

702 Distributed Systems

Distributed File Systems

Generic Distributed File System

NFS Network File System Developed at Sun (1984)

AFS Andrew File System Developed at CMU (1980's)

Google File System (**GFS**) (2004)

HDFS Open Source Hadoop Distributed File System(2008)

directory module:	relates file names to file IDs
file module:	relates file IDs to particular files
access control module:	checks permission for operation requested
file access module:	reads or writes file data or attributes
block module:	accesses and allocates disk blocks
device module:	disk I/O and buffering

Typical non-distributed file system's layered organization. Each module depends only on the layer below it.

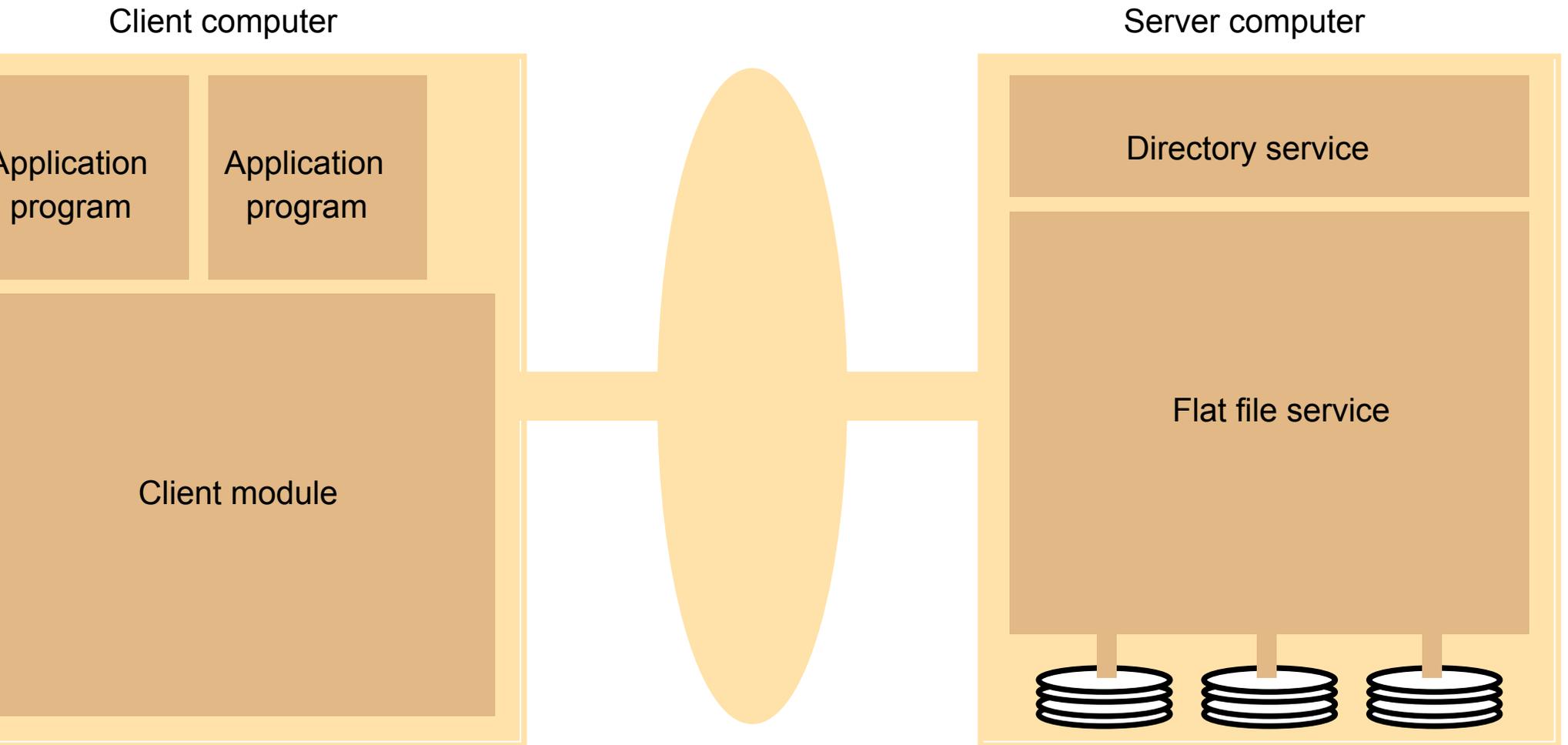
File length
Creation timestamp
Read timestamp
Write timestamp
Attribute timestamp
Reference count
Owner
File type
Access control list

Files contain both data and attributes.

The shaded attributes managed file system not normally directly managed by user programs.

<i>filedes = open(name, mode)</i>	Opens an existing file with the given <i>name</i> .
<i>filedes = creat(name, mode)</i>	Creates a new file with the given <i>name</i> .
	Both operations deliver a file descriptor referencing the open file. The <i>mode</i> is <i>read</i> , <i>write</i> or both.
<i>status = close(filedes)</i>	Closes the open file <i>filedes</i> .
<i>count = read(filedes, buffer, n)</i>	Transfers <i>n</i> bytes from the file referenced by <i>filedes</i> to <i>buffer</i> .
<i>count = write(filedes, buffer, n)</i>	Transfers <i>n</i> bytes to the file referenced by <i>filedes</i> from <i>buffer</i> .
	Both operations deliver the number of bytes actually transferred and advance the read-write pointer.
<i>pos = lseek(filedes, offset, whence)</i>	Moves the read-write pointer to offset (relative or absolute, depending on <i>whence</i>).
<i>status = unlink(name)</i>	Removes the file <i>name</i> from the directory structure. If the file has no other names, it is deleted.
<i>status = link(name1, name2)</i>	Adds a new name (<i>name2</i>) for a file (<i>name1</i>).
<i>status = stat(name, buffer)</i>	Gets the file attributes for file <i>name</i> into <i>buffer</i> .

These operations are implemented in the Unix kernel. These are operations available in the non-distributed case. Programs cannot server any discrepancies between cached copies and stored data after update. This is called **strict one copy semantics**.



The client module provides a single interface used by applications. It emulates traditional fs

Flat file service and dir. both provide an RPC interface.

Read(FileId, i, n) -> Data
throws *BadPosition*

If $1 \leq i \leq \text{Length}(\text{File})$: Reads a sequence of up to n items from a file starting at item i and returns it in *Data*.

Write(FileId, i, Data)
throws *BadPosition*

If $1 \leq i \leq \text{Length}(\text{File})+1$: Writes a sequence of *Data* to a file, starting at item i , extending the file if necessary.

Create() -> FileId

Creates a new file of length 0 and delivers a UFID for it.

Delete(FileId)

Removes the file from the file store.

