



Exercise 4

Introduction to High-Performance Computing WS 2019

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Exercise Tasks



- Hello World
- 2. Exploiting Architectures: STREAM
- 3. First steps with Tasks
- 4. Reasoning about Work-Distribution
- 5. OpenMP Puzzles
- 6. Dry runs on various aspects

1. Hello World



Problem 1. Hello World

Go to the hello directory. Compile the hello code via 'make [debug|release]' and execute the resulting executable via 'OMP_NUM_THREADS=procs make run', where procs denotes the number of threads to be used.

a) Change the code that the thread number (thread id) and the total number of threads in the team are printed. Re-compile and execute the code in order to verify your changes.

C/C++: In order to print a decimal number, use the %d format specifier with printf().

```
int i1 = value;
int i2 = other_value;
printf("Value of i1 is: %d, and i2 is: %d", i1, i2);
```

b) In which order did you expect the threads to print out the Hello World message? Did your expectations meet your observations? If not, is that wrong?

LIVE DEMO

2. Exploiting Architectures: STREAM





Problem 2. Exploiting Architectures: STREAM

The STREAM memory benchmark is a widely used instrument for memory bandwidth measurements. It tries to measure the bandwidth that is actually available to a program, which typically differs from what a vendor specifies as the maximum memory bandwidth.

On Linux, the taskset command can be used to restrict a process to a subset of all the cores in a system. The syntax that should be used in this exercise is as follows:

\$ taskset -c cpu-list cmd [args]

where cmd denotes the program to be executed under the specified restriction. The program lstopo can be used to get a graphical representation of the machine architecture.

- a) Go to the stream directory. Take a look at the source code and examine the operation that is done in order to measure the performance of the memory subsystem $(A = B + scalar \cdot C)$. Parallelize this operation with OpenMP. Take care of producing correct time measurements.
- b) Compile the stream code via 'make release' and execute the program on login-t (or the mpi backend) via 'OMP_NUM_THREADS=2 taskset -c binding ./stream.exe', for each binding in the table below. With an array size of 40,000,000, each array is approximately 300 MiB in size. How do you explain the performance variations? Hint: Consider NUMA effects.

#Threads	Binding (no blanks in between)	SAXPYing [MiB/s]
2	0,1	
2	0,2	
2	0,12	
2	0,24	

