- Flushing file content (e.g. for checkpoints)
 - MPI_File_sync(MPI_File fh)

High-Level I/O Libraries

- A file is more than just an array of bytes!
 - Abstract description of datatypes and I/O
 - Self describing data formats
 - Architecture independent (portability)
- Examples: NetCDF-4 and HDF5
- Recent development: ADIOS
 - API explicitly expresses non-blocking I/O
 - Adjust I/O method by changing XML
 - MPI (collective, individual), asynchronous I/O, NULL
 - Without recompiling!
 - BP Dataformat relaxes HDF5 dataformat for performance

What is HDF5

- A versatile data model that can represent very complex data objects and a wide variety of metadata.
- A completely portable file format with no limit on the number or size of data objects in the collection.
- A software library that runs on a range of computational platforms, from laptops to massively parallel systems, and implements a high-level API with C, C++, Fortran 90, and Java interfaces.
- A rich set of integrated performance features that allow for access time and storage space optimizations.
- Tools and applications for managing, manipulating, viewing, and analyzing the data in the collection.

HDF5

- More sophisticated than NetCDF
 - Relaxes shortcomings
 - Datatype concept similar to MPI, but references to data
- Uses its own portable file format
 - Basically its a file system inside a file
 - HDF5 group: directory with metadata contains datasets and other groups
 - HDF5 dataset: a multidimensional array of data elements with additional metadata
 - NetCDF-4 allows to store data in the HDF5 format!
- Parallel processing is possible, it can use MPI-IO

Other Features

- Filters can be added per dataset
 - Compression, integrity protection via MD5, ...
- File formats allows different file layouts
 - contiguous, compact, chunked
 - Depending on the workload the right one can be chosen
- References between datasets
- Attributes describe content of a dataset
- Tools:
 - h5perf tool measures performance of a N-D matrix
 - h5dump prints metadata and data of a file
 - h5repack copy & modify internal storage format

HDF5 – Creating a Dataset

```
hid t dataset, datatype, dataspace; /* declare identifiers */
/* Create dataspace: Describe the size of the array and create the data space for
fixed size dataset. */
dimsf[0] = NX; dimsf[1] = NY;
dataspace = H5Screate simple(RANK, dimsf, NULL);
/* Define datatype for the data in the file. Store little endian integer numbers. */
datatype = H5Tcopy(H5T_NATIVE_INT);
status = H5Tset order(datatype, H5T ORDER LE);
/* Create a new dataset within the file using defined dataspace and datatype and
default dataset creation properties. NOTE: H5T NATIVE INT can be used as
datatype if conversion to little endian is not needed. */
dataset = H5Dcreate(file, DATASETNAME, datatype, dataspace, H5P_DEFAULT);
```

[Example taken from http://www.hdfgroup.org/]

h5dump example

```
HDF5 "h5ex t cpxcmpdatt.h5" {
GROUP "/" {
 DATASET "Ambient Temperature" {
   DATATYPE H5T IEEE F64LE
   DATASPACE SIMPLE { (32, 32) / (32, 32) }
   DATA {
   (0,0): 66.8, 66.9, 67, 67.1, 67.2, 67.3, 67.4, 67.5, 67.6, 67.7, 67.8,
   (0,22): 69, 69.1, 69.2, 69.3, 69.4, 69.5, 69.6, 69.7, 69.8, 69.9,
   (31,22): 72.1, 72.2, 72.3, 72.4, 72.5, 72.6, 72.7, 72.8, 72.9, 73
 }}
 GROUP "Land Vehicles" {}
 DATASET "DS1" {
   ATTRIBUTE "A1" {
     DATATYPE H5T COMPOUND {
       H5T VLEN { H5T COMPOUND {
       H5T STD I32LE "Serial number";
       H5T STRING {
         STRSIZE H5T VARIABLE;STRPAD H5T STR NULLTERM;CSET
H5T_CSET_ASCII;CTYPE H5T_C_S1;
       } "Location";
       H5T IEEE F64LE "Temperature (F)";
       H5T IEEE F64LE "Pressure (inHg)";
      }} "Sensors";
```

```
H5T STRING {
   STRSIZE H5T_VARIABLE...
 } "Name":
 H5T ENUM {
   H5T STD I32LE; "Red" 0; "Green" 1; "Blue" 2;
 } "Color";
 H5T_ARRAY { [3] H5T_IEEE_F64LE } "Location";
 H5T_REFERENCE "Group";
 H5T REFERENCE "Surveyed areas";
DATASPACE SIMPLE { (2)/(2) }
DATA {
(1): { ({
       3244.
      "Roof".
      83.82,
      29.92
    }),
   "Automobile",
   Red.
   [ 326734, 221568, 432.36 ],
   GROUP 1400 /Land Vehicles,
   DATASET /Ambient_Temperature {(8,26)-(11,28)}
 } } } }
```

[Example taken from http://www.hdfgroup.org/]

I/O Programming Recommendation

- Design a On-Disk-Format
 - Think about later access patterns / use cases
 - Must allow large blocks to be accessed
 - Add a self describing header (version etc.)
- Develop a domain specific wrapper library
 - Abstract from application tasks e.g. start_checkpoint()
 - Don't use any I/O library directly in your code!
 - Maybe a more suitable library appears
 - Use HDF5 and MPI-I/O if appropriate
 - Add optimizations like write-behind / read-ahead later

Outlook

- SSDs will be deployed
 - Reduces the penalty of random I/O
 - Writing sequential blocks of data out remains important
- HDF5 will improve (grants from US government)
- No changes to I/O in MPI-3

- Long term perspective:
 - MPI will be aware of system topology
 - (Better) Automatic tuning of MPI towards application

Excercise

- Parallelization of a Matrix Vector multiplication
 - Each process multiplies a part of the matrix
 - Matrix/Result is read/written in MPI by two different ways:
 - The first process reads/writes the whole Matrix and distributes/gathers data
 - Each process uses a file view to read/write portions of its data
 The user can choose to use individual or collective calls
 - The data format should be variable between:
 - Your own simple file format
 - HDF5
 - Therefore, use functions to abstract from the real input/output
 - Identify performance factors and bottlenecks for your cluster
 - How could you estimate the time for I/O and computation?
- You can program a Matrix Matrix multiplication instead ;-)

Literature

- MPI:
 - http://www.mcs.anl.gov/research/projects/mpi/mpi-s
- HDF5: http://www.hdfgroup.org/
- ADIOS: http://adiosapi.org/
- Thomas Ludwig, Lecture HEAS0809 and HR10
- Brent Welch, NSC08, online presentation:
 Scalable Performance of the Panasas Parallel File System