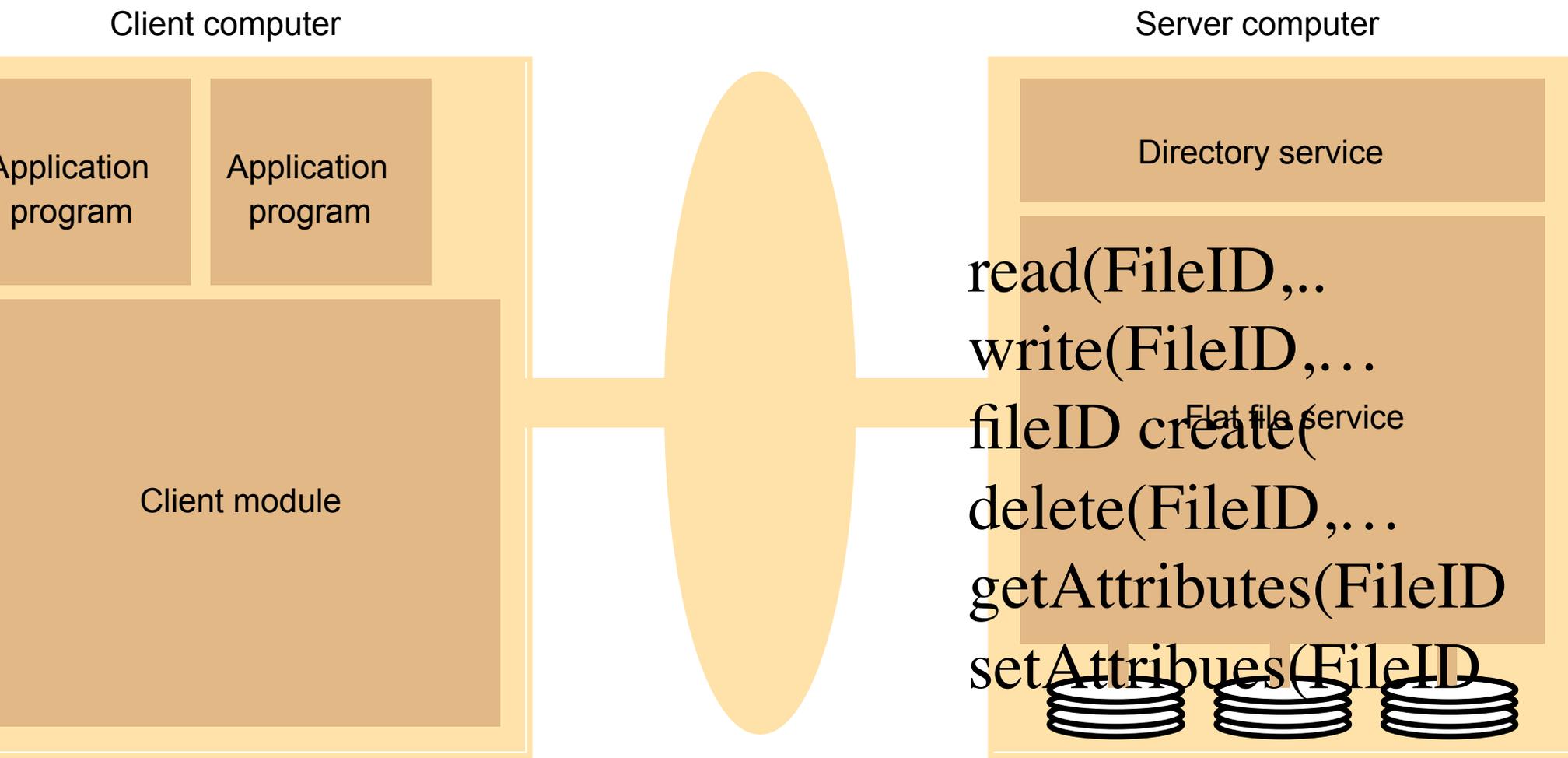
GetAttributes(FileId) -> Attr

Returns the file attributes for the file.

SetAttributes(FileId, Attr)

Sets the file attributes (only those attributes that are not shaded in Figure 12.3).

The client module will make calls on these operations and so will the directory service act as a client of the flat file service. Unique File Identifiers (UFID's) are passed in on all operations except create()



lookup(Dir, Name) -> FileId
- throws *NotFound*

Locates the text name in the directory and returns the relevant UFID. If *Name* is not in the directory, throws an exception.

addName(Dir, Name, FileId)
- throws *NameDuplicate*

If *Name* is not in the directory, adds (*Name, File*) to the directory and updates the file's attribute record.
If *Name* is already in the directory: throws an exception.

removeName(Dir, Name)
- throws *NotFound*

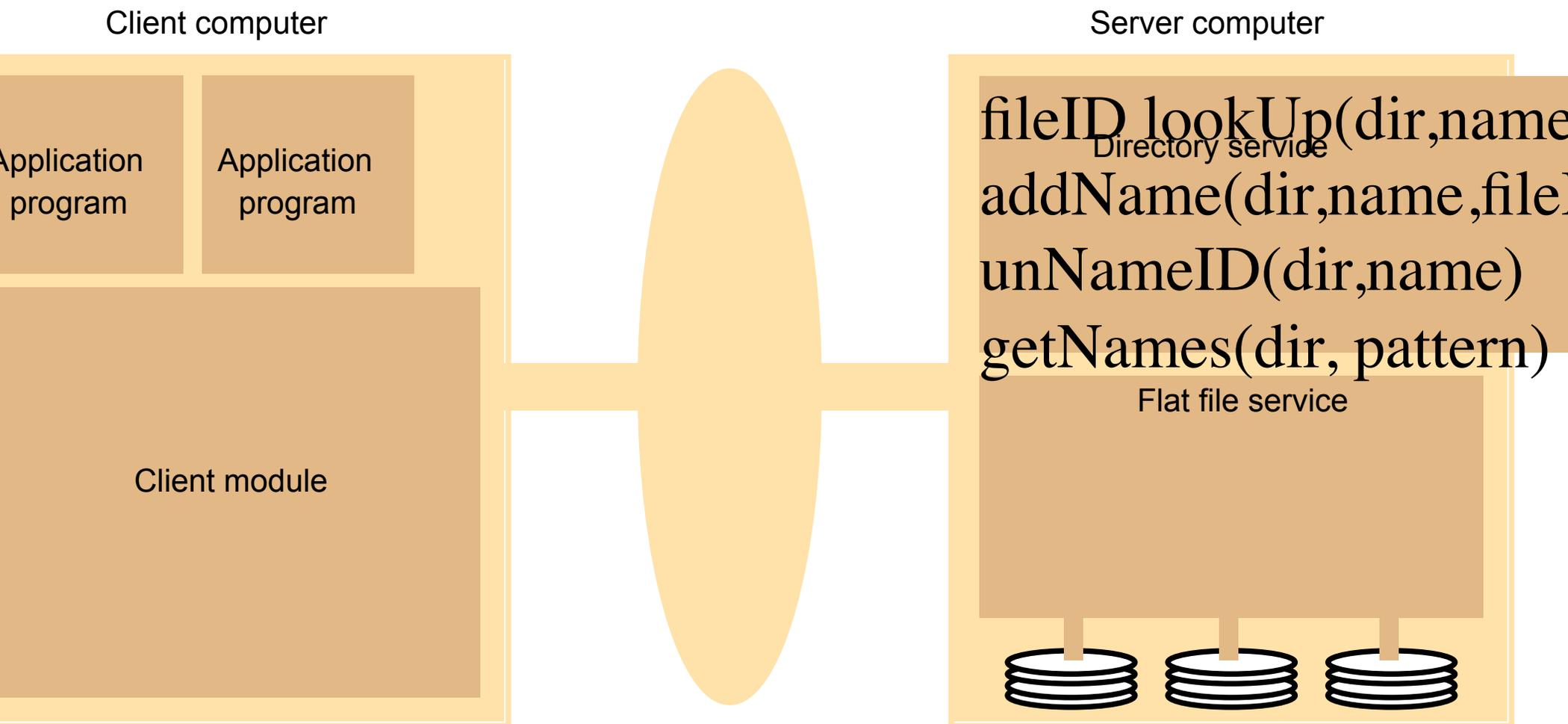
If *Name* is in the directory: the entry containing *Name* is removed from the directory.
If *Name* is not in the directory: throws an exception.

