14. Practical Concurrency

We begin our study of concurrency by describing how to use it in practice; later, in handout 17 on formal concurrency, we shall study it more formally. First we explain where the concurrency in a system comes from, and discuss the main ways to express concurrency. Then we describe the difference between 'hard' and 'easy' concurrency: the latter is done by locking shared data before you touch it, the former in subtle ways that are so error-prone that simple prudence requires correctness proofs. We give the rules for easy concurrency using locks, and discuss various issues that complicate the easy life: scheduling, locking granularity, and deadlocks.

Sources of concurrency

Before studying concurrency in detail, it seems useful to consider how you might get concurrency in your system. Obviously if you have a multiprocessor or a distributed system you will have concurrency, since in these systems there is more than one CPU executing instructions. Similarly, most hardware has separate parts that can change state simultaneously and independently. But suppose your system consists of a single CPU running a program. Then you can certainly arrange for concurrency by multiplexing that CPU among several tasks, but why would you want to do this? Since the CPU can only execute one instruction at a time, it isn’t entirely obvious that there is any advantage to concurrency. Why not get one task done before moving on to the next one?

There are only two possible reasons:

1. A task might have to wait for something else to complete before it can proceed, for instance for a disk read. But this means that there is some concurrent task that is going to complete, in the example an I/O device, the disk. So we have concurrency in any system that has I/O, even when there is only one CPU.

2. Something else might have to wait for the result of one task but not for the rest of the computation, for example a human user. But this means that there is some concurrent task that is waiting, in the example the user. Again we have concurrency in any system that has I/O.

In the first case one task must wait for I/O, and we can get more work done by running another task on the CPU, rather than letting it idle during the wait. Thus the concurrency of the I/O system leads to concurrency on the CPU. If the I/O wait is explicit in the program, the programmer can know when other tasks might run; this is often called a ‘non-preemptive’ system, because it has sequential semantics except when the program explicitly allows concurrent activity by waiting. But if the I/O is done at some low level of abstraction, higher levels may be quite unaware of it. The most insidious example of this is I/O caused by the virtual memory system: every instruction can cause a disk read. Such a system is called ‘preemptive’;

for practical purposes a task can lose the CPU at any point, since it’s too hard to predict which memory references might cause page faults.

In the second case we have a motivation for true preemption: we want some tasks to have higher priority for the CPU than others. An important special case is interrupts, discussed below.

A concurrent program is harder to write than a sequential program, since there are many more possible paths of execution and interactions among the parts of the program. The canonical example is two concurrent executions of

\[ x := x + 1 \]

Since this command is not atomic (either in Spec, or in C on most computers), \( x \) can end up with either 1 or 2, depending on the order of execution of the expression evaluations and the assignments. The interleaved order

\[
\begin{align*}
\text{evaluate } x + 1 \\
\text{evaluate } x + 1 \\
x := \text{result} \\
x := \text{result}
\end{align*}
\]

leaves \( x = 1 \), while doing both steps of one command before either step of the other leaves \( x = 2 \).

Since concurrent programs are harder to understand, it’s best to avoid concurrency unless you really needed it for one of the reasons just discussed.\(^2\)

One good thing about concurrency, on the other hand, is that when you write a program as a set of concurrent computations, you can defer decisions about exactly how to schedule them.

Ways to package concurrency

In the last section we used the word ‘task’ informally to describe a more-or-less independent, more-or-less sequential part of a computation. Now we shall be less coy about how concurrency shows up in a system.

The most general way to describe a concurrent system is in terms of a set of atomic actions with the property that usually more than one of them can occur (is enabled); we will use this viewpoint in our later study of formal concurrency. In practice, however, we usually think in terms of several ‘threads’ of concurrent execution. Within a single thread at most one action is enabled at a time; in general one action may be enabled from each thread, though often some of the threads are waiting or ‘blocked’, that is, have no enabled actions.

The most convenient way to do concurrent programming is in a system that allows each thread to be described as an execution path in an ordinary-looking program with modules, routines, commands, etc., such as Spec, C, or Java. In this scheme more than one thread can execute the code of the same procedure; threads have local state that is the local variables of the procedures

\(^2\) This is the main reason why threads with RPC or synchronous messages are good, and asynchronous messages are bad. The latter force you to have concurrency whenever you have communication, while the former let you put in the concurrency just where you really need it. Of course if the implementation of threads is clumsy or expensive, as it often is, that may overwhelm the inherent advantages.
they are executing. All the languages mentioned and many others allow you to program in this way.

In fault-tolerant systems there is a conceptual drawback to this thread model. If a failure can occur after each atomic command, it is hard to understand the program by following the sequential flow of control in a thread, because there are so many other paths that result from failure and recovery. In these systems it is often best to reason strictly in terms of independent atomic actions. We will see detailed examples of this when we study reliable messages, consensus, and replication. Applications programmed in a transaction system are another example of this approach: each application runs in response to some input and is a single atomic action.

The biggest drawback of this kind of ‘official’ thread, however, is the costs of representing the local state and call stack of each thread and of a general mechanism for scheduling the threads. There are several alternatives that reduce these costs: interrupts, control blocks, and SIMD computers. They are all based on restricting the freedom of a thread to block, that is, to yield the processor until some external condition is satisfied, for example, until there is space in a buffer or a lock is free, or a page fault has been processed.

Interrupts
An interrupt routine is not the same as a thread, because:

- It always starts at the same point.
- It cannot wait for another thread.

The reason for these restrictions is that the execution context for an interrupt routine is allocated on someone else’s stack, which means that the routine must complete before the thread that it interrupted can continue to run. On the other hand, the hardware that schedules an interrupt routine is efficient and takes account of priority within certain limits. In addition, the interrupt routine doesn’t pay the cost of its own stack like an ordinary thread.

