

# CP Course Notes

June 17, 2020

## Parallel Architectures

### Flynn's Taxonomy

#### Instruction Level

- Single Instruction (SI) - System in which all processors execute the same instruction
- Multiple Instruction (MI) - System in which different processors execute may execute different instructions

#### Data Level

- Single Data (SD) - System in which all processors operate on the same data
- Multiple Data (MD) - System in which different processors may operate on different data

#### Existing Architectures

	SD	MD
SI	SISD	SIMD
MI	MISD	MIMD

Figure 1: Possible processor architectures. MISD is the exception.

**SISD** The classic Von Neumann architecture, present in serial processors

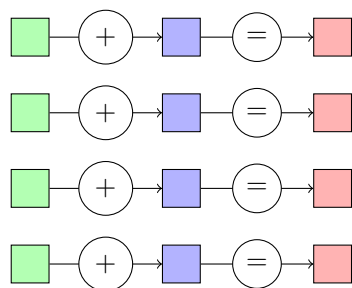


Figure 2: SISD Processing Example. One instruction is applied to each data item.

**SIMD** The data is divided among the processors and each item is subjected to the same sequence of instructions. Present in modern CPUs through Advanced Vector Extensions (AVX), as well as GPUs.

**MISD** Possible to envision as a pipeline at most, since it is not possible for multiple instructions to execute at the same time over the same data item.

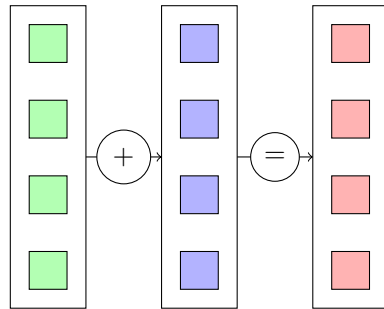


Figure 3: SIMD Processing Example. One instruction is applied to a block of data items.

*MIMD* Collection of autonomous processors executing independent programs, such as a distributed network or cluster. MIMD architectures can be categorized under four designations, each with different purposes, some of these architectures may be composed to form more complex systems.

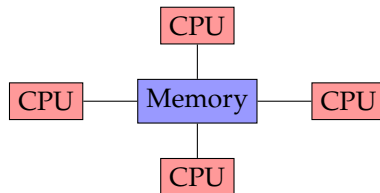


Figure 4: Uniform Memory Access (UMA). This architecture suffers from the need of synchronization among memory accesses. We can picture this as several CPUs accessing common RAM.

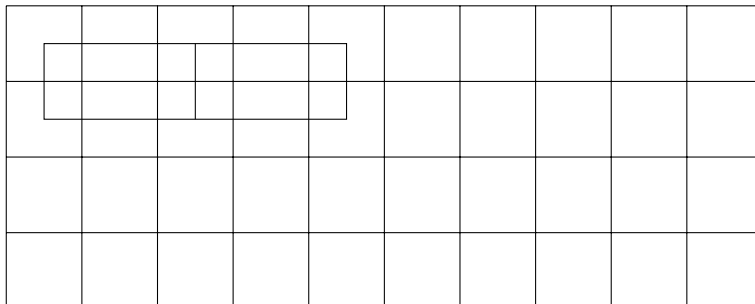


Figure 5: Non-Uniform Memory Access (NUMA). Common in distributed architectures.

## Software Taxonomies

Applying Flynn's taxonomy to software we observe the following patterns.

- Data Parallel (SIMD) - Parallelism that is a result of identical operations applied concurrently on different data items. This approach is difficult to apply to complex problems.
- Single Program, Multiple Data (SPMD) - A single program is run across multiple processing units (computers, processors, threads). Processing units execute the same code but do not work in lock-step.

## Memory

Memory can be split into shared memory or distributed memory.

### *Shared Memory*

Shared memory represents a global memory space, it is accessed by every processor. Each processor has a local cache (L1, L2, L3) which keep portions of the memory for faster access. The L1 and L2 caches are available at a core level while the L3 cache is shared among cores.

Shared memory is easier to program with, several frameworks are available (e.g. OpenMP, OpenACC). Low latency for data sharing between tasks. However the programmer is responsible for synchronization, furthermore the memory access is non-uniform (NUMA) and accesses to global memory (RAM) can be a bottleneck.

### *Distributed Memory*

Memory is shared between processes using messages (OpenMPI), this allows to exploit distributed computing architectures in a cost effective way.

Memory scales based on the number of processors and the access to local memory is fast.

This approach however is complex due to difficulties in programming communication effectively and data structures may be difficult to distribute. Parallel Performance

### *Performance*

In computing performance can be defined by two factors, computational requirements or resources. Computational requirements can be thought of "what needs to be done?", or efficacy, and computational resources can be thought in terms of "how much will it cost?", or efficiency.

$$Performance \sim \frac{1}{Resources \text{ for solution}} \quad (1)$$

### *Expectations*

In a scenario where we have  $p$  processors and each processor is rated at  $f$  MFLOP, should we see  $f \times p$  MFLOPS performance?