getNames(Dir, Pattern) -> NameSeq

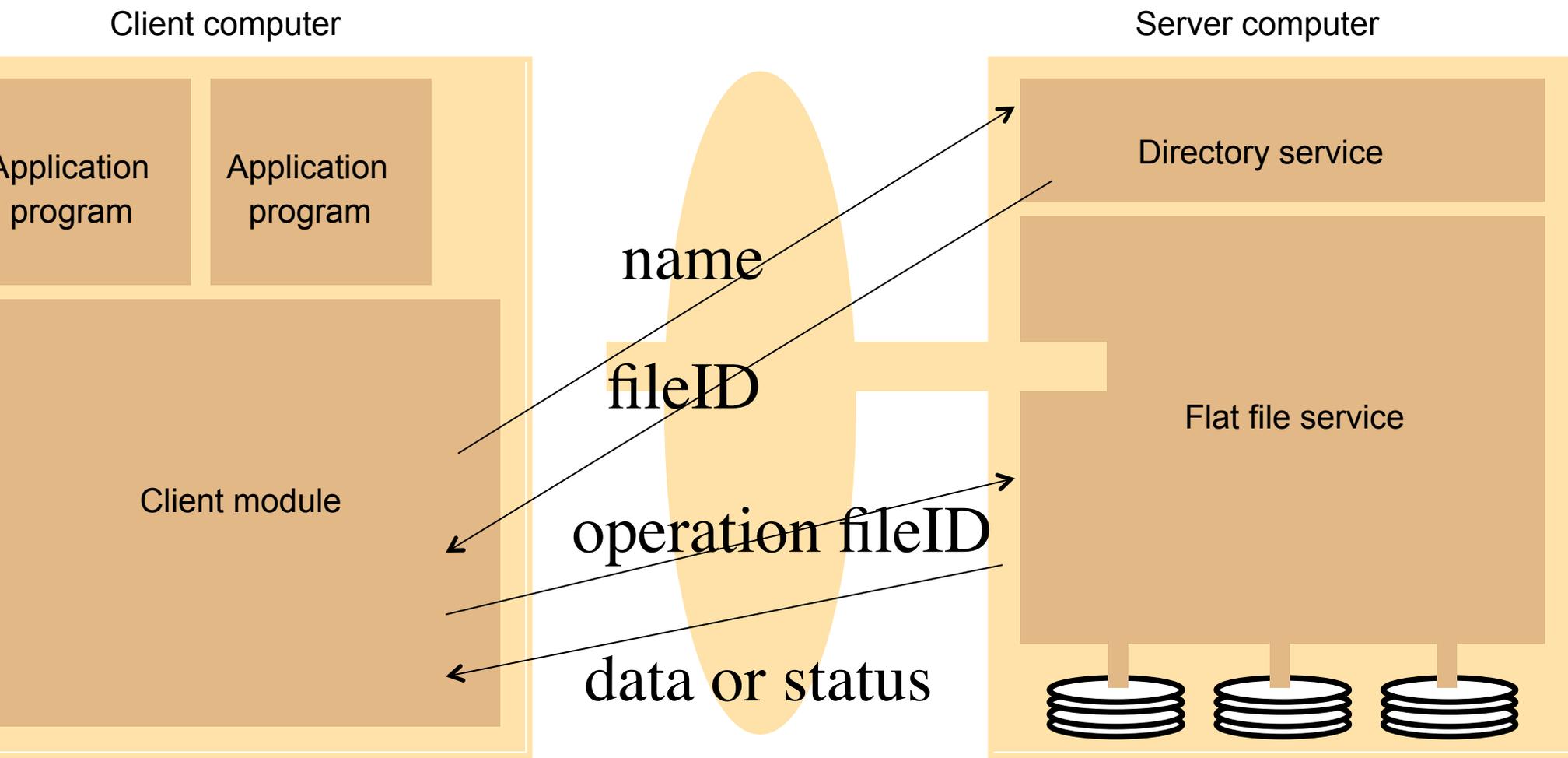
Returns all the text names in the directory that match the regular expression *Pattern*.

Primary purpose: translate text names to UFID's. Each directory is stored as a conventional file and so this is a client of the flat file service.

Once a flat file service and directory service is in place, it



have seen this pattern before.



l: Be unsurprising and look like a UNIX FS.

l: Implement full POSIX API. The **P**ortable **O**perating **S**ystem Interface is an IEEE family of standards that describe how Unix like Operating Systems should behave.

l: Your files are available from any machine.

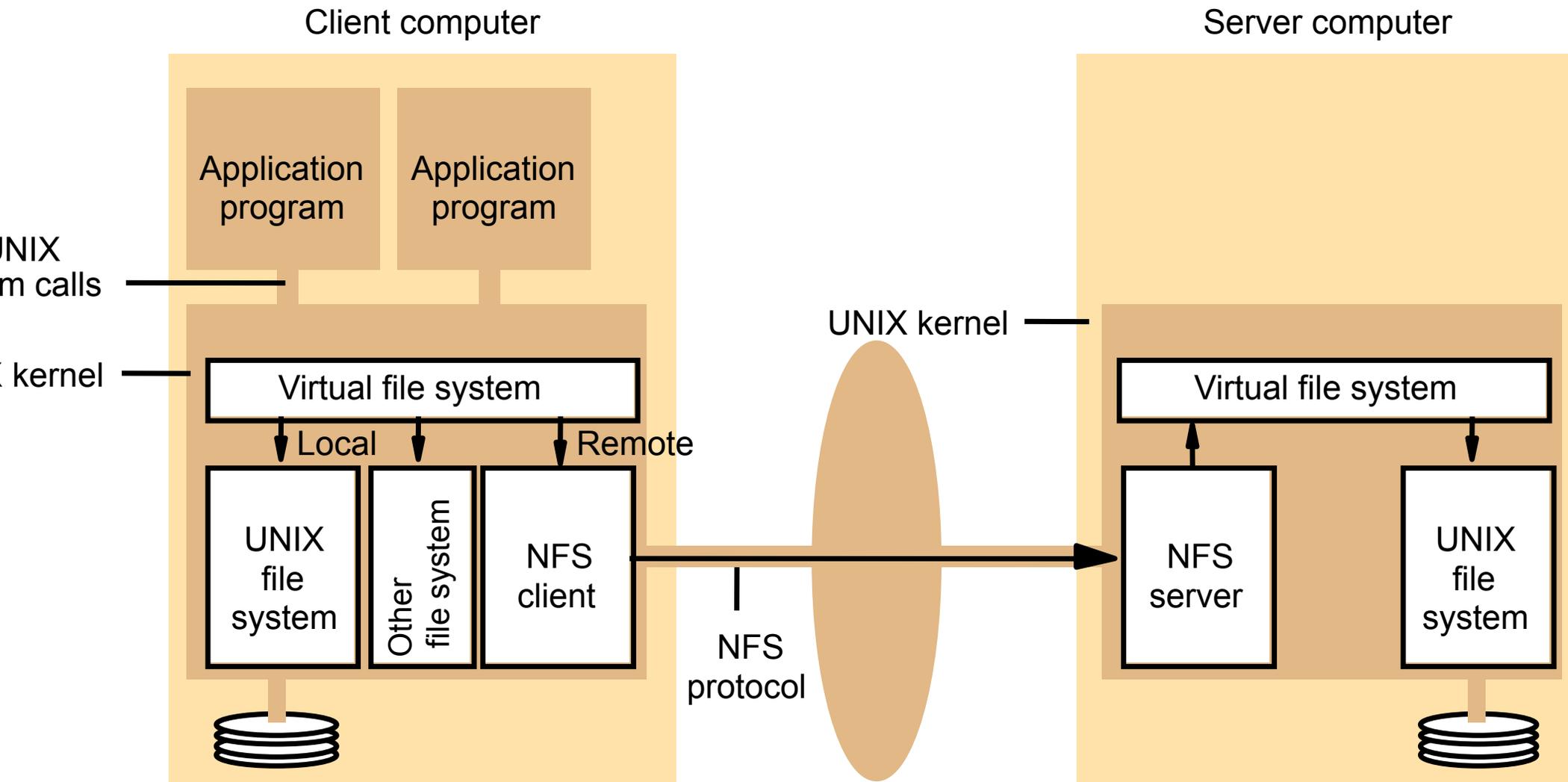
l: Distribute the files and we will not have to implement n protocols.

s has been a major success.

s was originally based on UDP and was stateless.

p added later.

e defines a virtual file system. The NFS client pretends to



NFS uses RPC over TCP or UDP.
External requests are translated into
RPC calls on the server. The virtual

<code>lookup(dirfh, name) -> fh, attr</code>	Returns file handle and attributes for the file <i>name</i> in the directory <i>dirfh</i> .
<code>create(dirfh, name, attr) -> newfh, attr</code>	Creates a new file name in directory <i>dirfh</i> with attributes <i>attr</i> and returns the new file handle and attributes.
<code>remove(dirfh, name) status</code>	Removes file name from directory <i>dirfh</i> .
<code>stat(fh) -> attr</code>	Returns file attributes of file <i>fh</i> . (Similar to the UNIX <i>stat</i> system call.)
<code>chmod(fh, attr) -> attr</code>	Sets the attributes (mode, user id, group id, size, access time and modify time of a file). Setting the size to 0 truncates the file.
<code>read(fh, offset, count) -> attr, data</code>	Returns up to <i>count</i> bytes of data from a file starting at <i>offset</i> . Also returns the latest attributes of the file.
<code>write(fh, offset, count, data) -> attr</code>	Writes <i>count</i> bytes of data to a file starting at <i>offset</i> . Returns the attributes of the file after the write has taken place.
<code>rename(dirfh, name, todirfh, toname) -> status</code>	Changes the name of file <i>name</i> in directory <i>dirfh</i> to <i>toname</i> in directory <i>todirfh</i>
<code>link(newdirfh, newname, dirfh, name) -> status</code>	Creates an entry <i>newname</i> in the directory <i>newdirfh</i> which refers to file <i>name</i> in the directory <i>dirfh</i> .

link(newdirfh, newname, string)
-> *status*

Creates an entry *newname* in the directory *newdirfh* of type symbolic link with the value *string*. The server does not interpret the *string* but makes a symbolic link file to hold it.

link(fh) -> *string*

Returns the string that is associated with the symbolic link file identified by *fh*.

mkdir(dirfh, name, attr) ->
newfh, attr

Creates a new directory *name* with attributes *attr* and returns the new file handle and attributes.

rmdir(dirfh, name) -> *status*

Removes the empty directory *name* from the parent directory *dirfh*. Fails if the directory is not empty.

