



Introduction to OpenMP

Christian Terboven, Dirk Schmidl

IT Center, RWTH Aachen University {terboven,schmidl}@itc.rwth-aachen.de

History



- De-facto standard for Shared-Memory Parallelization.
- 1997: OpenMP 1.0 for FORTRAN
- 1998: OpenMP 1.0 for C and C++
- 1999: OpenMP 1.1 for FORTRAN (errata)
- 2000: OpenMP 2.0 for FORTRAN
- 2002: OpenMP 2.0 for C and C++
- 2005: OpenMP 2.5 now includes both programming languages.
- 05/2008: OpenMP 3.0 release
- 07/2011: OpenMP 3.1 release
- 07/2013: OpenMP 4.0 release
- 11/2015: OpenMP 4.5 release



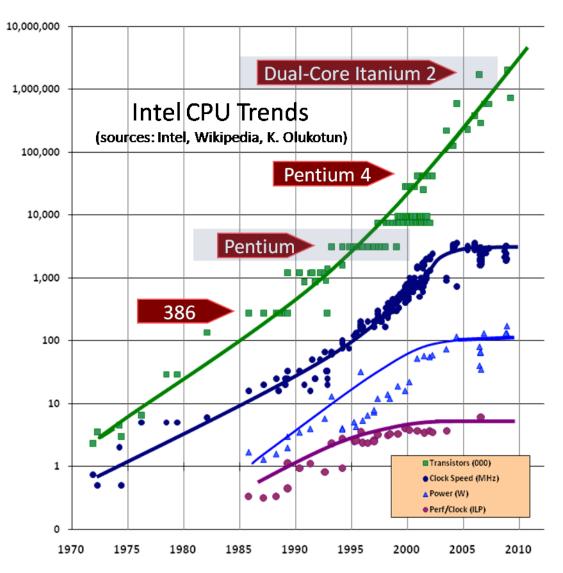
RWTH Aachen University is a member of the OpenMP Architecture Review Board (ARB) since 2006.



Multi-Core System Architecture

Moore's Law still holds!





The number of transistors on a chip is still doubling every 24 months ...

... but the clock speed is no longer increasing that fast!

Instead, we will see many more cores per chip!

Source: Herb Sutter

www.gotw.ca/publications/concurrency-ddj.htm

Example for a SMP system

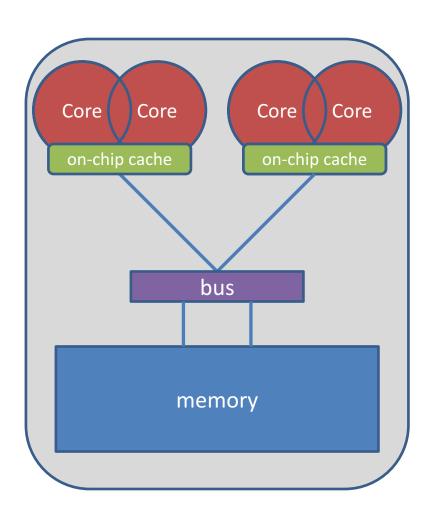


Dual-socket Intel Woodcrest (dual-core) system

- → Two cores per chip, 3.0 GHz
- → Each chip has 4 MB of L2 cache on-chip, shared by both cores
- → No off-chip cache
- → Bus: Frontsidebus

SMP: Symmetric Multi Processor

- Memory access time is uniform on all cores
- → Limited scalabilty



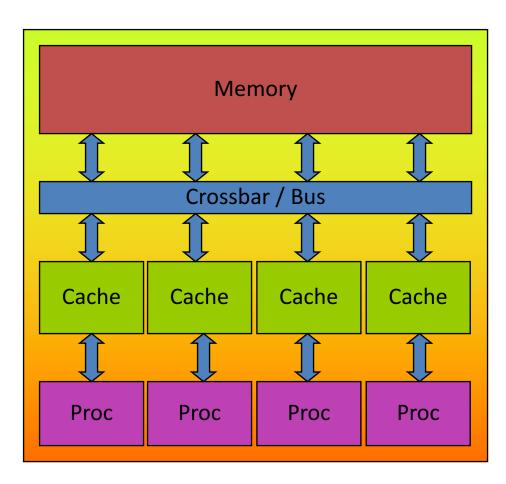


OpenMP Overview & Parallel Region

OpenMP's machine model



OpenMP: Shared-Memory Parallel Programming Model.



All processors/cores access a shared main memory.

