Technology trends are causing a paradigm shift in how computer systems are designed: Instead of steadily getting faster single processors, the entire semiconductor industry is turning to multiprocessor designs to improve performance. Future real-time and embedded systems will be using multiple-processor hardware, and the software is expected to adapt to the situation. Writing parallel software programs is known to be very difficult and fraught with unexpected problems, and now parallel programming is expected to go mainstream.

This class will discuss how to debug parallel software running on top of multiprocessor hardware, and the types of errors that occur due to parallelism. A range of techniques and tools are covered and the goal is to prime real-time and embedded software developers for multiprocessor integration and debugging.

1 Introduction

Since the early 1980s, embedded software developers have been taking advantage of the steady progress of processor performance. New processors were steadily being made available that increased the performance of a system just by dropping in a new processor without needing much change in the software. If you used a 500 MHz processor last year, you could buy a 1000 MHz processor next year. Software designers implemented applications that were intentionally too heavy for the current hardware, expecting faster processors to be available when the product hit the market.

This comfortable state was driven by the steady progress of the semiconductor industry that kept packing more transistors into smaller packages at higher clock frequencies, enabling performance to increase by both architectural innovations like more parallel pipelines, by adding resources like on-chip caches and memory controllers, and by increasing the clock frequency. All the while maintaining software compatibility with previous generations of processors.

The 32-bit PowerPC family offers a typical case study on this progression. Starting with a 603 processor, users have been able to upgrade with full application-software compatibility to the “G3”/750 processor series (rising from 166 to 1100 MHz over time). Next came the “G4”/7400 series, and then the 64-bit “G5”/970 series, providing software engineers a performance increase in a single processor.

However, in 2004, single-processor performance increases began slowing considerably. Due to a number of chip design and manufacturing issues, the clock-frequency increase has almost stopped, and we cannot expect dramatic increases in performance of single processors any more. Instead, the semiconductor industry has turned to parallelism to increase performance [1][2]. Using multiple processor cores (known as a multicore) on the same chip, the theoretical performance per chip can increase dramatically, even if the performance per core is only improving slowly. Also, the power efficiency of a multicore implementation is much better than traditional single-core implementations, which is a factor as important as absolute performance [3]. Basically, two cores at half frequency use half as much power as one full-speed core.

Consequently, every high-performance processor family in the embedded world and outside is moving to multiprocessor designs. PMC-Sierra has been selling dual-core RM9200 MIPS64 processors for some time. In 2005 and 2006, a flood of single-chip multiprocessors have been launched, including Cavium’s 16-way MIPS64-based processors, PA Semi’s 8-way PowerPC PA6T, ARM’s 4-way MPCore, and Freescale’s 4-way PPC8574. Parallel machines are here in force and are the only choice for designing the highest-performing embedded systems.
As illustrated above, we can in general expect future embedded systems to contain several shared-memory multiprocessor nodes, connected using one or more networks. The nodes can be located on the same or different boards as other nodes, depending on the hardware design.

The move to parallel hardware, however, creates problems for software developers. Applications that have traditionally used single processors will now have to be parallelized over multiple processors in order to take advantage of the multiple cores. Programs that used to run on single processors will have to run on multiprocessors. Apart from the act of creating parallel software, debugging is greatly complicated by the combination of parallel execution and tighter packaging [5][6].

Note that parallel processing is not really new to embedded systems. Many embedded applications have been using parallel processing with great success for a long time. For example, signal-processing in telecommunications makes use of arrays of digital signal processors (DSPs) to handle wireless and wired signals. Routers use multiple input and output ports with multiple traffic engines to scale to extreme traffic volumes. Automotive applications use large numbers of small processors spread throughout a vehicle. Control applications are distributed across multiple processor cards. Mobile phones contain a multitude of processors. All of these applications have been in domains where parallelism has been relatively easy to handle. The computations to be performed naturally divide into independent tasks and are “embarrassingly parallel” [9][10]. For example, in a mobile phone the user interface can run in parallel to the radio interface, and a mobile phone base station has one thread (or more) per active mobile terminal. Such parallel computation is easier to understand and design than general shared-memory processing [5].

Thus, moving to shared-memory parallelism within individual processing nodes will be the greatest challenge to the embedded software industry for a very long time. Tools, methodology and design thinking need to change. The force of this change is comparable to the introduction of multitasking operating systems, virtual memory and object-oriented programming.

Seymour Cray says, “If you were plowing a field, which would you rather use: two strong oxen or 1,024 chickens?” It seems that we are forced to start harnessing these chickens, since there are no oxen available…

### 1.1 Terminology

Some terminology before we get started on the technical details.

- **Multitasking** means that a single processor is used to run several different software tasks at the same time, basically time-sharing a single processor as scheduled by a (real-time) operating system.

- A **multiprocessor** (MP) is any computer system using more than one processor.

- **SMP, shared-memory** or **symmetric multiprocessing**, is a design where the processors in a multiprocessor use the same memory, can access the same data, and run the same tasks. In contrast, an **asymmetric multiprocessor** (AMP) gives each processor its own local memory.

- **Homogeneous** MP means that all processors are of the same type, while **heterogeneous** MP uses processors of several different types (for example, and ARM and a DSP). Most SMP designs are homogeneous, while most AMP designs are heterogeneous.