2. Exploiting Architectures: STREAM



- Cluster frontend: login18-1.hpc.itc.rwth-aachen.de
- lstopo --no-io

```
Machine (383GB total)
  Package L#0 + L3 L#0 (33MB)
    NUMANode L#0 (P#0 95GB)
      L2 L#0 (1024KB) + L1d L#0 (32KB) + L1i L#0 (32KB) + Core L#0 + PU L#0 (P
      L2 L#1 (1024KB) + L1d L#1 (32KB) + L1i L#1 (32KB) + Core L#1 + PU L#1 (P
                                                                                                              Core
                                                                                                                         Core
                                                                                    Core
                                                                                              Core
      L2 L#2 (1024KB) + L1d L#2 (32KB) + L1i L#2 (32KB) + Core L#2 + PU L#2
     L2 L#3 (1024KB) + L1d L#3 (32KB) + L1i L#3 (32KB) + Core L#3 + PU L#3 (P
      L2 L#4 (1024KB) + L1d L#4 (32KB) + L1i L#4 (32KB) + Core L#4 + PU L#4 (P
                                                                                              on-chip
                                                                                  on-chip
                                                                                                             on-chip
                                                                                                                          on-chip
      L2 L#5 (1024KB) + L1d L#5 (32KB) + L1i L#5 (32KB) + Core L#5 + PU L#5 (P
                                                                                                                          cache
      L2 L#6 (1024KB) + L1d L#6 (32KB) + L1i L#6 (32KB) + Core L#6 + PU L#6 (P
                                                                                   cache
                                                                                               cache
                                                                                                              cache
      L2 L#7 (1024KB) + L1d L#7 (32KB) + L1i L#7 (32KB) + Core L#7 + PU L#7 (P
      L2 L#8 (1024KB) + L1d L#8 (32KB) + L1i L#8 (32KB) + Core L#8 + PU L#8 (P
      L2 L#9 (1024KB) + L1d L#9 (32KB) + L1i L#9 (32KB) + Core L#9 + PU L#9 (P
      L2 L#10 (1024KB) + L1d L#10 (32KB) + L1i L#10 (32KB) + Core L#10 + PU L#
      L2 L#11 (1024KB) + L1d L#11 (32KB) + L1i L#11 (32KB) + Core L#11 + PU L#
                                                                                                interconnect
    NUMANode L#1 (P#1 96GB)
      L2 L#12 (1024KB) + L1d L#12 (32KB) + L1i L#12 (32KB) + Core L#12 + PU L#
      L2 L#13 (1024KB) + L1d L#13 (32KB) + L1i L#13 (32KB) + Core L#13 + PU L#
      L2 L#14 (1024KB) + L1d L#14 (32KB) + L1i L#14 (32KB) + Core L#14 + PU L#
      L2 L#15 (1024KB) + L1d L#15 (32KB) + L1i L#15 (32KB) + Core L#15 + PU L#
      L2 L#16 (1024KB) + L1d L#16 (32KB) + L1i L#16 (32KB) + Core L#16 + PU L#
      L2 L#17 (1024KB) + L1d L#17 (32KB) + L1i L#17 (32KB) + Core L#17 + PU L#
     L2 L#18 (1024KB) + L1d L#18 (32KB) + L1i L#18 (32KB) + Core L#18 + PU L#
      L2 L#19 (1024KB) + L1d L#19 (32KB) + L1i L#19 (32KB) + Core L#19 + PU L#
      L2 L#20 (1024KB) + L1d L#20 (32KB) + L1i L#20 (32KB) + Core L#20 + PU L#
                                                                                      memory
                                                                                                                 memory
      L2 L#21 (1024KB) + L1d L#21 (32KB) + L1i L#21 (32KB) + Core L#21 + PU L#
      L2 L#22 (1024KB) + L1d L#22 (32KB) + L1i L#22 (32KB) + Core L#22 + PU L#
      L2 L#23 (1024KB) + L1d L#23 (32KB) + L1i L#23 (32KB) + Core L#23 + PU L#
  Package L#1 + L3 L#1 (33MB)
    NUMANode L#2 (P#2 96GB)
      L2 L#24 (1024KB) + L1d L#24 (32KB) + L1i L#24 (32KB) + Core L#24 + PU L#24
                                                                                             (illustration from the lecture)
[...]
```

2. Exploiting Architectures: STREAM



login18-1

# Threads	Binding	SAXPYing [MiB/s]	SAXPYing [MiB/s] (NUMA optimized)
2	0,1	16663.4	16568.9
2	0,2	16802.6	16567.8
2	0,12	16892.2	16605.8
2	0,24	14119.9	17300.5

- → Hardware threads 0 and {1,2} are on the same socket
- → Hardware threads 0 and 12 are on the same socket on different numa dom.
- → Hardware threads 0 and 24 are on different sockets

OpenMP Places

→ Use OpenMP Places / first touch for adequate thread binding / NUMA optimization (refer to lecture)



Problem 3. First steps with OpenMP tasks

The code below performs a traversal of a dynamic list and for each list element the process() function is called. The for-loop continues until e->next points to null. Such a loop could not be parallelized in OpenMP until the task directive was introduced, as the number of loop iterations (= list elements) could not be computed. Parallelize this code using the task concept of OpenMP. State the scope of each variable explicitly.

```
List 1;
Element e;

for(e = 1->first; e; e = e->next)
{
    process(e);
}
```

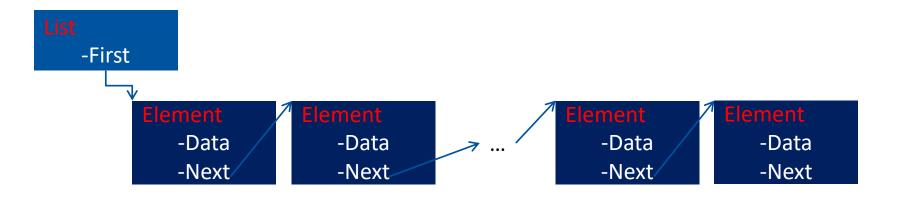


Traversal of dynamic list

```
List l;
Element e;
for (e = l-> first ; e; e = e-> next)
{
    process (e);
}
```