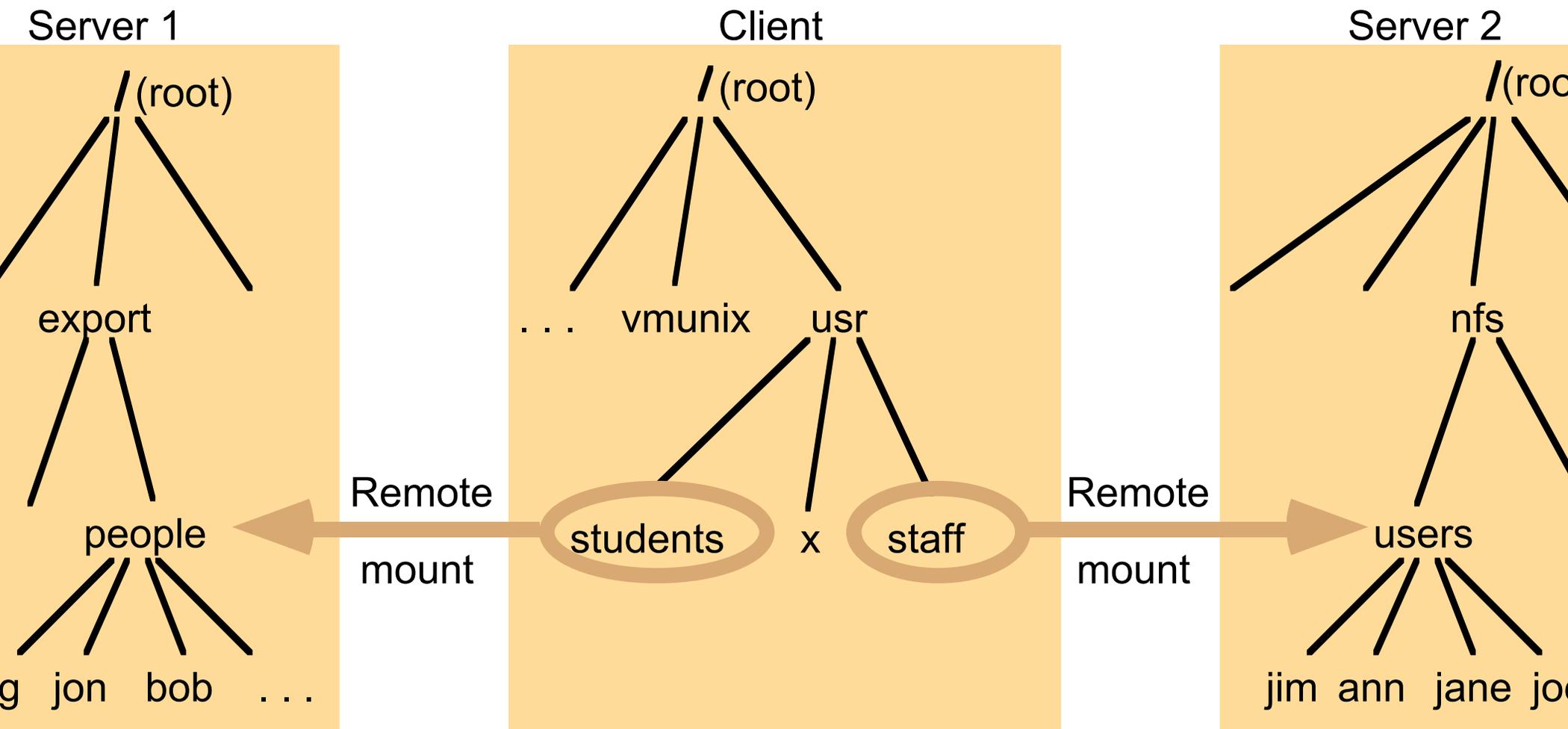
readdir(dirfh, cookie, count) ->
entries

Returns up to *count* bytes of directory entries from the directory *dirfh*. Each entry contains a file name, a file handle, and an opaque pointer to the next directory entry, called a *cookie*. The *cookie* is used in subsequent *readdir* calls to start reading from the following entry. If the value of *cookie* is 0, reads from the first entry in the directory.

fsstat(fh) -> *fsstats*

Returns file system information (such as block size, number of free blocks and so on) for the file system containing a file *fh*.

and remote file systems accessible on an NFS client



Note: The file system mounted at `/usr/students` in the client is actually the sub-tree located at `export/people` in Server 1; the file system mounted at `/usr/staff` in the client is actually

Like NFS, the most important design goal is scalability.

To achieve scalability, whole files are cached in client nodes. Why does this help with scalability?

It reduces client server interactions.

The client cache would typically hold several hundred files most recently used on that computer.

The cache is permanent, surviving reboots.

When the client opens a file, the cache is examined and used if the file is available there.

When the client code tries to open a file the client cache is tried first. If not there, a server is located and the server is called for the file.

A copy is stored on the client side and is opened.

Subsequent reads and writes hit the copy on the client.

When the client closes the file - if the file has changed it is sent back to the server. The client side copy is retained for possible more use.

Consider UNIX commands and libraries copied to the client.

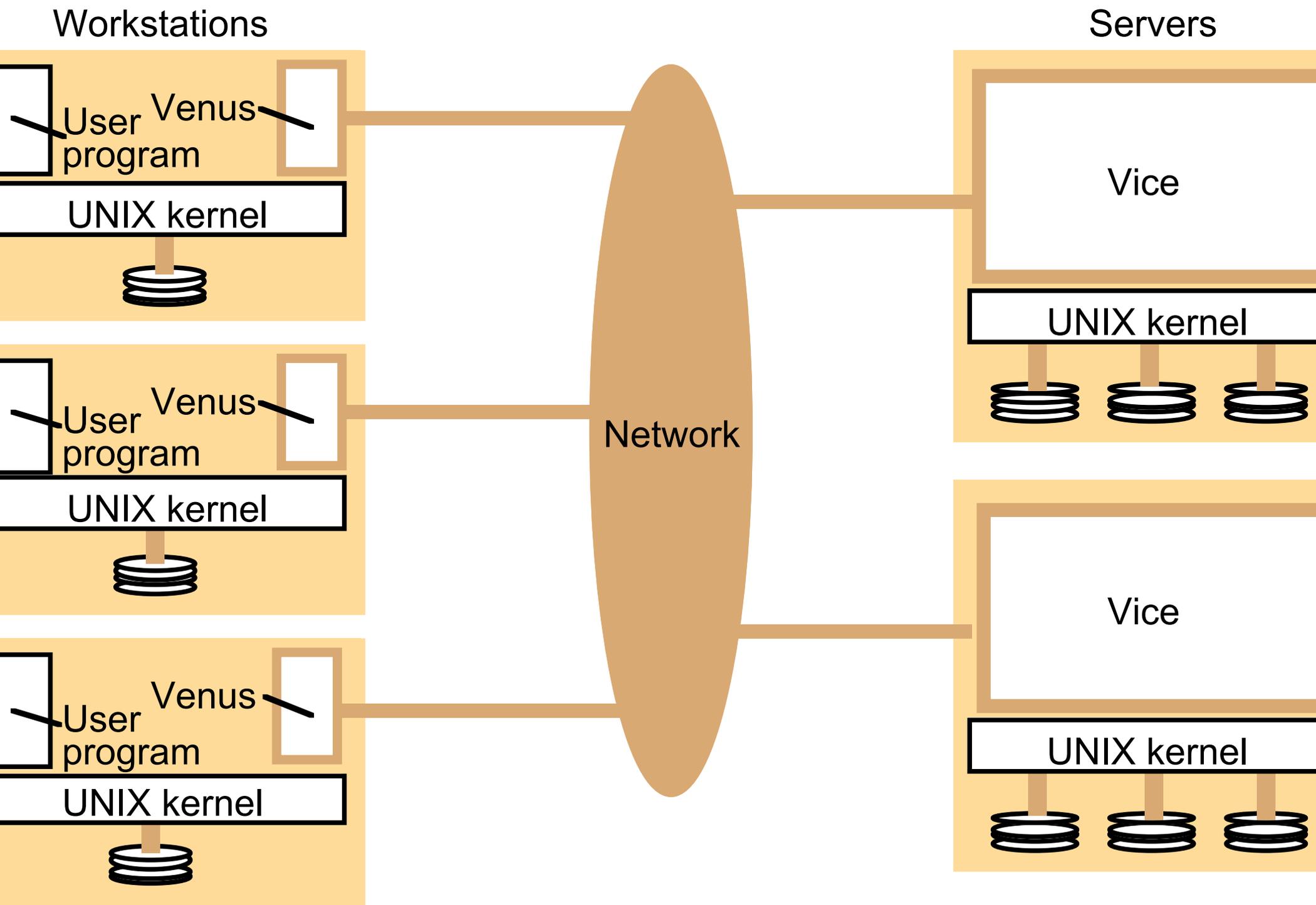
Consider files only used by a single user.