It’s possible to have a hybrid system in which an interrupt routine that needs to wait turns itself into an ordinary thread by copying its state. This is tricky if the wait happens in a subroutine of the main interrupt routine, since the relevant state may be spread across several stack frames. If the copying doesn’t happen too often, the interrupt-thread hybrid is efficient. The main drawbacks are that the copying usually has to be done by hand, which is error-prone, and that without compiler and runtime support it’s not possible to reconstruct the call stack, which means that the thread has to be structured differently from the interrupt routine.

A simpler strategy that is widely used is to limit the work in the interrupt routine to simple things that don’t require waits, and to wake up a separate thread to do anything more complicated.

Control blocks and message queues
Another, related strategy is to package all the permanent state of a thread, including its program counter, in a record (usually called a ‘control block’) and to explicitly schedule the execution of the threads. When a thread runs, it starts at the saved program counter (usually a procedure entry point) and runs until it explicitly gives up control or ‘yields’. During execution it can call procedures, but when it yields its stack must be empty so that there’s no need to save it, because all the state has to be in the control block. When it yields, a reference to the control block is saved where some other thread or interrupt routine can find it and queue the thread for execution when it’s ready to run, for instance after an I/O operation is complete.\footnote{3}

The advantages of this approach are similar to those of interrupts: there are no stacks to manage, and scheduling can be carefully tuned to the application. The main drawback is also similar: a thread must unwind its stack before it can wait. In particular, it cannot wait to acquire a lock at an arbitrary point in the program.

It is very common to code the I/O system of an operating system using this kind of thread. Most people who are used to this style do not realize that it is a restricted, though efficient, case of general programming with threads.

In ‘active messages’, a recent variant of this scheme, you break your computation down into non-blocking segments; as the end of a segment, you package the state into an ‘active message’ and send it to the agent that can take the next step. Incoming messages are queued until the receiver has finished processing earlier ones.\footnote{4}

There are lots of other ways to use the control block idea. In ‘scheduler activations’, for example, kernel operations are defined so that they always run to completion; if an operation can’t do what was requested, it returns intermediate state and can be retried later.\footnote{5} In ‘message queuing’ systems, the record of the thread state is stored in a persistent queue whenever it moves from one module to another, and a transaction is used to take the state off one queue, do some processing, and put it back onto another queue. This means that the thread can continue execution in spite of failures in machines or communication links.\footnote{6}

**SIMD or data-parallel computing**
This acronym stands for ‘single instruction, multiple data’, and refers to processors in which several execution units all execute the same sequence of instructions on different data values. In a ‘pure’ SIMD machine every instruction is executed at the same time by all the processors (except that some of them might be disabled for that instruction). Each processor has its own memory, and the processors can exchange data as part of an instruction. A few such machines were built between 1970 and 1993, but they are now out of favor.\footnote{7} The same programming paradigm is still used in many scientific problems however, at a coarser grain, and is called ‘data-parallel’ computing. In one step each processor does some computation on its private data.

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\footnote{5}{T. Anderson et al., Scheduler activations: Effective kernel support for the user-level management of parallelism. ACM Transactions on Computer systems 10, 1 (Feb. 1992), pp 54-79.}
\footnote{6}{See www.messageq.com or A. Dickman, Designing Applications With Msmq: Message Queuing for Developers, Addison-Wesley, 1998.}
\footnote{7}{The term ‘SIMD’ has been recycled in the Intel MMX instruction set, and similar designs from several other manufacturers, to describe something much more prosaic: doing 8 8-bit adds in parallel on a 64-bit data path.}
When all of them are done, they exchange some data and then take the next step. The action of
detecting that all are done is called ‘barrier synchronization’.

**Easy concurrency**

Concurrency is easy when you program with locks. The rules are simple:

- Every shared variable must be protected by a lock. A variable is shared if it is touched by
  more than one thread. Alternatively, you can say that *every* variable must be protected by a
  lock, and think of data that is private to a thread as being protected by an implicit lock that is
  always held by the thread.

- You must hold the lock for a shared variable before you touch the variable. The essential
  property of a lock is that two threads can’t hold the same lock at the same time. This property
  is called ‘mutual exclusion’; the abbreviation ‘mutex’ is another name for a lock.

- If you want an atomic operation on several shared variables that are protected by different
  locks, you must not release any locks until you are done. This is called ‘two-phase locking’,
  because there is a phase in which you only acquire locks and don’t release any, followed by a
  phase in which you only release locks and don’t acquire any.

Then your computation between the point that you acquire a lock and the point that you release it
is equivalent to a single atomic action, and therefore you can reason about it sequentially. This
atomic part of the computation is called a ‘critical section’. To use this method reliably, you
should annotate each shared variable with the name of the lock that protects it, and clearly
bracket the regions of your program within which you hold each lock. Then it is a mechanical
process to check that you hold the proper lock whenever you touch a shared variable.\(^8\) It’s also
possible to check a running program for violations of this discipline.\(^9\)

Why do locks lead to big atomic actions? Intuitively, the reason is that no other well-behaved
thread can touch any shared variable while you hold its lock, because a well-behaved thread
won’t touch a shared variable without itself holding its lock, and only one thread can hold a lock at
a time. We will make this more precise in handout 17 on formal concurrency, and give a proof of
atomicity. Another way of saying this is that locking ensures that concurrent operations
*commute*. Concurrency means that we aren’t sure what order they will run in, but commuting
says that the order doesn’t matter because the result is the same in either order.

