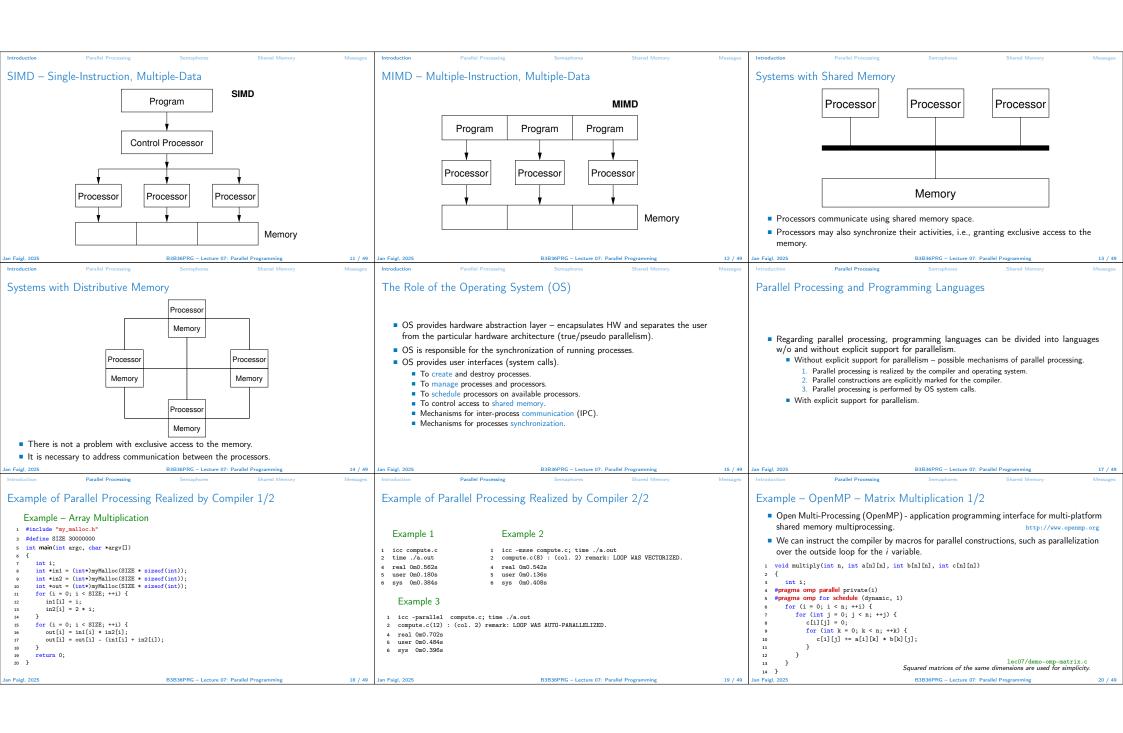
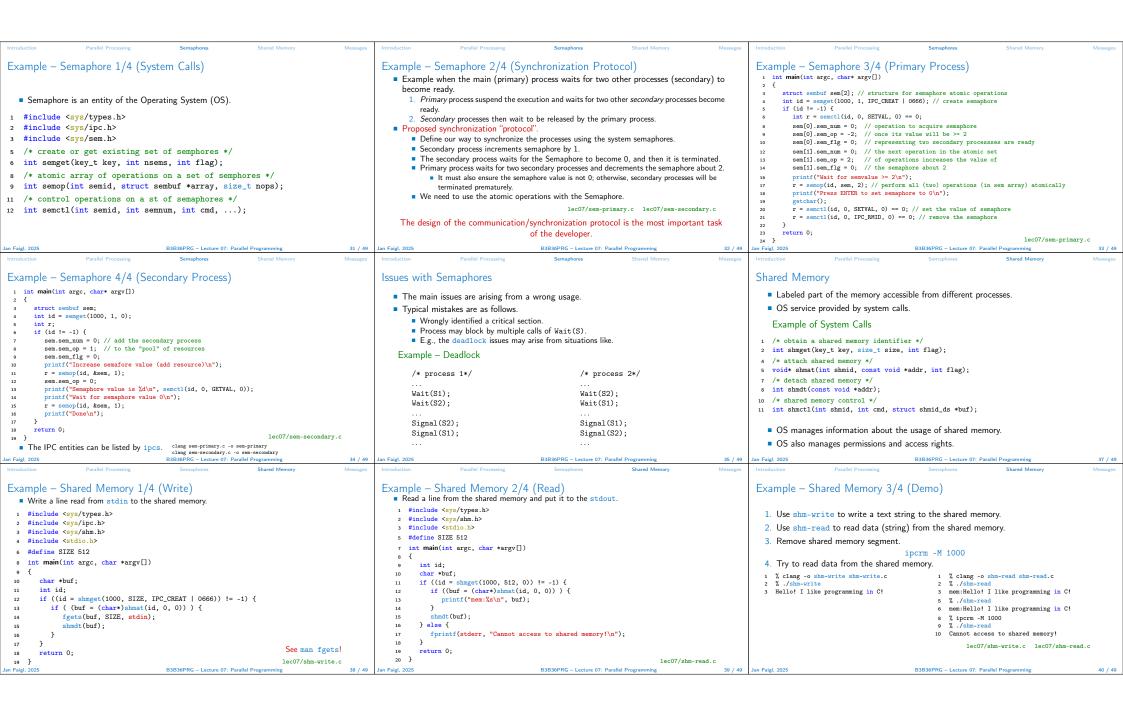
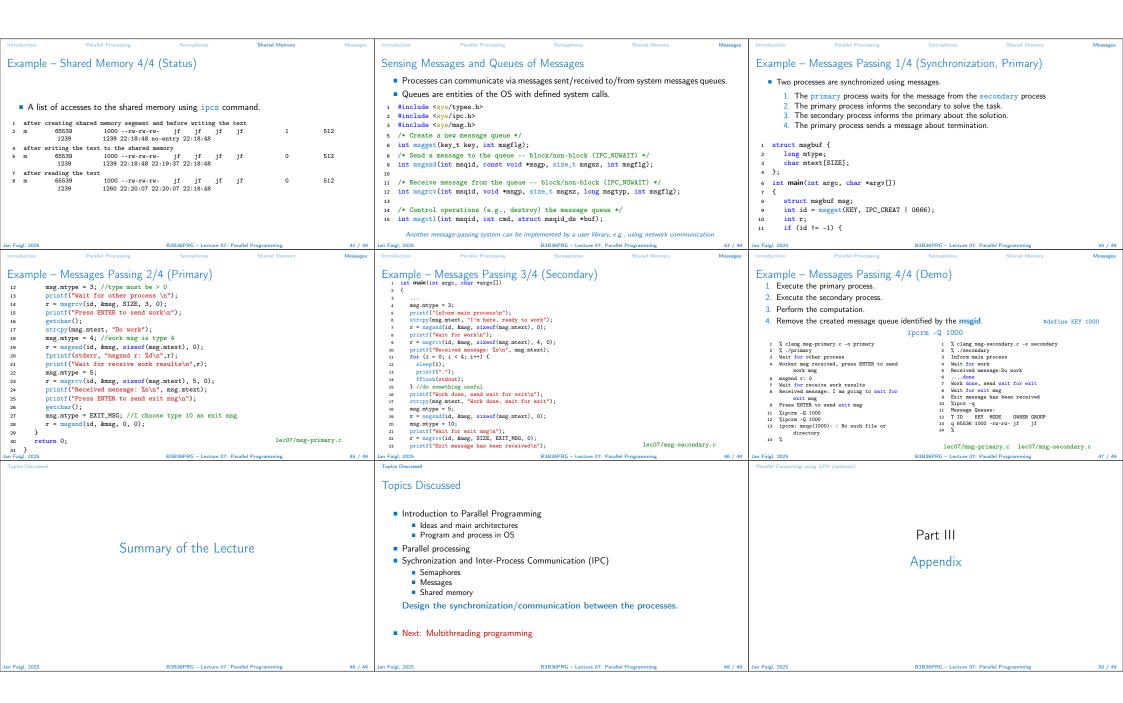
## Overview of the Lecture **Parallel Programming** ■ Part 1 - Introduction to Parallel Programming Introduction Part I Jan Faigl Parallel Processing Part 1 – Introduction to Parallel Programming Department of Computer Science Faculty of Electrical Engineering Semaphores Czech Technical University in Prague Shared Memory Lecture 07 B3B36PRG - Programming in C Messages Introduction Parallel Programming Motivation Why to Deal with Parallel Programming Process - Executed Program Process is an executed program running in a dedicated memory space. Process is an entity of the Operating System (OS) that is scheduled for independent Increase computational power. ■ The idea of parallel programming comes from the 60s with the first multi-program and Having a multiprocessor system, we can solve the computational problem faster. pseudo-parallel systems. Efficient usage of the computational power. Process is usually in one of three basic states: Parallelism can be hardware or software-based. Even a running program may wait for data. Executing – currently running on the processor (CPU); Hardware-based – true hardware parallelism of multiprocessor systems. Blocked – waiting for the periphery; Software based – pseudo-parallelism. E.g., a usual program with user interaction typically waits for the user input. Waiting – waiting for the processor. Simultaneous processing of many requests. ■ Pseudo-parallelism - A program with parallel constructions may run in the pseudo- Handling requests from individual clients in client/server architecture. A process is identified in the OS by its identifier, e.g., Process IDentificator PID. parallel environment on single or multiprocessor systems. Scheduler of the OS manages running processes to be allocated to the available proces-**Process States** Multi-processor Systems Possible Architectures for Parallel Executions Data become ready External event Ready Control of individual instructions. to run the process processes ■ SIMD - Single-Instruction, Multiple-Data - same instructions are simultaneously performed on different data. · "Processors" are identical and run synchronously. Multiprocessor systems allow true parallelism. Transition E.g., "Vectorization" such as MMX, SSE, 3Dnow!, and AVX, AVX2, etc. It is necessary to synchronize processors and support data communication. to the head ■ MIMD - Multiple-Instruction, Multiple-Data - processors run independently and queue with Blocked Resources for activity synchronization. asynchronously ready processes Scheduler picks processes Resources for communication between processors (processes) another process Memory Control Access. Systems with shared memory – central shared memory. Running