- A **multicore** processor is a **single chip** containing **multiple** processor cores. Essentially, it is a multiprocessor on a single chip. Arguably, the first multicore design was the TI C80 from 1995.
- **Manycore** is a term becoming popular for processors with more than about ten cores on a single chip. It is pertinent since multiprocessor systems featuring tens or hundreds of cores have some unique issues compared to current multicore processors featuring two or four cores. The line between multicore and manycore is not absolutely clear, but the distinction makes eminent sense.

- **Multithreading (MT)** is a technique where a single processor pipeline supports several parallel threads of computation. It is a relatively efficient way to increase the utilization of the execution resources of a core. Most MT designs are used to create small SMPs, but it can also be used in AMP mode [7].

In the rest of this paper, we will use the term “multiprocessor” to denote any kind of system with more than one processing thread in hardware.

On the software side, we prefer to use the generic term “task” to mean a single thread of computation. We want to avoid confusion between “process” and “processor,” and the overloaded meaning of “thread.” A number of tasks can share memory in order to implement an application, and this is really what a parallel program is: a number of tasks cooperating to perform the function of a particular application.

## 2 Programming Parallel Machines

The move to multiprocessing constitutes a break with the mainstream (and embedded) software-development tradition. Most programmers have been taught how to program single-threaded programs. Few real-time operating systems have supported SMP processing in the past. Debuggers (with a few exceptions) operate under the assumption that we are debugging a program that is single-threaded. There is no reason to despair, however. Parallel machines can be tamed. For example, the computer game business complained loudly about the difficulty of programming recent multicore game consoles like the Xbox 360 and Playstation3, but have since come through and created frameworks that take advantage of multiprocessors [11].

There are several ways in which to write code to run on a multiprocessor machine. The most important distinction is between *message passing* and *shared memory* programming styles. In message-passing, communication is explicitly programmed, while shared-memory makes communication implicit in the reading and writing of variables.  

The traditional way to program a parallel machine is to use a sequential language like C, C++, Java, C#, Fortran, Ada, or assembler, and add in parallelism using operating system APIs, class libraries, compiler directives, or language extensions. Some interesting and more or less common ways to add parallelism to programming are the following:

- **OpenMP** is a set of compiler directives which are added to a program to make it execute using parallel processors using shared memory. It requires support in the compiler, generating code based on the underlying operating system (as well as working directly towards the underlying machine for some operations). It is very successful in the supercomputing and server markets, and it also available for many embedded systems [11].

- **MPI** is a popular standardized library for message-passing parallelism. It has been used for supercomputing, creating programs successfully scaling to tens of thousands of processors. Just like OpenMP, it uses the underlying operating system and hardware-specific tweaks to access parallelism.

- **pthreads** is the POSIX standard for threading, and is available for all Linux operating systems and many embedded real-time operating systems.

- **OS APIs** offer the base support for other approaches like OpenMP, MPI, and pthreads. They can also be used directly, and most operating systems offer support for parallelism and distribution which goes beyond what is offered by the standards.

- **Middleware APIs** like Multicore association CAPI, Enea LINX, and Polycore Messenger provide a communications API that in principle insulate each task in an application from where the other tasks are running. Most of these are based on some kind of message-passing paradigm, since that makes supporting distributed and AMP processing easier.

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1 Message-passing is often used even on shared-memory machines, where messages are implemented using shared memory. Shared memory can also be emulated on machines without shared memory in hardware, so there is no clear association between hardware design and programming paradigm.
Ada offers built-in concurrency support in the programming language, which generates calls to the underlying operating system. Java also has concurrency handling defined as part of the language and standard.

Erlang is one of some languages designed explicitly for parallel computation. Programs are structured to use hundreds or thousands of lightweight tasks each with their own local storage, communicating using message-passing. Tasks are used as the primary means of program structuring — rather like objects are used on object-oriented programming. With a system like Erlang, hardware parallelism can be exploited by the run-time system invisibly to application software [13]. The Occam language from the 1980’s Transputer project is another example of an explicitly parallel language.

Performance libraries implement common mathematical operations (like FFTs) in a way tailored to a particular machine. Each operation from a performance library is executed in parallel using multiple processors, even if the originating program is not really using parallelism. They are typically provided by manufacturers to make exploiting parallel machines easier and more efficient.

Proprietary language extensions to sequential languages like C, C++, and Java are being offered by several compiler vendors such as CodePlay with their “sieve” extensions to C and C++. Such extensions typically modify the language syntax and add some keywords to make expressing parallelism easier, and also provide clearer parallel semantics. They are basically novel languages built on the base of a familiar sequential language.

Parallel programming platforms that add library calls to a sequential language, and then provide a powerful runtime system that automatically use multiple cores to accelerate computations. Most of the proposed approaches come from the stream programming field, and can use graphics processors and other massively parallel computation devices as well as multicore processors. Such platforms are offered by RapidMind and PeakStream [14], among others.

Coordination languages add another language or layer on top of existing sequential programming languages. This coordination layer dictates how parallelism is handled and how data access is synchronized. The sequential code expresses the sequential computation of each thread, without mixing in concurrency concerns [15].

Transactional memory is a hardware-supported approach currently being researched, where the shared memory of a machine is made to function much like a classic database. A “transaction” encompasses a sequences of memory accesses that either complete successfully or abort as a group. In this way, inconsistencies in program state resulting from concurrent data access can be reduced. Hardware support can be relatively easily introduced extending cache coherence protocols that are already necessary to build shared memory machines. No current hardware supports transactional memory.

Writing parallel programs seems to be easier where large data sets are being manipulated in a fairly regular manner, like supercomputing applications and databases. Parallelizing programs with smaller data sets is harder by experience. However, there is one mitigating factor in the current move to multicore implementations: inside a multicore chip, communication between cores is much faster than in a traditional multi-chip multiprocessor. This helps programmers write efficient programs, and should allow beneficial parallelization of more types of programs.

