

IC621: Distributed and Parallel
Computing

Lecture 07 : Shared-Memory Programming with OpenMP (I)

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OpenMP vs. Pthreads (I)

■ OpenMP : API for shared-memory parallel programming

- Pthreads are also API for shared-memory programming

■ Pthreads

- requires that the programmer explicitly specify the behavior of each thread
- is a library of functions that can be linked to a C program
 - can be used with any C compiler

■ OpenMP

- allows the programmer to simply state that a block of code should be executed in parallel
 - the precise determination of the tasks and which thread should execute them is left to the compiler and the run-time system
- requires compiler support for some operations

OpenMP vs. Pthreads (II)

Pthreads

- provides us with the power to program virtually any conceivable thread behavior
- **benefit** : lower-level API
- **cost** : it's up to us to specify every detail of the behavior of each thread

OpenMP

- allows the compiler and run-time system to determine some of the details of thread behavior
- **benefit**: simpler to code some parallel behaviors
- **cost** : some low-level thread interactions can be more difficult to program

Purpose of developing OpenMP

- Writing large-scale high-performance programs using APIs such as Pthreads was regarded to be **too difficult**
- OpenMP was explicitly designed to allow programmers to **incrementally parallelize existing serial programs**
- That is **virtually impossible** with MPI and fairly difficult with Pthreads

Directives-based API

■ **pragma**

- there are special preprocessor instructions known as `pragmas`
- `pragmas` are typically added to a system to allow behaviors that aren't part of the basic C/C++ specification
- compilers that don't support the `pragmas` are free to ignore them
- syntax: **# pragma**
 - # : preprocessor directives
 - if a `pragma` won't fit on a single line, the newline needs to be "escaped" by a backslash \

■ **Allowing a program that uses the pragmas to run on platforms that don't support them**

- carefully written OpenMP program can be compiled and run on any system with a C compiler
 - regardless of whether the compiler supports OpenMP

```

1  #include <stdio.h>
2  #include <stdlib.h>
3  #include <omp.h>
4
5  void Hello(void); /* Thread function */
6
7  int main(int argc, char* argv[]) {
8      /* Get number of threads from command line */
9      int thread_count = strtol(argv[1], NULL, 10);
10
11     #pragma omp parallel num_threads(thread_count)
12     Hello();
13
14     return 0;
15 } /* main */
16
17 void Hello(void) {
18     int my_rank = omp_get_thread_num();
19     int thread_count = omp_get_num_threads();
20
21     printf("Hello from thread %d of %d\n", my_rank, thread_count);
22
23 } /* Hello */

```

Compiling and running

■ Compiling

```
$ gcc -g -Wall -fopenmp -o omp_hello omp_hello.c
```

■ Running

```
$ ./omp_hello 4
```

■ Output

```
Hello from thread 0 of 4  
Hello from thread 1 of 4  
Hello from thread 2 of 4  
Hello from thread 3 of 4
```

or

```
Hello from thread 1 of 4  
Hello from thread 2 of 4  
Hello from thread 0 of 4  
Hello from thread 3 of 4
```

A “Hello” program using OpenMP

- OpenMP pragmas always begin with “# pragma omp”

- **parallel directive**

- specifies that the **structured block** of code that follows should be executed by multiple threads

- **Structured block**

- *definition* : C statement or a compound C statement with **one point of entry and one point of exit**
- it prohibits code that branches into or out of the middle of the **structured block**
- the number of threads running it is determined by the run-time system
 - the algorithm used is fairly complicated
 - the system typically runs one thread on each available core

■ **num_threads** clause

- it allows the programmer to specify the number of threads that should execute the structured block

```
# pragma omp parallel num_threads(thread_count)
```

- there may be system-defined limitations on the number of threads that a program can start
 - OpenMP Standard doesn't guarantee that `thread_count` threads will be actually started
 - most current systems can start hundreds or even thousands of threads

■ **When the program reaches the parallel directive**

- the original thread continues executing
- `thread_count-1` additional threads are started