The answer is not as simple as "yes". Several causes may affect performance, while causes may interact with each other, they need to be understood separately.

### *Embarrassingly Parallel Computations*

Or for short EPCs, are computations that can be trivially divided into several independent parts able to be executed simultaneously. For *truly* EPCs there should be no interaction between processes, while in *nearly* EPCs the input and output is required to be distributed and combined in some way.

EPCs have potential to achieve maximal speedups in parallel platforms.

### Scalability

As previously discussed the performance of a parallel solution is subjected to several factors, namely scalability. Scalability is the ability of a parallel algorithm of achieving performance gains proportional to the number of processors and the size of the problem.

*Evaluation* To evaluate scalability we can start from the following metrics.

- Sequential Runtime ( $T_{seq}$ ) which is a function of problem size and architecture.
- Parallel Runtime ( $T_{par}$ ) which is a function of problem size and parallel architecture, that is the number of processors used in execution.

With that in mind we can define the speedup  $S_p$  as Equation 2, efficiency  $E_p$  as Equation 3 and finally the cost  $C_p$  as Equation 4, where  $p$  is the number of available processors and  $T_p$  is the execution time on a  $p$  processor system<sup>1</sup>.

<sup>1</sup> A parallel algorithm is cost-optimal if  $C_p = T_1$  or equivalently  $E_p = 1$

$$S_p = \frac{T_1}{T_p} \quad (2)$$

$$E_p = \frac{S_p}{p} \quad (3)$$

$$C_p = p \times T_p \quad (4)$$

### Amdahl's Law

Amdahl's law fixes the problem size and varies the number of processors, hence the denomination of Fixed Size Speedup. The law relates at the reduction of the execution time. Let  $f$  be the sequential fraction of a program,  $1 - f$  is the part that can be parallelized.

$$\begin{aligned} S_p &\leq \frac{T_1}{T_p} = \frac{T_1}{(f \cdot T_1) + \frac{(1-f)T_1}{p}} \\ S_p &\leq \frac{1}{f + \frac{1-f}{p}} \\ S_{p \rightarrow \infty} &\leq \frac{1}{f} \end{aligned} \quad (5)$$

*Scalability* When considering scalability, Amdahl's law refers to the ability of a parallel algorithm to achieve performance gains proportional to the number of processors and the size of the problem.

*Application* Amdahl's law is applicable under the following circumstances:

- When the problem size is fixed.
- Strong scaling ( $p \rightarrow \infty, S_p = S_\infty \rightarrow \frac{1}{f}$ ).
- Speedup bound is determined by the degree of sequential execution time in the computation.

*Example* If 90% of the computation can be parallelized, what is the maximum speedup achievable using 8 processors?

We first calculate the sequential part of the program:

$$\begin{aligned} 0.9 &= 1 - f \\ f &= 0.1 \end{aligned}$$

And then apply Amdahl's law:

$$\begin{aligned} S_8 &\leq \frac{1}{0.1 + \frac{1-0.1}{8}} \\ S_8 &\leq 4.7 \end{aligned}$$

### *Gustafson-Barsis' Law*

...SPEEDUP SHOULD BE MEASURED BY SCALING THE PROBLEM TO THE NUMBER OF PROCESSORS, NOT BY FIXING THE PROBLEM SIZE.  
- JOHN GUSTAFSON

Also denominated as Scaled Speedup, it is interested in larger problems when scaling. In other words, for the same time, how much more work can we do?

Considering  $a$  the non-parallelizable part and  $b$  as the parallelizable part, we calculate  $T_P$ , for  $P$  processors, as follows:

$$\begin{aligned} T_P &= a + P \cdot b \\ T_1 &= a + 1 \cdot b \\ &= a + b \end{aligned} \tag{6}$$

Given that the wall-clock execution time is always the same, the scaled speedup is calculated on the volume of data processed.

$$\begin{aligned} S_P &\leq \frac{T_P}{T_1} \\ &\leq \frac{a + P \cdot b}{a + b} \end{aligned} \tag{7}$$

Consider  $\alpha = \frac{a}{a+b}$  as the sequential fraction of the parallel execution time. We can then define the scaled speedup as follows:

$$\begin{aligned} S_P &\leq \alpha + P \cdot (1 - \alpha) \\ &\leq P - \alpha \cdot (P - 1) \end{aligned} \tag{8}$$

*Scalability* The ability of a parallel algorithm to achieve performance gains proportional to the number of processors and the size of the problem.

*Application* The Gustafson's law applies under the following scenarios:

- When the problem size can increase.
- When the number of processors increases.
- Speedup function include the number of processors.
- Can maintain or increase parallel efficiency as the problem scales.

*Example* An application executing on 64 processors spends 5% of the total time on non-parallelizable computations. What is the scaled speedup?

$$\begin{aligned} S_{64} &\leq P - \alpha \cdot (P - 1) \\ &\leq 64 - 0.05 \cdot (64 - 1) \\ &\leq 60.85 \end{aligned} \tag{9}$$

### Computational DAG

A parallel program maybe be represented as a directed acyclic graph (DAG), this graph has several properties, such as the span, which can be used to evaluate the potential performance of an algorithm.