However, debugging an asymmetric parallel system is still harder than debugging a single-processor system. Fundamentally, humans are poor at thinking about parallel systems; we seem to be wired to handle a single flow of events better than multiple simultaneous flows. In this paper, we will mainly consider the case of using sequential programming languages like C and C++ with shared memory, as that is the model with the most subtle debugging problems, and the one that most embedded programmers will encounter when multicore processors replace old single-core processors at the core of their products.

3 The Software Breaks

Ignoring the issues of creating efficient parallel programs for the moment, even getting programs to function correctly in an SMP environment is harder than for a single processor. Existing software that has been developed on a single processor might not work correctly when transitioned onto a multiprocessor. That an application works correctly in a multitasking environment does not imply that it works correctly in a multiprocessing environment; serious new issues occur with true concurrency.
True parallel execution (or concurrency) makes it hard to establish the precise order of events in different concurrent tasks. The propagation time of information between processors in shared-memory multiprocessors is not zero (even if it is fairly short, a few thousand clock cycles at most), and this is sufficient to create subtle random variations in the execution, which can snowball. A multiprocessor by nature exhibits chaotic behavior where a small change in initial parameters gives rise to large differences in system state over time. The system is actually inherently unpredictable, at least in terms of timing, and correct function can only be achieved by allowing for this and designing code, which works even in what seems like bizarre circumstances.

The following catalogue of problems attempts to highlight the many new and interesting ways in which software can break on a multiprocessor.

### 3.1 Latent Concurrency Problems

There can be latent problems in an existing, proven, multitasking workload that runs just fine on a single processor. The presence of true concurrency makes mistakes in protecting against concurrent accesses much more likely to trigger and cause program errors. As the timing of tasks becomes more variable, and they run in parallel for longer periods of time, the task set is simply subjected to more stress. This effect is similar to optimizing a program in a C compiler: optimizations might expose bugs in the program that were previously hidden. The error was always there; it just didn’t manifest itself.

### 3.2 Missing Reentrancy

To make efficient use of a multiprocessor, all code that is shared between multiple tasks has to support reentrant execution. This means using locks to protect shared data and to allocate local data for each time a function is called. Shared state between invocations of the same function has to be avoided. In a multiprocessor environment, actual occurrences of multiple tasks using the same shared function simultaneously will occur much more frequently (and thus trigger bugs according to Section 3.1).

This effect is especially important for operating systems and shared libraries, as such code will be used heavily by multiple tasks.

### 3.3 Priorities do not Provide Mutual Exclusion

In application code written for a traditional single-processor, strictly priority-scheduled RTOS, a common design pattern to protect shared data is to make all tasks that access the same data run at the same priority level. With a strict priority-driven scheduler, each process will run to completion before the next process can run, giving exclusive access without any locking overhead. This will fail when true concurrency is introduced in a multiprocessor; the picture below illustrates a typical case:

![Diagram](image)

This code is multitasking-safe on a single processor, but will break on a multiprocessor. This means that even existing proven code will have to be reviewed and tested before it can be assumed to run correctly on a multiprocessor. Explicit locking has to be introduced in order to handle the access to shared data.

One solution proposed for maintaining the semantics of single-processor priority scheduling on an SMP is to only run the task(s) with highest priority. Thus, if there is only a single highest-priority task in a system, only one processor will be used and the rest left idle. This ensures that high-priority tasks do not need to worry
about simultaneous execution of lower-priority tasks, but does not solve the problem for tasks with the same priority when priority is used (incorrectly) as a serialization device.

### 3.4 Interrupts are not Locks

In the operating system and device driver code, you can no longer simply assume that you get exclusive access to shared data and devices by turning off all interrupts. Instead, SMP-safe locking mechanisms have to be used. Redesigning the locking mechanisms in an OS kernel or driver (or user application, in case it makes use of interrupt management) is a major undertaking in the change to multiprocessors, and getting an operating system to run efficiently on an SMP will take time [19][20].

Note that device drivers can sometimes be helped by hardware architecture. By steering all external interrupts from devices onto a single processor, that processor can often run single-processor device drivers without change. Disable/enable interrupts will work to integrate various parts of the device driver just like on a single-processor machine.

### 3.5 Race Conditions

Race conditions are situations where the outcome of a computation differs depending on which participating task gets to a certain point first. They typically trigger when some piece of code takes longer than expected to execute or timing is disturbed in some other way. Since race conditions are inherently timing-related, they are among the hardest bugs to find. The name comes from the fact that the tasks are racing forward in parallel, and the result depends on who gets to a certain point first.

The picture on the left below illustrates the classic data race, where a piece of data shared between two tasks and the tasks do not protect the common data with a lock. In this case, both tasks can be editing the data at the same time and will not correctly account for the updates from the other task. If task 1 was fast enough, it could finish its editing before task 2 begins, but there are no guarantees for this. Note that shared data is often a complex data structure, and the net result of a race is that different parts of the structure have been updated by different tasks, leading to an inconsistent data state.

**Classic data race**

<table>
<thead>
<tr>
<th>Task 1</th>
<th>Task 2</th>
</tr>
</thead>
<tbody>
<tr>
<td>read</td>
<td>read</td>
</tr>
<tr>
<td>edit</td>
<td>edit</td>
</tr>
<tr>
<td>write</td>
<td>write</td>
</tr>
</tbody>
</table>

Task 1 and task 2 work on the same data. Update from task 2 gets overwritten by task 1.

