

Introduction to OpenMP

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Tasking

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 one 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...

Recursive approach to compute Fibonacci



```
int main(int argc,  
        char* argv[])  
{  
    [...]  
    fib(input);  
    [...]  
}
```

```
int fib(int n) {  
    if (n < 2) return n;  
    int x = fib(n - 1);  
    int y = fib(n - 2);  
    return x+y;  
}
```

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

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:

→ `shared(list)`

→ `private(list)` `firstprivate(list)`

→ `default(shared | none)`

- **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

- **So far no verification tool is available to check Tasking programs for correctness!**

First version parallelized with Tasking (omp-v1)



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

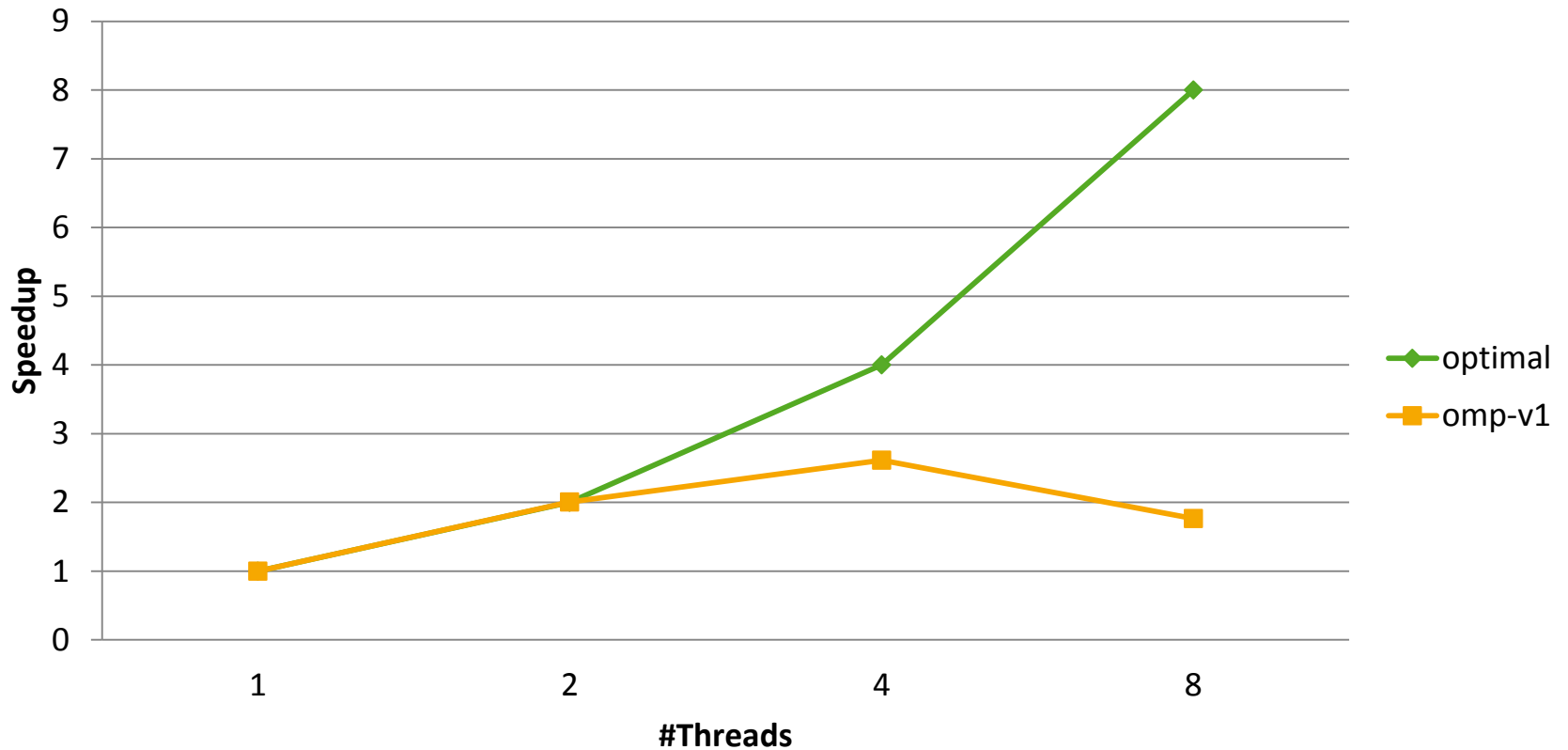
```
int fib(int n) {
    if (n < 2) return n;
    int x, y;
#pragma omp task shared(x)
{
    x = fib(n - 1);
}
#pragma omp task shared(y)
{
    y = fib(n - 2);
}
#pragma omp taskwait
    return x+y;
}
```

- **Only one Task / Thread enters fib () from main (), it is responsible for creating the two initial work tasks**

Taskwait is required, as otherwise x and y would be lost

Overhead of task creation prevents better scalability!