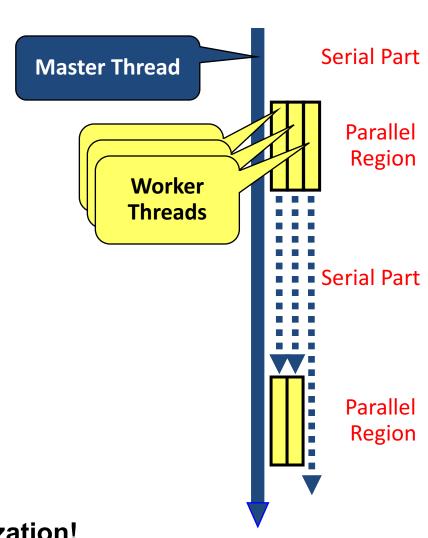
Real architectures are more complex, as we will see later / as we have seen.

Parallelization in OpenMP employs multiple threads.

OpenMP Execution Model



- OpenMP programs start with just one thread: The Master.
- Worker threads are spawned at Parallel Regions, together with the Master they form the Team of threads.
- In between Parallel Regions the Worker threads are put to sleep. The OpenMP Runtime takes care of all thread management work.
- Concept: Fork-Join.
- Allows for an incremental parallelization!



Parallel Region and Structured Blocks



The parallelism has to be expressed explicitly.

```
C/C++
#pragma omp parallel
{
    ...
    structured block
    ...
}
```

```
Fortran

!$omp parallel

...

structured block

...

$!omp end parallel
```

Structured Block

- → Exactly one entry point at the top
- → Exactly one exit point at the bottom
- → Branching in or out is not allowed
- → Terminating the program is allowed (abort / exit)

Specification of number of threads:

Environment variable:

```
OMP_NUM_THREADS=...
```

Or: Via num_threads clause:
add num_threads (num) to the
parallel construct



Hello OpenMP World



Hello orphaned OpenMP World

Starting OpenMP Programs on Linux



From within a shell, global setting of the number of threads:

From within a shell, one-time setting of the number of threads:



For Worksharing Construct

For Worksharing



- If only the parallel construct is used, each thread executes the Structured Block.
- Program Speedup: Worksharing
- OpenMP's most common Worksharing construct: for

```
C/C++
int i;
#pragma omp for
for (i = 0; i < 100; i++)
{
   a[i] = b[i] + c[i];
}</pre>
```

```
Fortran

INTEGER :: i
!$omp do

DO i = 0, 99

a[i] = b[i] + c[i];

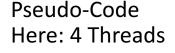
END DO
```

- → Distribution of loop iterations over all threads in a Team.
- → Scheduling of the distribution can be influenced.
- Loops often account for most of a program's runtime!

Worksharing illustrated









do i = 0, 24

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

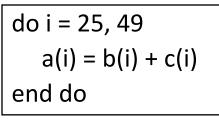
end do

Thread 2

Serial

Thread 3

Thread 4

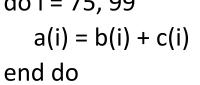


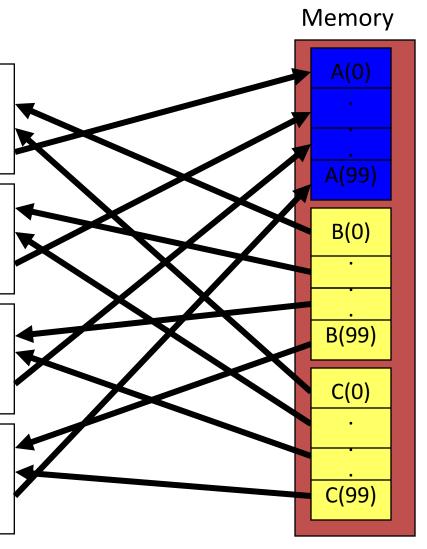
do i = 50, 74

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

end do

do i = 75, 99







Vector Addition

Influencing the For Loop Scheduling



- for-construct: OpenMP allows to influence how the iterations are scheduled among the threads of the team, via the schedule clause:
 - → schedule(static [, chunk]): Iteration space divided into blocks of chunk size, blocks are assigned to threads in a round-robin fashion. If chunk is not specified: #threads blocks.
 - → schedule (dynamic [, chunk]): Iteration space divided into blocks of chunk (not specified: 1) size, blocks are scheduled to threads in the order in which threads finish previous blocks.
 - → schedule (guided [, chunk]): Similar to dynamic, but block size starts with implementation-defined value, then is decreased exponentially down to chunk.
- Default on most implementations is schedule (static).