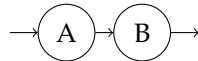


Figure 6: Serial Composition

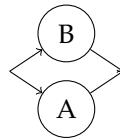


Figure 7: Parallel Composition

*Work* The total number of tasks of the graph is called *work*, this is equivalent to running the algorithm with a single processing unit ( $T_1$ ).

$$T_p \geq \frac{T_1}{P} \tag{10}$$

The work is equal whether considering  $T_1$  or  $T_p$ , that is:

$$T_1(A \cup B) = T_1(A) + T_1(B) \tag{11}$$

*Span* The span, or the critical-path length, is the minimum of sequential steps the algorithm must execute.

$$T_p \geq T_\infty \quad (12)$$

The span is different when comparing  $T_1$  and  $T_P$ .

$$\begin{aligned} T_\infty(A \cup B) &= T_\infty(A) + T_\infty(B) \\ T_\infty(A \cup B) &= \max\{T_\infty(A), T_\infty(B)\} \end{aligned} \quad (13)$$

### *Speedup*

The speedup on  $P$  processors is given by:

$$\frac{T_1}{T_P} \quad (14)$$

If  $\frac{T_1}{T_P} = P$  we have *linear* speedup. When  $\frac{T_1}{T_P} \geq P$  we have *superlinear* speedup, this however is not possible due to the work law (Equation 10).

### *Parallelism*

Parallelism can be defined as the average amount of work per step along the span, that is:

$$\frac{T_1}{T_\infty} \quad (15)$$

### *Work-Span Model*

Considers a greedy scheduler, that is, no worker is idle while there are tasks to execute.

- $T_P$  - running time with  $P$  workers.
- $T_1$  - running time with 1 worker, serial execution.
- $T_\infty$  - the time taken along the critical path.

The critical path is the sequence of tasks that has the longest execution time.

*Lower-Bound* A lower-bound can be calculated with the following formula:

$$\max\left(\frac{T_1}{P}, T_\infty\right) \leq T_P \quad (16)$$

### *Upper-Bound*

$$T_P \leq \frac{T_1 - T_\infty}{P} + T_\infty \quad (17)$$

### Parallel Programming

Parallel programs often start as sequential programs. This approach has the advantage of having a previous codebase to validate the new approach, furthermore it is easier to start with a sequential approach given the ease of programming and debugging.

To parallelize a program we first have to evaluate if it is worth the effort, to do so we first identify the hotspots with a profiler.

The parallelization process starts with the hotspots, applying small changes with testing in-between.

### *Finding Concurrency*

The first step in parallelizing a candidate program is to find concurrency, in other words, we have to find tasks that can be executed at the same time.

To do so, we need to consider flexibility, efficiency and simplicity.

### *Task Decomposition Guidelines*

*Flexibility* The program design should afford flexibility in the number and size of the tasks generated.

*Efficiency* Tasks should have enough work as to amortize the cost of creating and managing them. furthermore, they should be sufficiently independent so that managing dependencies does not become the bottleneck.

*Simplicity* The code must remain as simple and readable as possible so as to be debuggable and quick to understand and modify.

### *Data Decomposition Guidelines*

Often implied by task decomposition, data decomposition is a good starting point when the main computation is organized around manipulation of large data structures or similar operations are applied to different parts of the data structure.

*Flexibility* The size and number of data chunks should support a wide range of executions.

*Efficiency* Data chunks should generate considerable amounts of work to minimize the communication and management impact.

*Simplicity* Complex data compositions can get difficult to manage and debug.

### *Algorithmic Structure Design Space*

The high level approach of an algorithm can be split in to three possibilities, all approaches can then be split into two sub approaches, the linear or recursive approach.



- Organize by Tasks - the linear approach to this problem is to make use of task parallelism, the recursive approach is through divide and conquer.
- Organize by Data Decomposition - the linear approach of data decomposition is by making use of geometric decomposition (arrays), the recursive approach is through recursive data structures.
- Organize by Data Flow - the linear solution is to make use of the pipeline pattern, the recursive approach is based on event coordination.

## *Implementation Environment*

### *Program Structures*

*Single Program Multiple Data* In the SMPD approach all tasks execute the same program in parallel but each has its own set of data.

*Master/Worker Pattern* The master/worker is works by having a thread distribute work amongst the others, the workers execute concurrently and each worker takes tasks from the bag of tasks. It is suited for embarrassingly parallel problems.

*Loop Parallelism Pattern* This pattern exploits the iterative nature of some programs by parallelizing the execution of loop iterations. This requires the loop not to have dependencies among iterations.

*Fork/Join Pattern* A main task forks off some number of other tasks that then continue in parallel, the main task then waits for completion before moving on.