Speedup of Fibonacci with Tasks



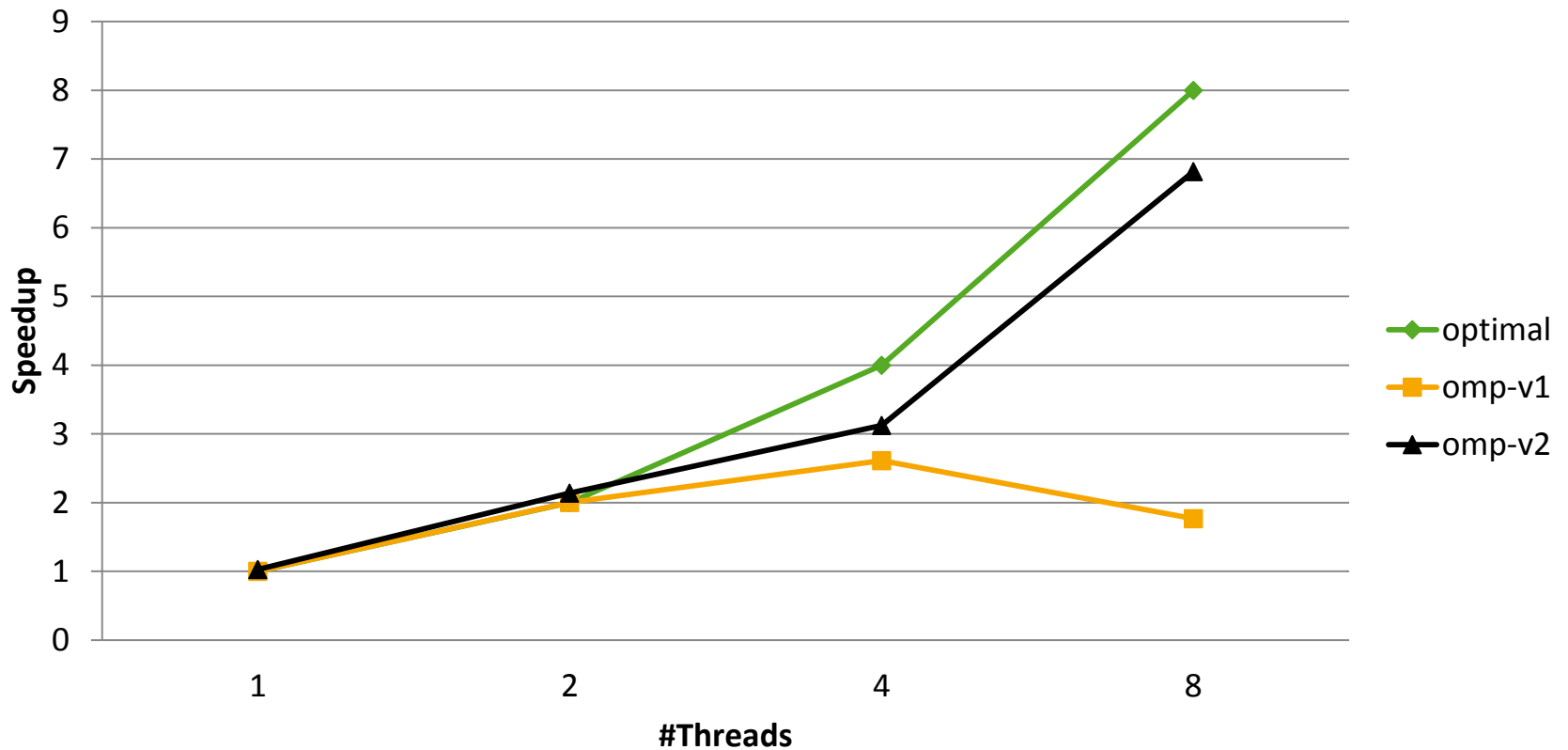
- **Improvement: Don't create yet another task once a certain (small enough) n is reached**

```
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) \
        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;
}
```