Synchronization Overview



- Can all loops be parallelized with for-constructs? No!
 - → Simple test: If the results differ when the code is executed backwards, the loop iterations are not independent. BUT: This test alone is not sufficient:

```
C/C++
int i, int s = 0;
#pragma omp parallel for
for (i = 0; i < 100; i++)
{
    s = s + a[i];
}</pre>
```

Data Race: If between two synchronization points at least one thread writes to a memory location from which at least one other thread reads, the result is not deterministic (race condition).

Synchronization: Critical Region



A Critical Region is executed by all threads, but by only one thread simultaneously (Mutual Exclusion).

```
C/C++
#pragma omp critical (name)
{
    ... structured block ...
}
```

Do you think this solution scales well?



Data Scoping

Scoping Rules



- Managing the Data Environment is the challenge of OpenMP.
- Scoping in OpenMP: Dividing variables in shared and private:
 - → private-list and shared-list on Parallel Region
 - → private-list and shared-list on Worksharing constructs
 - → General default is shared for Parallel Region, firstprivate for Tasks.
 - → Loop control variables on *for*-constructs are *private*
 - → Non-static variables local to Parallel Regions are *private*
 - → private: A new uninitialized instance is created for each thread
 - → firstprivate: Initialization with Master's value
 - → lastprivate: Value of last loop iteration is written back to Master
 - → Static variables are shared

Privatization of Global/Static Variables



- Global / static variables can be privatized with the threadprivate directive
 - → One instance is created for each thread
 - → Before the first parallel region is encountered
 - →Instance exists until the program ends
 - → Does not work (well) with nested Parallel Region
 - → Based on thread-local storage (TLS)
 - →TIsAlloc (Win32-Threads), pthread_key_create (Posix-Threads), keyword __thread (GNU extension)

```
C/C++
static int i;
#pragma omp threadprivate(i)
```

```
Fortran

SAVE INTEGER :: i
!$omp threadprivate(i)
```

Privatization of Global/Static Variables



- Global / static variables can be privatized with the threadprivate → Before the first parallel region is encountered read private
 → Instance exists until the program ends
 Does not work ' directive
 - → One instance is created for each thread

 - → Does not work (well) with nester
 - → Based on thread-local sto
 - →TIsAlloc (Win324) ead_key_create (Posix-Threads), keyword

```
threadprivate(i)
```

```
Fortran
```

```
INTEGER
!$omp threadprivate(i)
```



The Barrier Construct

The Barrier Construct



- OpenMP barrier (implicit or explicit)
 - → Threads wait until all threads of the current *Team* have reached the barrier

```
C/C++
#pragma omp barrier
```

All worksharing constructs contain an implicit barrier at the end



Back to our bad scaling example

It's your turn: Make It Scale!





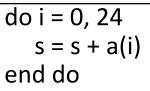
```
#pragma omp parallel
{
```

```
#pragma omp for
  for (i = 0; i < 99; i++)
  {</pre>
```

$$s = s + a[i];$$

}

} // end parallel



do i = 25, 49

do i = 0, 99

$$s = s + a(i)$$
 \Rightarrow
 $s = s + a(i)$
 \Rightarrow

end do s = s + a(i) s = s + a(i) s = s + a(i) s = s + a(i)end do

The Reduction Clause



- In a reduction-operation the operator is applied to all variables in the list. The variables have to be shared.
 - → reduction (operator:list)
 - → The result is provided in the associated reduction variable

```
C/C++
int i, s = 0;
#pragma omp parallel for reduction(+:s)
for(i = 0; i < 99; i++)
{
    s = s + a[i];
}</pre>
```

→ Possible reduction operators with initialization value:

```
+ (0), * (1), - (0), & (~0), | (0), && (1), || (0), 
^ (0), min (largest number), max (least number)
```

False Sharing



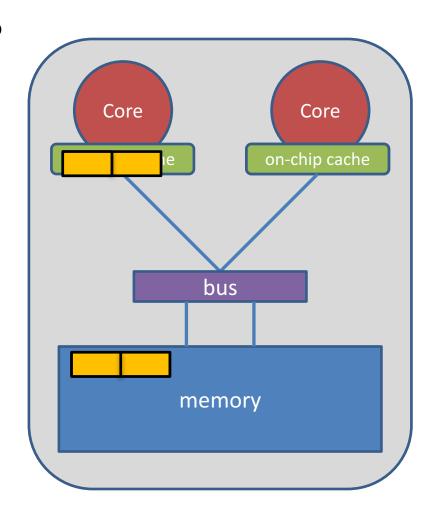


```
double s priv[nthreads];
#pragma omp parallel num threads(nthreads)
  int t=omp get thread num();
  #pragma omp for
  for (i = 0; i < 99; i++)
        s priv[t] += a[i];
} // end parallel
for (i = 0; i < nthreads; i++)
      s += s priv[i];
```