*Pipeline Pattern* Tasks are applied in sequence to data. Parallel Patterns

A parallel pattern is a recurring combination of task distribution and data access that solves a specific design problem in parallel algorithm design.

### *Nesting Pattern*

Nesting is the ability to hierarchically compose patterns. This pattern appears in both serial and parallel algorithms.

It is a compositional pattern, allowing other patterns to be composed in such way that task blocks can be replaced with another pattern.

## *Control Patterns*

### *Serial Control Patterns*

Structured serial programming is based on following patterns.

*Sequence* Ordered list of tasks that are executed in a specific order.

*Selection* Condition  $c$  is first evaluated. Either task  $a$  or  $b$  is executed depending on the true or false result of  $c$ .

*Iteration* Condition  $c$  is evaluated, if it is true  $a$  is evaluated again, afterwards  $c$  is evaluated again. This repeats until  $c$  is false.

*Recursion* Dynamic form of nesting allowing functions to call themselves.

### *Parallel Control Patterns*

Parallel control patterns extend serial control patterns. Each parallel control pattern is related to at least one serial control pattern.

*Fork-Join* Allows control flow to fork into multiple parallel flows, then rejoin later.

*Map* Performs a function over every element of a collection, it replicates a serial iteration pattern.

*Stencil* Elemental function accesses a set of "neighbors", stencil is a generalization of map.

*Reduction* Combines every element of a collection using an associative "combiner function".

*Scan* Computes all partial reductions of a collection. For every output in a collection, a reduction of the input up to that point is computed.

*Recurrence* More complex version of map, where loop iterations can depend on one another.

## *Data Management Patterns*

### *Serial Data Management Patterns*

*Random Read & Write* Memory locations indexed with addresses, used with pointers. Aliasing can cause problems when parallelizing the code.

*Stack Allocation* Useful for dynamically allocate data in LIFO manner. It preserves locality and when parallelized, since each threads gets its own stack thread locality is preserved.

*Heap Allocation* Useful when data cannot be allocated in LIFO manner. Slower and more complex when compared with stack allocations. A parallelized allocator should be used when allocating memory in parallel.

*Objects* Objects are language constructs to associate data with code to manipulate and manage that data.

### *Parallel Data Management Patterns*

*Pack* Used to eliminate unused space in a collection.

*Pipeline* Connects tasks in a producer-consumer manner.

*Geometric Decomposition* Arranges data into subcollections. Overlapping and non-overlapping compositions are possible.

*Gather* Reads a collection of data given a collection of indices.

*Scatter* The inverse of gather, given a collection of indices, it writes to such indices. Parallel Models & Dependencies

### *Parallelism, Correctness & Dependencies*

Parallel execution shall always be constrained by the sequence of operations needed to be performed for a correct result. Such execution must address control, data and system dependencies.

A *dependency* arises when one operation depends on an earlier operation to complete and produce a result before this later operation can be performed.

#### *Sequential Consistency*

Sequential consistency implies that the execution of two statements do not interfere with each other, meaning their running order is irrelevant. This means that the statements are *independent*, if the order of execution affects the computation outcome they are deemed *dependent*.

#### *Dependencies*

*True Dependency* A statement  $S_2$  has a true dependency on statement  $S_1$  if and only if  $S_2$  reads a value written by  $S_1$ . This dependency is also known as Read After Write (RAW).

Formally we can describe a true dependency as follows:

$$out(S_1) \cap in(S_2) \neq \emptyset$$

*Anti-Dependency* A statement  $S_2$  has an anti-dependency in statement  $S_1$  if and only if  $S_2$  writes a value read by  $S_1$ . This dependency is also known as Write After Read (WAR).

Formally we can describe an anti-dependency as follows:

$$in(S_1) \cap out(S_2) \neq \emptyset$$

*Output Dependency* A statement  $S_2$  has an output dependency on  $S_1$  if and only if  $S_2$  writes a variable written by  $S_1$ . This dependency is also known as Write After Write (WAW).

Formally we can describe an output dependency as follows:

$$out(S_1) \cap out(S_2) \neq \emptyset$$

*Loop-Carried Dependencies* A loop-carried dependency is a dependency between two statement instances in two different loop iterations. Synchronization

### *Algorithms, Programs & Processes*

*Sequential Algorithm* A sequential algorithm is a formal description of the behavior of a sequential state machine.

*Concurrent Algorithm* The description of a set of sequential state machines that cooperate through a communication medium is called a concurrent algorithm.

*Program* When the algorithm is written in a specific programming language.

*Process* The running instance of an algorithm and thus of a program as well.

### *Process Synchronization*

Process synchronization occurs when the progress of one or several processes depends on the behavior of other processes. Two types of process interaction require synchronization, competition and cooperation.

#### *Competition*

Competition occurs when processes compete to execute some statements and only one process at a time is allowed to execute them.

#### *Cooperation*

Cooperation occurs when one process can only progress after some event on another process.

*Barrier* A barrier is a set of control points, one per process involved in the barrier, such that each process is allowed to pass its control point only when all other processes have attained their own control points. From an operation point of view, each process has to stop until all other processes have arrived at their control point.