- Speedup is ok, but we still have some overhead when running with 4 or 8 threads

Speedup of Fibonacci with Tasks



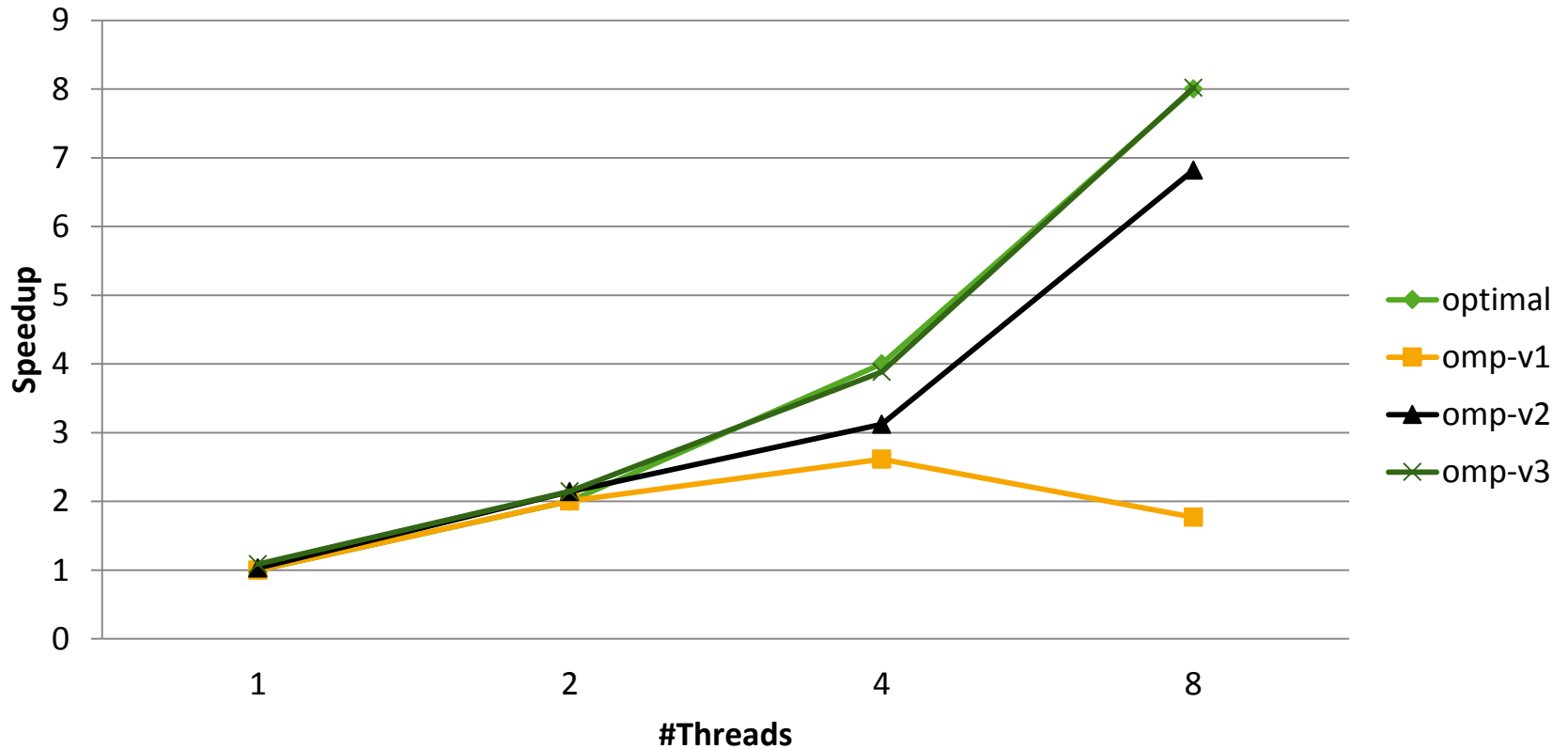
- **Improvement: Skip the OpenMP overhead once a certain n is reached (no issue w/ production compilers)**

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

```
int fib(int n)  {
    if (n < 2) return n;
    if (n <= 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;
}
```

■ Everything ok now 😊

Speedup 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
        }
    }
}
```

■ 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 explained:

