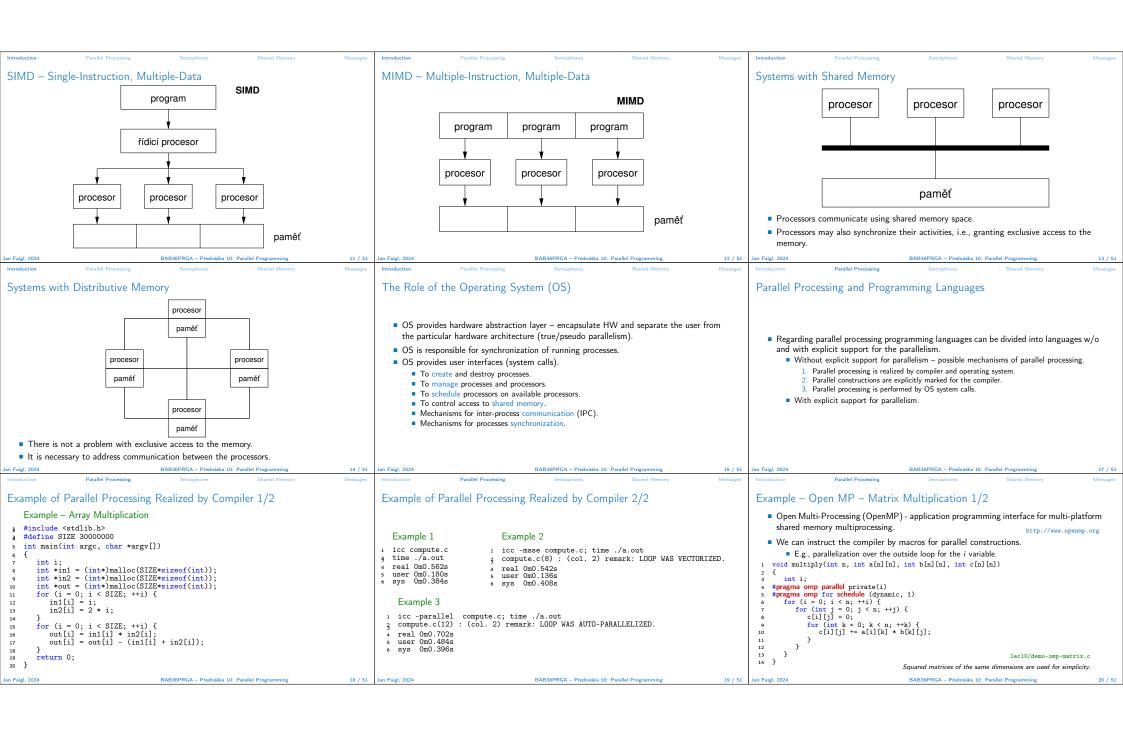
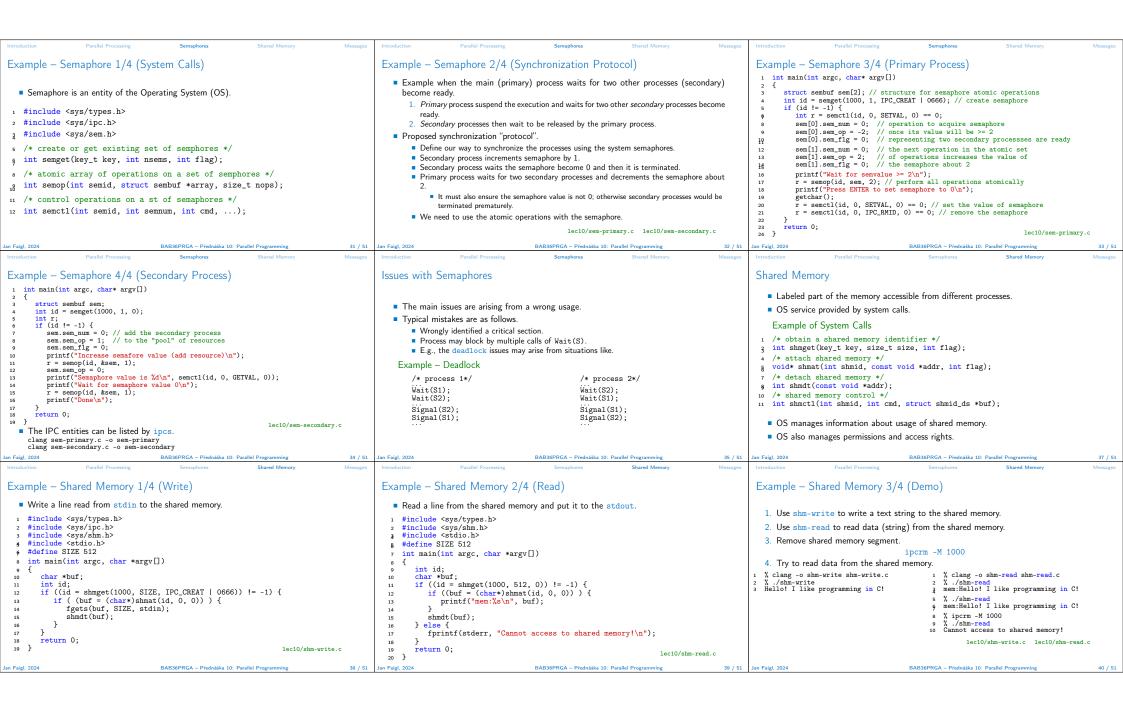
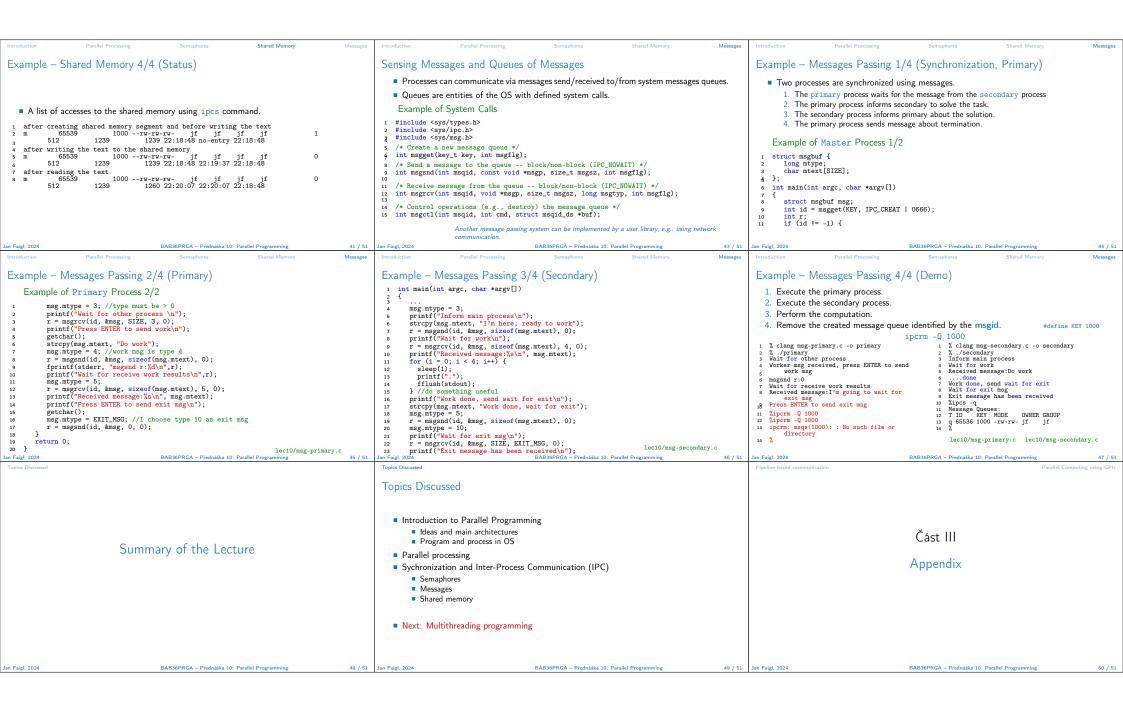
## Overview of the Lecture **Parallel Programming** ■ Part 1 - Introduction to Parallel Programming Introduction Část I Jan Faigl Parallel Processing Part 1 – Introduction to Parallel Programming Katedra počítačů Fakulta elektrotechnická Semaphores České vysoké učení technické v Praze Shared Memory Přednáška 10 BAB36PRGA - Programování v C Messages Parallel Programming Motivation Why to Deal with Parallel Programming Process - Executed Program Process is executed program running in a dedicated memory space. Process is an entity of the Operating System (OS) that is schedule for independent Increase computational power. ■ The idea of parallel programming comes from the 60s with the first multi-program and Having multi-processor 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 pseudo-parallel Handling requests from individual clients in client/server architecture. A process is identified in the OS by its identifier, e.g., Process IDentificator PID. environment on single or multi-processor systems. Scheduler of the OS manage running processes to be allocated to the available proces-**Process States** Multi-processor Systems Possible Architectures for Parallel Executions Důvod čekání pominul Spuštění procesu Control of individual instructions. Připravené z vnější