### *The Producer/Consumer Problem*

In the producer/consumer problem we have a producer that loops forever, producing data items, and a consumer that loops forever, consuming said data items.

It is required to ensure that, only produced data items are consumed and each produced item is consumed exactly once.

*Synchronization Barrier* One approach to the problem consists in using a synchronization barrier to ensure that only when a data item is produced the consumer is able to consume it. While this approach works it is far from efficient.

*Shared Buffer* The buffer has size greater than 1 and the underlying structure can be a queue or a circular array. Whenever the producer has a new item, it adds the item at the end of the structure, the consumer then withdraws items from the head of the structure.

This way the producer only waits whenever the structure is full, the consumer will only wait whenever the structure is empty.

### *The Mutual Exclusion Problem*

*Critical Section* A part (or several parts) that are required to be executed by a single process at a time.

*Solution* We are required to provide an entry algorithm as well as an exit algorithm, these are used to respectively enter and exit the critical section, ensuring that the critical section is only executed once.

*Definition* The mutual exclusion problem consists in implementing the operations to acquire and release a mutex in such a way that the following properties are always satisfied:

- **Mutual Exclusion** - at most one process at a time executes the critical section code.
- **Starvation-Freedom** - each invocation of `acquire_mutex` for any process  $p$ , made by  $p$  eventually terminates.

### *Solving Mutual Exclusion - Software Level*

This discusses how to solve mutual exclusion from a software point of view.

### Atomic Read/Write Registers

An atomic shared register satisfies the following properties:

- Each invocation  $op$  of a read or write operation:
  - Appears as if it was executed at a single point  $T(op)$  of the timeline.
  - $T(op)$  is such that  $T_{start}(op) \leq T(op) \leq T_{end}(op)$
  - For any two operation invocations  $op_1$  and  $op_2$  we have that  $op_1 \neq op_2 \Rightarrow T(op_1) \neq T(op_2)$ .
- Each read invocation returns the value written by the closest preceding write invocation in the sequence defined by the  $T(\dots)$  instants associated with the operation invocations.

### After You

We assume there are only two processes trying to access the critical region.

```

1 fn acquire_mutex(i) {
2     AFTER_YOU = i;
3     wait(AFTER_YOU != i);
4 }
```

Listing 1: Acquire Mutex Operation.

```

1 fn release_mutex(i) { }
```

Listing 2: Release Mutex Operation.

While this approach provides mutual exclusion, it does not guarantee progress. Furthermore this approach can cause starvation, for example, if process  $p_0$  tries to access the critical region and the other process  $p_1$  is busy,  $p_0$  will not be able to access the critical region since the variable `AFTER_YOU` will not change.

### Flag Array

```

1 fn acquire_mutex(i) {
2     FLAG[i] = up;
3     wait(FLAG[j] == down);
4 }
```

Listing 3: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     FLAG[i] = down;
3 }
```

Listing 4: Release Mutex Operation.

This approach provides mutual exclusion, however, once again it does not guarantee progress as it may cause a deadlock<sup>2</sup>. As we can see in Equation 18, if both processes are able to put their flag to up they will be waiting for each other indefinitely.

<sup>2</sup> A *deadlock* happens when two processes are waiting on each other.

$$\begin{aligned}
 p_0 &\rightarrow \text{FLAG}[0] = \text{up} \\
 p_1 &\rightarrow \text{FLAG}[1] = \text{up} \\
 p_0 &\rightarrow \text{wait} \\
 p_1 &\rightarrow \text{wait}
 \end{aligned}
 \tag{18}$$

### *Flag Array with Delayed Wait*

```

1 fn acquire_mutex(i) {
2     FLAG[i] = up;
3     while (FLAG[j] == up) {
4         FLAG[i] = down;
5         sleep(random());
6         FLAG[i] = up;
7     }
8 }

```

Listing 5: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     FLAG[i] = down;
3 }

```

Listing 6: Release Mutex Operation.

This solution provides mutual exclusion but is not able to guarantee progress as well. However only in specific cases where the processes are synchronized and the result from the `random()` call is consistently the same for both processes the program will be stuck, creating a livelock<sup>3</sup>.

This solution may also suffer from starvation, in the case that a process  $p_0$  acquires the mutex and  $p_1$  is waiting, it is possible for  $p_0$  to release and re-acquire the mutex without  $p_1$  getting access.

<sup>3</sup> A *livelock* is similar to a deadlock, except that the states of the processes involved in the livelock constantly change with regard to one another, none progressing.