Data in Caches



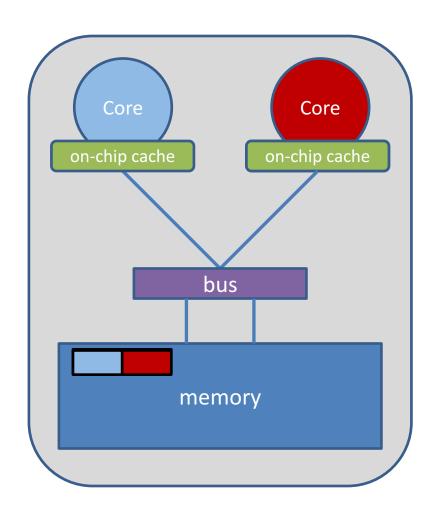
- When data is used, it is copied into caches.
- The hardware always copies chunks into the cache, so called cache-lines.
- This is useful, when:
 - the data is used frequently (temporal locality)
 - consecutive data is used which is on the same cache-line (spatial locality)



False Sharing

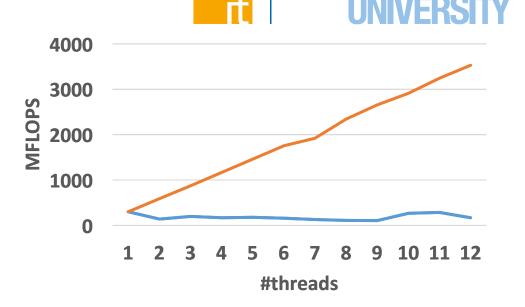


- False Sharing occurs when
 - different threads use elements of the same cache-line
 - one of the threads writes to the cache-line
- As a result the cache line is moved between the threads, also there is no real dependency
- Note: False Sharing is a performance problem, not a correctness issue

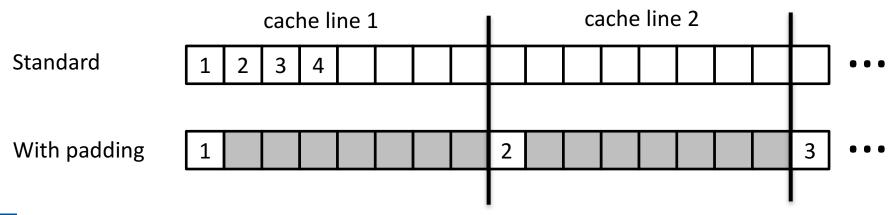


False Sharing

- no performance benefit for more threads
- Reason: false sharing of s_priv
- Solution: padding so that only one variable per cache line is used



—with false shawith false whith ring false sharing



False Sharing avoided





```
double s priv[nthreads * 8];
#pragma omp parallel num threads(nthreads)
  int t=omp get thread num();
  #pragma omp for
  for (i = 0; i < 99; i++)
        s priv[t * 8] += a[i];
} // end parallel
for (i = 0; i < nthreads; i++)
      s += s priv[i * 8];
```

Example





34

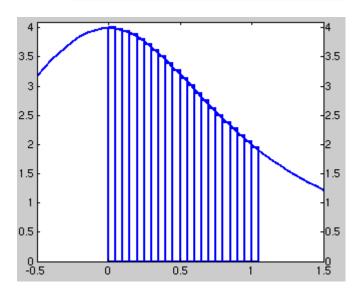
Example: Pi (1/2)

```
double f(double x)
  return (4.0 / (1.0 + x*x));
double CalcPi (int n)
  const double fH = 1.0 / (double) n;
  double fSum = 0.0;
  double fX;
  int i;
#pragma omp parallel for
  for (i = 0; i < n; i++)
    fX = fH * ((double)i + 0.5);
    fSum += f(fX);
  return fH * fSum;
```





$$\pi = \int_{0}^{1} \frac{4}{1 + x^2}$$



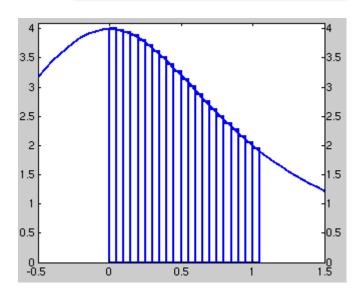
Example: Pi (1/2)

```
double f(double x)
  return (4.0 / (1.0 + x*x));
double CalcPi (int n)
  const double fH = 1.0 / (double) n;
  double fSum = 0.0;
  double fX;
  int i;
#pragma omp parallel for private(fX,i) reduction(+:fSum)
  for (i = 0; i < n; i++)
    fX = fH * ((double)i + 0.5);
    fSum += f(fX);
  return fH * fSum;
```





$$\pi = \int_{0}^{1} \frac{4}{1 + x^2}$$



Example: Pi (2/2)