■ **Team**

- the collection of threads executing the parallel block (master+slaves)
- **master** : the original thread
- **slaves** : the additional threads

■ Implicit barrier

- there is when the block of code is completed
 - i.e., when the threads return from the call to “Hello”
- a thread that has completed the block of code will wait for all the other threads in the team to complete the block
- when all the threads have completed the block
 - the slave threads will terminate
 - the master thread will continue executing the next codes

■ Local variables

- `my_rank` (rank or ID)
 - by calling `omp_get_thread_num`
 - the range is 0, 1, ..., `thread_count` - 1
- `thread_count` (the number of threads)
 - by calling `omp_get_num_threads`

Error checking

■ If the compiler doesn't support OpenMP

- it will just ignore the `parallel` directive
- however, the attempt to include `omp.h` and the calls to `omp_get_thread_num` and `omp_get_num_threads` will cause errors

■ Solution : using the preprocessor macro `_OPENMP`

```
#ifndef _OPENMP
#  include <omp.h>
#endif
```

```
#  ifdef _OPENMP
    int my_rank = omp_get_thread_num();
    int thread_count = omp_get_num_threads();
#  else
    int my_rank = 0;
    int thread_count = 1;
#  endif
```

Example : trapezoidal rule

```
1  #include <stdio.h>
2  #include <stdlib.h>
3  #include <omp.h>
4
5  void Trap(double a, double b, int n, double* global_result_p);
6
7  int main(int argc, char* argv[]) {
8      double  global_result = 0.0;
9      double  a, b;
10     int      n;
11     int      thread_count;
12
13     thread_count = strtol(argv[1], NULL, 10);
14     printf("Enter a, b, and n\n");
15     scanf("%lf %lf %d", &a, &b, &n);
16     # pragma omp parallel num_threads(thread_count)
17     Trap(a, b, n, &global_result);
18
19     printf("With n = %d trapezoids, our estimate\n", n);
20     printf("of the integral from %f to %f = %.14e\n",
21           a, b, global_result);
22     return 0;
23 } /* main */
```

```

25 void Trap(double a, double b, int n, double* global_result_p) {
26     double h, x, my_result;
27     double local_a, local_b;
28     int i, local_n;
29     int my_rank = omp_get_thread_num();
30     int thread_count = omp_get_num_threads();
31
32     h = (b-a)/n;
33     local_n = n/thread_count;
34     local_a = a + my_rank*local_n*h;
35     local_b = local_a + local_n*h;
36     my_result = (f(local_a) + f(local_b))/2.0;
37     for (i = 1; i <= local_n-1; i++) {
38         x = local_a + i*h;
39         my_result += f(x);
40     }
41     my_result = my_result*h;
42
43     # pragma omp critical
44     *global_result_p += my_result;
45 } /* Trap */

```

■ Race condition

- multiple threads are attempting to access a shared resource
- at least one of the accesses is an update
- then, the accesses can result in an error
- e.g., `global_result += my_result`

■ Critical section

- code executed by multiple threads that updates a shared resource
- shared resource can only be updated by one thread at a time
- e.g., mutexes and semaphore (in Pthreads)

■ **critical** directive

- tells the compiler that the system needs to arrange for the threads to have **mutually exclusive access** to the structured block of code

```
# pragma omp critical
global_result += my_result;
```

Scope of variables

■ Scope of a variable (in OpenMP)

- refers to the set of threads that can access the variable in a parallel block

■ **Shared scope** : a variable that can be accessed by all the threads in the team

- e.g., variables declared before a parallel block (`a`, `b`, `n`, `global_result`, and `thread_count`)
- e.g., `*global_result_p`

■ **Private scope**: a variable that can only be accessed by a single thread

- variables of private scope have no need for the `critical` directive
- e.g., variables used by each thread (`my_rank` and `thread_count`) are allocated from the thread's (private) stack