The last two cases represent the vast majority of cases.

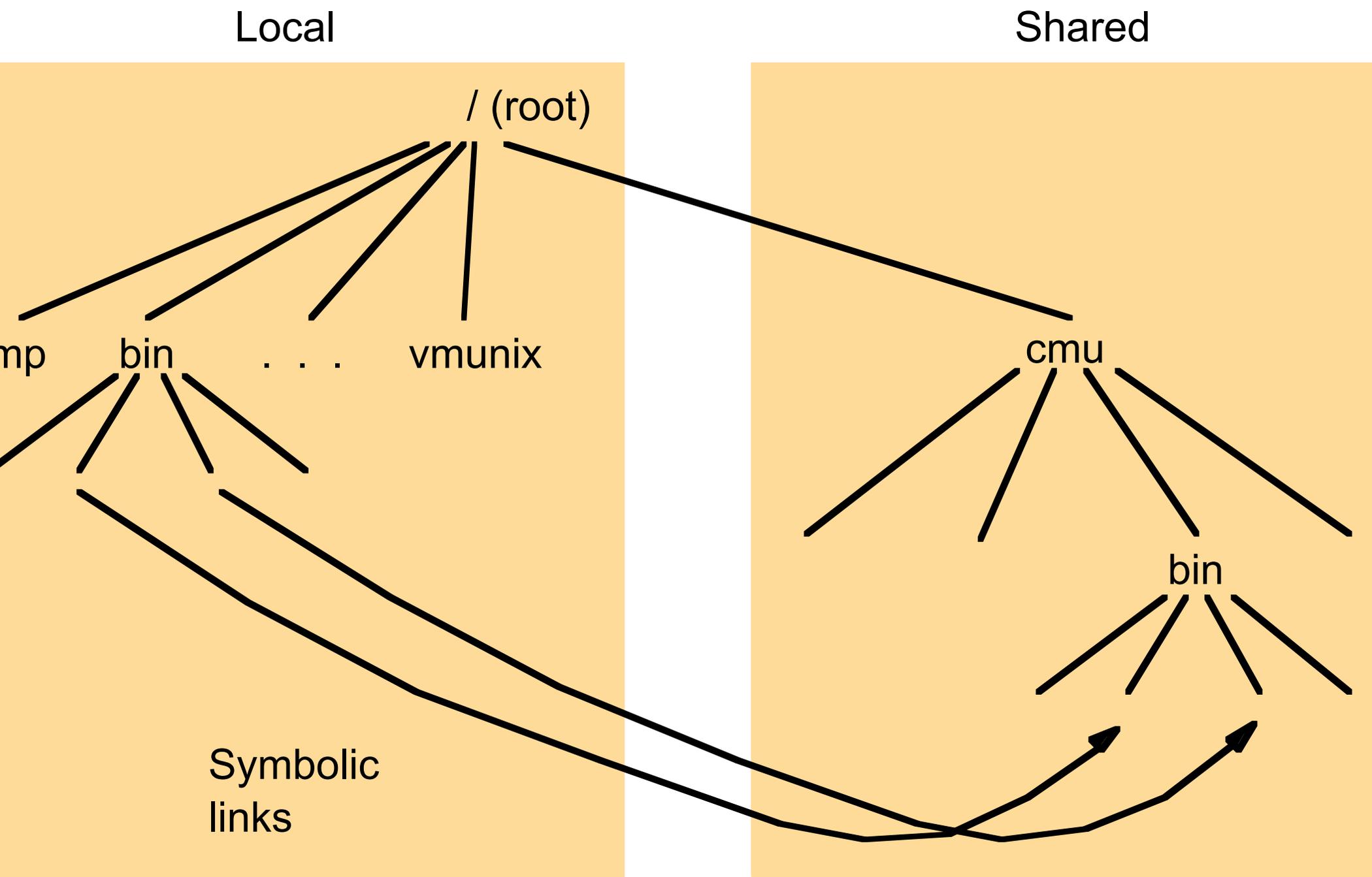
Goal: Your files are available from any workstation.

Principle: Make the common case fast. See Amdahl's Law.

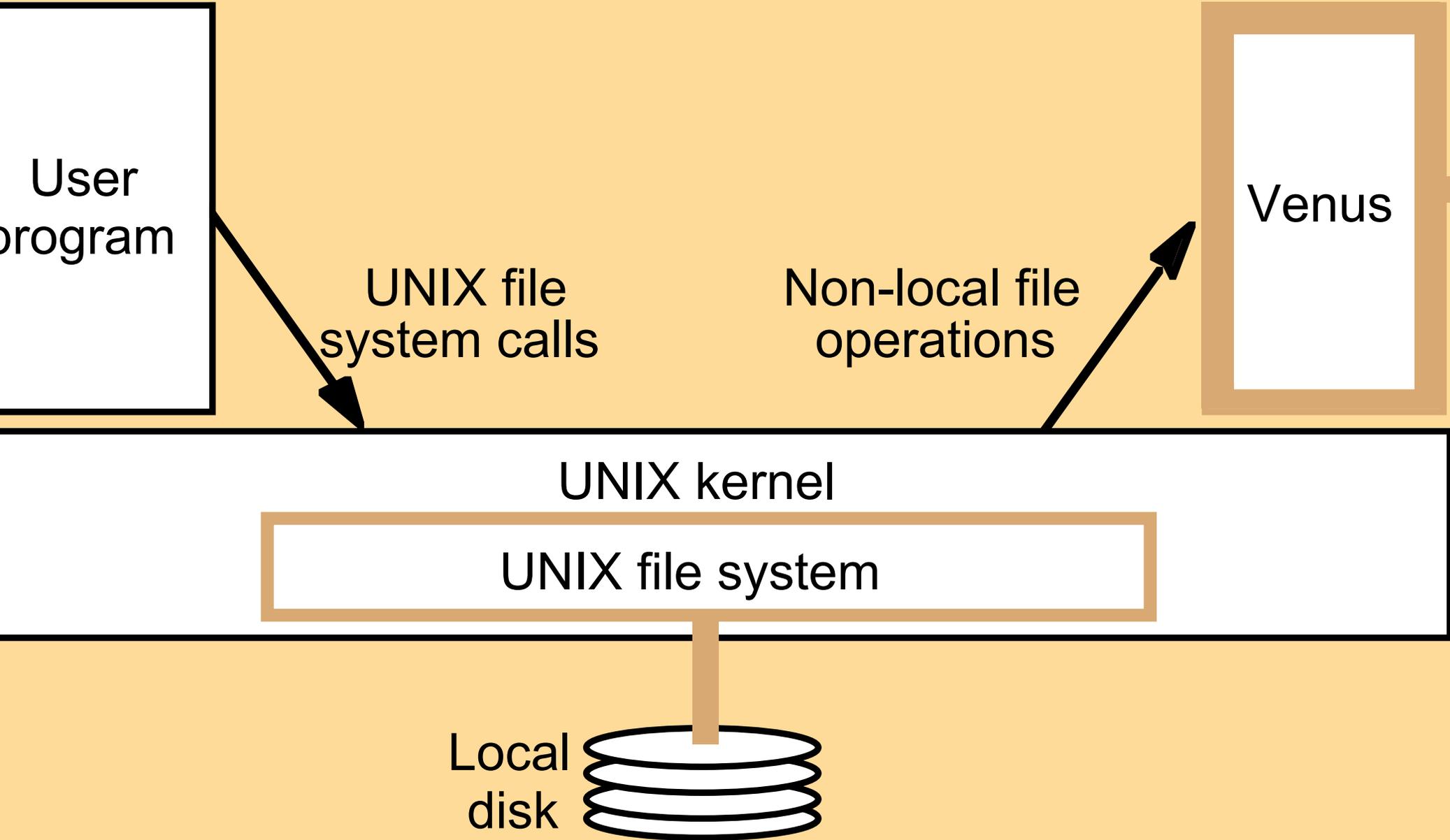
Distribution of processes in the Andrew File System