System call that cannot E.g., multi-core CPUs. be handled immediatelly Systems with distributed memory – each processor has its own memory. process E.g., computational grids. Process asked for termination System call that can be performed immediatelly



Parallel Processing Example - OpenMP - Matrix Multiplication 2/2 Languages with Explicit Support for Parallelism Parallelism – Statement Level Example - parbegin-parend block ■ Comparison of matrix multiplication with 1000× 1000 matrices using OpenMP on iCore5 parbegin (2 cores with HT) It has support for the creation of new processes.  $S_1$ ; Running process creates a copy of itself.  $S_2$ ; gcc -std=c99 -02 -o demo-omp demo-omp-matrix Both processes execute the identical code (copied). ./demo-omp 1000  $S_n$ Size of matrices 1000 x 1000 naive • The parent process and child process are distinguished by the process identifier (PID). multiplication with O(n^3) narend The code segment is explicitly linked with the new process. c1 == c2: 1 Multiplication single core 9.33 sec Statement S<sub>1</sub> are S<sub>n</sub> executed in parallel. Regardless of how a new process is created, the most important is the relation to the Multiplication multi-core 4.73 sec parent process execution and memory access. • Execution of the main program is interrupted until all statements  $S_1$  to  $S_n$  are terminated. 9 export OMP\_NUM\_THREADS=2 Does the parent process stop its execution till the end of the child process? ./demo-omp 1000 Statement S<sub>1</sub> are S<sub>n</sub> executed in parallel. Size of matrices 1000 x 1000 naive Is the memory shared by the child and parent processes? Example - doparallel multiplication with O(n^3) Granularity of the processes – parallelism ranging from the level of the instructions to 1 for i = 1 to n doparalel { Multiplication single core 9.48 sec the parallelism of programs. for j = 1 to n do {
 c[i,j] = 0; 15 Multiplication multi-core 6.23 sec Parallel execution of the outer loop over all i. Use, e.g., top program for a list of running processes/threads. c[i,j] = c[i,j] + a[i,k]\*b[k,j];lec07/demo-omp-matrix.c 6 } } } E.g., OpenMP in C. an Faigl, 2025 Parallel Processing Parallel Processing Parallelism - Procedure Level Parallelism - Program (Process) Level Example - fork() 1 #define NUMPROCS 4 A procedure is coupled with the execution process. 2 for (int i = 0; i < NUMPROCS; ++i) {</pre> A new process can be only a whole program. pid\_t pid = fork(); if (pid == 0) { procedure P: • A new program is created by a system call, which creates a complete copy of itself, compute(i, n); including all variables and data at the moment of the call. exit(0):  $PID x_{pid} = newprocess(P);$ printf("Child %d created\n", pid); Example - Creating a copy of the process by fork system call clang demo-fork.c && ./a.out Child 2049 created Process myid 0 start computing 10 }  $killprocess(x_{pid});$ 1 if (fork() == 0) { Process myid 0 start computing Child 2050 created Process myid 1 start computing Process myid 2 start computing Child 2051 created Child 2052 created printf("All processes created\n"); /\* code executed by the child process \*/ P is a procedure and x<sub>pid</sub> is a process identifier. 