Reduction clause

■ Alternative code for trapezoidal rule

```
double Local_trap(double a, double b, int n);
```

```
    global_result = 0.0;
#   pragma omp parallel num_threads(thread_count)
    {
#       pragma omp critical
        global_result += Local_trap(double a, double b, int n);
    }
```

■ It may actually be slower with multiple threads than one thread

```
    global_result = 0.0;
#   pragma omp parallel num_threads(thread_count)
    {
        double my_result = 0.0; /* private */
        my_result += Local_trap(double a, double b, int n);
#       pragma omp critical
        global_result += my_result;
    }
```


■ Reduction variable

- a cleaner alternative that also avoids serializing execution of `Local_trap`
- we can specify that `global_result` is a *reduction variable*
- *reduction operator* : a *binary operation* (such as addition or multiplication)
- *reduction* : a computation that repeatedly applies the same reduction operator to a sequence of operands to get a single result
- note: all of the intermediate results of the operation should be stored in the same variable, i.e., the reduction variable

```
global_result = 0.0;
# pragma omp parallel num_threads(thread_count) \
    reduction(+: global_result)
global_result += Local_trap(double a, double b, int n);
```

- `global_result` : reduction variable
- “+” : reduction operator (addition)
- the calls to `Local_trap` can take place in parallel

■ Syntax of the reduction clause

```
reduction(<operator>: <variable list>)
```

■ Reduction operators (in C)

➤ `+, *, -, &, |, ^, &&, ||`

■ Subtraction is a bit problematic

➤ it isn't associative or commutative

➤ e.g.,

```
result = 0;
for (i = 1; i <= 4; i++)
    result -= i;
```

➤ two threads : thread 0 subtracts 1 and 2, thread 1 subtracts 3 and 4, then thread 0 will compute -3 and thread 1 will compute -4 and, of course, $-3 - (-7) = 4$

➤ however, correct answer : -10

■ Reduction variables of float or double type

- the results may differ slightly when different numbers of threads are used
- it is due to the fact that floating point arithmetic isn't associative
- e.g., if a, b, and c are floats, then $(a+b)+c$ may not be exactly equal to $a+(b+c)$

■ Variable in a reduction clause is shared

- private variable is created for each thread in the team
- threads' private variables are initialized to 0
- so, the reduction version is effectively identical to the non-reduction version

```
global_result = 0.0;
#pragma omp parallel num_threads(thread_count)
{
    double my_result = 0.0; /* private */
    my_result += Local_trap(double a, double b, int n);
#pragma omp critical
    global_result += my_result;
}
```

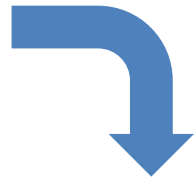
```
global_result = 0.0;
#pragma omp parallel num_threads(thread_count) \
    reduction(+: global_result)
    global_result += Local_trap(double a, double b, int n);
```

The parallel for directive

■ Simply placing a directive immediately before the for loop

- the structured block following the `parallel for` directive must be a for loop
- the system parallelizes the for loop by dividing the iterations of the loop among the threads

```
h = (b-a)/n;  
approx = (f(a) + f(b))/2.0;  
for (i = 1; i <= n-1; i++)  
    approx += f(a + i*h);  
approx = h*approx;
```



```
h = (b-a)/n;  
approx = (f(a) + f(b))/2.0;  
# pragma omp parallel for num_threads(thread_count) \  
    reduction(+: approx)  
for (i = 1; i <= n-1; i++)  
    approx += f(a + i*h);  
approx = h*approx;
```

■ Partitioning work among the threads

- partitioning is up to the system
- most systems use roughly a **block partitioning**
 - if there are m iterations,
 - then roughly the first $m/\text{thread_count}$ are assigned to thread 0
 - the next $m/\text{thread_count}$ are assigned to thread 1, and so on

■ Note: `approx` should be a reduction variable

- otherwise, it would have been an ordinary shared variable
- so, the body of the loop would be an **unprotected critical section**

```
approx += f(a + i*h);
```