Actually locks give you a bit more atomicity than this. If a well-behaved thread acquires a
sequence of locks and then releases them (not necessarily in the same order), the entire
computation from the first acquire to the last release is atomic. Once you have done a release,
however, you can’t do another acquire without losing atomicity.

The simple locks we have been describing are also called ‘mutexes’; this is short for ‘mutual
exclusion’. As we shall see, more complicated kinds of locks are often useful.

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\(^8\) This process is mechanized in ESC; see http://www.research.digital.com/SRC/esc/ESC.html.

first you put your hand over some shared variables, say $x$ and $y$, so that no one else can change them,

then you look at them and perhaps do something with them, and

finally you take your hand away.

The reason $x$ and $y$ can’t change is that the rest of the program obeys some conventions; in particular, it acquires locks before touching shared variables. There are other, trickier conventions that can keep $x$ and $y$ from changing; we will see some of them later on.

This viewpoint sheds light on why fault-tolerant programming is hard: Crash is no respecter of conventions, and the invariant must be maintained even though a Crash may stop an update in mid-flight and reset all or part of the volatile state.

**Scheduling: Condition variables**

If a thread can’t make progress until some condition is established, and therefore has to release a lock so that some other thread can establish the condition, the simplest idiom is

```plaintext
VAR c : SET Thread := {} % Each condition variable is the set of waiting threads.
PROC wait(m) =
    << c /\ t := (SELF); m.rel >>; % m.rel=HAVOC unless SELF IN m
    << ~ (SELF IN c) => m.acq >>

APROC signal() = << c := {} >>

APROC broadcast() = << c := {} >>
```

This code is correct, but reacquiring the lock immediately makes it more difficult for another thread to get it, and going around the loop while the condition remains false wastes processor cycles. Even if you have your own processor, this isn’t a good scheme because of the system-wide cost of repeatedly acquiring the lock.

The way around these problems is an optimization that replaces $m\rel; m\acq$ in the box with $c.wait(m)$, where $c$ is a ’condition variable’. The $c.wait(m)$ releases $m$ and then blocks the thread until some other thread does $c.signal$. Then it reacquires $m$ and returns. If several threads are waiting, $signal$ picks one or more to continue in a fair way. The variation $c.broadcast$ continues all the waiting threads.

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Here is the spec for condition variables. It says that the state is the set of threads waiting on the condition, and it allows for lots of $c$’s because it’s a class. The $wait$ method is especially interesting, since it’s the first procedure we’ve seen in a spec that is not atomic (except for the clumsy non-atomic specs for disk and file writes, and $ObjNames$). This is because the whole point is that during the $wait$ other threads have to run, access the variables protected by the mutex, and signal the condition variable. Note that $wait$ takes an extra parameter, the mutex to release and reacquire.

```plaintext
CLASS Condition EXPORT wait, signal, broadcast =
    TYPE M = Mutex
```

Note that you are not entitled to assume that the condition is true just because $wait$ returns. That would be a little more efficient for the waiter, but it would be much more error prone, and it would require a tighter spec for $wait$ and $signal$ that is often less efficient to code. You are supposed to think of $c.wait(m)$ as just an optimization of $m\rel; m\acq$. This idiom is very robust. Warning: many people don’t agree with this argument, and define stronger condition variables; when reading papers on this subject, make sure you know what religion the author embraces.

More generally, after $c.wait(m)$ you cannot assume anything about $x$ beyond its invariant, since the $wait$ unlocks $m$ and then locks it again. After a $wait$, only the invariant is guaranteed to hold, not anything else that was true about $x$ before the wait.

**Really easy concurrency**

An even easier kind of concurrency uses buffers to connect independent modules, each with its own set of variables disjoint from those of any other module. Each module consumes data from some predecessor modules and produces data for some successor modules. In the simplest case the buffers are FIFO, but they might be unordered or use some other ordering rule. A little care is needed to program the buffers’ put and get operations, but that’s all. This is often called ‘pipelining’. The fancier term ‘data flow’ is used if the modules are connected not linearly but by a more general DAG.

Another really easy kind of concurrency is provided by transaction processing or TP systems, in which an application program accepts some input, reads and updates a shared database, and generates some output. The transaction mechanism makes this entire operation atomic, using techniques that we will describe later. The application programmer doesn’t have to think about

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concurrency at all. In fact, the atomicity usually includes crash recovery, so she doesn’t have to think about fault-tolerance either.

In the pure version of TP, there is no state preserved outside the transaction except for the shared database. This means that the only invariants are invariants on the database; the programmer doesn’t have to worry about mistakenly keeping private state that records something about the shared state after locks are released. Furthermore, it means that a transaction can run on any machine that can access the database, so the TP system can take care of launching programs and doing load balancing as well as locking and fault tolerance. How easy can it get?

**Hard concurrency**

If you don’t program according to the rules for locks, then you are doing hard concurrency, and it will be hard. Why bother? There are three reasons:

- You may have to code mutexes and condition variables on top of something weaker, such as the atomic reads and writes of memory that a basic processor or file system gives you. Of course, only the low-level runtime implementer will be in this position.

- It may be cheaper to use weaker primitives than mutexes. If efficiency is important, hard concurrency may be worth the trouble. But you will pay for it, either in bugs or in careful proofs of correctness.

- It may be important to avoid waiting for a lock to be released. Even if a critical section is coded carefully so that it doesn’t do too much computing, there are still ways for the lock to be held for a long time. If the thread holding the lock can fail independently (for example, if it is in a different address space or on a different machine), then the lock can be held indefinitely. If the thread can get a page fault while holding the lock, then the lock can be held for a disk access time. A concurrent algorithm that prevents one slow (or failed) thread from delaying other threads too much is called ‘wait-free’.