**Message races**

<table>
<thead>
<tr>
<th>Task 1</th>
<th>Task 2</th>
<th>Task 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>msg1</td>
<td>calc</td>
<td>msg2</td>
</tr>
<tr>
<td>msg2</td>
<td>calc</td>
<td>msg1</td>
</tr>
</tbody>
</table>

Task 2 expects data from first task 1 and then task 3. Messages can also arrive in a different order.

The picture on the right illustrates another type of race, the message race, where one task is expecting a series of messages from other tasks. Such messages can come in any order, as there is no synchronization between the tasks; if task 2 implicitly expects a certain order, it will get the wrong results if task 3 happens to send before task 1. Whenever the ordering of events is important, explicit synchronization has to be in place.

Note that races occur in multitasking single-processor systems too, but there they are less likely to trigger, as they require a task switch to occur at an unlucky time. In a true concurrent system, races will happen more often, as discussed in Section 3.1.

### 3.6 Deadlocks

When all shared data is protected by locks, you get into another situation where the locking itself can be the cause of errors. If two tasks are taking multiple locks in different order, they can get stuck, both waiting for the other task to release the other lock. A simple but fairly typical example is given below:
Task T1 locks lock L1 first, which protects the variable V1. After awhile, T1 also needs to work on variable V2, protected by lock L2, and thus needs to lock L2 while still holding L1. This code in itself is sound, as all accesses to shared data is correctly protected by locks. In task T2, work is being performed on variable V2, and lock L2 is taken. When calling the function foo(), V1 is also accessed, and locking is in place.

Assume a scenario where T1 and T2 start at the same time, and manage to obtain one lock each. When T1 gets to the point lock(L2), it stops and waits as T2 is holding that lock. Slightly later, T2 gets to the call to function foo(), and duly tries to lock L1. At this point, we are stuck, as T1 and T2 are mutually waiting for the other task to release a lock, which will never happen. We have a deadlock.

The example above illustrates a common cause of deadlocks: calling into functions like foo(), which do locking invisible to the caller. In a complex software system, it happens easily that locks get taken in a bad order if no measures are taken to ensure consistent locking orders.

### 3.7 Partial Crashes

When a computation involves multiple tasks that compute different parts of an overall solution, it is possible that one (or more) of the tasks crash during the computation. This problem in a single task will then escalate to a parallel problem, as the other tasks wait forever for the crashed task to get back with its part of the result.

### 3.8 Silent Parallelization Errors

Parallel code written with explicit calls to threading libraries will naturally have to check whether thread creation and other operations succeeded, but when using indirect parallel programming by compiler directives like OpenMP, error handling is often suboptimal. The compiler will create extra code in the application to start tasks and handle synchronization, and this code does not have a defined way to report problems to the main user code – since it is not part of the program text per se. If a program fails to create the tasks it needs at run-time, it might just crash or hang or fail silently with no message to the user or programmer indicating a problem. Such behavior makes it unsuitable for high-reliability code where error recovery is necessary.

### 3.9 Bad Timing Assumptions

Missing synchronization is a common theme in parallel programming bugs. One particular variant to be wary of is the assumption that one task is going to finish its next work long before some other task needs the results of this work. In essence, as a task T1 is doing something short and simple, we expect that work to be completed before a task T2 finishes something long and complicated, and we know that they start out at the same time. A simple example is the following:

<table>
<thead>
<tr>
<th>Task T1</th>
<th>Task T2</th>
</tr>
</thead>
<tbody>
<tr>
<td>create(T2)</td>
<td>initialize()</td>
</tr>
<tr>
<td>write(V)</td>
<td>read(V)</td>
</tr>
</tbody>
</table>

Here, we expect T1 to finish writing V long before T2 is done with its initialization. However, it is possible that T2 in some situation completes its initialization before T1 can complete its write. This sort of bug typically triggers under heavy load or unusual situations in a shipping system.

### 3.10 Relaxed Memory Ordering Wreaks Havoc

For performance reasons, a multiprocessor employs various forms of relaxed memory orders (also known as weak memory consistency models). Basically, the computer architecture specifically allows memory operations performed to be reordered and queued locally on a processor in order to avoid stalling a processor when the memory system is slow. In most such models, memory operations from one processor may be
seen in a different order from the viewpoint of another processor. This is necessary in order to get any efficiency out of a multiprocessor system. There are a number of relaxed memory orderings\(^2\), differing in how operations can bypass each other [22]. Understanding relaxed memory orders is probably one of the hardest parts of understanding parallel computing [10], and trying to understand the details of a particular memory consistency model is quite headache-inducing. Unfortunately, the memory consistency model of a system is visible to a programmer since it affects data-transfer operations between parallel threads.

Many data exchange algorithms that are correct on multitasking single processor systems break when used on multiprocessors. For example, reads can almost always be performed out of order with respect to the program, and sometimes writes might be seen in different orders on different processors, especially when they originate from different other processors. In general, it is necessary to use special operations like memory barriers to guarantee that all processors in a system have seen a certain memory operation and that the state is thus globally consistent. Failing to deal with memory consistency will result in intermittent timing-sensitive bugs caused by processors observing the same execution in different orders.

One example is the classic Dekker locking algorithm shown below, which is perfectly valid on a single processor (assuming that each read of a flag variable is atomic), no matter how tasks are interleaved.