Results:

# Threads	Runtime [sec.]	Speedup
1	1.11	1.00
2		
4		
8	0.14	7.93

Scalability is pretty good:

- → About 100% of the runtime has been parallelized.
- → As there is just one parallel region, there is virtually no overhead introduced by the parallelization.
- → Problem is parallelizable in a trivial fashion ...



Single and Master Construct

The Single Construct





```
C/C++
#pragma omp single [clause]
... structured block ...
```

```
Fortran
```

```
!$omp single [clause]
... structured block ...
!$omp end single
```

- The single construct specifies that the enclosed structured block is executed by only on thread of the team.
 - → It is up to the runtime which thread that is.
- Useful for:
 - → I/O
 - → Memory allocation and deallocation, etc. (in general: setup work)
 - → Implementation of the single-creator parallel-executor pattern as we will see now...

The Master Construct





```
C/C++
#pragma omp master[clause]
... structured block ...
```

```
Fortran
```

```
!$omp master[clause]
... structured block ...
!$omp end master
```

- The master construct specifies that the enclosed structured block is executed only by the master thread of a team.
- Note: The master construct is no worksharing construct and does not contain an implicit barrier at the end.



Section and Ordered Construct

How to parallelize a Tree Traversal?



How would you parallelize this code?

```
void traverse (Tree *tree)
{
   if (tree->left) traverse(tree->left);
   if (tree->right) traverse(tree->right);
   process(tree);
}
```

One option: Use OpenMP's parallel sections.

The Sections Construct



```
C/C++

#pragma omp sections [clause]
{
    #pragma omp section
    ... structured block ...
    #pragma omp section
    ... structured block ...
}
```

```
Fortran
!$omp sections [clause]
 !$omp section
 ... structured block ...
 !$ omp section
 ... structured block ...
 !$ omp section
 ... structured block ...
!$omp end sections
```

The sections construct contains a set of structured blocks that are to be distributed among and executed by the team of threads.

How to parallelize a Tree Traversal?!





How would you parallelize this code?

```
void traverse (Tree *tree)
                                                 Nested Parallel Regions
#pragma omp parallel sections
#pragma omp section
       if (tree->left)
                         traverse(tree->left);
#pragma omp section
       if (tree->right) traverse(tree->right);
} // end omp parallel
      process(tree);
```

Barrier here!

- Downsides of this option:
 - → Unneccessary overhead and synchronization points
 - → Not always well supported (how many threads to be used?)

The ordered Construct



- Allows to execute a structured block within a parallel loop in sequential order
 - → In addition, an ordered clause has to be added to the for construct which any ordered construct may occur

Use Cases:

- → Can be used e.g. to enforce ordering on printing of data
- → May help to determine whether there is a data race



Runtime Library

Runtime Library



C and C++:

- → If OpenMP is enabled during compilation, the preprocessor symbol _OPENMP is defined. To use the OpenMP runtime library, the header omp.h has to be included.
- → omp_set_num_threads (int): The specified number of threads will be used for the parallel region encountered next.
- → int omp_get_num_threads: Returns the number of threads in the current team.
- → int omp_get_thread_num(): Returns the number of the calling thread in the team, the Master has always the id 0.
- Additional functions are available, e.g. to provide locking functionality.



Tasking

Recursive approach to compute Fibonacci



On the following slides we will discuss three approaches to parallelize this recursive code with Tasking.

The Task Construct



```
C/C++
#pragma omp task [clause]
    structured block ...
```

```
Fortran
!$omp task [clause]
... structured block ...
!$omp end task
```

Each encountering thread/task creates a new Task

- → Code and data is being packaged up
- → Tasks can be nested
 - → Into another Task directive
 - →Into a Worksharing construct

Data scoping clauses:

- \rightarrow shared(*list*)
- → private(*list*) firstprivate(*list*)
- → default(shared | none)

Tasks in OpenMP: Data Scoping



Some rules from Parallel Regions apply:

- → Static and Global variables are shared
- → Automatic Storage (local) variables are private

If shared scoping is not derived by default:

