**Blue Waters Petascale Semester Curriculum v1.0**

**Unit 4: OpenMP**

**Lesson 1: Race Conditions**

**Exercise Instructions for Students**

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**Race conditions in OpenMP**

**(using the Mandelbrot set as an example)**

In this lesson, you will learn about race conditions using the Mandelbrot set.

**Race conditions**

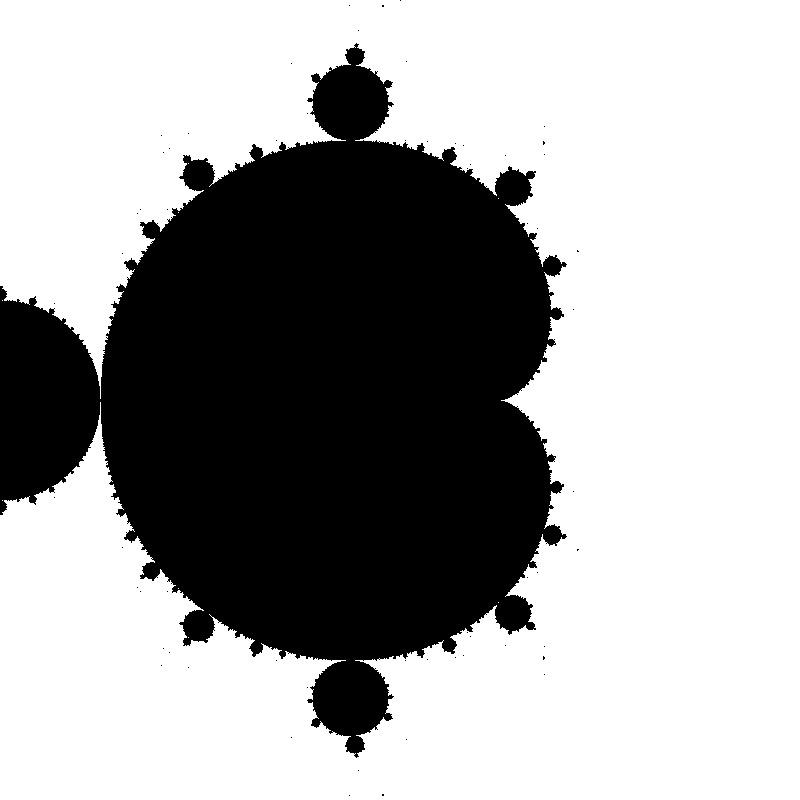
A race condition is when different tasks in the program have operations that can interfere with each other and the correctness of the program depends on the order in which those operations are performed. Typically, a race condition occurs when multiple tasks use the same variable and the order in which they do it affects the result. This lesson looks at two types of race conditions and their solutions. We’ll start by illustrating each kind with non-programming examples.

For the first example, consider the kitchen in a restaurant as an example of a parallel system. The dishes to be prepared are the tasks that need to be run and the cooks are the processors or cores available to run those jobs. Representing the shared variable is a single cutting board which is needed to prepare several of the dishes. This might be ok if only one cook at a time ends up using it, but problems could occur if 2 cooks attempt to use it at the same time, resulting in cross contamination between the dishes they are making or a shouting match between them. The situation is easily resolved by getting separate cutting boards for each cook, corresponding to using separate variables in each task.

As an example of the second kind of race condition, consider a group of chefs making an opera cake (shown to the right; image by [Arnold Gatilao](https://www.flickr.com/photos/58819758@N00), 2007, from <https://en.wikipedia.org/wiki/Opera_cake> and shared with a CC 2.0 license). The defining feature of this cake is its large number of layers. In principle, these layers could be made by different chefs and then added together to form the final product. However, the tray on which the cake is built (and the partially-built cake on it) cannot be used simultaneously by multiple chefs since the layers must be deposited one at a time. Because the layers made by each chef must be combined, this situation cannot be entirely resolved by providing more resources and is thus different from the first type of race condition. The need to combine different parts means that resolving this sort of situation requires what is called a reduction rather than simply providing additional resources.