### *Mixed Approach*

Just like the previous schemes, this one is also applicable for two processes only.

```

1 fn acquire_mutex(i) {
2     FLAG[i] = up;
3     AFTER_YOU = i;
4     wait(FLAG[j] == down || AFTER_YOU != i);
5 }

```

Listing 7: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     FLAG[i] = down;
3 }

```

Listing 8: Release Mutex Operation.

By mixing the previous approaches we are able to achieve progress, since even if both flags are up the AFTER\_YOU variable will make one proceed to the critical region. While this solution provides mutual exclusion and progress, it will only work for two processes.

### *Mixed Approach for N processes*

```

1 fn acquire_mutex(i) {
2     for idx in 1 to (n-1) {
3         FLAG_LEVEL[i] = idx;
4         AFTER_YOU[idx] = i;
5         wait(
6             forall k != i : FLAG_LEVEL[k] < idx ||
7             AFTER_YOU[idx] != i
8         );
9     }
10 }

```

Listing 9: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     FLAG_LEVEL[i] = 0;
3 }

```

Listing 10: Release Mutex Operation.

This approach generalizes the one described in subsection , allowing for  $N$  processes to use the mutex. The intuition behind it so that whenever  $n$  processes try to access the critical region, they will go through  $n - 1$  levels, the process reaching the last one will gain access to the region.

$p_i$  is allowed to progress to level  $idx + 1$  if, from its point of view, one of the following is true:

- All other processes are at a lower level ( $\forall k \neq i : \text{FLAG\_LEVEL}[k] < idx$ ).
- It was not the last one to enter the level  $idx$  ( $\text{AFTER\_YOU}[idx] \neq i$ ).

### *Solving Mutual Exclusion - Hardware Level*

#### *The test&set & reset primitives*

Consider that both primitives are atomic and  $X$  is a shared register initialized to 1.



- test&set will set  $X$  to 0 and return the previous value.
- reset will write 1 to  $X$

```

1 fn acquire_mutex(i) {
2     while (r != 1) {
3         r = X.test&set();
4     }
5 }

```

Listing 11: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     X.reset();
3 }

```

Listing 12: Release Mutex Operation.

The implementation is based on the following invariant:

$$X + \sum_{i=1}^n r_i = 1$$

### *The swap primitive*

Consider a shared register  $X$ , the  $\text{swap}(v)$  primitive will assign  $v$  to  $X$  and return the previous value of  $X$ .

We can think of the previous instructions, test&set and reset as swap operations with fixed values, respectively,  $\text{swap}(0)$  and  $\text{swap}(1)$ .

```

1 fn acquire_mutex(i) {
2     r = 0;
3     while (r != 1) {
4         r = X.swap(r);
5     }
6 }

```

Listing 13: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     X.swap(r);
3 }

```

Listing 14: Release Mutex Operation.

The invariant from test&set applies to swap as well. Since this approach is just a re-implementation of the previous ones with a different instruction, it offers the same guarantees.

### *The compare&swap primitive*

Let  $X$  be a shared register and old and new be two values. The compare&swap primitive returns a boolean value and may be defined as follows:

```

1 fn compare&swap(old, new) {
2     if (X == old) {
3         X = new;
4         return true;
5     } else {
6         return false;
7     }
8 }

```

Listing 15: compare&amp;swap implementation.

The code above is considered to be executed atomically.

Consider  $X$  to be an atomic register initialized to 1, implementing mutual exclusion with compare&swap we have.

```

1 fn acquire_mutex(i) {
2     while (r != 1) {
3         r = X.compare&swap(1, 0);
4     }
5 }

```

Listing 16: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     X = 1;
3 }

```

Listing 17: Release Mutex Operation.

This and the previous algorithms are not starvation free.

### *Deadlock & Starvation Free Algorithm*

```

1 fn acquire_mutex(i) {
2     FLAG[i] = up;
3     wait(TURN == i || FLAG[TURN] == down);
4     LOCK.acquire_lock(i);
5 }

```

Listing 18: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     FLAG[i] = down;
3     if (FLAG[TURN] == down) {
4         TURN = (TURN % n) + 1;
5     }
6     LOCK.release_lock();
7 }

```

Listing 19: Release Mutex Operation.

While this approach provides mutual exclusion, progress and no starvation, it does not provide fairness, meaning that we may wait for a long time until we acquire the lock.

*The fetch&add primitive*

Let  $X$  be a shared register, the primitive fetch&add will atomically increment the register  $X$  and return the new value. There are a couple of variants of this primitive in which the value return is the previous or the incremented value is not 1 but rather passed as an argument.

```

1 fn acquire_mutex(i) {
2     my_turn = TICKET.fetch&add();
3     while(NEXT != my_turn) { }
4 }
```

Listing 20: Acquire Mutex Operation.

```

1 fn release_mutex(i) {
2     NEXT = NEXT + 1;
3 }
```

Listing 21: Release Mutex Operation.

While the `release_mutex` is not atomic, there is no problem since only one process will execute `release_mutex`.

*Locking Strategies**Coarse-Grained Synchronization*

Simple approach, use a single lock, methods are always executed in mutual exclusion, eliminating all concurrency within the object.

This approach wastes resources, for example, if there exists two objects and three threads, whenever an object is being used the other threads cannot use the other.

This approach causes false conflicts, only works under fault-free scenarios and "reverts" the system into a sequential system.

*Fine-Grained Synchronization*

Split the object into multiple independently-synchronized components.