In fact, the “put out your hand” way of looking at things applies to hard concurrency as well. The difference is that instead of preventing \(x\) and \(y\) from changing at all, you do something to ensure that some predicate \(p(x, y)\) will remain true. The convention that the rest of the program obeys \(p(x, y)\) is called ‘wait-free’.

Problems in easy concurrency: Deadlock

The biggest problem for easy concurrency is deadlock, in which there is a cycle of the form

- Lock \(a\) is held by thread 1.
- Thread 1 is waiting for lock \(b\).
- Lock \(b\) is held by thread 2.
- ...
- Lock \(h\) is held by thread 8.
- Thread 8 is waiting for lock \(a\).

All the locks and threads are nodes in a lock graph with the edges “lock \(a\) is held by thread 1”, “thread 1 is waiting for lock \(b\)”, etc.

The way to deal with this that is simplest for the application programmer is to detect a deadlock and automatically roll back one of the threads, undoing any changes it has made and releasing its locks. Then the rolled-back thread retries; in the meantime, the others can proceed. Unfortunately, this approach is only practical when automatic rollback is possible, that is, when all the changes are done as part of a transaction. Handout 19 on sequential transactions explains how this works.

Note that from inside a module, absence of deadlock is a safety property: something bad doesn’t happen. The “bad” thing is a loop of the kind just described, which is a well-defined property of certain states, indeed, one that is detected by systems that do deadlock detection. From the outside, however, you can’t see the internal state, and the deadlock manifests itself as the failure of the module to make any progress.

The main alternative to deadlock detection and rollback is to avoid deadlocks by defining a partial order on the locks, and abiding by a rule that you only acquire a lock if it is greater than every lock you already hold. This ensures that there can’t be any cycles in the graph of threads and locks. Note that there is no requirement to release the locks in order, since a release never has to wait.

To implement this idea you

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11 M. Herlihy. Wait-free synchronization. *ACM Transactions on Programming Languages and Systems* **13**, 1 (Jan. 1991), pp 124-149. There is a general method for implementing wait-free concurrency, given a primitive at least as strong as compare-and-swap; it is described in M. Herlihy. A methodology for implementing highly concurrent data objects. *ACM Transactions on Programming Languages and Systems* **15**, 9 (Nov. 1993), pp 745-770. The idea is the same as optimistic concurrency control (see handout 20): do the work on a separate version of the state, and then install it atomically with compare-and-swap, which detects when someone else has gotten ahead of you.

annotate each shared variable with its protecting lock (which you are supposed to do anyway when practicing easy concurrency),

state the partial order on the locks, and

annotate each procedure or code block with its 'locking level' \( l_1 \), the maximum lock that can be held when it is entered, like this: \( l_1 <= x \).

Then you always know textually the biggest lock that can be held (by starting at the procedure entry with the annotation, and adding locks that are acquired), and can check whether an \( acq \) is for a bigger lock as required, or not. With a stronger annotation that tells exactly what locks are held, you can subtract those that are released as well. You also have to check when you call a procedure that the current locking level is consistent with the procedure’s annotation. This check is very similar to type checking.

Having described the basic method, we look at some examples of how it works and where it runs into difficulties.

If resources are arranged in a tree and the program always traverses the tree down from root to leaves, or up from leaves to root (in the usual convention, which draws trees upside down, with the root at the top), then the tree defines a suitable lock ordering. Examples are a strictly hierarchical file system or a tree of windows. If the program sometimes goes up and sometimes goes down, there are problems; we discuss some solutions shortly. If instead of a tree we have a DAG, it still defines a suitable lock ordering.

Often, as in the file system example, this graph is actually a data structure whose links determine the accessibility of the nodes. In this situation you can choose when to release locks. If the graph is static, it’s all right to release locks at any time. If you release each lock before acquiring the next one, there is no danger of deadlock regardless of the structure of the graph, because a flat ordering (everything unordered) is good enough as long as you hold at most one lock at a time. If the graph is dynamic and a node can disappear when it isn’t locked, you have to hold on to one lock at least until after you have acquired the next one. This is called ‘lock coupling’, and a cyclic graph can cause deadlock. We will see an example of this when we study hierarchical file systems in handout 15.

Here is another common locking pattern. Consider a program that manipulates objects named by handles and maintains a set of these objects. For example, the objects might be buffers, and the set the buffers that are non-empty. One thread works on an object and sometimes needs to mess with the set, for instance when a buffer changes from empty to non-empty. Another thread processes the set and needs to mess with some of the objects, for instance to empty out the buffers at regular intervals. It’s natural to have a lock \( h.m \) on each object and a lock \( ms \) on the set. How should they be ordered? We work out a solution in which the ordering of locks is every \( h.m < ms \).

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Note that each piece of state that is not a mutex is annotated with the lock that protects it: \( x \) with \( ms \) and \( y \) with \( m \). The ‘object table’ \( ot \) is fixed and therefore doesn’t need a lock.