**Mandelbrot set**

To illustrate race conditions with code, we’ll use the Mandelbrot set. This is a famous fractal which is defined by a mathematical operation. The details are not crucial to understanding the example, but the basic idea is to treat each 2D point as a complex number (i.e. one with real and imaginary parts) and seeing for which points the result of an iterative process stays bounded. (Refer to the Wikipedia page: <https://en.wikipedia.org/wiki/Mandelbrot_set> for more detail.) The following is a depiction of the Mandelbrot set created using the given code:



The black points are in the Mandelbrot set and the white points are not. The image is centered around the value 0, which represents the point (0,0).

To create this image yourself, get the given code mandelbrot.c and compile and run it:

gcc -Wall -o mandelbrot mandelbrot.c

./mandelbrot.c output.ppm

output.ppm is the name of the output file to create. You can view it by clicking on it.

**Parallel for loops in OpenMP**

The only part of the program we need to focus on to demonstrate race conditions, however, is the following loop in main:

//set pixels

for (int j = 0; j < numRows; j++) {

for (int i = 0; i < numCols; i++) {

x = ((double)i / numCols -0.5) \* 2;

y = ((double)j / numRows -0.5) \* 2;

pixels[i][j] = mandelbrot(x,y);

}

}

This loop sets the color for each pixel of the output image. The pixel coordinates are i and j. First, the loop converts each coordinate to a real value between -1 and 1. Then, it calls the mandelbrot function and stores the result in the 2D array pixels.

We will use OpenMP to parallelize the j loop here. The idea is that different iterations of the j loop are independent since the color of each pixel depends only on its coordinates and thus all the iterations are independent. (This idea is nearly correct, but our first attempt won’t quite work.) Converting the for loop into a parallel for loop partitions the loop iterations between several tasks, each of which can run in parallel.

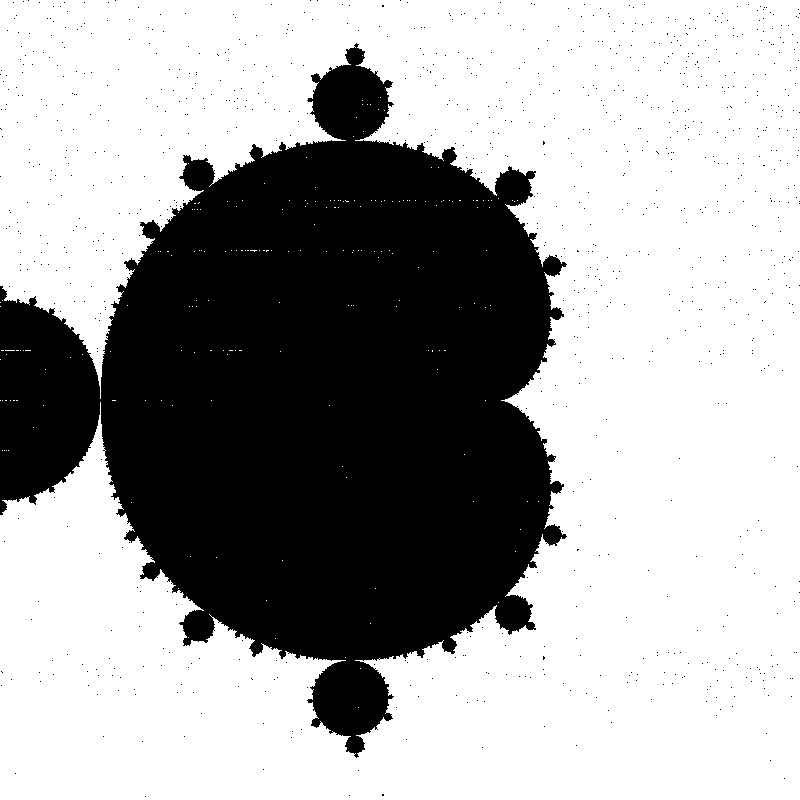
The syntax to make a parallel for loop in OpenMP is to add a pragma (basically a suggestion to the compiler) right before it. For the loop above, this looks like

#pragma omp parallel for

for (int j = 0; j < numRows; j++) {

The omp part specifies that this pragma concerns OpenMP. The words parallel for mean that you want to create multiple tasks and split the iterations of the following line (which must be a for loop) between them. By default, this creates one task per core; other instructions can be added to create a different number of tasks or to customize them in other ways (as we’ll see).