- → Orphaned Task variables are firstprivate by default!
- → Non-Orphaned Task variables inherit the shared attribute!
- → Variables are firstprivate unless shared in the enclosing context

First version parallelized with Tasking (omp-v1)



```
int main (int argc,
         char* argv[])
   [...]
   #pragma omp parallel
        #pragma omp single
                fib (input);
   [...]
```

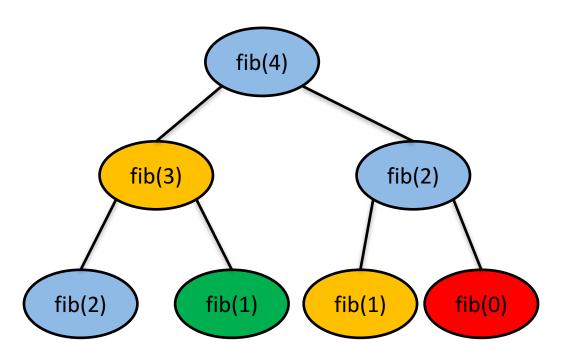
```
int fib(int n) {
   if (n < 2) return n;
  int x, y;
  #pragma omp task shared(x)
       x = fib(n - 1);
  #pragma omp task shared(y)
       v = fib(n - 2);
  #pragma omp taskwait
       return x+y;
```

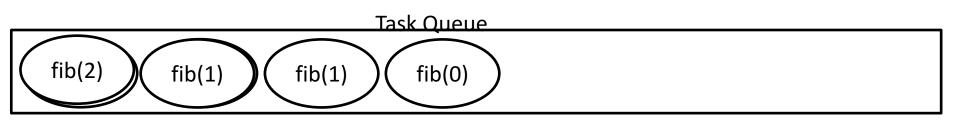
- Only one Task / Thread enters fib() from main(), it is responsable for creating the two initial work tasks
- Taskwait is required, as otherwise x and y would be lost

Fibonacci Illustration



- T1 enters fib(4)
- T1 creates tasks for fib(3) and fib(2)
- T1 and T2 execute tasks from the queue
- T1 and T2 create 4 new tasks
- T1 T4 execute tasks

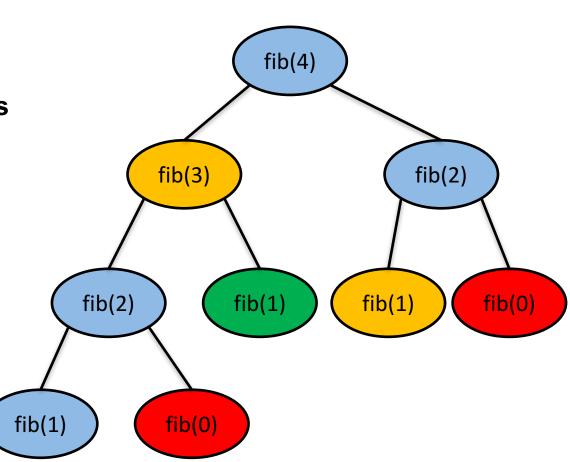




Fibonacci Illustration



- T1 enters fib(4)
- T1 creates tasks for fib(3) and fib(2)
- T1 and T2 execute tasks from the queue
- T1 and T2 create 4 new tasks
- T1 T4 execute tasks



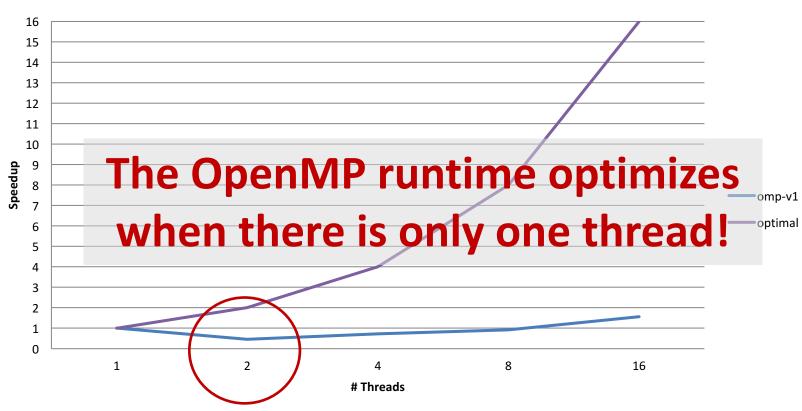
Scalability measurements (1/3)





Overhead of task creation prevents scalability!