Traversal of dynamic list

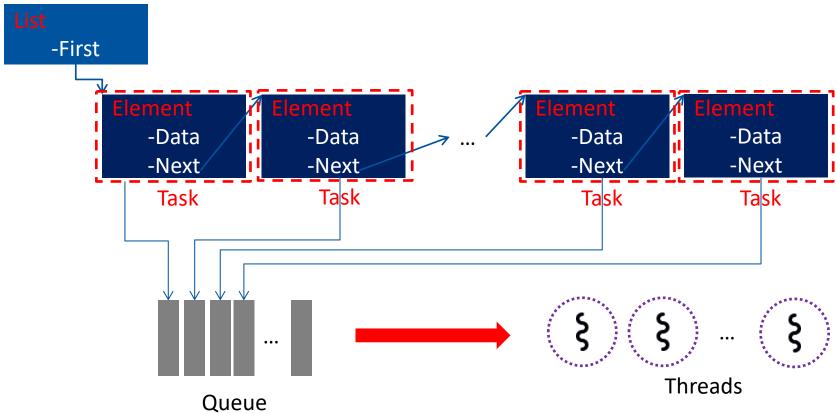


How can this be parallelized?



Traversal of dynamic list

- → Number of elements unknown before traversal
- → Tasks are created and subsequently processed by threads







How to parallelize it?

```
List I;
Element e;
#pragma omp parallel shared(l,e)
     #pragma omp single shared(l,e)
          for (e = I-> first ; e; e = e-> next )
              #pragma omp task firstprivate(e)
                    process (e);
```

4. Reasoning about Work-Distribution



Problem 4. Reasoning about Work-Distribution

Go to the for directory. Compile the for code via 'make [debug|release]' and execute the resulting executable via 'OMP_NUM_THREADS=procs make run', where *procs* denotes the number of threads to be used.

- a) Examine the code and think about where to put the parallelization directive(s).
- b) Measure the speedup and the efficiency of the parallelize code. How good does the code scale and which scaling did you expect?

#Threads	Runtime [sec]	Speedup	Efficiency
1			

LIVE DEMO



Problem 5. OpenMP Puzzles

Try to parallelize the following loops by inserting the missing OpenMP directives. If a loop cannot be parallelized, state reasons why you think so.

a) Insert missing OpenMP directives to parallelize the given loop.

```
double A[N] = { ... }, B[N] = { ... }, D[N];
const double c = ...;
const double x = ...;
double y;

for (int i = 0; i < N; i++ )
{
    y = sqrt(A[i]);

    D[i] = y + A[i] / (x*x);
}</pre>
```



Puzzle 1:



b) Can you parallelize this loop — if yes, how? if not, why?

```
double A[N] = { ... }, B[N] = { ... }, D[N];
const double c = ...;
const double x = ...;
double y;

for (int i = 1; i < N; i++ )
{
    A[i] = B[i] - A[i - 1];
}</pre>
```



Puzzle 2:

→ Can you parallelize this loop?

Recap: Data dependencies



- Dependencies within loops may prevent efficient software pipelining
 - → Software pipelining: interleaving of loop iterations to meet latency requirements (done by the compiler)

No Dependency

```
//FORTRAN
DO I=1,N
   A(I)=A(I)*c
END DO

//C
for(i=0;i<N;++i)
   A[i]=A[i]*c;
```

Real Dependency

```
//FORTRAN
DO I=2,N
   A(I)=A(I-1)*c
END DO

//C
for(i=1;i<N;++i)
   A[i]=A[i-1]*c;
```

Pseudo Dependency

```
//FORTRAN
DO I=1,N-1
   A(I)=A(I+1)*c
END DO

//C
for(i=0;i<N-1;++i)
   A[i]=A[i+1]*c;</pre>
```



Puzzle 2:

→ Can you parallelize this loop?

1	A[i]
1	B[1] - A[0]
2	B[2] - A[1] = B[2] - B[1] + A[0]
3	B[3] - A[2] = B[3] - B[2] + B[1] - A[0]
4	B[4] - A[3] = B[4] - B[3] + B[2] - B[1] + A[0]
5	



Puzzle 2:

→ Can you parallelize this loop?