příčiny. ■ SIMD - Single-Instruction, Multiple-Data - same instructions are simultaneously procesy performed on different data. · "Processors" are identical and run synchronously. Multi-processor systems allow true parallelism. Přechod E.g., "Vectorization" such as MMX, SSE, 3Dnow!, and AVX, AVX2, etc. do čela fronty It is necessary to synchronize processors and support data communication. MIMD – Multiple-Instruction, Multiple-Data – processors run independently and připravených Čekající (blokované) Resources for activity synchronization. asynchronously. procesy Procesu je odňat procesor. Resources for communication between processors (processes) Memory Control Access. Systems with shared memory – central shared memory. Proces zažádal o službu, Aktivní na kterou musí čekat E.g., multi-core CPUs. proces Systems with distributed memory – each processor has its memory Proces zažádal o ukončení. E.g., computational grids. Proces zažádal o službu, kterou lze vyřídit okamžitě.



Example - Open MP - Matrix Multiplication 2/2 Languages with Explicit Support for Parallelism Parallelism - Statement Level Example - parbegin-parend block parbegin ■ Comparison of matrix multiplication with 1000× 1000 matrices using OpenMP on S<sub>1</sub>; It has support for creation of new processes. iCore5 (2 cores with HT). Running process create a copy of itself. S, gcc -std=c99 -02 -o demo-omp demo-omp-matrix.c Both