We would like to maintain the invariant "object is non-empty" = "in set": \(- h.empty = h \in s \). This requires holding both \( h.m \) and \( ms \) when the emptiness of an object changes. Actually we maintain "\( h.m \) is locked \( \land (\neg h.empty \rightarrow h \in s) \)". which is just as good. The \( Fill \) procedure that works on objects is very straightforward; \( Add \) and \( Drain \) are functions that compute the new state of the object in some unspecified way, leaving it non-empty and empty respectively. Note that \( Fill \) only acquires \( ms \) when it becomes non-empty, and we expect this to happen on only a small fraction of the calls.

```
PROC Fill(h, x: Any) =
% Update the object h using the data x
    h.acq;
    IF h.empty => ms.acq; s \:=\: \{h\}; ms.rel [*] SKIP FI;
    ot(h).y := Add(h.y, x);
    h.rel
```

The \( Demon \) thread that works on the set is less straightforward, since the lock ordering keeps it from acquiring the locks in the order that is natural for it.

```
THREAD Demon() = DO
    ms.acq;
    IF VAR h | h \in s =>
        ms.rel;
        h.acq; ms.acq;
        IF h \in s =>
            s \:=\: \{h\}; ot(h).y := Drain(h.y)
                [*] SKIP FI;
            ms.rel; h.rel
                [*] ms.rel FI
        OD
```

\( Drain \) itself does no locking, so we don’t show its body.

The general idea, for parts of the program like \( Demon \) that can’t acquire locks in the natural order, is to collect the information you need, one mutex at a time. Then reacquire the locks according to the lock ordering, check that things haven’t changed (or at least that your conclusions still hold), and do the updates. If it doesn’t work out, retry. Version numbers can make the ‘didn’t change’ check cheap. This scheme is closely related to optimistic concurrency control, which we discuss later in connection with concurrent transactions.

It’s possible to use a hybrid scheme in which you keep locks as long as you can, rather than preparing to acquire a lock by always releasing any larger locks. This works if you can acquire a lower lock ‘cautiously’, that is, with a failure indication rather than a wait if you can’t get it. If you fail in getting a lower lock, fall back to the conservative scheme of the last paragraph. This doesn’t simplify the code (in fact, it makes the code more complicated), but it may be faster.
Deadlock with condition variables: Nested monitors

Since a thread can wait on a condition variable as well as on a lock, it’s possible to have a deadlock that involves condition variables as well as locks. Usually this isn’t a problem because there are many fewer conditions than locks, and the thread that signals a condition is coupled to the thread that waits on it only through the single lock that the waiting thread releases. This is fortunate, because there is no simple rule like the ordering rule for locks that can avoid this kind of deadlock. The lock ordering rule depends on the fact that a thread must be holding a lock in order to keep another thread waiting for that lock. In the case of conditions, the thread that will signal can’t be distinguished in such a simple way.

The canonical example of deadlock involving conditions is known as “nested monitors”. It comes up when there are two levels of abstraction, H and M (for high and medium; low would be confused with the L of locks), each with its own lock lH and lM. M has a condition variable cM. The code that deadlocks looks like this, if two threads 1 and 2 are using H, 1 needs to wait on cM, and 2 will signal cM.

```
H1: lH.lock; call M1
M1: lM.lock; cM.wait(lM)
H2: lH.lock; call M2
M2: lM.lock; cM.signal
```

This will deadlock because the wait in M1 releases lM but not lH, so that H2 can never get past lH.lock to reach M2 and do the signal. This is not a lock-lock deadlock because it involves the condition variable cM, so a straightforward deadlock detector will not find it. The picture below illustrates the point.

![Deadlock Diagram](image)

To avoid this deadlock, don’t wait on a condition with any locks held, unless you know that the `signal` can happen without acquiring any of these locks. The ‘don’t wait’ is simple to check, given the annotations that the methodology requires, but the ‘unless’ may not be simple.

People have proposed to solve this problem by generalizing `wait` so that it takes a set of mutexes to release instead of just one. Why is this a bad idea? Aside from the problems of passing the right mutexes down from n to m, it means that any call on m might release lH. The m programmer must be careful not to depend on anything more than the lH invariant across any call to m. This style of programming is very error-prone.

Problems in easy concurrency: Scheduling

If there is a shortage of processor resources, there are various ways in which the simple easy concurrency method can go astray. In this situation we may want some threads to have priority over others, but subject to this constraint we want the processor resources allocated fairly. This means that the amount of time a task takes should be roughly proportional to the amount of work it does; in particular, we don’t want short tasks to be blocked by long ones.

Priority inversion

When there are priorities there can be “priority inversion”. This happens when a low-priority thread λ acquires a lock and then loses the CPU, either to a higher-priority thread or to round-robin scheduling. Now a high-priority thread µ tries to acquire the lock and ends up waiting for λ. Clearly the priority of λ should be temporarily increased to that of µ until λ completes its critical section, so that µ can continue. Otherwise µ may wait for a long time while threads with priorities between λ and µ run, which is not what we had in mind when we set up the priority scheme. Unfortunately, many thread systems don’t raise λ’s priority in this situation.

Granularity of locks

A different issue is the ‘granularity’ of the locks: how much data each lock protects. A single lock is simple and cheap, but doesn’t allow any concurrency. Lots of fine-grained locks allow lots of concurrency, but the program is more complicated, there’s more overhead for acquiring locks, and there’s more chance for deadlock (discussed earlier). For example, a file system might have a single global lock, one lock on each directory, one lock on each file, or locks only on byte ranges within a file. The goal is to have fine enough granularity that the queue of threads waiting on a mutex is empty most of the time. More locks than that don’t accomplish anything.

It’s possible to have an adaptive scheme in which locks start out fine-grained, but when a thread acquires too many locks they are collapsed into fewer coarser ones that cover larger sets of variables. This process is called ‘escalation’. It’s also possible to go the other way: a process keeps track of the exact variables it needs to lock, but takes out much coarser locks until there is contention. Then the coarse locks are ‘de-escalated’ to finer ones until the contention disappears. Closely related to the choice of granularity is the question of how long locks are held. If a lock that protects a lot of data is held for a long time (for instance, across a disk reference or an interaction with the user) concurrency will obviously suffer. Such a lock should protect the minimum amount of data that is in flux during the slow operation. The concurrent buffered disk example in handout 15 illustrates this point.