12 for (int i = 0; i < NUMPROCS; ++i) { pid t pid = wait(&r): 3 } else { Assignment of the procedure/function to the process at the declaration printf("Wait for pid %d return: %d\n", pid, r); /\* code executed by the parent process \*/ Process myid 3 start computing 15 } PID  $x_{pid}$  process(P). All processes created Process myid 1 finished Process myid 0 finished 16 void compute(int myid, int n) ■ The process is created at the creation of the variable x. 17 printf("Process myid %d start computing\n", myid); Wait for pid 2050 return: E.g., fork() in C The process is terminated at the end of x or sooner. 18 Process myid 3 finished Process myid 2 finished Wait for pid 2049 return E.g., Threads (pthreads) in C. printf("Process myid %d finished\n", myid); 20 21 } lec07/demo-fork.c Semaphores Implementation Semaphore Usage of Semaphores Semaphores can be utilized for defining a critical sections. ■ E. W. Dijkstra – Semaphore is a mechanism to synchronize parallel processes with Critical sections are a part of the program where exclusive access to the shared Semaphore is an integer variable with the following operations. memory (resources) must be guaranteed. InitSem - initialization. Example of critical section protected by a semaphore Operations with a semaphore must be atomic. InitSem(S,1); The processor cannot be interrupted during the execution of the operation. If S > 0 then  $S \leftarrow S - 1$  (resources are available, in this case, acquire one). Wait(S): Otherwise suspend execution of the calling process (wait for S become S > 0). Machine instruction TestAndSet reads and stores the content of the addressed /\* Code of the critical section \*/ memory space and sets the memory to a non-zero value. If there is a waiting process, awake it (let the process acquire one resource). Signal(S); Signal During execution of the TestAndSet instructions, the processor holds the system bus, Otherwise increase value of S by one, i.e.,  $S \leftarrow S + 1$  (release one resource). Synchronization of the processes using semaphores. Semaphores can be used to control access to shared resources. and access to the memory is not allowed for any other processor Example of synchronization of processes. ■ S < 0 - shared resource is in use. The process asks for access to the resources and waits /\* process p \*/ for its release /\* process q \*/ S > 0 - shared resource is available. The process releases the resource. InitSem(S,0) The value of the Semanhore can represent the number of available resources. Then we Signal(S); Wait(S); ... can acquire (or wait for) k resources – wait(k):  $S \leftarrow S - k$  for S > k, and also releases exit(); k resources – signal(k):  $S \leftarrow S + k$ . exit();