```
#pragma omp parallel num_threads (np)
{
#pragma omp task
    function_A ();
#pragma omp barrier
#pragma omp single
    {
#pragma omp task
        function_B ();
    }
}
```

np Tasks created here, one for each thread

All Tasks guaranteed to be completed here

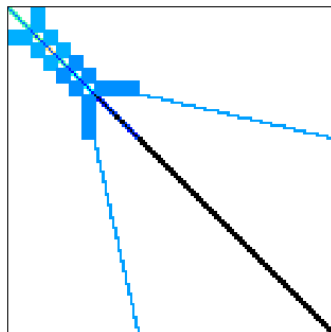
1 Task created here

B-Task guaranteed to be completed here

Load Balancing

■ Sparse Linear Algebra

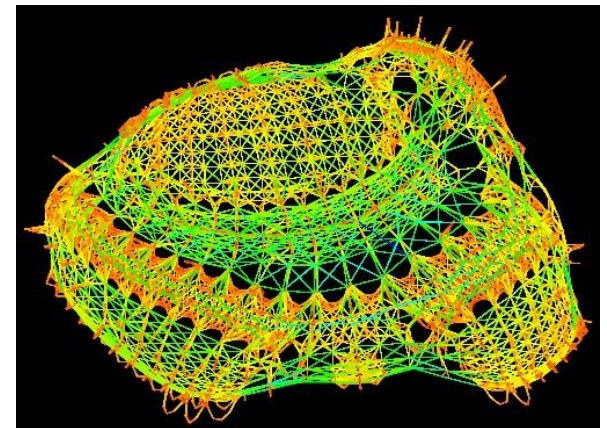
- Sparse Linear Equation Systems occur in many scientific disciplines.