On the other hand, sometimes you want to minimize the amount of communication needed for acquiring and releasing the same lock repeatedly. To do this, you hold onto the lock for longer than is necessary for correctness. Another thread that wants to acquire the lock must somehow signal the holder to release it. This scheme is commonly used in distributed coherent caches, in which the lock only needs to be held across a single read, write, or test-and-set operation, but one thread may access the same location (or cache line) many times before a different thread touches it.
Lock modes

Another way to get more concurrency at the expense of complexity is to have many lock ‘modes’. A mutex has one mode, usually called ‘exclusive’ since ‘mutex’ is short for ‘mutual exclusion’. A reader/writer lock has two modes, called exclusive and ‘shared’. It’s possible to have as many modes as there are different kinds of commuting operations. Thus all reads commute and therefore need only shared mode (reader) locks. But a write commutes with nothing and therefore needs an exclusive mode (write) lock. The commutativity of the operations is reflected in a ‘conflict relation’ on the locks. For reader/writer or shared/exclusive locks this matrix is:

<table>
<thead>
<tr>
<th></th>
<th>None</th>
<th>Shared (read)</th>
<th>Exclusive (write)</th>
</tr>
</thead>
<tbody>
<tr>
<td>None</td>
<td>OK</td>
<td>OK</td>
<td>OK</td>
</tr>
<tr>
<td>Shared (read)</td>
<td>OK</td>
<td>OK</td>
<td>Conflict</td>
</tr>
<tr>
<td>Exclusive (write)</td>
<td>OK</td>
<td>Conflict</td>
<td>Conflict</td>
</tr>
</tbody>
</table>

Just as different granularities bring a need for escalation, different modes bring a need for ‘lock conversion’, which upgrades a lock to a higher mode, for instance from shared to exclusive, or downgrades it to a lower mode.

Explicit scheduling

In simple situations, queuing for locks is an adequate way to schedule threads. When things are more complicated, however, it’s necessary to program the scheduling explicitly because the simple first-come first-served queuing of a lock isn’t what you want. A set of printers with different properties, for example, can be optimized across a set of jobs with different priorities, requirements for paper handling, paper sizes, color, etc. There have been many unsuccessful attempts to build general resource allocation systems to handle these problems. They fail because they are too complicated and expensive for simple cases, and not flexible enough for complicated ones. A better strategy is to program the scheduling as part of the application, using as many condition variables as necessary to queue threads that are waiting for resources. Application-specific data structures can keep track of the various resource demands and application-specific code, perhaps written on top of a library, can do the optimization.

Just as you must choose the granularity of locks, you must also choose the granularity of conditions. With just a few conditions (in the limit, only one), it’s easy to figure out which one to wait on and which ones to signal. The price you pay is that a thread (or many threads) may wake up from a wait only to find that it has to wait again, and this is inefficient. On the other hand, with many conditions you can make useless wakeups very rare, but more care is needed to ensure that a thread doesn’t get stuck because its condition isn’t signaled.

Simple vs. fancy locks

We have described a number of features that you might want in a locking system:

- multiple modes with conversion, for instance from shared to exclusive;
- multiple granularities with escalation from fine to coarse and de-escalation from coarse to fine;
- deadlock detection.

Database systems typically provide these features. In addition, they acquire locks automatically based on how an application transaction touches data, choosing the mode based on what the operation is, and they can release locks automatically when a transaction commits. For a thorough discussion of database locking see Jim Gray and Andreas Reuter, Transaction Processing: Concepts and Techniques, Morgan Kaufmann, 1993, Chapter 8, pages 449-492.

The main reason that database systems have such elaborate locking facilities is that the application programmers are quite naive and can’t be expected to understand the subtleties of concurrent programming. Instead, the system does almost everything automatically, and the programmers can safely assume that execution is sequential. Automatic mechanisms that work well across a wide range of applications need to adapt in the ways listed above.

By contrast, a simple mutex has only one mode (exclusive), only one granularity, and no deadlock detection. If these features are needed, the programmer has to provide them using the mutex and condition primitives. We will study one example of this in detail in handout 17 on formal concurrency: building a reader/writer lock from a simple mutex. Many others are possible.

Summary of easy concurrency

There are four simple steps:

1. Protect each shared data item with a lock, and acquire the lock before touching the data.
2. Write down the invariant which holds on shared data when a lock isn’t held, and don’t depend on any property of the shared unless it follows from the invariant.
3. If you have to wait for some other thread to do something before you can continue, avoid busy waiting by waiting on a condition; beware of holding any locks when you do this. When you take some action that might allow a waiting thread to continue, signal the proper condition variable.
4. To avoid deadlock, define a partial order on the locks, and acquire a lock only if it is greater in the order than any lock you already hold. To make this work with procedures, annotate a procedure with a pre-condition: the maximum set of locks that are held whenever it’s called.
15. Concurrent Disks and Directories

In this handout we give examples of more elaborate concurrent programs:

Code for Disk.read using the same kind of caching used in BufferedDisk from handout 7 on file systems, but now with concurrent clients.

Code for a directory tree or graph, as discussed in handout 12 on naming, but again with concurrent clients.

Concurrent buffered disk

The ConcurrentDisk module below is similar to BufferedDisk in handout 7 on file systems; both implement the Disk spec. For simplicity, we omit the complications of crashes. As in handout 7, the buffered disk is based on underlying code for disk called UDisk, and calls on UDisk routines are in bold so you can pick them out easily.

We add a level of indirection so that we can have names (called b's) for the buffers; a b is just an integer, and we keep the buffers in a sequence called bv. b has methods that let us write b.db for bv(b).db and similarly for other fields.