Parallel Computing using GPU (optional) Parallel Computing using GPU (optional) Parallel Computing using GPU (optional) Massive parallelism using graphics cards Computational Power (2008) **CUDA** • What is the reported processor computational power? NVIDIA Compute Unified Device Architecture. Graphics (stream) processors. Image rendering performed pixel-by-pixel can be easily parallelized. Extension of the C to access the parallel computational units of the GPU. 96 GigaFLOPs • Graphics Processing Units (GPU) have a similar (or even higher) degree of integration 102 GigaFLOPs Computation (kernel) is executed by the GPU. with the main processors (CPU). GeForce 8800 GTX 518 GigaFLOPs (including texture units) Radeon HD 4670 480 GigaFLOPs Kernel is performed in parallel using available computational units. They have a huge number of parallel processors. GeForce RTX 4060 15 110 GigaFLOPs E.g., GeForce GTX 1060 ~ 1280 cores. Main processors : Host - Main processor (process). Peak catalogue values. • The computational power can also be used in another application. Phenom X4 9950 (@2.6 GHz) 21 GigaFLOPs Device - GPU. • Processing stream of data (SIMD instructions - processors). Core 2 Duo E8600 (@3.3 GHz) 22 GigaFLOPs Data must be in the memory accessible by the GPU. Cure 2 Quad QX9650 (@3.3 GHz) 35 GigaFLOPs GPGPU - General Purpose computation on GPU. http://www.gpgpu.org Cure 2 Quad QX9650 (@3.3 GHz) 35 GigaFLOPs Host memory → Device memory OpenCL (Open Computing Language) – GPGPU abstract interface. Core i7 970 (@3.2 GHz) 42 GigaFLOPs CUDA - Parallel programming interface for NVIDIA graphics cards. Core i9-13900 (@2.00-5.60 GHz) 846 GigaFLOPs (2023) The result (of the computation) is stored in the GPU memory. Test linpack 32-bit. http://www.nvidia.com/object/cuda\_home.html Host memory ← Device memory Is the reported power really achievable? (float vs double) How about other indicators? E.g., computational power / power consumption. CSX700 has typical power consumption around 9W. B3B36PRG - Lecture 07: Parallel Programming Parallel Computing using GPU (optional) Parallel Computing using GPU (optional) Parallel Computing using GPU (ontional) CUDA - Example - Matrix Multiplication 1/8 CUDA - Computational Model CUDA - Grid, Blocks, Threads, and Memory Access NVIDIA CUDA SDK - Version 2.0, matrixMul. Kernel (computation) is divided into blocks. ■ Simple matrix multiplication. • Each block represents a parallel computation of the part of the result.  $\mathbf{C} = \mathbf{A} \cdot \mathbf{B}$  Matrices have identical dimensions n × n, E.g., a part of the matrix multiplication. where n is the multiple of the block size. ■ Each block consists of computational threads. Comparison Parallel computations are synchronization within the block. ■ naive implementation in C (3× for loop), Blocks are organized into the grid. naive implementation in C with matrix transpose. CUDA implementation. • Scalability is realized by dividing the computation into blocks. Blocks may not necessarily be computed in parallel. Based on the available number of Hardware CPU - Intel Core 2 Duo @ 3 GHz, 4 GB RAM, Access time to the memory. GPU - NVIDIA G84 (GeForce 8600 GT), 512 MB RAM. Collisions for simultaneous access of several threads. Parallel Computing using GPU (optional) Parallel Computing using GPU (optional) Parallel Computing using GPU (optional) CUDA - Example - Matrix Multiplication 2/8 CUDA - Example - Matrix Multiplication 3/8 CUDA – Example – Matrix Multiplication 4/8 Naive implementation with transpose void simple\_multiply\_trans(const int n, const float \*a, const float \*b, float \*c) Naive implementation CUDA – computation strategy Divide matrices into blocks. void simple\_multiply(const int n, for (int i = 0; i < n; ++i) { ■ Each block computes a single const float \*A, const float \*B, float \*C) bT[i\*n + i] = b[i\*n + i]; for (int j = i + 1; j < n; ++j) { sub-matrix C<sub>sub</sub>. 3 { bT[i\*n + j] = b[j\*n + i]; bT[j\*n + i] = b[i\*n + j]; Each thread of the individual blocks for (int i = 0; i < n; ++i) {</pre> computes a single element of  $C_{sub}$ . for (int j = 0; j < n; ++j) { float prod = 0; for (int i = 0; i < n; ++i) { for (int k = 0; k < n; ++k) { for (int j = 0; j < n; ++j) { prod += A[i \* n + k] \* B[k \* n + j]; float tmp = 0; for (int k = 0; k < n; ++k) { tmp += a[i\*n + k] \* bT[j\*n + k]; C[i \* n + j] = prod;c[i\*n + j] = tmp; 13 } free(bT); Jan Faigl, 2025

```
Parallel Computing using GPU (optional)
CUDA - Example - Matrix Multiplication 5/8
    CUDA - Implementation - main function
 void cuda_multiply(const int n, const float *hostA, const float *hostB, float *hostC)
 3 const int size = n * n * sizeof(float);
 4 float *devA, *devB, *devC;
 6 cudaMalloc((void**)&devA, size);
 7 cudaMalloc((void**)&devB, size);
s cudaMalloc((void**)&devC, size);
10 cudaMemcpy(devA, hostA, size, cudaMemcpyHostToDevice);
cudaMemcpy(devB, hostB, size, cudaMemcpyHostToDevice);
13 dim3 threads(BLOCK_SIZE, BLOCK_SIZE); // BLOCK_SIZE == 16
14 dim3 grid(n / threads.x, n /threads.y);
16 // Call kernel function matrixMul
17 matrixMul<<<<grid, threads>>>(n, devA, devB, devC);
19   cudaMemcpy(hostC, devC, size, cudaMemcpyDeviceToHost);
21 cudaFree(devA);
22 cudaFree(devB);
23 cudaFree(devC);
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Parallel Computing using GPU (optional)