If you add this line, recompile,[[1]](#footnote-1) and then run the program, it runs noticeably faster. Prepending the command lines with time lets us measure the running time of each version. On my system,[[2]](#footnote-2) the original (serial) version took 3.39 seconds and the parallel version took only 0.58 seconds, for a speedup of 3.39/0.58=5.84. Unfortunately, it also gets the wrong answer since it suffers from a race condition based on unnecessarily shared variables. Specifically, the program produces an image such as the following:



This image is largely correct, but has a number of artifacts, namely the “haze” of dark pixels around the dark figure and the partially white streaks in the figure. Both of these are caused by a race condition. Examine the code with this in mind and see if you can figure it out before continuing.

**Fixing the race condition**

The issue is that the variables x and y are shared between the tasks. This means that sometimes the values of x and y are changed during the computation and the wrong color is selected for a given pixel.

In order to fix the race condition, we need to end the interference between tasks by having each task use its own copy of the variables. There are two ways to do this with the given code. The first is simply to delete the existing declaration of these variable and to declare them inside the loop as follows:

#pragma omp parallel for

for (int j = 0; j < numRows; j++) {

for (int i = 0; i < numCols; i++) {

double x = ((double)i / numCols -0.5) \* 2;

double y = ((double)j / numRows -0.5) \* 2;

pixels[i][j] = mandelbrot(x,y);

}

}

Now x and y are local to the loop. The compiler knows to create separate variables for each task in this case.

The other solution is to explicitly say that the variables shouldn’t be shared. This is done by adding the qualifier private to the pragma as follows:

#pragma omp parallel for private(x,y)

for (int j = 0; j < numRows; j++) {

for (int i = 0; i < numCols; i++) {

x = ((double)i / numCols -0.5) \* 2;

y = ((double)j / numRows -0.5) \* 2;

pixels[i][j] = mandelbrot(x,y);

}

}

Now the compiler knows that these variables should be private to each task. (There is also a qualifier shared that tells it that variables should be shared between tasks.)

Try each of these changes by modifying the code, recompiling, and checking that the correct image is created.

**Using a reduction**

To see a race condition that can be resolved using a reduction, we’re going to modify the program to count the number of black pixels in the image, which corresponds to the number of times the mandelbrot function returns 0. To count these pixels, we’ll create a counter variable numBlack and add an if statement to check the return value of mandelbrot. Modify the main loop so it matches the following (the modifications are shown with underlines):

int numBlack = 0; //number of black pixels in image

#pragma omp parallel for private(x,y)

for (int j = 0; j < numRows; j++) {

for (int i = 0; i < numCols; i++) {

x = ((double)i / numCols -0.5) \* 2;

y = ((double)j / numRows -0.5) \* 2;

int pixel = mandelbrot(x,y);

pixels[i][j] = pixel;

if(pixel == 0)

numBlack++;

}

}

printf(“number of black pixels: %d\n”, numBlack);

When creating an 800 by 800 image, the correct number of black pixels (computed by removing the pragma line) is 222,680. Compile and run your your code; you should get a slightly smaller number than this. Again, examine the code and see if you can figure out where the race condition occurs.

Now the race condition is on the variable numBlack when it is incremented. Although numBlack++ is a single statement of C code, it requires multiple operations: loading the current value into a register, incrementing that register, and then storing the value from that register into memory. An undercount occurs when one thread performs these operations and another thread performs its load operation before the first thread finishes.

This code pattern where the results of many computations (the color calculation for each pixel) is combined into a single value is called a reduction. Other common examples include adding up all the numbers in an array or finding the minimum (or maximum) value in an array.

Because reductions are fairly common, parallel for loops in OpenMP provide a special type of variable for them. To inform the compiler that a variable is participating in a reduction over iterations of a loop, use the reduction qualifier. It is also necessary to specify the variable and the kind of reduction being performed. Here is the pragma for our program using this qualifier to denote that numBlack is being used in a summing reduction:

#pragma omp parallel for private(x,y) reduction(+:numBlack)

Make this change, re-compile, and run the program again. Now you should get the correct number of black pixels. In particularly, your parallel program should get the same number as the serial version (222,680 for an 800 by 800 image).

1. You need to add -fopenmp to the compile command in order to compile with OpenMP. [↑](#footnote-ref-1)
2. A 2.1GHz AMD Opteron with 16 cores running Fedora Linux version 31. The code was compiled with gcc 9.2.1. [↑](#footnote-ref-2)