Speedup of Fibonacci with Tasks



if Clause



- If the expression of an if clause on a task evaluates to false
 - → The encountering task is suspended
 - → The new task is executed immediately
 - → The parent task resumes when the new task finishes
 - → Used for optimization, avoids queuing of small tasks

Improved parallelization with Tasking (omp-v2)



Improvement: Don't create yet another task once a certain (small enough) n is reached

```
int main (int argc,
         char* arqv[])
   [...]
#pragma omp parallel
#pragma omp single
   fib(input);
   [...]
```

```
int fib(int n) {
   if (n < 2) return n;
int x, y;
#pragma omp task shared(x) \
  if(n > 30)
   x = fib(n - 1);
#pragma omp task shared(y) \
  if(n > 30)
   y = fib(n - 2);
#pragma omp taskwait
   return x+y;
```

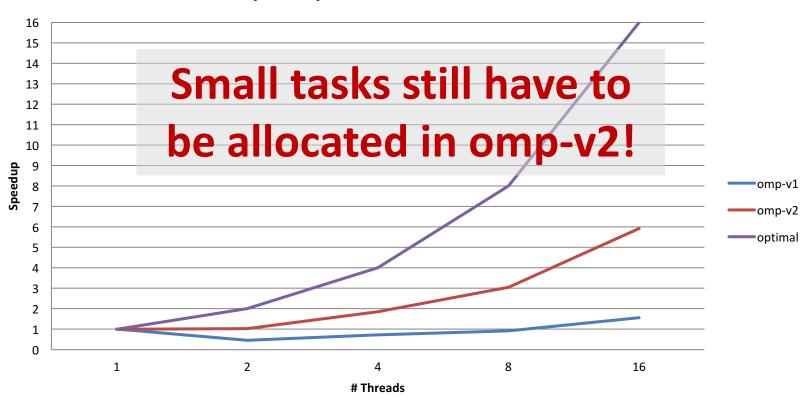
Scalability measurements (2/3)





Speedup is better, but still not great

Speedup of Fibonacci with Tasks



Improved parallelization with Tasking (omp-v3)



Improvement: Skip the OpenMP overhead once a certain n is reached

```
int main (int argc,
         char* arqv[])
   [...]
#pragma omp parallel
#pragma omp single
   fib (input);
}
   [...]
```

```
int fib(int n) {
   if (n < 2) return n;
   if (n \le 30)
      return serfib(n);
int x, y;
#pragma omp task shared(x)
  x = fib(n - 1);
#pragma omp task shared(y)
  y = fib(n - 2);
#pragma omp taskwait
   return x+y;
```

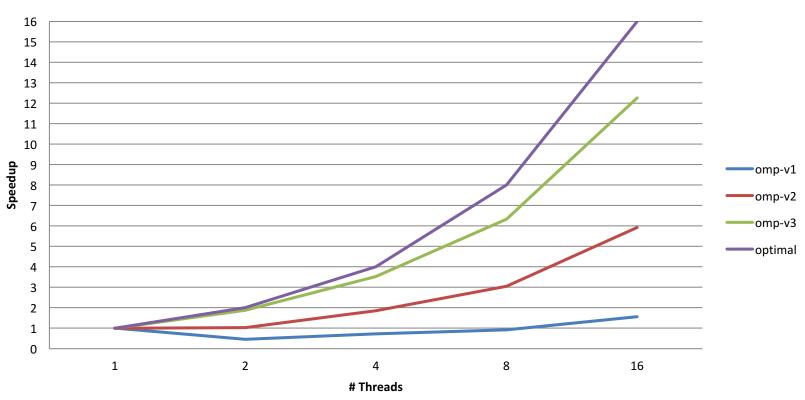
Scalability measurements (3/3)





Looks promising...

Speedup of Fibonacci with Tasks

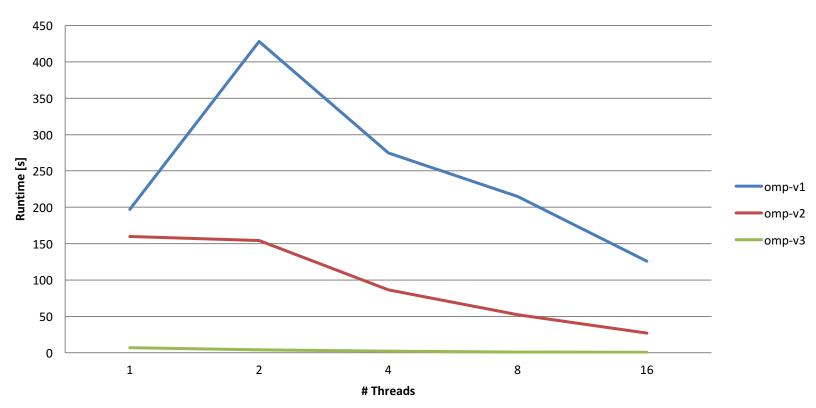


Runtime measurements (1/2)



First two versions were slow because of overhead!