name space seen by clients of AFS



Workstation



Implementation of file system calls in AFS

<i>User process</i>	<i>UNIX kernel</i>	<i>Venus</i>	<i>Net</i>	<i>Vice</i>
<i>open(FileName, mode)</i>	<p>If <i>FileName</i> refers to a file in shared file space, pass the request to Venus.</p> <p>Open the local file and return the file descriptor to the application.</p>	<p>Check list of files in local cache. If not present or there is no valid <i>callback promise</i>, send a request for the file to the Vice server that is custodian of the volume containing the file.</p> <p>Place the copy of the file in the local file system, enter its local name in the local cache list and return the local name to UNIX.</p>		<p>Transfer a copy of the file and a <i>callback promise</i> to the workstation. Log the callback promise.</p>
<i>read(FileDescriptor, Buffer, length)</i>	Perform a normal UNIX read operation on the local copy.			
<i>write(FileDescriptor, Buffer, length)</i>	Perform a normal UNIX write operation on the local copy.			
<i>close(FileDescriptor)</i>	Close the local copy and notify Venus that the file has been closed.	<p>If the local copy has been changed, send a copy to the Vice server that is the custodian of the file.</p>		<p>Replace the file contents and send a <i>callback</i> to all other clients holding <i>callback promises</i> on the file.</p>

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main components of the Vice service Interface

<i>h(fid) -> attr, data</i>	Returns the attributes (status) and, optionally, the contents of file identified by the <i>fid</i> and records a callback promise on it.
<i>e(fid, attr, data)</i>	Updates the attributes and (optionally) the contents of a specified file.
<i>te() -> fid</i>	Creates a new file and records a callback promise on it.
<i>ove(fid)</i>	Deletes the specified file.
<i>ock(fid, mode)</i>	Sets a lock on the specified file or directory. The mode of the lock may be shared or exclusive. Locks that are not removed expire after 30 minutes.
<i>aseLock(fid)</i>	Unlocks the specified file or directory.
<i>oveCallback(fid)</i>	Informs server that a Venus process has flushed a file from its cache.
<i>kCallback(fid)</i>	This call is made by a Vice server to a Venus process. It cancels the callback promise on the relevant file.

What is Hadoop?

Sort of the opposite of virtual machines where one machine may act like many. Instead, with Hadoop many machines act as one.

Hadoop is an open source implementation of GFS

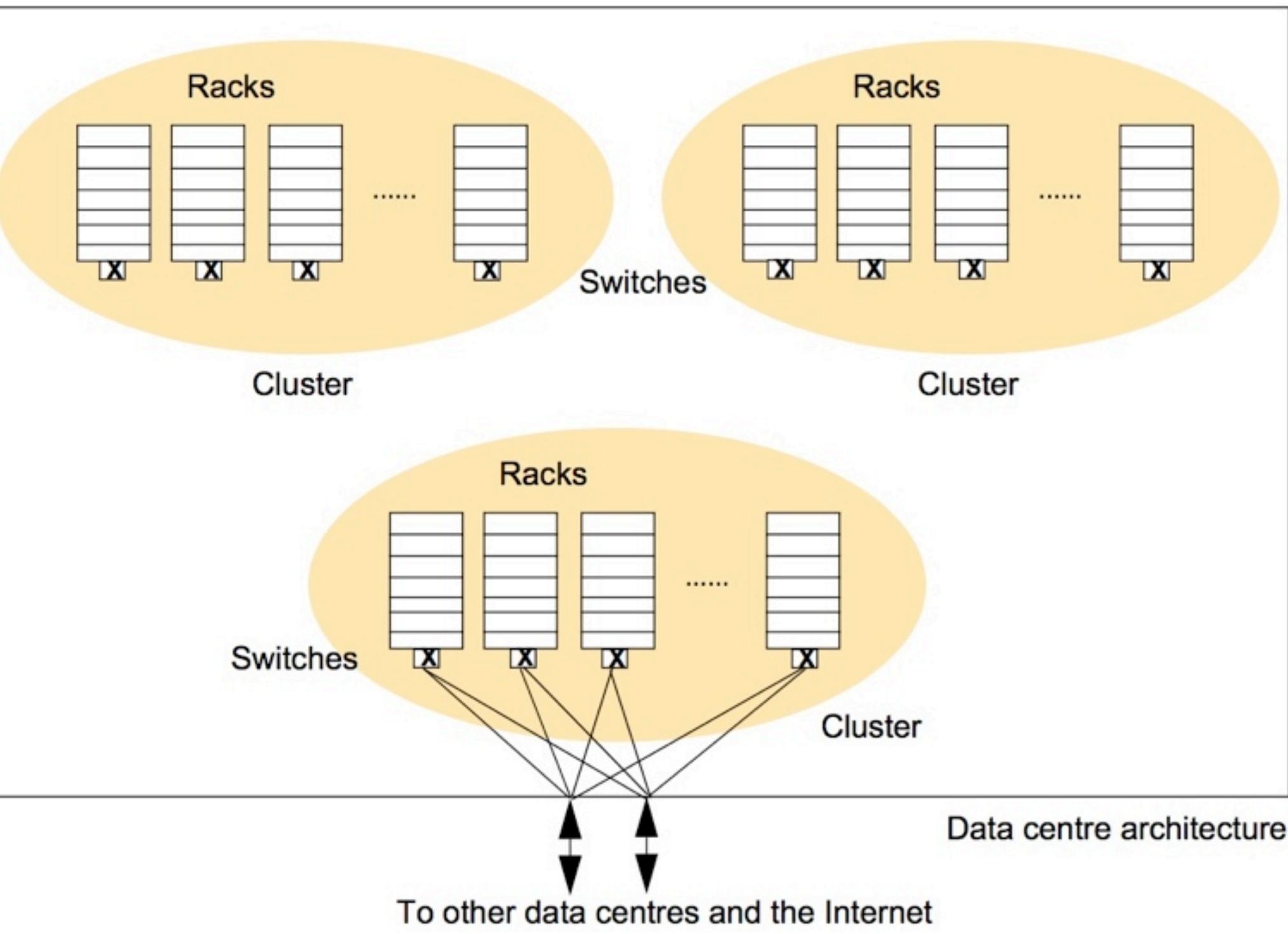
Microsoft has Dryad with similar goals.

At its core, an operating system (like Hadoop) is all about:

a) storing files

b) running applications on top of files

Organization of the Google physical Infrastructure



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Work reliably with component failures.

Have problems that Google needs solved – not a massive number of files but massively large files are common.

Workload is dominated by long sequential streaming reads and sequential appends. No need for caching on the client.

Throughput more important than latency.

Think of very large files each holding a very large number of HTML documents scanned from the web. These need read and analyzed.

Not your everyday use of a distributed file

Each file is mapped to a set of fixed size chunks.