The cache is protected by a mutex mc. Each cache buffer is protected by a mutex b.m; when this is held, we say that the buffer is locked. Each buffer also has a count users of the number of b's to the buffer that are outstanding. This count is also protected by mc. It plays the role of a readers lock on the cache reference to the buffer during a disk transfer: if it's non-zero, it is not OK to reassign the buffer to a different disk page. GetBufs increments users, and InstallData decrements it. No one waits explicitly for this lock. Instead, read just waits on the condition moreSpace for more space to become available.

Thus there are three levels of locking, allowing successively more concurrency and held for longer times:

mc is global, but is held only during pointer manipulations;

b.m is per buffer, but exclusive, and is held during data transfers;

b.users is per buffer and shared; it keeps the assignment of a buffer to a DA from changing.

There are three design criteria for the code:

1. Don't hold mc during an expensive operation (a disk access or a block copy).
2. Don't deadlock.
3. Handle additional threads that want to read a block being read from the disk.

You can check by inspection that the first is satisfied. As you know, the simple way to ensure the second is to define a partial order on the locks, and check that you only acquire a lock when it is greater than one you already have. In this case the order is mc < every b.m. The users count takes care of the third.

The loop in read calls GetBufs to get space for blocks that have to be read from the disk (this work was done by MakeCacheSpace in handout 7). GetBufs may not find enough free buffers, in which case it returns an empty set to read, which waits on moreSpace. This condition is signaled by the demon thread FlushBuf. A real system would have signaling in the other direction too, from GetBufs to FlushBuf, to trigger flushing when the number of clean buffers drops below some threshold.

The boxes in ConcurrentDisk highlight places where it differs from BufferedDisk. These are only highlights, however, since the code differs in many details.

CLASS ConcurrentDisk

EXPORT read, write, size, check, sync =

| TYPE | % Data, DB, Blocks, Dsk, DA as in Disk |
| I = Int |  |
| J = Int |  |
| Buf = [db, B, users: Int, clean: Bool] | % db, B, clean is protected by mc. The rest |
| M = Mutex |  |
| B = Int WITH {m := (bv(b).m), db := (bv(b).db), users := (bv(b).users), clean := (bv(b).clean)} | % index in bv |
| BS = SET B |  |

CONST

DSOSize := Disk.DSOLsize
nBufs := 100

FlushDiskRead := 3

% wait for this many Bufs

VAR

% uses UDisk's disk, so there's no state for that
udisk := Disk

% protected by mc

pause := (cache(da) -> b){} % protected by mc

% protected cache, users

mc := M

% wait for more space

moreSpace := Condition.C

% see Buf for protection

bv := (B -> Buf)

% only for the AF

flushing := (DA + Null) := nil

% ABSTRACTION FUNCTION Disk.disk(0) = \( \forall \ da | (\ cache(da) \land (\ cache(da).m \ not \ held \lor \ da = flushing) \lor \ cache(da).db \lor \ cache(da).db \lor \ cache(da).db ) \))

% INVARIANT 1: ( ALL da :IN cache.dom, b | b = cache(da) \land \ b.m \ not \ held \lor \ b.m \ not \ held \lor \ b.m \ not \ held \lor \ b.m \ not \ held )

A buffer in the cache, not locked, and clean agrees with the disk (if it's locked, the code in ReadBuf and the caller of GetBufs is responsible for keeping track of whether it agrees with the disk).
% INVARIANT 2: (ALL b | {da | cache!da \ cache(da) = b}.size \leq 1)
A buffer is in the cache at most once.

% INVARIANT 3: mc not held \Rightarrow (ALL b :IN bv.dom | b.clean \& b.users = 0 \Rightarrow b.m not held)
If mc is not held, a clean buffer with users = 0 isn’t locked.

PROC new(size: Int) -> Disk =
  self := StdNew();
  udisk := udisk.new(size);
  mc.acq; DO VAR b | ~ bv!b => VAR m := m.new() |
    bv(b) := Buf{m := m, db := {}, users := 0, clean := true}
  OD; mc.rel
  RET self

PROC read(e) -> Data RAISES {notThere} =
  udisk := udisk.check(e);
  VAR data := Data{}, da := e.da, upto := da + e.size, i |
    mc.acq; % Note that we release mc before a slow operation (bold below)
    % and reacquire it afterward.
    DO da < upto => VAR data, bs |
      IF cache!da =>
        b := cache(da);
        % Must increment users before releasing mc.
        bv(b).users + := 1; mc.rel;
        % Now acquire m before copying the data.
        % May have to wait for m if the block is being read.
        b.m.acq; data + := b.db; b.m.rel;
        mc.acq bv(b).users - := 1;
        da + := 1
    [*] i := RunNotInCache(da, upto);
    % da not in the cache
    bs := GetBufs(da, i); i := bs.size; % GetBufs is fast
    IF i > 0 =>
      mc.rel; data + := InstallData(da, i); mc.acq;
      da + := i
    [*] moreSpace.wait(mc)
    FI
    OD;
  mc.rel; RET data

PROC GetBufs(da, i) -> BS =
  % mc locked. Return some buffers assigned to da, da+1, ..., locked, and
  % with users = 1 or 0 if there's not enough space. No slow operations.
  VAR bs := {b | b.users = 0 \& b.clean} |
    % the usable buffers
  IF bs.size >= {i, minDiskRead}.min =>
    i := {i, bs.size}.min;
    DO VAR b | b IN bs \& b.users = 0 =>
      % Remove the buffer from the cache if it's there.
      IF VAR da' | cache(da') = b => cache := cache{da' -> } 
        [*] SKIP FI;
      VAR b.m.users := 1; cache(da) := b; da + := 1
    OD;
  [*] RET {} % too few; caller must wait
  FI

In handout 7, InstallData is done inline in read.

