# Lecture 16—Automatic Parallelization

ECE 459: Programming for Performance

February 9, 2015

### Road Map

- Past: manual parallelization with threads
- Now: compilers & automatic parallelization (when)
- Future: under the hood (how compilers parallelize))

### Example Code to Parallelize

```
#include < stdlib . h>
1
3
    void setup(double *vector, int length) {
4
        int i:
5
        for (i = 0; i < length; i++)
6
            vector[i] += 1.0;
8
9
    }
10
11
    int main()
12
13
        double *vector;
14
        vector = (double*) malloc(sizeof(double)*1024*1024);
15
        for (int i = 0; i < 1000; i++)
16
17
            setup (vector, 1024*1024);
18
19
```

### Automatic Parallelization of Example Code

Let's try automatic parallelization.

Compiling with solarisstudio and automatic parallelization yields the following:

```
% solarisstudio—cc —O3 —xautopar —xloopinfo omp_vector.c "omp_vector.c", line 5: PARALLELIZED, and serial version generated "omp_vector.c", line 15: not parallelized, call may be unsafe
```

How will this code compare to our manual efforts? (If you weren't in class, you'll have to try it yourself.)

**Note:** solarisstudio generates two versions of the code, and decides, at runtime, if the parallel code would be faster.

### Autoparallelization implementation: OpenMP

Under the hood, most parallelization frameworks use OpenMP, which we'll see next lecture.

For now: you can control the number of threads with the OMP\_NUM\_THREADS environment variable.

# Automatic Parallelization in gcc

gcc (since 4.3) can also auto-parallelize loops. However, there are a few problems:

- It will not tell you which loops it parallelizes (nicely).
- It only operates with a fixed number of threads.
- The profitability metrics are quite simple.
- Only operates in simple cases.

Use the flags -floop-parallelize-all -ftree-parallelize-loops=N where N is the # of threads.

Note: gcc also uses OpenMP but ignores OMP\_NUM\_THREADS.

# Understanding Automatic Parallelization in gcc

Flag -fdump-tree-parloops-details shows what the automatic parallelizations were, but it's quite unreadable.

Instead, you can look at the assembly code to see the parallelizations (obviously, impractical for a large project).

```
% gcc -std=c99 -O3 -ftree-parallelize-loops=4 omp_vector_gcc.c -S -o omp_vector_gcc_auto.s
```

The resulting .s file contains the following code:

```
call GOMP_parallel_start
leaq 80(%rsp), %rdi
call setup._loopfn.0
call GOMP_parallel_end
```

**Note:** gcc also parallelizes main.\_loopfn.2 and main.\_loopfn.3, although it looks like it serves little purpose.

What can we do to parallelize this code? Option 1: Option 2: Option 3:

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Option 1: horizontal ====

 Create 4 threads; each thread does 1000 iterations on its own sub-array.

Option 2:

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 Create 4 threads; each thread does 1000 iterations on its own sub-array.

Option 2: bad horizontal ====

• 1000 times, create 4 threads which each operate once on the sub-array.

### Option 3:

What can we do to parallelize this code?

Option 1: horizontal ====

 Create 4 threads; each thread does 1000 iterations on its own sub-array.

Option 2: bad horizontal ====

• 1000 times, create 4 threads which each operate once on the sub-array.

**Option 3:** vertical |||| ||| |||

 Create 4 threads; for each element, the owning thread does 1000 iterations on that element.

### Manual Parallelization Demo

I'll show a demo of three example PThread parallelizations.

Methodology: compiling with solarisstudio, flags -03 -1pthread.

Which manual option performs better?

# Comparing Parallelization Results

How does autoparallelization compare to manual parallelization?

### Case Study 2: Multiplying a Matrix by a Vector

Let's see how automatic parallelization does on a more complicated program (could we parallelize this?):

```
1
   void matVec (double **mat, double *vec, double *out,
2
                  int *row, int *col)
4
      int i, j;
5
      for (i = 0; i < *row; i++)
6
        for (j = 0; j < *col; j++)
8
9
          out[i] += mat[i][j] * vec[i];
10
11
12
13
```

Reminder: 
$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 14 \\ 32 \end{bmatrix}$$

### Case Study: Automatic Parallelization, Attempt 1

Well, based on our knowledge, we could parallelize the outer loop.

Let's see what solarisstudio will do for us...

```
% solarisstudio—cc —xautopar —xloopinfo —O3—c fploop.c
"fploop.c", line 5: not parallelized, not a recognized for loop
"fploop.c", line 8: not parallelized, not a recognized for loop
```

...it refuses to do anything, guesses?