<table>
<thead>
<tr>
<th>Task T1</th>
<th>Task T2</th>
<th>Example sequence</th>
</tr>
</thead>
<tbody>
<tr>
<td>Basic algorithm</td>
<td></td>
<td></td>
</tr>
<tr>
<td>flag1 = 1 if(flag2 == 0) critical section</td>
<td>flag2 = 1 if(flag1 == 0) critical section</td>
<td></td>
</tr>
<tr>
<td>Problem example</td>
<td></td>
<td></td>
</tr>
<tr>
<td>x = 6 y = 5 flag1 = 1</td>
<td>while(flag1==0) loop; read x read y</td>
<td></td>
</tr>
</tbody>
</table>

Assuming that all variables are zero to begin with, it is quite possible to read old values for x and y in task T2 despite the use of a “lock,” since there is no guarantee that the writes to variables x and y are seen by task T2 before the write to flag1. With a relaxed memory ordering, you cannot assume that writes complete in the same order as they are specified in the program text. For example, if the write to flag1 hit the local cache, and x and y missed the cache, the net result could be that the write to flag1 takes a longer time to propagate. The execution is illustrated on the right in the picture above (in which we assume that each task runs on its own processor). Note that an additional task 3 can see the writes in the “expected” order – which is the common case. The rare case is the problem case. The programming solution is to put in barriers, forcing memory writes and reads to complete before the tasks continue beyond the lock. The cost is in performance, as processors stall when memory operations run to completion.

A more complex example that fails for the same reason is the following scheduling code:

```c
while(there are workunits) {
    allocate new workunit
    fill in workunit data
    insert workunit into linked list
}  
Head = first workunit
while(MyWorkunit == NULL) {
    lock critical section
    if(Head!=NULL) {
        MyWorkunit = Head
        Head = Head -> next
    }
    release lock for critical section
}  
read data from MyWorkunit
```

\(^2\) Note that the potential operation reorderings allowed by a specific memory consistency model are precisely defined, but that even so, understanding the implications of a particular memory consistency model requires deep considerations. University students usually find memory consistency the hardest part of computer architecture.
Here, the worker tasks T2…Tn wait for the linked lists of work units to become available from the master thread T1, and then grab work units off of it protected by a lock. However, considering the possibility of write reordering, the work unit data read by a worker task need not be complete just because the linked list has been updated. And if the locking is not multiprocessor-aware, the whole task queue can be destroyed by processors having a different idea of the value of Head.

Yet another example of the problems caused by memory consistency is the need to explicitly flush memory writes so that they become visible to all other processors in the system. On certain processor architectures, programs will just deadlock since necessary data for progress is queued on a processor that does not think it needs to write back data to memory.

Note that using the C keyword “volatile” has no effect on memory consistency; it does guarantee that the compiled code writes variable values back to memory, but provides no guarantees as to when other processors will see the write. Some kind of explicit synchronization operation is needed additionally. This is a symptom of a general problem in that most language definitions do not consider the implications of multiprocessors.

Even when multiprocessing is considered, programming complexity is not necessarily reduced. For example, the Java language specifies a “Java Memory Model” which is fairly weak. This has the advantage of making it possible to optimize performance on a multiprocessor host. It also has the disadvantage of forcing application programmers to deal with memory consistency issues and to include barriers and locks even into regular application code.

4 Debugging Parallel Programs

Debugging parallel programs is generally acknowledged as a difficult problem. Most of the literature on parallel programming focuses on the constructive part of creating a parallel algorithm for a particular problem and how to implement it, and mostly ignores issues of debugging and troubleshooting [11]. Despite the fact that multiprocessning has been on the horizon for a long time, very little in terms of practical tool support for parallel programming and debugging has been produced [5][16][19]³.

Debugging a software problem on a multiprocessor problem involves three main steps:

- Provoking problems
- Reproducing problems that have been provoked
- Diagnosing and resolving the reproduced problems

What makes multitasking programming and debugging difficult is that provoking and reproducing problems is much harder than with single-threaded programs. In classic single-tasking programs, most bugs are deterministic and caused by particular variations of input. Such deterministic bugs will still occur in each task in a parallel system, and will be solved in a traditional manner.

The parallel program adds a new category of bugs caused by the interaction of multiple tasks, often depending on the precise timing of their execution and communications. Such bugs are the most difficult bugs to provoke and reproduce, not to mention understand and resolve. They are often called Heisenbugs, meaning bugs that change behavior (usually, by moving somewhere else or disappearing from sight) when you try to observe the program in order to discover them [6] [16].

This section discusses techniques for debugging multiprocessor programs in spite of this. Some techniques focus on provoking and reproducing bugs and others on diagnosis of the problems.

4.1 Breakpoints on Single Tasks

A traditional single-task debugger can be applied to a parallel software system by debugging a single task in the traditional manner, setting breakpoints and watchpoints, and investigating state. This sometimes does work well in a parallel system, but carries some obvious risks:

³ Part of the problem could be that parallel programming has been on the horizon for a very long time, always in line as the next big thing but never actually arriving. It’s simply a case of “crying wolf.”
Breakpoints disturb the timing. Activating a breakpoint or watchpoint and just observing that it was triggered (i.e. not stopping execution) will make the timing of the task different. As discussed above, timing changes can make the program take a different path, avoiding the problem being investigated.

A stopped task can be swamped with traffic. If a single task is stopped and the rest of the parallel program continues to run, the other tasks in the system might keep queuing up requests for work and communications with the stopped task. This might cause an avalanche of problems in the rest of the parallel program if queues fill up and other tasks get time-outs on replies. Also, when the stopped task is resumed, it might be faced with an unusually large amount of communications from other parts of the application, leading to a different behavior.