PROC InstallData(da, i) = VAR data, j := 0 |
  % Pre: cache(da) .. cache(da+i-1) locked by SELF with users > 0.
  data := udisk.read(E{da, i});
  DO j < i => VAR b := cache(da + j) |
    bv(b).db := udisk.DToB(data).sub(j); b.m.rel;
    mc.acq; bv(b).users - := 1; mc.rel;
  j + := 1
  OD;
  RET data

PROC write is omitted. It sets clean to false for each block it writes. The background thread
FlushBuf does the writes to disk. Here is a simplified version that does not preserve write order.
Note that, like read, it releases mc during a slow operation.

THREAD FlushBuf() =
  % flush a dirty buffer
  mc.acq;
  IF VAR da, b | b = cache(da) \& b.users = 0 \& ~ b.clean =>
    flushing := true;
    b.m.acq bv(b).clean := true; mc.rel
    udisk.write(da, b.db);
    flushing := false;
    b.m.rel moreSpace.signal
    [*] mc.rel
  [*] mc.rel
  OD

% Other procedures omitted

END ConcurrentDisk

GetBufs tries to return i buffers, but it returns at least minDiskRead buffers (unless i is less than this)
so that read won’t do lots of tiny disk transfers. It’s tempting to make GetBufs always
succeed, but this means that it must do a Wait if there’s not enough space. While mc is released in the
Wait, the state of the cache can change so that we no longer want to read i pages. So the
choice of i must be made again after the Wait, and it’s most natural to do the Wait in read.

If users and clean were protected by = (as do is) rather than by mc, GetBufs would have to
acquire pages one at a time, since it would have to acquire the = to check the other fields. If it
couldn’t find enough pages, it would have to back out. This would be both slow and clumsy.
Concurrent directory operations

In handout 12 on naming we gave an ObjNames spec for looking up path names in a tree of graph of directories. Here are the types and state from ObjNames:

```plaintext
TYPE D = Int  % Just an internal name
WITH {get:=GetFromS, set:=SetInS}
% get returns nil if undefined
Link = [d: (D + Null), pn]  % d=nil means the containing D
Z = (V + D + Link + Null)  % nil means undefined
DD = N -> Z

CONST
root : D := 0  % initially empty root
s := (D -> DD){}{root -> DD{}}

APROC GetFromS(d, n) -> Z =
% d.get(n)
<< RET s(d)(n) [^] RET nil >>

APROC SetInS(d, n, z) =
% d.set(n)
% If z = nil, SetInS leaves n undefined in s(d).
<< IF z # nil => s(d)(n) := z [^] s(d) := s(d){n -> } FI >>
```

We wrote the spec to allow the bindings of names to change during lookups, but it never reuses a D value or an entry in s. If it did, a lookup of /a/b might obtain the D for /a, say dA, and then /a might be deleted and dA reused for some entirely different directory. When the lookup continues it will look for b in that directory. This is definitely not what we have in mind.

Code, however, will represent a DD by some data structure on disk (and cached in RAM), and if the directory is deleted it will reuse the space. This code needs to prevent the anomalous behavior we just described. The simplest way to do so is similar to the users device in ConcurrentDisk above: a shared lock that prevents the directory data structure from being deleted or reused.

The situation is trickier here, however. It’s necessary to make sufficiently atomic the steps of first looking up a to obtain dA, and then incrementing s(dA).users. To do this, we make users a true readers lock, which prevents changes to its directory. In particular, it prevents an entry from being deleted or renamed, and thus prevents a subdirectory from being deleted. Then it’s sufficient to hold the lock on dA, look up b to obtain dB, and acquire the lock on dB before releasing the lock on dA. This is called ‘lock coupling’.

As we saw in handout 12, the amount of concurrency allowed there makes it possible for lookups done during renames to produce strange results. For example, Read(/a/x) can return 3 even though there was never any instant at which the path name /a/x had the value 3, or indeed was defined at all. We copy the scenario from handout 12. Suppose:

- initially /a is the directory d1 and /b is undefined;
- initially x is undefined in d1;
- concurrently with Read(/a/x) we do Rename(/a, /b); Write(/b/x, 3).

The following sequence of actions yields Read(/a/x) = 3:

- In the Read, Get(root, a) = d1
- Rename(/a, /b) makes /a undefined and d1 the value of /b
- Write(/b/x, 3) makes 3 the value of x in d1

In the Read, RET d1.get(x) returns 3.

Obviously, whether this possibility is important or not depends on how clients are using the name space.

To avoid this kind of anomaly, it’s necessary to hold a read lock on every directory on the path. When the directory graph is cyclic, code that acquires each lock in turn can deadlock. To avoid this deadlock, it’s necessary to write more complicated code. Here is the idea.

Define some arbitrary ordering on the directory locks (say based on the numeric value of D). When doing a lookup, if you need to acquire a lock that is less than the biggest one you hold, release the bigger locks, acquire the new one, and then repeat the lookup from the point of the first released lock to reacquire the released locks and check that nothing has changed. This may happen repeatedly as you look up the path name.

This can be made more efficient (and more complicated, alas) with a ‘tentative’ Acquire that returns a failure indication rather than waiting if it can’t acquire the lock. Then it’s only necessary to backtrack when another thread is actually holding a conflicting write lock.
16. Paper: Programming with Threads


Read it as an adjunct to the lecture on practical concurrency. It explains how to program with threads, mutexes, and condition variables, and it contains a lot of good advice and examples.