

Basics of Parallel Computing 2024S Assignment 2

May 26, 2024

2 Person Group 13

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1 Exercise 1

```
#include <stdio.h>
#include <stdlib.h>
#include <omp.h>
int main(int argc, char *argv[]) {
   int nenv = 3;
    omp_set_num_threads(nenv); // set number of threads
    printf("nenv: %d\n", nenv);
    int chunk = 5:
    omp_set_schedule(omp_sched_static, chunk);
    // omp_set_schedule(omp_sched_dynamic, chunk);
    // omp_set_schedule(omp_sched_guided, chunk);
   printf("chunk size: %d\n", chunk);
   int i = 0;
    int n = 17;
    int a[n];
    int t[nenv];
    #pragma omp parallel for schedule(runtime)
    for (i=0; i<n; i++) {
       a[i] = omp_get_thread_num(); // chosen thread per iteration
       t[omp_get_thread_num()]++; // parallel increment
   printf("a (schedule): ");
    for (i=0; i<n; i++) {
       printf("%d ", a[i]);
   printf("\n");
   printf("t (counter): ");
   for (i=0; i<nenv; i++) {</pre>
       printf("%d ", t[i]);
   printf("\n");
```

1.1 What do a and t count?

The variable a stores the selected thread number for each parallel iteration, while t stores a non-atomic counter that all threads with the same ID increment. Unless no two threads are assigned the same iteration, the final value of t will be non-deterministic as each var++ operation is in fact a read-modify-write operation:

```
movl -4(%rbp), %eax # load var into eax addl $1, %eax # increment eax by 1 movl %eax, -4(%rbp) # store eax back into var
```

1.2 Values for all elements in a and t

Table 1: Values of array a for different scheduling strategies

case / a	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
static, 0	0	0	0	0	0	0	1	1	1	1	1	1	2	2	2	2	2
static, 1	0	1	2	0	1	2	0	1	2	0	1	2	0	1	2	0	1
dynamic, 1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
dynamic, 2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
guided, 5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

 $Table\ 2:\ Values\ of\ array\ t\ for\ different\ scheduling\ strategies\ -\ keep\ in\ mind\ that\ these\ values\ are\ not\ reproducible\ /\ deterministic.$

case / t	0	1	2
static, 0	74307862	7	1806905557
static, 1	8591638	7	1872621781
dynamic, 1	6150416	18	1875062992
dynamic, 2	40737057	1	1840476368
guided, 5	51370273	1	1829843168

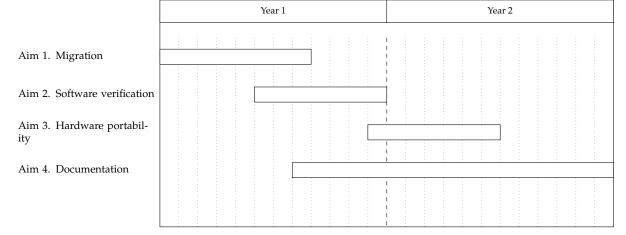
Table 3: Duration of independent tasks we want to schedule optimally.

Task ID	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
Task duration	1	2	1	2	1	2	1	2	1	2	1	2	1	2	4	3	3

2 Exercise 2

2.1 Optimal Schedule

Assuming the given tasks can be executed independently and the goal is to minimize the total execution time, with 4 workers available, the optimal schedule can be calculated as follows:



- 2.2 Schedule static,3
- 2.3 Schedule dynamic, 2
- 3 Exercise 3
- $3.1\,$ Fix the problems with this OpenMP code
- 4 Exercise 4
- 4.1 What is the output of the three different versions?
- 4.2 How often is the function omp_{tasks} called?
- 5 Exercise 5
- 5.1 Parallelize the pixel computation
- 5.2 Running time analysis
- 5.3 Influence of schedule parameter
- 6 Exercise 6
- 6.1 Parallelize the filter computation
- 6.2 Strong scaling analysis
- 6.3 Weak scaling analysis
- 7 Exercise 7
- 7.1 Convert OpenMP code to CUDA
- 7.2 Running time analysis
- 7.3 Impact of block size
- 7.4 Running time: CPU vs GPU code

8 Addendum: Raw Data

1168	1	1	0.0603872
1168	1	1	0.0607409
1168	1	1	0.0600319
1168	2	1	0.196807
1168	2	1	0.2452
1168	2	1	0.19003
1168	4	1	3.45923
1168	4	1	3.90704
1168	4	1	3.45583
1168	8	1	5.395
1168	8	1	5.45436
1168	8	1	4.53896
1168	16	1	10.7055
1168	16	1	10.5507
1168	16	1	10.2593
1168	24	1	17.3402
1168	24	1	18.5362
1168	24	1	17.2604
1168	32	1	26.1056
1168	32	1	25.1663
1168	32	1	27.9486

Figure 1: Raw output from "filter strong" job.

1168	1	1	0.060196
1168	1	1	0.0609
1168	1	1	0.060195
1168	2	2	0.401089
1168	2	2	0.635222
1168	2	2	1.18221
1168	4	4	14.4383
1168	4	4	13.3359
1168	4	4	9.2267
1168	8	8	44.0875
1168	8	8	44.8141
1168	8	8	42.5354

 $Figure\ 2: Raw\ output\ from\ "weak\ scaling"\ job.\ Timed\ out\ on\ \textit{slurmstepd}\ due\ to\ time\ out\ /\ time\ limit.$

90	1	0.110155						
90	1	0.109749						
90	1	0.109885						
90	2	0.056617						
90	2	0.056599						
90	2	0.056612						
90	4	0.045880						
90	4	0.045966						
90	4	0.045863						
90	8	0.031120						
90	8	0.031132						
90	8	0.031170						
90	16	0.018182						
90	16	0.018227						
90	16	0.018220						
90	24	0.013238						
90	24	0.013257						
90	24	0.013180						
90	32	0.014816						
90	32	0.017296						
90	32	0.014814						
1100	1	16.306608						
1100	1	16.316588						
1100	1	16.284397						
1100	2	8.175213						
1100	2	8.178992						
1100	2	8.170321						
1100	4	6.621239						
1100	4	6.678632						
1100	4	6.639713						
1100	8	4.557337						
1100	8	4.554004						
1100	8	4.586490						
1100	16	2.447131						
1100	16	2.448894						
1100	16	2.447200						
1100	24	1.731222						
1100	24	1.718731						
1100	24	1.718424						
1100	32	1.312658						
1100	32	1.313263						
1100	32	1.320209						

Figure 3: Raw output from "juliap" job.

"static"	1100	16	2.450491
"static"	1100	16	2.448260
"static"	1100	16	2.449136

Figure 4: Raw output from "juliap2" job.