A stopped processor in a real-time system will quickly cause a system crash or watchdog reset, as the computer system is expected to keep up with the sensors and actuation needs of the controlled system.

Many tools in the market today support running multiple debug sessions in parallel, originally in order to support asymmetric multiprocessors (typically one window to the DSP, one window to the control processor). Such solutions typically feature various forms of synchronizations between the debugger sessions, so that when one processor stops, the other stops. Such global stops have a certain skid time, since it takes a small but non-zero time for the debugger to catch one breakpoint and order the other processors to stop. Also, doing a global stop might not be feasible in a system connected to the outside world. For example, the operating system will still need to run in order to handle interrupts from peripheral devices. Stopping a processor does not mean stopping the world.

Such tools are better than a single sequential debugger in an SMP environment, but the hardware support is not really there to make them work as well as expected. Consider the issue of using hardware breakpoints: A hardware breakpoint set in the debug facilities of one particular processor will not trigger if the task is scheduled onto another processor before executing. Debuggers for SMPs must work around such issues, as the hardware itself does not provide the facilities to migrate breakpoints around with tasks.

### 4.2 Tracing

Tracing is an indirect debug that gathers knowledge: If an error is provoked or reproduced when tracing is enabled, the trace will hopefully contain hints to the sequence of events leading up to the error. Using this information, you can then construct a hypothesis as to what went wrong.

Traces can be gathered in a number of ways:

- **Printf:** The most common trace tool (and probably debug tool overall) is to use printouts in a program to determine its execution path. The advantage of using printouts inside the application code is that the information is typically application-level and semantically rich; a debug printout generally states that a program has reached a certain point and shows the values of relevant data at that point. The obvious disadvantage is that adding printouts to a program disturbs its timing, potentially changing its behavior so that the bug disappears. There have been cases where the printouts had to be left inside shipping systems (albeit silenced), as the system did not work without them.

- **Monitor code:** A more structured way to trace is to use special monitor programs that monitor system execution, outside of the actual application code. Such monitors will have less timing impact on the overall system, but also provide less information compared with a printout approach. Some multiprocessing libraries provide hooks for monitors. Embedded operating systems typically offer tools that trace operating system-level events like task scheduling, pre-emption and messages, which can be very useful to understand the behavior of a system. Sun’s DTrace tool is one example of such a tool (available for the Solaris operating system).

- **Instrumented code:** Some debug and software analysis tools instrument the source code or binary code of a program to trace and profile its behavior. Such instrumentation will change the program behavior, but a tool can use intelligent algorithms to minimize this effect. Instrumenting binaries makes it possible to specifically investigate shared memory communications, as each load and store instruction can be traced.

- **Bus trace:** Since basically all multiprocessor systems use caches, tracing system activity on the memory bus will only provide a partial trace. Also, some systems employ split buses or rings, where there is no single point of observation. Continued hardware integration is making bus analysis less and less feasible as time goes by.
Hardware trace using core support: A hardware trace buffer drawing on processor core abilities like embedded trace features and JTAG interfaces is a standard embedded debug tool. Applying a hardware trace to one processor in a multiprocessor system will provide a good trace of its low-level behavior, and provide minimal timing disturbance (provided the trace can keep up with the processor). The information will have to be digested by a debugger to provide a good insight into the system behavior. See the discussion in Section 4.3 for more on how hardware support can help multicore debugging.

Simulation: Simulation (as discussed in Section 4.8) offers a way to trace the behavior of a system at a low level without disturbing it. The overall functionality is very similar to hardware tracing, with the added ability to correlate traces from multiple processors running in parallel. Simulation avoids the correlation problem.

A general problem with tracing is that when tracing execution on a multiprocessor system, it can be hard to reconstruct a global precise ordering of events (in fact, such an order might not even theoretically exist). Thus, even if the event log claims that event A happened before event B based on local time stamps, the opposite might be true from a global perspective. In a distributed system where nodes communicate over a network, a consistent global state and event history is known to be very hard to accurately reconstruct. This applies in particular to low-level hardware traces that need to match up traces from different processors on a cycle-by-cycle basis, considering all the possible delays and jitter from data transmissions over JTAG or other debug ports and trace units. Failing to match traces correctly will present an incorrect picture of the system behavior.

A limitation of most forms of tracing is that they only capture certain events. No information is provided as to what happens between the events in the trace, or to parts of the system that are not being traced. If more fine-grained or different information is needed, the tracing has to be updated correspondingly and the program re-executed.

Detailed traces of the inputs and outputs of a particular processor, program, task or system can be used to replay execution, as discussed in Section 4.9 [6].

### 4.3 Hardware Multicore Debug Support

Hardware is beginning to provide support for debugging tightly-coupled multicore architectures. Typically, this involves adding monitoring, tracing, and breaking facilities into the heart of the multicore SoC. Such a debug facility can listen both to the traffic on the internal bus between cores, and possibly also to the traffic between a core and its caches. A complicating issue is that the bandwidth available on the external debug port of an SoC is limited compared to the rate of data generated by a multicore system running at gigahertz speeds. Thus, intelligence and storage for traces has to be integrated into the debug facility. An important debug support feature is cross-triggering, where events observed on one core can cause execution to stop or tracing to start or stop on other cores [17][18].

Note that the traffic visible on the external bus of an integrated multicore SoC is not very helpful, since it is filtered by several layers of caches. Much of the communication between cores never needs to go off-chip at all (which is one important benefit of multicore designs).