CUDA - Example - Matrix Multiplication 8/8
    Computational time (in milliseconds)
          Naive
                Transp. CUDA
                                                                             CUDA
    208
304
            11
                   11
                           82
                                                         1104
                                                               6360
                                                                       1628
                                                                               235
308
            35
                    33
                                                               9763
                                                                       2485
   Matlab 7.6.0 (R2008a):
n=1104: A=rand(n,n); B=rand(n,n); tic; C=A*B; toc
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```

```
Parallel Computing using GPU (optional)
Parallel Computing using GPU (optional)
```

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## CUDA - Example - Matrix Multiplication 6/8

```
CUDA implementation - kernel function
_global__void matrixful(int n, float* A, float* B, float* C) {
    int bx * blockfdx x; int by = blockfdx, y;
    int tx * threadfdx x; int ty = threadfdx y;
    int aBegin = n * BLOCK_SIZE * by; //beginning of sub-matrix in the block
    int aBegin = n * BLOCK_SIZE * by; //beginning of sub-matrix in the block

     float Csub = 0;
for (
                 int a = aBegin, b = BLOCK_SIZE * bx;
               a <= aEnd;
a += BLOCK_SIZE, b += BLOCK_SIZE * n
            __shared__ float As[BLOCK_SIZE][BLOCK_SIZE]; // shared memory within __shared__ float Bs[BLOCK_SIZE][BLOCK_SIZE]; // the block
           As[ty][tx] = A[a + n * ty + tx]; // each thread reads a single element Bs[ty][tx] = B[b + n * ty + tx]; // of the matrix to the memory
             __syncthreads(); // synchronization, sub-matrix in the shared memory
           for (int k = 0; k < BLOCK_SIZE; ++k) { // each thread computes
                Csub += As[ty][k] * Bs[k][tx]; // the element in the sub-matrix
              _syncthreads();
     int c = n * BLOCK_SIZE * by + BLOCK_SIZE * bx;
C[c + n * ty + tx] = Csub; // write the results to memory
```

## CUDA - Example - Matrix Multiplication 7/8

CUDA source codes

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Example - Dedicated source file cuda\_func.cu

1. Declaration of the external function.

```
extern "C" { // declaration of the external function (cuda kernel)
void cuda_multiply(const int n, const float *A, const float *B, float *C);
```

- 2. Compile the CUDA code to the C++ code.
- 1 nvcc --cuda cuda\_func.cu -o cuda\_func.cu.cc
- 3. Compilation of the cuda\_func.cu.cc file using standard compiler.

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