■ Loop variable

- the default scope for all variables in a `parallel` directive is shared
- but, the **default scope of the loop variable in a `parallel for` directive is private**
 - each thread in the team has its own copy of `i`

Caveats

■ Simple method of parallelizing a serial program

- successively placing `parallel for` directives before each loop of a serial program

■ Caveat 1

- OpenMP will only parallelize for loops
- it **won't parallelize while loops or do-while loops**
- but, a `while` loop or a `do-while` loop can be converted to equivalent code that uses a `for` loop

■ Caveat 2

- OpenMP will only parallelize for loops for which the number of iterations can be determined
 - from the for statement itself (that is, the code for (. . . ; . . . ; . . .)),
 - prior to execution of the loop
- e.g., infinite loop cannot be parallelized
- e.g., the following loop cannot be parallelized, since the number of iterations can't be determined from the for statement alone
 - this `for` loop is also not a structured block, since the `break` adds another point of exit from the loop

```
for (i = 0; i < n; i++) {  
    if ( . . . ) break;  
    . . .  
}
```

■ OpenMP only parallelizes **for** loops that are in **canonical form**

- variable `index` must have integer or pointer type
 - e.g., it can't be a float
- expressions `start`, `end`, and `incr` must have a compatible type
 - e.g., if `index` is a pointer, then `incr` must have integer type
- expressions `start`, `end`, and `incr` must not change during execution of the loop
- during execution of the loop, the variable `index` can only be modified by the “**increment expression**” in the `for` statement

```
for ( index = start ; index < end      ; index++  
      index <= end   ; index--  
      index >= end   ; --index  
      index > end    ; index += incr  
                      ; index -= incr  
                      ; index = index + incr  
                      ; index = incr + index  
                      ; index = index - incr )
```


Data dependences

■ Example : computing the first n fibonacci numbers

- the compiler will create an executable without complaint

```
fibo[0] = fibo[1] = 1;  
for (i = 2; i < n; i++)  
    fibo[i] = fibo[i-1] + fibo[i-2];
```



```
fibo[0] = fibo[1] = 1;  
# pragma omp parallel for num_threads(thread_count)  
for (i = 2; i < n; i++)  
    fibo[i] = fibo[i-1] + fibo[i-2];
```

- with more than one thread, the results are unpredictable

1 1 2 3 5 8 0 0 0 0 *instead of* 1 1 2 3 5 8 13 21 34 55

■ What happened?

- the run-time system assigned
 - the computation of fibo[2], fibo[3], fibo[4], and fibo[5] to one thread
 - the computation of fibo[6], fibo[7], fibo[8], and fibo[9] to the other
- the second thread is using the values fibo[4] = 0 and fibo[5] = 0 to compute fibo[6]
 - then goes on to use fibo[5] = 0 and fibo[6] = 0 to compute fibo[7], and so on

■ Data dependences (sometimes a loop-carried dependence)

- OpenMP compilers don't check for dependences among iterations in a loop that's being parallelized with a `parallel for` directive
 - it's up to the programmers to identify these dependences
- a loop in which the results of one or more iterations depend on other iterations cannot be correctly parallelized by OpenMP

Finding loop-carried dependences

■ We only need to worry about loop-carried dependences

- don't need to worry about more general data dependences
- e.g., data dependence between Lines 2 and 3
 - however, **no problem** with the parallelization since **the computation of $x[i]$ and its subsequent use will always be assigned to the same thread**

```
1   for (i = 0; i < n; i++) {  
2       x[i] = a + i*h;  
3       y[i] = exp(x[i]);  
4   }
```



```
1  # pragma omp parallel for num_threads(thread_count)  
2  for (i = 0; i < n; i++) {  
3      x[i] = a + i*h;  
4      y[i] = exp(x[i]);  
5  }
```

■ Observation

- at least one of the statements must write or update the variable in order for the statements to represent a dependence
- so, to detect a loop-carried dependence, we should only concern ourselves with variables that are updated by the loop body
- that is, we should look for **variables that are read or written in one iteration, and written in another**