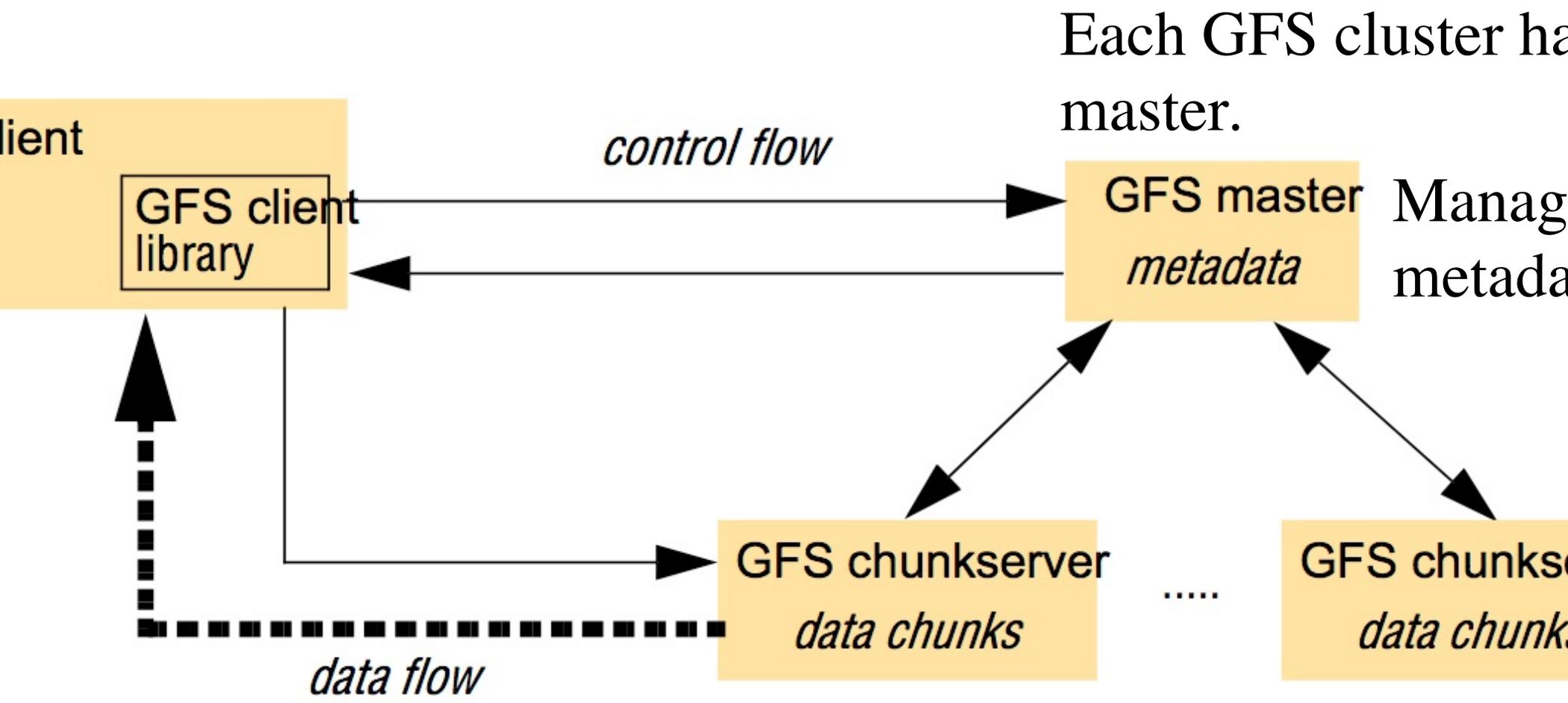
Each chunk is 64Mb in size.

Each cluster has a single master and multiple (usually hundreds) of chunk servers.

Each chunk is replicated on three different chunk servers.

The master knows the locations of chunk replicas.

The chunk servers know what replicas they have and are polled by the master on startup.



Each GFS cluster has a master.

Manages metadata

- Hundreds of chunkservers
- Data is replicated on three independent chunkservers
- Locations known by master.
- With log files, the master is restorable after failure

Reading a file sequentially

Suppose a client wants to perform a sequential read by processing a very large file from a particular byte offset.

The client can compute the chunk index from the byte offset.

Client calls master with file name and chunk index.

Master returns chunk identifier and the locations of replicas.

Client makes call on a chunk server for the chunk and it is processed sequentially with no caching. Client may ask for and receive several chunks.

◉ Mutation Operations

Suppose a client wants to perform sequential writes at the end of a file.

The client can compute the chunk index from the byte offset. This is the chunk holding End Of File.

Client calls master with file name and chunk index.

Master returns chunk identifier and the locations of replicas. One is designated as the primary.

The client sends all data to all replicas. The primary coordinates with replicas to update files consistently across replicas.

MapReduce Runs on Hadoop

Provide a clean abstraction on top of parallelization and fault tolerance.

Easy to program. The parallelization and fault tolerance is automatic.

Programmer implements two interfaces: one for mappers and one for reducers.

Map takes records from source in the form of key value pairs.

Map produces one or more intermediate values along with an output key from the input.

When Map is complete, all of the intermediate values for a given output key are combined into a list. The combiners in the mapper machines

duce combines the intermediate values into one or more final values for the same output key (usually one final value per key)

the master tries to place the mapper on the same machines as the data or nearby.

Map function, written by the user, takes an input pair and produces a **set of output key/value** pairs.

The MapReduce library groups together all intermediate values associated with the key k and passes them to the reduce function.

The Reduce function, also written by the user, accepts an intermediate key k and a set of values for that key. It merges together these values to form a possibly smaller set of values. Typically, just zero or one output value is produced per Reduce invocation.