### 4.4 Bigger Locks

If the problem in a parallel program appears to be the corruption of shared data, a common debug technique is to make the locked sections bigger, reducing available parallelism, until the program works. Alternatively, the program is first implemented using very coarse locking to ensure correct function initially. This is typically what has happened in previous moves to multiprocessing in the computer industry, for example in the multiprocessor ports and optimization of Linux, SunOS [19] and MacOS X [20].

After locks have been made coarser, the scope of locking is carefully reduced in scope to enhance performance, testing after each change to make sure that execution is still correct. Finally, since fine-grained locking is usually used to increase performance, performance has to be analyzed. If the wrong locks are made finer-grained, no performance increase might result from an extensive effort. Too fine-grained locking might also create inefficiencies when execution time is spent managing locks instead of performing useful computation.
4.5 **Apply a Heavy Load**

Reliably provoking and reproducing errors is one of the major problems in debugging parallel systems. Since errors typically depend on timing, stretching the timing is a good way to provoke errors, and this can be done by increasing the load on a system. At close to 100 percent CPU load, a system is much more likely to exhibit errors than at light loads. The best way to achieve this is to create a test system that can repeatedly apply a very heavy load to a system by starting a large number of tasks. The tasks should preferably be a mix of different types of programs, in order to exercise all parts of the software system. This method has proven to be very successful in ensuring high reliability in systems like IBM mainframes [21].

4.6 **Use a Different Machine**

For most embedded systems, the machine a piece of code will run on is usually known. The hardware design and part selection is part of the system development. Thus, running a piece of code on a different machine might seem pointless: We want it to work on the target system. However, testing on different hardware is a good way to find timing bugs, as the execution timing and task scheduling is likely to be different. It also serves to make the code more likely to work on future generations of the target system which are more than likely to exhibit different timing and a different number of processor cores.

A machine can be different in several relevant ways: Each processor might be faster or slower than the actual target. It might also have a different number of processors, more or fewer. Moving to different processor architecture or a different operating system is likely to take as much time under a tight schedule, but if it can be done, it is an excellent way of flushing out bad assumptions in the code. Portable code is in general more correct code.

Simulation, as discussed in Section 4.8, offers an interesting variant on using a different machine.

4.7 **Use a Different Compiler**

Just like using several different machines to run a program creates variation and makes problem exposure more likely, compiling a program using several different compilers also serves to find problems. Different compilers perform program analysis in different ways and typically generate different warnings (errors should be consistent across compilers unless a program is seriously borderline). Especially the complex features used in parallel programming like compiler directives and libraries (OpenMP and MPI in particular) vary across compilers. Thus, if a program runs well when compiled with several different compilers, it is more likely to be correct than if it is just compiled using a single compiler.

4.8 **Simulate the System**

The key problem with debugging a parallel system is lack of synchronization between parallel processors and determinism in execution. The inherent chaotic behavior makes cyclical debugging very hard. There is one technique that overcomes these problems: simulation of the target system. Traditionally, simulation has been used as a means to develop and run software before the real hardware was available. In the area of parallel computer system execution, simulation remains useful even after hardware becomes available – simulation of a parallel machine provides explicit control over the execution of instructions and propagation of information, which makes for a very powerful debugging tool.

A simulated system fundamentally provides determinism, as each simulation run repeats the same execution (unless nondeterministic inputs or timing variations are provided as external input). This enables classic cyclic debugging. It also makes reproducing errors easier, as once a bug has been provoked in a simulation, the same bug scenario can be replayed over and over again in exactly the same manner.

Simulators are commonly used as a backend to standard debuggers, in which a user simply connects to the simulator instead of to a JTAG probe, remote debug server or other debug interface. The simulator makes it possible to single-step one task while time is virtually standing still for other parts of the system, which solves the problems with breakpoints discussed in Section 4.1. For a real-time system where a simulation of the external world is available, simulation also makes it possible to single-step code and to stop the system without the external world running ahead.

Full-system simulators capable of running the complete software stack (from firmware to application software) also provide the ability to inspect and debug the interaction of an application with the operating system, as well as low-level code such as operating-system kernels and device drivers. Fundamentally, a full-
system simulator provides a controlling layer underneath the hardware, which enables capabilities that simply cannot be provided by traditional hardware-based debug tools.

A simulator will not be able to faithfully reproduce the detailed timing of a real system (in any non-trivial case). This is not really a problem for software development, as the goal is to find and eliminate software bugs: If they occur in the simulator, they could have happened in some real circumstance as well, and thus they are valid bugs that should be fixed [25].

Furthermore, a simulator offers an interesting variant of running on a different system, as discussed in Section 4.6. The simulator can be used to increase the number of processors present beyond that available on any real-world machine in order to stress the software, and checks its behavior on future machines with more cores. Simulation can make different processors run at different speeds in order to provoke errors. Performing such corner-case chasing in simulators is common in hardware design, and it has great potential as a tool for software developers as well.

Several vendors offer simulation solutions for embedded systems of varying scope. The investment to create a simulation model can be quite high. But for larger development projects, the benefits typically outweigh the costs, especially when considering the added value of reverse debugging, as discussed in Section 4.9.

4.9 Replay Execution

A recurring theme is the problem that re-executing a multitasking, multiprocessor program will result in a different execution, potentially failing to trigger bugs. If a program can be forced to execute in the same way multiple times, debugging will be much easier. Such techniques are known as record-replay techniques. Replay is different from tracing, as the amount of data needed to facilitate a deterministic replay can be made much smaller than a full trace. The trace recorded only needs to contain the non-deterministic events from the real system, like task switches, message arrivals and other communication between tasks [6][16].