*no
loop-carried
dependence*

```
1  # pragma omp parallel for num_threads(thread_count)
2      for (i = 0; i < n; i++) {
3          x[i] = a + i*h;
4          y[i] = exp(x[i]);
5      }
```

*loop-carried
dependence*

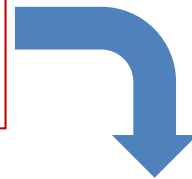
```
    fibo[0] = fibo[1] = 1;
    # pragma omp parallel for num_threads(thread_count)
    for (i = 2; i < n; i++)
        fibo[i] = fibo[i-1] + fibo[i-2];
```

Example : estimating π

$$\pi = 4 \left[1 - \frac{1}{3} + \frac{1}{5} - \frac{1}{7} + \dots \right] = 4 \sum_{k=0}^{\infty} \frac{(-1)^k}{2k+1}$$

- the update to factor in Line 7 in iteration k
- the subsequent increment of sum in Line 6 in iteration k+1

```
1    double factor = 1.0;
2    double sum = 0.0;
3    for (k = 0; k < n; k++) {
4        sum += factor/(2*k+1);
5        factor = -factor;
6    }
7    pi_approx = 4.0*sum;
```



```
1    double factor = 1.0;
2    double sum = 0.0;
3    # pragma omp parallel for num_threads(thread_count) \
4        reduction(+:sum)
5    for (k = 0; k < n; k++) {
6        sum += factor/(2*k+1);
7        factor = -factor;
8    }
9    pi_approx = 4.0*sum;
```

■ Loop-carried dependence in estimating π

- if iteration k is assigned to one thread
- iteration $k+1$ is assigned to another thread
- then, no guarantee that the value of factor in Line 6 will be correct

■ We can fix the problem by examining the series

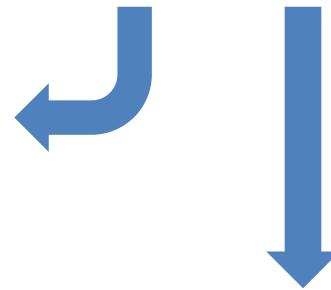
$$\sum_{k=0}^{\infty} \frac{(-1)^k}{2k+1}$$

- in iteration k , the value of factor should be $(-1)^k$
 - +1 if k is even, and -1 if k is odd
- so, if we replace the code

```
sum += factor/(2*k+1);  
factor = -factor;
```

- by

```
if (k % 2 == 0)  
    factor = 1.0;  
else  
    factor = -1.0;  
sum += factor/(2*k+1);
```



- or, if you prefer the **?:** operator

```
factor = (k % 2 == 0) ? 1.0 : -1.0;  
sum += factor/(2*k+1);
```
- we will eliminate the loop dependency

■ Things still aren't quite right

- e.g., two threads and $n = 1000$, the result is consistently wrong

```
1      With n = 1000 terms and 2 threads,  
2          Our estimate of pi = 2.97063289263385  
3      With n = 1000 terms and 2 threads,  
4          Our estimate of pi = 3.22392164798593
```

- on the other hand, only one thread get a correct result

```
1      With n = 1000 terms and 1 threads,  
2          Our estimate of pi = 3.14059265383979
```

■ Reasons

- by default, any variable declared before the loop (except the loop variable) is shared among the threads
- `factor` is shared
 - thread 0 might assign it the value 1, but before it can use the value in the update to sum, thread 1 could assign it the value -1

■ Solution

- in addition to eliminating the loop-carried dependence
- we need to insure that each thread has its own copy of `factor` by adding a **private clause** to the `parallel for` directive

```
1      double sum = 0.0;
2  #    pragma omp parallel for num_threads(thread_count) \
3      reduction(+:sum) private(factor)
4      for (k = 0; k < n; k++) {
5          if (k % 2 == 0)
6              factor = 1.0;
7          else
8              factor = -1.0;
9          sum += factor/(2*k+1);
10     }
```