Runtime of Fibonacci with Tasks

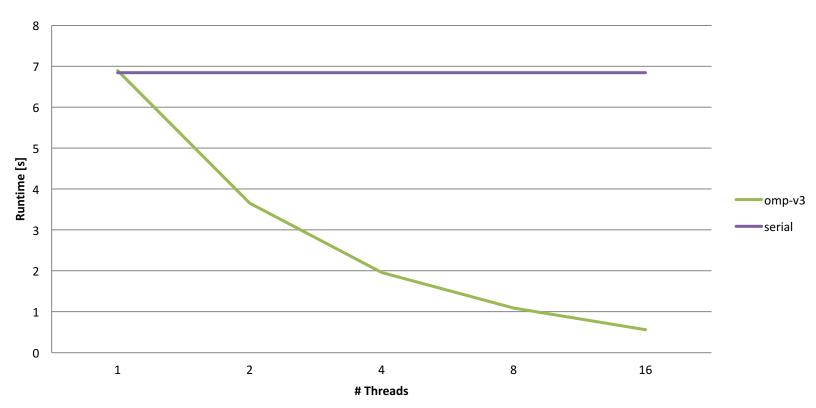


Runtime measurements (2/2)



Third version is comparable to serial version w/o OpenMP ©

Runtime of Fibonacci with Tasks



Data Scoping Example (1/7)



```
int a = 1;
void foo()
{
   int b = 2, c = 3;
   #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a:
               // Scope of b:
               // Scope of c:
               // Scope of d:
               // Scope of e:
```

Data Scoping Example (2/7)



```
int a = 1;
void foo()
{
   int b = 2, c = 3;
   #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a: shared
               // Scope of b:
               // Scope of c:
               // Scope of d:
               // Scope of e:
```

Data Scoping Example (3/7)



```
int a = 1;
void foo()
{
   int b = 2, c = 3;
   #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a: shared
               // Scope of b: firstprivate
               // Scope of c:
               // Scope of d:
               // Scope of e:
```

Data Scoping Example (4/7)



```
int a = 1;
void foo()
{
   int b = 2, c = 3;
   #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a: shared
               // Scope of b: firstprivate
               // Scope of c: shared
               // Scope of d:
               // Scope of e:
```

Data Scoping Example (5/7)



```
int a = 1;
void foo()
{
   int b = 2, c = 3;
   #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a: shared
               // Scope of b: firstprivate
               // Scope of c: shared
               // Scope of d: firstprivate
               // Scope of e:
```

Data Scoping Example (6/7)

```
int a = 1;
void foo()
{
  int b = 2, c = 3;
  #pragma omp parallel shared(b)
   #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
               int e = 5;
               // Scope of a: shared
               // Scope of b: firstprivate
               // Scope of c: shared
               // Scope of d: firstprivate
               // Scope of e: private
```



Hint: Use default(none) to be forced to think about every variable if you do not see clear.

Data Scoping Example (7/7)



```
int a = 1;
void foo()
{
  int b = 2, c = 3;
  #pragma omp parallel shared(b)
  #pragma omp parallel private(b)
       int d = 4;
       #pragma omp task
              int e = 5;
              // Scope of a: shared,
                                           value of a: 1
              // Scope of b: firstprivate, value of b: 0 / undefined
              // Scope of c: shared,
                                        value of c: 3
              // Scope of d: firstprivate, value of d: 4
              // Scope of e: private, value of e: 5
```

The Barrier and Taskwait Constructs



- OpenMP barrier (implicit or explicit)
 - → All tasks created by any thread of the current *Team* are guaranteed to be completed at barrier exit

```
C/C++
#pragma omp barrier
```

- Task barrier: taskwait
 - → Encountering Task suspends until child tasks are complete
 - →Only direct childs, not descendants!

```
C/C++
#pragma omp taskwait
```

Task Synchronization



Task Synchronization explained:

```
#pragma omp parallel num threads(np)
                              np Tasks created here, one for each thread
#pragma omp task 🕢
   function A();
                              All Tasks guaranteed to be completed here
#pragma omp barries
#pragma omp single
#pragma omp task <
                                                1 Task created here
       function B();
                               B-Task guaranteed to be completed here
```



Questions?