To enable replay, the system has to contain a monitor that can record relevant data as well as force the execution to replay the same behavior. Typically, this has to work on the operating-system level, but it can also be applied at the application level if a program has sufficiently well-defined interfaces to the outside world (a state machine is a good example of this: The time and type of messages arriving will determine its precise behavior).

In a parallel program setting, replay can be done on a single task, on a set of tasks (an entire application), or even on all software running on a single computer node. Recording-based replay allows the debugger to isolate one part of the system, as all input values and asynchronous timings are known. The rest of the system does not need to be running in the debug session, simplifying the problem.

A less precise form of replay is to capture system interactions with the environment in the field, by adding logging facilities to a deployed system. Such interaction capture ignores internal system state changes, and thus cannot force execution down a particular path. But if the root cause of a problem is an unusual order of external events, such capture and replay can help debugging. If nothing else, a scenario can be replayed in the lab into some other debugging system to provide a good starting point for more detailed and powerful debugging tools.

4.10 Reverse Debugging

One of the key problems in debugging parallel programs in general, and parallel programs running on multiprocessors in particular, is that re-executing a program to put a breakpoint to locate an error is likely not to reproduce an error. Instead, a programmer would like to be able to go back from the point at which an error occurs and investigate the events that immediately preceded the error. What is needed here is a tool that allows reverse debugging, i.e. the ability to back up in a program from the current state instead of (attempting to) reproducing the state by executing from the start [9].

Reverse debugging is particularly useful for bugs corrupting the state of the machine. If the stack is overwritten or a process has terminated, there is little material to use for post-mortem bug analysis. By backing up in time, the state prior to the bug can be examined.

Tools for reverse debugging do exist today in the market. They can be based on simulation (see Section 4.7), trace collection (see Section 4.2) or regular desktop virtual machine technology [23]. Such tools offer the ability to back up the execution to examine the path that led to a tricky bug. Tools based on traces will by necessity have a limited window of observation (limited by the size of the trace buffer), while simulation- and
virtual-machine-based tools can make use of the resources of a powerful workstation to allow much longer sequences of execution to be observed. Also, most tools do not offer multiprocessor support.

Compared to replay as discussed in Section 4.9, reverse debugging is typically faster in practical use, as backing up a short distance in the execution is faster than reloading and replaying a complete execution. One of the main advantages of reverse debugging is actually that a programmer can focus on a small context and go back and forth over it, applying different breakpoints and traces to understand the bug, without long turnaround times that break the flow.

4.11 Dynamic Analysis

Dynamic analysis tools trace a program as it runs and try to find potential problems even if no error occurs in that particular run. The idea is to trace the memory accesses and synchronizations between threads, and look for potential problems using various algorithms that generalize from a single execution. There is no guarantee that all problems are found, since only code paths similar to those in the underlying concrete set of paths can be investigated. For example, code that is never executed in the concrete test case will not be subject to analysis. Even so, it is much more effective than regular testing that has to actually provoke a concrete error in order to spot it. It is also very important to note that a dynamic tool targets a specific type of error, and that each type of error typically requires a custom algorithm to detect.

One example is Intel’s ThreadChecker, a tool that attempts to check that all shared data accesses follow a correct locking regime. Basically, the tool checks that locks are always acquired in the same order by all parts of the program. If this makes sense for your program, it can offer very good help for locating that particular (and important) class of bugs. The open-source Valgrind tool has support for finding errors in MPI programs. The downside of such tools is that execution runs about 100 times slower than normal execution due to the instrumentation overhead.

4.12 Formal Methods

No discussion on debugging is complete without mentioning formal methods. In contrast to dynamic analysis, formal methods explore all possible executions of a program. Thus, formal methods promise to find all errors regardless of whether they are actually encountered in a concrete test run.

Program-checking tools from vendors like Polyspace and Coverity have been on the market for several years. They start by analyzing the behavior of single threads, which is a hard problem in its own right. They also are extending into analysis of parallel applications, but such coverage is currently limited. Also, the analysis time can be prohibitively large for large programs [26].

In general, real-world formal method tools applicable to real programs will behave analogously to lint: they will initially produce many false warnings, that are removed by customizing the rules set and making the code clearer. They will also only find the types of bugs they were designed to find. For example, a wild pointer will not be found by a tool checking for deadlock in message passing. But when used, they can dramatically increase the quality of code.

Another way to apply formal methods is to focus on the parallel algorithms being used rather than their concrete implementations in code. Especially for communications-intense systems featuring custom protocols at various levels, checking the correctness of the protocol itself is a powerful way to find fundamental bugs before they are committed to harder-to-fix code [27].

5 Summary

This paper has discussed the software implications of the undergoing hardware paradigm shift to multiprocessor, shared-memory computers. This is a fundamental change in how computers are constructed, and will affect both existing code and the creation of new code. We have taken inventory of problems that occur when moving to multiprocessor environments, and discussed a number of debug techniques available today on the market. The future is parallel, and we really have no choice but to prepare our code and ourselves for handling that parallelism. Even if the current situation in programming tools for multiprocessors is fairly bleak, we should expect new tools to hit the market in coming years that will make multiprocessor programming and debugging easier.
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More Reading

If you want to know more about this subject, here are some recommended articles and books that will provide insight into the issues of parallel programming and debugging.