### Case Study: Automatic Parallelization, Attempt 2

 The loop bounds are not constant, since one of the variables may alias row or col, even though int ≠ double.

So, let's add restrict to row and col and see what happens...

```
\% solarisstudio — cc — O3 — xautopar — xloopinfo — c fploop.c "fploop.c", line 5: not parallelized, unsafe dependence "fploop.c", line 8: not parallelized, unsafe dependence
```

Now it recognizes the loop, but still won't parallelize it. Why?

# Case Study: Automatic Parallelization, Attempt 3

• out might alias mat or vec, which would make this unsafe Let's add another restrict to out:

```
% solarisstudio -cc -O3 -xautopar -xloopinfo -c fploop.c "fploop.c", line 5: PARALLELIZED, and serial version generated "fploop.c", line 8: not parallelized, unsafe dependence
```

Now, we can get the outer loop to parallelize.

 Parallelizing the outer loop is almost always better than inner loops, and usually it's a waste to do both, so we're done.

**Note:** We can parallelize the inner loop as well (it's similar to Assignment 1). We'll see that solarisstudio can do it automatically.

# Loops That gcc's Automatic Parallelization Can Handle

### Single loop:

```
for (i = 0; i < 1000; i++)
 \times[i] = i + 3;
```

#### Nested loops with simple dependency:

### Single loop with not-very-simple dependency:

```
for (i = 0; i < 10; i++)
 X[2*i+1] = X[2*i];
```

### Loops That gcc's Automatic Parallelization Can't Handle

#### Single loop with if statement:

```
for (j = 0; j \le 10; j++)
if (j > 5) X[i] = i + 3;
```

### Triangle loop:

```
for (i = 0; i < 100; i++)
for (j = i; j < 100; j++)
X[i][j] = 5;
```

#### Examples from:

http://gcc.gnu.org/wiki/AutoparRelated

### Summary of Conditions for Automatic Parallelization

From Chapter 10 of Oracle's Fortran Programming Guide <sup>1</sup> translated to C, a loop must:

- have a recognized loop style, e.g., for loops with bounds that don't vary per-iteration;
- have no dependencies between data accessed in loop bodies for each iteration;
- not conditionally change scalar variables read after the loop terminates, or change any scalar variable across iterations; and
- have enough work in the loop body to make parallelization profitable.

<sup>1</sup> http://download.oracle.com/docs/cd/E19205-01/819-5262/index.html

### Reductions

- Reductions combine input data into a smaller (summary) set.
- We'll see a more complete definition when we touch on functional programming.
- Simplest instance: computing the sum of an array.

#### Consider the following code:

```
double sum (double *array, int length)
{
  double total = 0;
  for (int i = 0; i < length; i++)
    total += array[i];
  return total;
}</pre>
```

Can we parallelize this?

### Reduction Problems

#### Barriers to parallelization:

- value of total depends on previous iterations;
- addition is actually non-associative for floating-point values (is this a problem?)

Recall that "associative" means:

$$a + (b + c) = (a + b) + c.$$

n this case, the program probably isn't sensitive to rounding, but you should always consider if an operation is associative.

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### Automatic Parallelization via Reduction

If we compile the program with solarisstudio and add the flag -xreduction, it will parallelize the code:

```
% solarisstudio—cc —xautopar —xloopinfo —xreduction —O3—c sum.c "sum.c", line 5: PARALLELIZED, reduction, and serial version generated
```

**Note:** If we try to do the reduction on fploop.c with restricts added, we'll get the following:

```
\% solarisstudio – cc – O3 – xautopar – xloopinfo – xreduction – c fploop.c "fploop.c", line 5: PARALLELIZED, and serial version generated "fploop.c", line 8: not parallelized, not profitable
```

# Dealing with Function Calls

- A general function could have arbitary side effects.
- Production compilers tend to avoid parallelizing any loops with function calls.

Some built-in functions, like sin(), are "pure", have no side effects, and are safe to parallelize.

**Note:** this is why functional languages are nice for parallel programming: impurity is visible in type signatures.

# Dealing with Function Calls in solarisstudio

- For solarisstudio you can use the -xbuiltin flag to make the compiler use its whitelist of "pure" functions.
- The compiler can then parallelize a loop which uses sin() (you shouldn't replace built-in functions with your own if you use this option).

### Other options which may work:

- Crank up the optimization level (-x04).
- Explicitly tell the compiler to inline certain functions (-xinline=, or use the inline keyword).

# Summary of Automatic Parallelization

To help the compiler, we can:

- use restrict (make a restricted copy); and,
- make sure that loop bounds are constant (temporary variables).

Some compilers automatically create different versions for the alias-free case and the (parallelized) aliased case.

At runtime, the program runs the aliased case if correct.