- Sparse matrix-vector multiplications (SpMxV) are the dominant part in many iterative solvers (like the CG) for such systems.
- number of non-zeros $\ll n \cdot n$



Beijing Botanical Garden

Oben Rechts: Original Gebäude
Unten Rechts: Modell
Unten Links: Matrix

(Quelle: Beijing Botanical Garden and University of Florida, Sparse Matrix Collection)



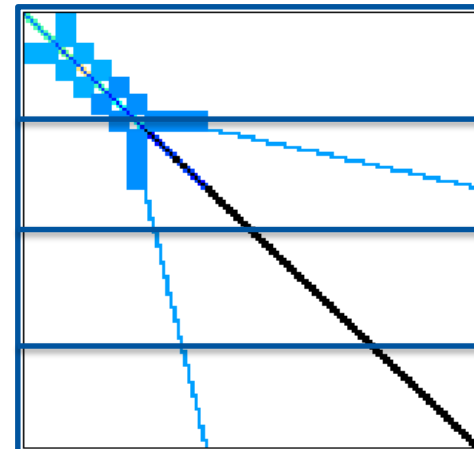
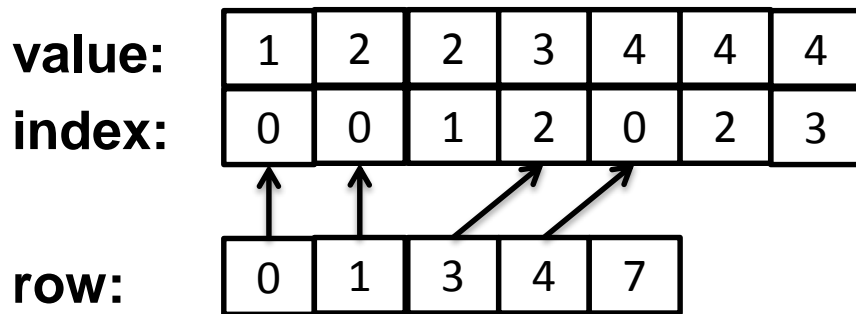
Case Study: CG



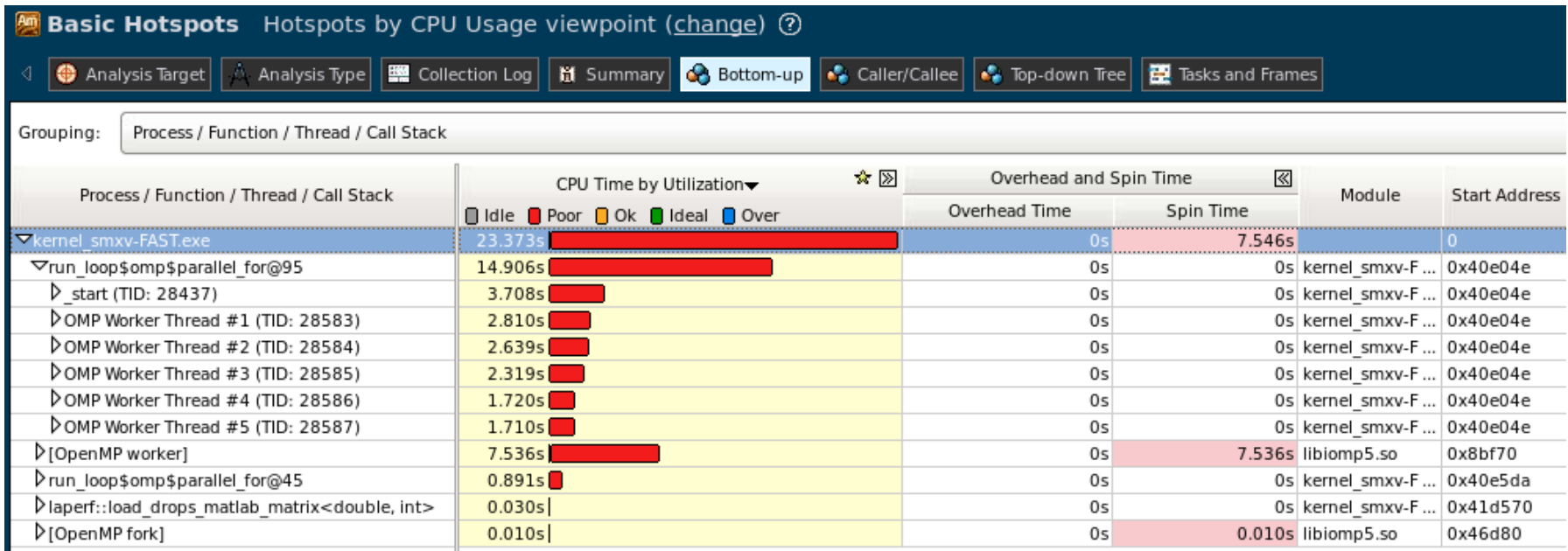
■ $A = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 4 & 0 & 4 & 4 \end{pmatrix}$