- the updates of one thread to `factor` won't affect the value of `factor` in another thread

Note

- the value of a variable with private scope is unspecified at the beginning of a parallel block or a parallel for block
- its value is also unspecified after completion of a parallel or parallel for block
- e.g., in the first `printf` statement and the final `printf`

```
1  int x = 5;
2  # pragma omp parallel num_threads(thread_count) \
3      private(x)
4  {
5      int my_rank = omp_get_thread_num();
6      printf("Thread %d > before initialization, x = %d\n",
7            my_rank, x);
8      x = 2*my_rank + 2;
9      printf("Thread %d > after initialization, x = %d\n",
10           my_rank, x);
11 }
12 printf("After parallel block, x = %d\n", x);
```

■ **default(none)** clause

- Usually, we need to think about the scope of each variable in a `parallel block` or a `parallel for block`
 - rather than letting OpenMP decide on the scope of each variable
- `default(none)` clause will require that we specify the scope of each variable in the block and outside the block
- e.g.,
 - `sum` (reduction variable): has properties of both private and shared scope

```
double sum = 0.0;
#pragma omp parallel for num_threads(thread_count) \
    default(none) reduction(+:sum) private(k, factor) \
    shared(n)
for (k = 0; k < n; k++) {
    if (k % 2 == 0)
        factor = 1.0;
    else
        factor = -1.0;
    sum += factor/(2*k+1);
}
```

More about loops in OpenMP : Sorting

■ Bubble sort

➤ outer loop

- first, finds the largest element in the list and stores it in $a[n-1]$
- then, finds the next-to-the-largest element and stores it in $a[n-2]$, ...

➤ inner loop

- compares consecutive pairs of elements in **the current list**
- when a pair is out of order ($a[i] > a[i+1]$), it swaps them

```
for (list_length = n; list_length >= 2; list_length--)  
    for (i = 0; i < list_length-1; i++)  
        if (a[i] > a[i+1]) {  
            tmp = a[i];  
            a[i] = a[i+1];  
            a[i+1] = tmp;  
        }
```

■ Loop-carried dependences

- in **outer loop**: the contents of the current list depends on the previous iterations of the outer loop
- in **inner loop**: the elements that are compared in iteration i depend on the outcome of iteration $i-1$

■ Keep in mind that

- even though we can always find loop-carried dependences
- it may be difficult or impossible to remove them
- `parallel for` directive is not a universal solution to the problem of parallelizing `for` loops

■ Odd-even transposition sort

- even phase (`phase % 2 == 0`)
 - each odd-subscripted element, `a[i]`, is compared to the element to its “left,” `a[i-1]`
 - if they’re out of order, they’re swapped
- odd phase (`phase % 2 == 1`)
 - each odd-subscripted element is compared to the element to its right
 - if they’re out of order, they’re swapped
- it is guaranteed that after `n` phases, the list will be sorted

```
for (phase = 0; phase < n; phase++)  
    if (phase % 2 == 0)  
        for (i = 1; i < n; i += 2)  
            if (a[i-1] > a[i]) Swap(&a[i-1], &a[i]);  
    else  
        for (i = 1; i < n-1; i += 2)  
            if (a[i] > a[i+1]) Swap(&a[i], &a[i+1]);
```

■ Loop-carried dependence

- in **outer loop**: if phase 0 and phase 1 are executed simultaneously, the pair that's checked in phase 1 might be (7,8) rather than (9,6)
- in **inner loop**: no loop-carried dependences

Table 5.1 Serial Odd-Even Transposition Sort

Phase	Subscript in Array						
	0		1		2		3
0	9	↔	7		8	↔	6
	7		9		6		8
1	7		9	↔	6		8
	7		6		9		8
2	7	↔	6		9	↔	8
	6		7		8		9
3	6		7	↔	8		9
	6		7		8		9