*Optimistic Synchronization*

Check if the operation can be done, if the operation can be done, lock and recheck if the operation can still be done.

While this strategy is cheaper than hand-over-hand locking it is not starvation free and mistakes are expensive as safety may be compromised.

*Lazy Synchronization*

Make common operations fast and delay the hard work. Separate the complex operations into two phases, logical, marking the com-

ponent, and physical, actually executing the operation.

## *Lock-Free Synchronization*

### *Progress Conditions*

*Obstruction Freedom* At any point a single thread executed in isolation will complete its operation in a bounded number of steps. All lock-free algorithms are obstruction-free.

This approach is the simplest one however it is too weak to guarantee progress.

*Lock Freedom* When the program threads are run sufficiently long, at least one of the threads makes progress (for a sensible definition of progress).

In most cases it is strong enough, while more complex than an obstruction free approach it is simpler than wait free approaches. With limited contention it will behave as wait free.

*Wait Freedom* Every operation has a bounded number of steps the algorithm will take before the operation completes. Being the stronger approach it is also the most complex. Concurrency Errors

### *Data Races*

Code is supposed to be executed atomically since there are multiple dependent instructions manipulating the same data. This creates the possibility to several possible interleavings.

```

1 void Bank::Deposit(int amount)
2 {
3     int balance = getBalance();
4     setBalance(balance + amount);
5 }
```

Listing 22: Deposit Operation.

```

1 void Bank::Withdraw()
2 {
3     int balance = getBalance();
4     setBalance(balance - amount);
5 }
```

Listing 23: Withdraw Operation.

As we can see in the above example, if the code is run in different threads the withdraw may not be reflected on the final balance, this happens because a thread (Thread 1) may read the balance, then the other thread (Thread 2) will update it, when Thread 1 continues the balance value it has is now outdated.

### Data Race Detection

Data race detection may be executed by static or dynamic analysis. Static analysis depends on compiler features and may report false positives. Dynamic analysis happens at runtime, all reported data races, are data races.

*Happens-Before Algorithm* The Happens-Before algorithm defines a partial order for events in a set of concurrent threads. In a single thread, happens-before reflects the temporary order of event occurrence. Between threads,  $A$  happens before  $B$  if  $A$  is an unlock access in one threads and  $B$  is a lock access in a different thread<sup>4</sup>.

<sup>4</sup> Data races between threads are possible if accesses to shared variables are not ordered.

$$\text{if } (a = \text{unlock}(\mu) \wedge b = \text{lock}(\mu)) \Rightarrow a \rightarrow b \quad (19)$$

*Lock-Set Algorithm* The algorithm may fail and the program still be data race free (false positives), this happens since the algorithm check a sufficient condition for data-race freedom. This algorithm does not check for data races but rather for a consistent lock discipline. The algorithm defines two data structures,  $\text{LocksHeld}(t)$ , the set of locks currently held by the thread  $t$  and  $\text{LockSet}(x)$ , the set of locks that could potentially be protecting variable  $x$ .

- When thread  $t$  acquires lock  $l$  we have:

$$\text{LocksHeld}(t) = \text{LocksHeld}(t) \cup \{l\}$$

- When thread  $t$  releases lock  $l$  we have:

$$\text{LocksHeld}(t) = \text{LocksHeld}(t) \setminus \{l\}$$

- When thread  $t$  accesses location  $x$  we have:

$$\text{LockSet}(x) = \text{LockSet}(x) \cap \text{LocksHeld}(t)$$

A data race is detected if  $\text{LockSet}(x)$  becomes empty.

### High-Level Data Races

Also known as high-level atomicity violations. This happens when seemingly atomic code has interleavings that lead to a data race.

```

1 class Pair {
2     synchronized int getX() {
3         return pair.x;
4     }
5
6     synchronized int getY() {
7         return pair.y;
8     }
9
10    synchronized void setPair(int x, int y) {
```

```

11     this.x = x;
12     this.y = y;
13 }
14
15 boolean areEqual() {
16     int x = this.getX(); // synchronized
17     int y = this.getY(); // synchronized
18     return x == y;
19 }
20 }

```

Listing 24: Class Pair definition, notice the synchronized methods.

Seemingly the methods are well synchronized, however if a thread calls `areEqual`, it is possible for the pair class to be changed between get calls, running an interleaving like the following.

```

1  getX();
2  setPair(x', y');
3  getY(); // y is now y'

```

Listing 25: Possible incorrect interleaving. Assuming  $y \neq y'$ .

### Stale-Value Errors

Stale-Value errors are caused by the privatization of data.

```

1  void setYToXTimesTwo() {
2      int x = getX();
3      setY(2 * x); // x may have changed
4  }
5
6  @Atomic
7  synchronized int getX() {
8      return x;
9  }
10
11 @Atomic
12 synchronized int setY(int y) {
13     this.y = y;
14 }

```

Listing 26: Stale-Value error example.