```
for (i = 0; i < num_rows; i++){
    sum = 0.0;
    for (nz=row[i]; nz<row[i+1]; ++nz){
        sum+= value[nz]*x[index[nz]];
    }
    y[i] = sum;
}
```

- Format: compressed row storage
- store all values and columns in arrays (length nnz)
- store beginning of a new row in a third array (length n+1)



- Grouping execution time of parallel regions by threads helps to detect load imbalance.
- Significant portions of Spin Time also indicate load balance problems.
- Different loop schedules might help to avoid these problems.



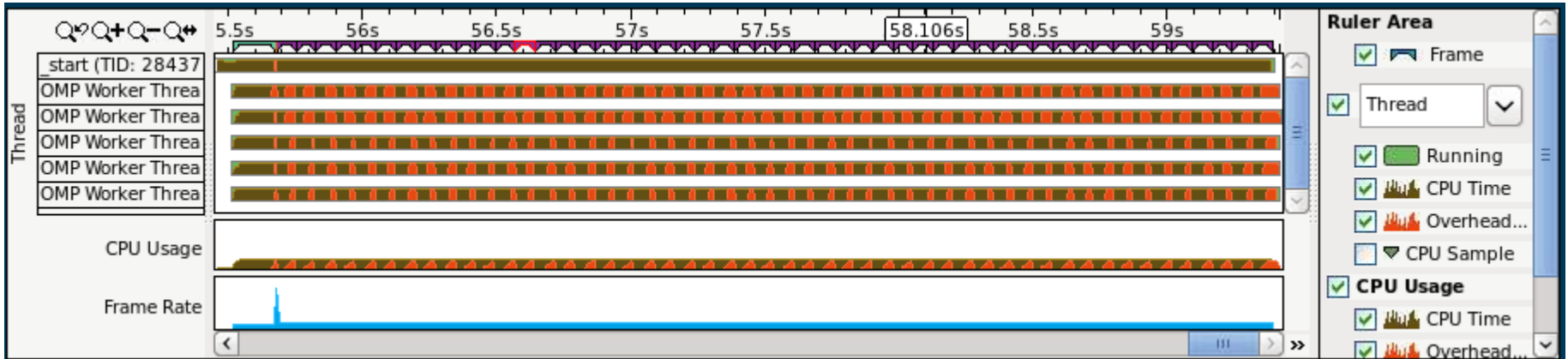
Basic Hotspots Hotspots by CPU Usage viewpoint (change) ?

Analysis Target Analysis Type Collection Log Summary Bottom-up Caller/Callee Top-down Tree Tasks and Frames

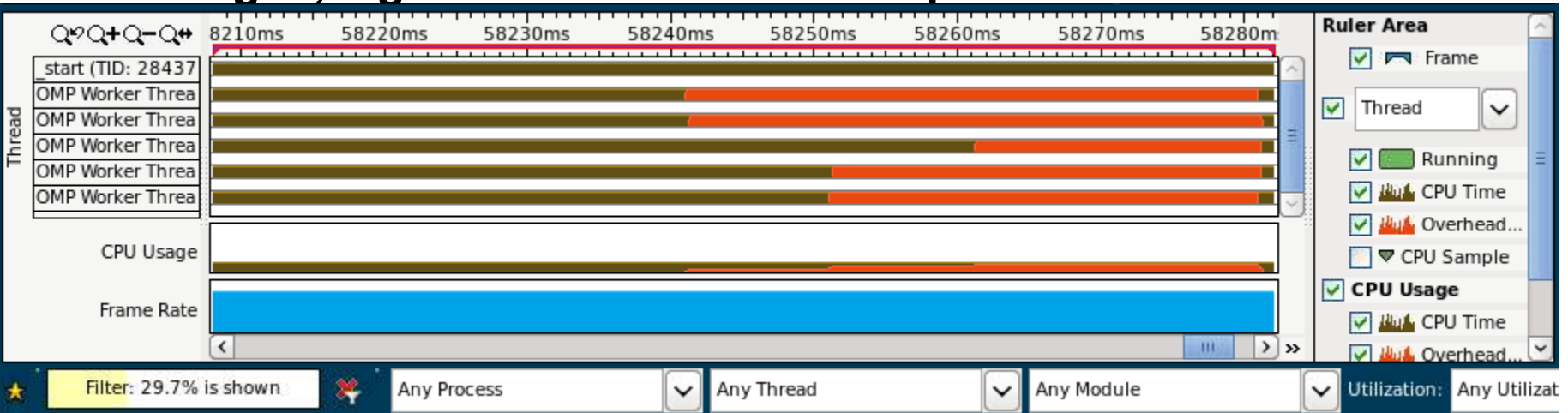
Grouping: Process / Function / Thread / Call Stack

Process / Function / Thread / Call Stack	CPU Time by Utilization				Overhead and Spin Time		Module	Start Address
	Idle	Poor	Ok	Ideal	Over	Overhead Time		
kernel_smxv-FAST.exe	23.373s					0s	7.546s	0
run_loop\$omp\$parallel_for@95	14.906s					0s	0s	kernel_smxv-F ... 0x40e04e
_start (TID: 28437)	3.708s					0s	0s	kernel_smxv-F ... 0x40e04e
OMP Worker Thread #1 (TID: 28583)	2.810s					0s	0s	kernel_smxv-F ... 0x40e04e
OMP Worker Thread #2 (TID: 28584)	2.639s					0s	0s	kernel_smxv-F ... 0x40e04e
OMP Worker Thread #3 (TID: 28585)	2.319s					0s	0s	kernel_smxv-F ... 0x40e04e
OMP Worker Thread #4 (TID: 28586)	1.720s					0s	0s	kernel_smxv-F ... 0x40e04e
OMP Worker Thread #5 (TID: 28587)	1.710s					0s	0s	kernel_smxv-F ... 0x40e04e
[OpenMP worker]	7.536s					0s	7.536s	libiomp5.so 0x8bf70
run_loop\$omp\$parallel_for@45	0.891s					0s	0s	kernel_smxv-F ... 0x40e5da
laperf::load_drops_matlab_matrix<double, int>	0.030s					0s	0s	kernel_smxv-F ... 0x41d570
[OpenMP fork]	0.010s					0s	0.010s	libiomp5.so 0x46d80

- The Timeline can help to investigate the problem further.



- Zooming in, e.g. to one iteration is also possible.



Parallel 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)`.**



Exercises 3 and 5