

# PHYS52015 Core Ib: Introduction to High Performance Computing (HPC)

Session III: OpenMP (2/3) Christopher Marcotte

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#### **Outline**





Race Conditions Thread communication Barriers

pollev.com/christophermarcotte820

#### **Race Conditions**



- ▶ Race conditions arise from interdependency on data access across threads
- ► These result from dependencies (read-write, write-read, write-write) on a position in memory
- ⇒ These manifest as indeterminacy in your program, and are sensitive to the memory model of your machine and how many threads are used.

#### **Race Conditions**



- ▶ Race conditions arise from interdependency on data access across threads
- ► These result from dependencies (read-write, write-read, write-write) on a position in memory
- ⇒ These manifest as indeterminacy in your program, and are sensitive to the memory model of your machine and how many threads are used.

Example: write a short program to compute this sum manually,

$$\sum_{n=1}^{N} n = \frac{N(N+1)}{2}$$

for some (not too large) N.

## Race condition example



```
#include <omp.h>
#include <stdio.h>
int main() {
   int sum=0:
   const int N=100:
   #pragma omp parallel for
   for (int n=1: n< N+1: n++){
                      // Race condition here
      sum +=n:
   printf("Result is %d. It should be %d\n", sum, N*(N+1)/2);
   return 0:
```

- ▶ The variable sum is accessed for both reading and writing in parallel
  - $\Rightarrow$  Thread 0 reads sum into (local) sum
  - $\Rightarrow$  Thread 1 reads sum into (local) sum
  - $\Rightarrow$  Thread 0 increments (local) sum by N/p
  - $\Rightarrow$  Thread 1 increments (local) sum by N/p
  - $\Rightarrow$  Thread 0 writes (local) sum to sum
  - $\Rightarrow$  Thread 1 writes (local) sum to sum
- ▶ So long as the variables are initialised correctly, the race condition means you'll have  $sum \le N(N+1)/2$ .

## Concept of building block: Race Conditions



- Content
  - Race conditions
- Expected Learning Outcomes
  - ► The student can identify potential race conditions in an OpenMP parallel code
  - ▶ The student can ameliorate race conditions in simple loops using alternative OpenMP constructions

#### Thread communication



We distinguish two different communication types:

- Communication through the join
- Communication inside the BSP part

Critical section: Part or code that is ran by at most one thread at a time – a manual serialisation block.

Reduction: Join variant, where all the threads reduce a value into one single value.

- $\Rightarrow$  Reduction maps a vector of  $(x_0, x_1, x_2, x_3, \dots, x_{p-1})$  onto one value x, i.e. we have an all-to-one data flow
- ⇒ Inter-thread communication realised by data exchange through shared memory
- $\Rightarrow$  Fork can be read as one-to-all information propagation (done implicitly by shared memory)

#### **Critical sections**



```
#pragma omp critical (mycriticalsection)
  x *= 2:
#pragma omp critical (anothersection)
  x *= 2:
#pragma omp critical (mycriticalsection)
  x /= 2;
```

- Name is optional (default name i.e. does not block with other sections with a name)
- ► Single point of exit policy ⇒ return, break, ... not allowed within critical section
- For operations on built-in/primitive data types, an atomic operation is usually the better, i.e. faster choice

## critical vs. single regions



- ritical sections specify that code is executed by one thread at a time
- single sections specify that a section of code should be executed by a single thread (not necessarily the manager thread)

```
for (int i=0: i<size: i++){
        a[i] = i;
#pragma omp critical
for (int i=0: i<size: i++){
        a[i] = a[i]*2:
#praama omp sinale
for (int i=0; i<size; i++){</pre>
        a[i] = a[i]/2;
```

With p = 1 threads, will a[i] == i? With p > 1 threads, will a[i] == i?

## Case study: scalar product Serial starting point:



```
double result = 0;
for( int i=0; i<size; i++ ) {
  result += a[i] * b[i];
}</pre>
```

#### A parallel variant:

```
double result = 0.0;
#pragma omp parallel
{
   double myResult = 0.0;
   #pragma omp for
   for( int i=0; i<size; i++ ) {
      myResult += a[i] * b[i];
   }
   #pragma omp critical
   result += myResult;
}</pre>
```

## Case study: scalar product A parallel variant:



```
double result = 0.0;
#pragma omp parallel
{
   double myResult = 0.0;
   #pragma omp for
   for( int i=0; i<size; i++ ) {
      myResult += a[i] * b[i];
   }
   #pragma omp critical
   result += myResult;
}</pre>
```

#### Observations:

- Avoid excessive synchronisation
- Type of operation is called *reduction* (as defined before)
- ▶ We may not use result to accumulate because of races
- We may not hide result as we then loose access to outer variable

## Recap: Requirements for parallel loops



```
#pragma omp parallel for
{
  for( int i=0; i < size; i++ ) {
    a[i] = a[i] * 2;
  }
}</pre>
```

- ► Loop has to follow plain initialisation-condition-increment pattern:
  - Only integer counters
  - Only plain comparisons
  - Only increment and decrement (no multiplication or any arithmetics)
- ► Loop has be countable (otherwise splitting is doomed to fail).
- ► Loop has to follow single-entry/single-exit pattern.
- Loop copies all share the memory.

#### Consistency observation:

- All attributes are shared
- ► Besides the actual loop counter (otherwise splitting wouldn't work)
- ⇒ There has to be support for non-shared data important for memory consistency!