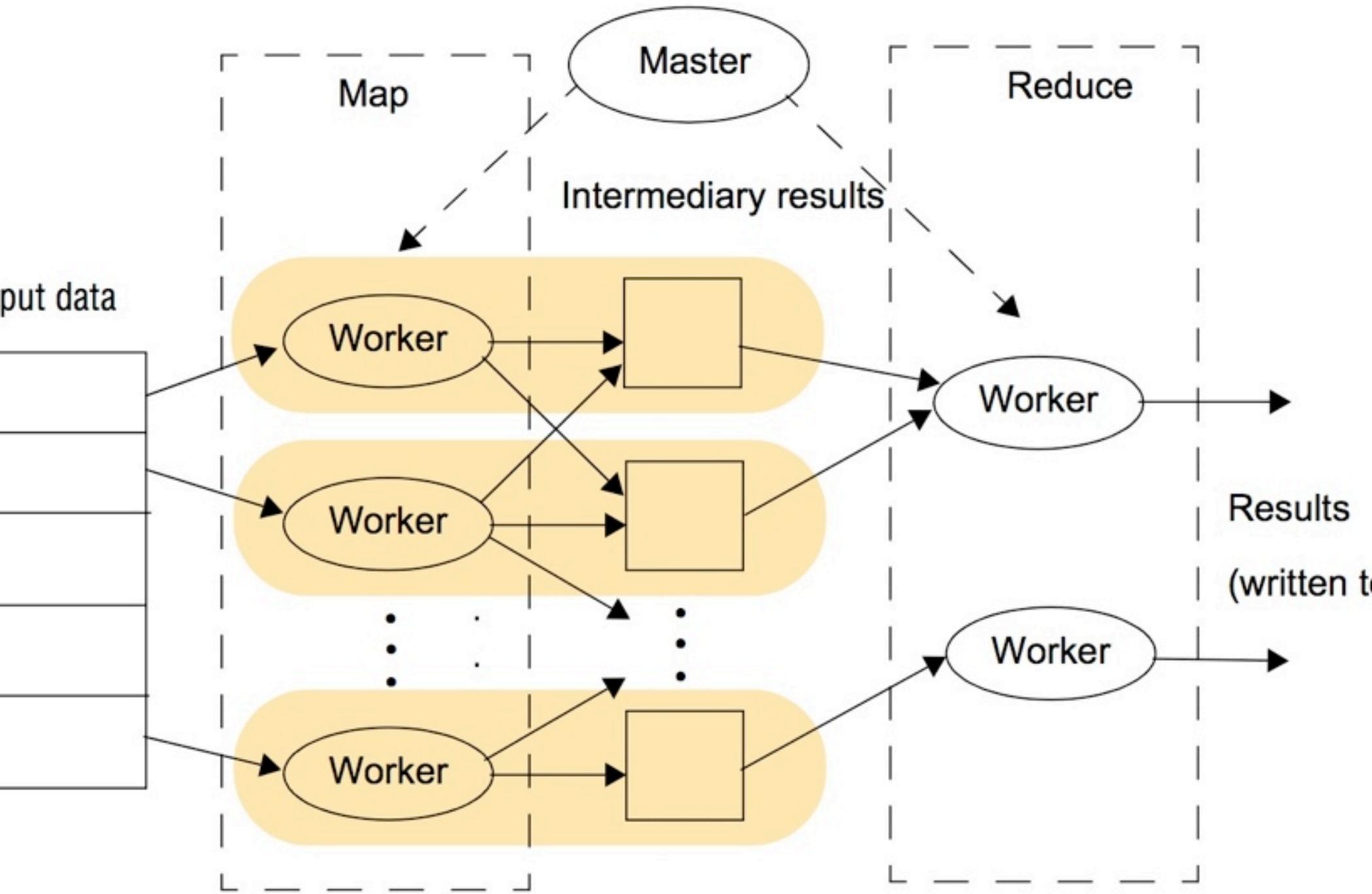
o $(k1, v1) \rightarrow list(k2, v2)$

duce $(k2, list(v2)) \rightarrow list(v2)$

Examples of the use of MapReduce

Function	Initial step	Map phase	Intermediate step	Reduce phase
Word count	Partition data into fixed-size chunks for processing	For each occurrence of word in data partition, emit $\langle \text{word}, 1 \rangle$	Merge/sort all key-value keys according to their intermediary key	For each word in the intermediate set, count the number of 1s
This heavily intermediate		Output a line if it matches a given pattern		Null
ted		For each entry in the input data, output the key-value pairs to be sorted		Null
		Parse the associated documents and output a $\langle \text{word}, \text{document ID} \rangle$ pair wherever that word exists		For each word, produce a list (sorted) of document IDs

Overall execution of a MapReduce program



Map Execution of MapReduce

Maps run on the input data scattered over n machines:

on Disk 1 => (key,value) => map₁

on Disk 2 => (key,value) => map₂

on Disk n => (key,value) => map_n

map tasks produce (key, value) pairs:

=> (key 1, value)

(key 2, value)

=> (key 1, value)

(key 2, value)

(key 3, value)

(key 1, value)

Output of each map task is collected and sorted on the key. These key, value pairs

are passed to the reducers:

(key, value list) => reducer1 => list(value)

(key, value list) => reducer2 => list(value)

Maps run in parallel.

Reducers run in parallel.

Map phase must be completed before the reduce phase can begin.

The combiner phase is run on mapper nodes after map phase.

This is a mini-reduce on local map output.

For complex activities, combining the output of a reducer

Reduce to Count Word Occurrences In Many Documents

=> (Document name, Document) => map₁ On machine near disk 1

=> (Document name, Document) => map₂ On machine near disk 2

=> (Document name, Document) => map_n

=> (ball, 1)

(game, 1)

=> (ball, 1)

(team, 1)

(ball, 1)

map output and sort by key. Send these pairs to reducers.

(ball, 1, 1) => reducer => (ball, 3)

(game, 1) => reducer => (game, 1)

(team, 1) => reducer => (team, 1)

The MapReduce Examples

Count the number of occurrences of each word in a large collection of documents.

Distributed GREP: Count the number of lines with a particular pattern.

From a web server log, determine URL access frequency.

Reverse a web link graph. For a given URL, find URL's of pages pointing to it.

For each word, create list of documents containing it. (Same as 4.)

Distributed sort of a lot of records with keys

MapReduce Example (1)

Count the number of occurrences of each word in a collection of documents.

```
// (K1,V1) → List(K2,V2)
```

```
map(String key, String value)
```

```
key: document name
```

```
value: document contents
```

```
for each word w in value
```

```
emitIntermediate(w,"1")
```

Doc1

car

bell

Doc2

car

(car,1),(bell,1),(car,1)

```
// (K2, List(V2)) → List(V2)
```

```
reduce(String key, Iterator values)
```

```
key: a word
```

```
values: a list of counts
```

```
result = 0
```

(bell,[1]), (car,[1,1])

MapReduce Example (2)

Distributed GREP: Count the number of lines with a particular pattern. Suppose searchString is "th".

$(K1, V1) \rightarrow List(K2, V2)$

$(fileOffset, lineFromFile)$

if searchString in lineFromFile

emitIntermediate(lineFromFile, 1)

$(K2, List(V2)) \rightarrow List(V2)$

for each $(k2, \text{iterator values})$

sum = sum up values

emit (sum, k2)

(0, the line) (8, a line) (14, the

(22, the line)

(the line, 1), (the store, 1), (the

(the line, [1,1]), (the store, [1])

(2 the line), (1 the store)

From a web server log, determine URL access frequency.

Page request log:

URL1 was visited

(0,URL1),(45,URL1),(90,URL2),(135,URL1)

URL1 was visited

URL2 was visited

URL1 was visited

(URL1,1),(URL1,1),(URL2,1),(URL1,1)

Map(URL1,1) → List(K2,V2)

Map(URL2,1)

(URL1, [1,1,1]), (URL2, [1])

Reduce(List(K2,V2))

Reduce(List(V2)) → List(V2)

Map(url, values)

(URL1, 3),(URL2,1)

Map(values into total)

Reverse a web link graph. For a given URL, find URL's of pages pointing to it.

$(K1) \rightarrow List(K2, V2)$ $(URL1, \{P1, P2, P3\})$ $(URL2, \{P1, P2, P3\})$

String SourceDocURL, sourceDoc) $(P1, URL1), (P2, URL1), (P3, URL1)$
for each target in the document $(P1, URL2), (P3, URL2)$
emitIntermediate(target, SourceDocURL) $(P1, URL2), (P3, URL2)$

$List(V2) \rightarrow List(V2)$ $(P1, (URL1, URL2)), (P2, (URL1, URL2)), (P3, (URL1, URL2))$
emit(target, listOfSourceURL's)
emit(target, listOfSourceURL's)

Distributed sort of a lot of records with keys.

$(K1, V1) \rightarrow \text{List}(K2, V2)$

$(0, k2, \text{data}), (20, k1, \text{data}), (30, k3, \text{data})$

$(\text{offset}, \text{record})$

$\text{sk} = \text{find sort key in record}$

$\text{emitIntermediate}(\text{sk}, \text{record})$

$(k2, \text{data}), (k1, \text{data}), (k3, \text{data})$

$(\text{List}(V2)) \rightarrow \text{List}(V2)$

$(k1, \text{data}), (k2, \text{data}), (k3, \text{data})$

Reducer emits records unchanged

Map Example 1: Word Count

Count the number of occurrences of each word in a collection of documents.

```
// (K1,V1) → List(K2,V2)
```

```
map(String key, String value)
```

```
key: document name
```

```
value: document contents
```

```
for each word w in value
```

```
emitIntermediate(w,"1")
```

Doc1

car

bell

Doc2

car

(car,1),(bell,1),(car,1)

```
=====  
// (K2, List(V2)) → List(V2)
```

```
reduce(String key, Iterator values)
```

```
key: a word
```

```
values: a list of counts
```

```
result = 0
```

(bell,[1]), (car,[1,1])

```
public static class MapClass extends MapReduceBase
implements Mapper<LongWritable, Text, Text, IntWritable> {

    private final static IntWritable one = new IntWritable(1);
    private Text word = new Text();

    public void map(LongWritable key, Text value,
        OutputCollector<Text, IntWritable> output,
        Reporter reporter) throws IOException {
        String line = value.toString();
        StringTokenizer itr = new StringTokenizer(line);
        while (itr.hasMoreTokens()) {
            word.set(itr.nextToken());
            output.collect(word, one);
        }
    }
}
```

```
public static class Reduce extends MapReduceBase
implements Reducer<Text, IntWritable, Text, IntWritable> {

    public void reduce(Text key, Iterator<IntWritable> values,
                      OutputCollector<Text, IntWritable> output,
                      Reporter reporter) throws IOException {
        int sum = 0;
        while (values.hasNext()) {
            sum += values.next().get();
        }
        output.collect(key, new IntWritable(sum));
    }
}
```

How do you think of an embarrassingly parallel approach to approximating the value of π ?

1000 monkeys to each throw one thousand darts at 1000 square 1 X 1 boards, all with inscribed circles.

Let N be the number of darts landing inside the circles and M those landing outside. Compute the area $A = \pi r^2 = \pi (1/2)^2 = 1/4 \pi$.

Then $A = \pi r^2 = \pi (1/2)^2 = 1/4 \pi$.

$N/M = 4A$.