No – There is a real dependency inside the loop which can't be resolved.





Problem 6. Dry runs on various aspects

The code snippet below implements a matrix times vector (MxV) operation, where a is a vector of \mathbb{R}^m , B is a matrix of $\mathbb{R}^{m \times n}$ and c is a vector of \mathbb{R}^n : $a = B \cdot c$.

```
void mxv_row(int m, int n, double *A, double *B, double *C)
 ^{2}
 3
        int i, j;
 4
 5
 6
        for (i = 0; i < m; i++)
 8
            A[i] = 0.0;
 9
10
11
12
13
            for (j = 0; j < n; j++)
14
                 A[i] += B[i * n + j] * C[i]:
15
16
17
        }
18
```

- a) Parallelize the for loop in line 07 by providing the appropriate OpenMP pragmas. Which variables have to be private and which variables have to be shared?
- b) Which schedule would you propose for this parallelization? Explain your answer and briefly list the pros and cons for each of the three OpenMP worksharing schedules (namely *static dynamic* and *guided*) for this specific case.
- c) Would it be possible to parallelize the for loop in line 13? If your answer is yes, provide the appropriate line in OpenMP and explain what scaling you would expect. If your answer is no, explain why you think it is not possible.



a) Parallelize the for loop in line 07 by providing the appropriate OpenMP pragmas. Which variables have to be private and which variables have to be shared?

```
void mxv_row ( int m, int n, double *A, double *B, double *C)
 2
 3
           int i, j;
 4
 5
           #pragma omp parallel for shared(A,B,C, m, n) private(i,j)
 6
 7
           for (i = 0; i < m; i++)
 8
 9
               A[i] = 0.0;
10
11
12
13
               for (j = 0; j < n; j++)
14
15
                   A[i] += B[i * n + j] * C[j];
16
           }
17
18
```



b. Which schedule would you propose for this parallelization? Explain your answer and briefly list the pros and cons for each of the three OpenMP worksharing schedules (namely static, dynamic and guided) for this specific case.

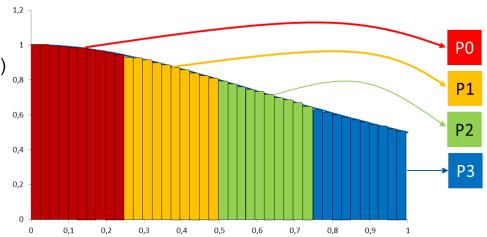


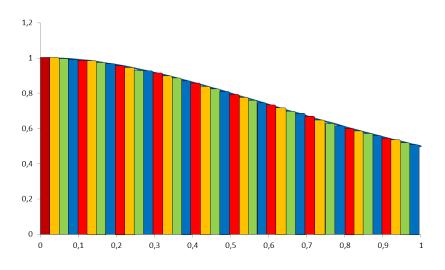
Static Schedule

- → schedule(static [, chunk])
- Decompositiondepending on chunksize
- → Equal parts of size 'chunksize' distributed in round-robin fashion

Pros?

- → No/low runtime overhead
- Cons?
 - → No dynamic workload balancing







Dynamic schedule

- → schedule(dynamic [, chunk])
- → Iteration space divided into blocks of chunk size
- → Threads request a new block after finishing the previous one
- → Default chunk size is 1

Pros ?

→ Workload distribution

Cons?

- → Runtime Overhead
- → Chunk size essential for performance
- → No NUMA optimizations possible



Guided schedule

- → Similar to dynamic: Threads request chunks as they require
- → Chunk size decreases over time until it reaches 1

Pros & Cons?

→ As for dynamic



b. Which schedule would you propose for this parallelization? Explain your answer and briefly list the pros and cons for each of the three OpenMP worksharing schedules (namely static, dynamic and guided) for this specific case.

Static, because work is evenly distributed among the threads if the MxV operation is parallelized among the matrix rows.



c. Would it be possible to parallelize the for loop in line 13? If your answer is yes, provide the appropriate line in OpenMP and explain what scaling you would expect. If your answer is no, explain why you think it is not possible.

It is possible with an OpenMP reduction. For performance reason one should in general try to parallelize the most outer loop