#### The shared default clause



```
double result = 0;
#pragma omp parallel for
for( int i=0; i<size; i++ ) {
   result += a[i] * b[i]; // race, don't worry about it here
}</pre>
```

```
double result = 0;
#pragma omp parallel for shared(result)
for( int i=0; i<size; i++ ) {
   result += a[i] * b[i]; // race, don't worry about it here
}</pre>
```

- By default, all OpenMP threads share all variables, i.e. variables declared outside are visible to threads
- ► This sharing can be made explicit through the clause shared
- Explicit shared annotation is good practice (improved readability)

#### Thread-local variables



```
double result = 0;
for( int i=0; i<size; i++ ) {
  double result = 0.0;
  result += a[i] * b[i];
}</pre>
```

#### Without OpenMP pragma:

- ► C/C++ allows us to "redefine" variable in inner scope
- ► Hides/shadows outer result
- We may not forget the second initialisation; otherwise garbage

#### Thread-local variables



```
double result = 0;
#pragma omp parallel for
for( int i=0; i<size; i++ ) {
  double result = 0.0;
  result += a[i] * b[i];
}</pre>
```

#### With OpenMP pragma:

- OpenMP introduces (concurrent) scope of its own
- Scope-local variables are thread-local variables
- ► These are *not* shared *private variables in local scope*

## shared **vs.** private **clauses**



```
double result = 0;
#pragma omp parallel for private(result)
for( int i=0; i<size; i++ ) {
  result += a[i] * b[i];
}</pre>
```

- private is the counterpart of shared, i.e. each thread works on its own copy
- ▶ Basically, the copying is similar to the following fragment:

```
double result = 0;
#pragma omp parallel
{
  double result;
  #pragma omp for
  for( int i=0; i<size; i++ ) {
    result += a[i] * b[i];
  }
}</pre>
```

## shared vs. private clauses



```
double result = 0;
#pragma omp parallel
{
   double result;
    #pragma omp for
   for( int i=0; i<size; i++ ) {
     result += a[i] * b[i];
   }
}</pre>
```

- ⇒ In this example, result within thread is not initialised (garbage)!
- ⇒ In this example, data within result is lost throughout join!
- ► This code will not do what we want (the inner product of a & b)
- ⇒ We will first look at other copy policies and then come back to properly parallelise the inner product calculation a few different ways next time!

## default(none) and scoping



```
double result = 0;
#pragma omp parallel for default(none) private(result) shared(a,b,size)
for( int i=0; i<size; i++ ) {
    result += a[i] * b[i];
}</pre>
```

- Using default(none) tells the compiler that, by default, the privacy/sharing of all variables is unspecified.
- ⇒ This means the programmer is responsible for all explicit sharing.
- ▶ This can dramatically improve legibility of your code when using a large number of temporary variables.

## **Example**



## Example



#### Observations:

- ▶ If we comment out x=i, x is not properly initialised.
- x in the last line always equals 40.

Now remove initialisation and change private(x) to firstprivate(x).

Does the final value of x change?

What if we use lastprivate(x)? What value is x on the last line?

## **Copy policies**



- default Specifies default visibility of variables
- ▶ firstprivate Variable is private, but initialised with value from surrounding
- ▶ lastprivate Variable is private, but value of very last iteration is copied over to outer variable

## Concept of building block



- Content
  - Introduce three types of data flow/usage of shared memory: making memory available to all threads throughout fork, sharing data throughout computation, reducing data at termination
  - Introduce private variables
  - Study semantics and usage of critical sections
- Expected Learning Outcomes
  - ► The student *can use* critical sections
  - ▶ The student *knows difference* between private and shared variables
  - ► The student can identify race conditions and resolve them through critical sections for given code snippet

#### **Barriers**



- ► Barriers are synchronisation points in your code
- ▶ Places where each threads waits for all other threads to arrive at the barrier before proceeding
- ► Lots of implicit barriers when using default constructs in OpenMP

#### **Barriers**



- ► Barriers are synchronisation points in your code
- ▶ Places where each threads waits for all other threads to arrive at the barrier before proceeding
- Lots of implicit barriers when using default constructs in OpenMP





```
Thread 0 prints 1
Thread 0 prints 2
Thread 1 prints 1
Thread 1 prints 2
Thread 3 prints 1
Thread 3 prints 2
Thread 2 prints 1
Thread 2 prints 2
```



```
// ...
    #pragma omp parallel
{
        printf("Thread %d prints 1\n", omp_get_thread_num());
        #pragma omp barrier
        printf("Thread %d prints 2\n", omp_get_thread_num());
}
// ...
```



```
// ...
    #pragma omp parallel
    {
        printf("Thread %d prints 1\n", omp_get_thread_num());
        #pragma omp barrier
        printf("Thread %d prints 2\n", omp_get_thread_num());
    }
// ...
```

```
Thread 0 prints 1
Thread 3 prints 1
Thread 2 prints 1
Thread 1 prints 1
Thread 3 prints 2
Thread 1 prints 2
Thread 2 prints 2
Thread 0 prints 2
```

#### Barriers and nowait clause



- ▶ The implicit barriers here mean all threads complete the first loop before starting the second loop
- $\Rightarrow$  What if we want the second loop to run as soon as there are threads available to do so?

#### Barriers and nowait clause



```
#pragma omp parallel
   #pragma omp for nowait
   for ... {
   } // no more implicit barrier here
   #pragma omp for
   for ... {
   } // still an implicit barrier here
```

- Adding the nowait clause to the first loop means that the threads will immediately progress to the second loop.
- ⇒ Since the #pragma omp for splits the range, the progressing threads will not necessarily work on the same range values in both loops, this can be wuite dangerous!

## Concept of building block: Barriers

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- Content
  - Barriers
- Expected Learning Outcomes
  - The student can identify implicit synchronisation points in an OpenMP parallel code
  - ► The student can **add** explicit synchronisation points in an OpenMP parallel code