■ Code for odd-even transposition sort using OpenMP

```
1   for (phase = 0; phase < n; phase++) {
2       if (phase % 2 == 0)
3   #       pragma omp parallel for num_threads(thread_count) \
4           default(none) shared(a, n) private(i, tmp)
5           for (i = 1; i < n; i += 2) {
6               if (a[i-1] > a[i]) {
7                   tmp = a[i-1];
8                   a[i-1] = a[i];
9                   a[i] = tmp;
10            }
11        }
12    else
13    #       pragma omp parallel for num_threads(thread_count) \
14        default(none) shared(a, n) private(i, tmp)
15        for (i = 1; i < n-1; i += 2) {
16            if (a[i] > a[i+1]) {
17                tmp = a[i+1];
18                a[i+1] = a[i];
19                a[i] = tmp;
20            }
21        }
22    }
```

■ Potential problem 1

- we need to be sure that all the threads have finished phase p before any thread starts phase $p+1$
- however, none of the threads will proceed to the phase $p+1$, until all of the threads have completed the phase p

■ Potential problem 2

- OpenMP implementation may fork and join `thread_count` threads on each pass through the body of the outer loop
 - overhead associated with forking and joining the threads
- e.g., the input list is of 20,000 elements

Table 5.2 Odd-Even Sort with Two parallel for Directives and Two for Directives (times are in seconds)

thread_count	1	2	3	4
Two parallel for directives	0.770	0.453	0.358	0.305
Two for directives	0.732	0.376	0.294	0.239

■ Solution for the potential problem 2

- it would be superior to fork the threads once and reuse the same team of threads for each execution of the inner loops
- we can fork our team of `thread_count` threads **before the outer loop** with **a `parallel` directive**
- **in the inner loop**, we use **a `for` directive** such that OpenMP parallelizes the for loop **with the existing team of threads**

■ `for` directive

- it doesn't fork any threads
- it uses whatever threads have already been forked in the enclosing `parallel` block
- there is an implicit barrier at the end of the loop

```

1  # pragma omp parallel num_threads(thread_count) \
2      default(none) shared(a, n) private(i, tmp, phase)
3      for (phase = 0; phase < n; phase++) {
4          if (phase % 2 == 0)
5              # pragma omp for
6                  for (i = 1; i < n; i += 2) {
7                      if (a[i-1] > a[i]) {
8                          tmp = a[i-1];
9                          a[i-1] = a[i];
10                         a[i] = tmp;
11                     }
12                 }
13             else
14                 # pragma omp for
15                     for (i = 1; i < n-1; i += 2) {
16                         if (a[i] > a[i+1]) {
17                             tmp = a[i+1];
18                             a[i+1] = a[i];
19                             a[i] = tmp;
20                         }
21                     }
22         }

```

Scheduling loops

- Exact assignment of loop iterations to threads is system dependent
- However, most OpenMP implementations use roughly a **block partitioning**
 - this assignment might be **less than optimal**
 - e.g., suppose that the time required by the call to f is proportional to the size of the argument i

```
sum = 0.0;  
for (i = 0; i <= n; i++)  
    sum += f(i);
```

- assign much more work to thread `thread_count-1` than it will assign to thread 0

■ A better assignment

- a **cyclic partitioning** of the iterations among the threads
- e.g., suppose `t = thread_count`
 - $n = 10,000$
 - single thread : **3.67** seconds
 - block partitioning (2 threads):
2.76 seconds (**x1.33**)
 - cyclic partitioning (2 threads):
1.84 seconds (**x1.99**)

Thread	Iterations
0	0, n/t , $2n/t$, ...
1	1, $n/t + 1$, $2n/t + 1$, ...
\vdots	\vdots
$t - 1$	$t - 1$, $n/t + t - 1$, $2n/t + t - 1$, ...