### Detecting High-Level Data Races

A view of an atomic block  $B$ , named  $V(B)$ , is the set of variables accessed inside the atomic code block  $B$ . We can define a *read view* of  $B$  as  $V_R(B)$  as the set of variables read inside the atomic code block  $B$ . Analogously the *write view* of  $B$ ,  $V_W(B)$  is the set of variables written inside the atomic code block  $B$ .

```

1 public int getX() {
2     return this.x;
3 }

```

Listing 27: The `getX` has the read view  $V_R(\text{getX}) = \{x\}$  and the write view  $V_W(\text{getX}) = \{\}$ .

*Direct Correlation* There is a direct correlation between a read variable  $x$  and a written variable  $y$  if in the dependency graph  $D$  there is a path from  $x$  to  $y$ .

*Common Correlation* There is a common correlation between read variables  $x$  and  $y$  if, in a dependency graph  $D$ , there is a written variable  $z$  such that:

$$z \neq x, z \neq y : \exists (x \rightarrow z, y \rightarrow z)$$

Example of an high-level data race detection:

$$\begin{aligned}
 V_{\max}(T_1) &= \{x, y\} \\
 V_{\max}(T_2) &= \{x\}, \{y\} \\
 V_A &= V_{\max}(T_1) \cap \{x\} \\
 V_B &= V_{\max}(T_1) \cap \{y\} \\
 V_A &\not\subseteq V_B \wedge V_B \not\subseteq V_A
 \end{aligned} \tag{20}$$

### Ordering Violation

Missing or incorrect synchronization between two processes.

```

1 work = null;
2 LaunchThread(new Thread(2));
3 work = new Work();

```

Listing 28: Simple Producer, running on Thread 1

```

1 ConsumeWork(work);

```

Listing 29: Simple Consumer, running on Thread 2

It is possible for `work` to be consumed before it was initialized, thus the correct operation ordering for Thread 1 would be the following.

```

1 work = new Work();
2 LaunchThread(new Thread(2));

```

Listing 30: Simple Correct Producer, running on Thread 1

## Unintended Sharing

This happens when state is shared between processes unintentionally.

```

1 void inc() {
2     static int local = 0;
3     local += 1;
4 }

```

Listing 31: Example of unintended sharing.

Since the static keyword is used the variable `local` will be kept among runs of the `inc` function, it also has the effect that the variable will be shared by threads running the same function.

## Deadlocks

Deadlocks happen when processes wait on each other, this can be expressed as a cycle in a task dependency graph.

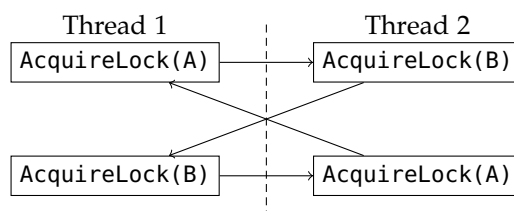


Figure 8: Deadlock dependency graph.

## System Model

- We have a finite number of resources.
- Resources are organized into classes.
- Processes compete for accessing resources.
- If a process requests an instance of a resource class, any instance of that class must satisfy the process.

## Resource Usage Protocol

- **Request** - the process either gets an instance of the resource immediately or waits until one is available.
- **Use** - the process can operate on its resource instance.
- **Release** - the process releases its resource instance.

## Deadlock Definition

*Informal Definition* A set of two or more processes are deadlocked if:

- They are blocked.



- Each is holding a resource.
- Each is waiting to acquire a resource held by another process in the set.

*Formal Definition* In a formal way the conditions necessary for a deadlock are the following:

- **Mutual Exclusion** - only one process can use a resource at a time.
- **Hold & Wait** - a process holding at least one resource is waiting to acquire additional resources which are currently held by other processes.
- **No Preemption** - a resource can only be released voluntarily by the process holding it.
- **Circular Wait** - a cycle of process requests exists ( $P_0 \rightarrow P_1 \rightarrow \dots P_{n-1} \rightarrow P_0$ ).

### *Deadlock Prevention*

Restrict the way requests can be made.

*Mutual Exclusion* Not required for shared resources, however it must hold for non-shareable files.

*Hold & Wait* Must guarantee that whenever a process requests a resource, it does not hold any other resources. Require the process to request and allocate all its resources before it begins execution. Low resource utilization; possibly resulting in starvation.

*No Preemption* If a process that is holding some resources requests another resource that cannot be immediately allocated to it, then all resources currently being held are released. Preempted resources are added to the list of resources for which the process is waiting. Process will be restarted only when it can regain its old resources, as well as the new ones that it is requesting.

*Circular Wait* Impose a total ordering of all resource types, requiring that each process requests resources in an increasing order of enumeration.

### *Deadlock Avoidance*

Requires the system to have some additional *à priori* information available.

- Requires that each process declare the maximum number of resources of each type that it may need.

- The deadlock-avoidance algorithm dynamically examines the resource-allocation state to ensure that there can never be a circular-wait condition.
- Resource-allocation state is defined by the number of available and allocated resources and the maximum demands of the processes.

### *References*