■ Scheduling : assigning iterations to threads

- a good assignment of iterations to threads can have a very significant effect on performance
- the **schedule** clause can be used to assign iterations in either a `parallel for` or a `for` directive

The schedule clause

■ Without the schedule clause

```
sum = 0.0;
# pragma omp parallel for num_threads(thread_count) \
    reduction(+:sum)
for (i = 0; i <= n; i++)
    sum += f(i);
```

■ With the schedule clause

```
sum = 0.0;
# pragma omp parallel for num_threads(thread_count) \
    reduction(+:sum) schedule(static,1)
for (i = 0; i <= n; i++)
    sum += f(i);
```

■ Syntax of the `schedule` clause

```
schedule(<type> [, <chunksize>])
```

type	meaning
static	the iterations can be assigned to the threads before the loop is executed
dynamic <i>or</i> guided	the iterations are assigned to the threads while the loop is executing, so after a thread completes its current set of iterations, it can request more from the run-time system
auto	the compiler and/or the run-time system determine the schedule
runtime	the schedule is determined at run-time

- `chunksize` is a positive integer
- only **static**, **dynamic**, and **guided** schedules can have a `chunksize`

The static schedule type

- The system assigns chunks of `chunksize` iterations to each thread in a **round-robin** fashion

➤ e.g., `schedule(static,1)`

Thread 0: 0,3,6,9

Thread 1: 1,4,7,10

Thread 2: 2,5,8,11

➤ e.g., `schedule(static,2)`

Thread 0: 0,1,6,7

Thread 1: 2,3,8,9

Thread 2: 4,5,10,11

➤ e.g., `schedule(static,4)`

Thread 0: 0,1,2,3

Thread 1: 4,5,6,7

Thread 2: 8,9,10,11

➤ if `chunksize` is omitted, it becomes approximately
`total_iterations/thread_count`

The `dynamic` and `guided` schedule types

- The iterations are broken up into chunks of `chunksize` consecutive iterations

- **Dynamic**

- each thread executes a chunk
- when a thread finishes a chunk, it requests another one from the run-time system
- this continues until all the iterations are completed
- if `chunksize` is omitted, 1 is used as it

- **Guided**

- as chunks are completed, the size of the new chunks decreases
- if no `chunksize` is specified, it decreases down to 1
- e.g., `n = 10,000` and `thread_count = 2`

Table 5.3 Assignment of Trapezoidal Rule Iterations 1–9999 using a guided Schedule with Two Threads

Thread	Chunk	Size of Chunk	Remaining Iterations
0	1–5000	5000	4999
1	5001–7500	2500	2499
1	7501–8750	1250	1249
1	8751–9375	625	624
0	9376–9687	312	312
1	9688–9843	156	156
0	9844–9921	78	78
1	9922–9960	39	39
1	9961–9980	20	19
1	9981–9990	10	9
1	9991–9995	5	4
0	9996–9997	2	2
1	9998–9998	1	1
0	9999–9999	1	0

The runtime **schedule type**

■ Environment variables

- named values that can be accessed by a running program
- available in the program's environment
- e.g., PATH, HOME, ...

```
$ echo $PATH
```

■ The system uses the environment variable **OMP_SCHEDULE** to determine at run-time how to schedule the loop

- OMP_SCHEDULE : can take on any of the values that can be used for a static, dynamic, or guided schedule
- e.g., in the bash shell

```
$ export OMP_SCHEDULE="static,1"
```

Which schedule?

- **There is some overhead associated with the use of a schedule clause**

- the overhead is greater for dynamic schedules than static schedules
- the overhead associated with guided schedules is the greatest of the three

- **If we're getting satisfactory performance without a schedule clause, we should go no further**

- however, if we suspect that the performance of the default schedule can be substantially improved
- then, we should probably experiment with some different schedules

■ Guidance for scheduling option

- if each iteration of the loop requires roughly the same amount of computation, then it's likely that the default distribution will give the best performance
- If the cost of the iterations decreases (or increases) linearly as the loop executes, then a static schedule with small chunk sizes will probably give the best performance
- If the cost of each iteration can't be determined in advance, then it may make sense to explore a variety of scheduling options

Thank you!