processes execute the identical code (copied). ./demo-omp 1000 parend • The parent process and child process are distinguished by the process identifier (PID). Size of matrices 1000 x 1000 naive Statement S<sub>1</sub> are S<sub>n</sub> executed in parallel. The code segment is explicitly linked with the new process. multiplication with O(n^3) • Execution of the main program is interrupted until all statements  $S_1$  to  $S_n$  are terminated. Multiplication single core 9.33 sec Multiplication multi-core 4.73 sec Regardless how a new process is created, the most important is the relation to the Statement S<sub>1</sub> are S<sub>n</sub> executed in parallel. parent process execution and memory access. Example – doparallel export OMP\_NUM\_THREADS=2 Does the parent process stops its execution till the end of the child process? ./demo-omp 1000 Size of matrices 1000 x 1000 naive for i = 1 to n doparalel { Is the memory shared by the child and parent processes? for j = 1 to  $n do {$ multiplication with O(n^3) c[i,j] = 0; Granularity of the processes – parallelism ranging from the level of the instructions to for k = 1 to n do { Multiplication single core 9.48 sec Multiplication multi-core 6.23 sec Use, e.g., top program for a list of running the parallelism of programs. c[i,j] = c[i,j] + a[i,k]\*b[k,j];6 } } } lec10/demo-omp-matrix.c Parallel execution of the outer loop over all i. F.g., OpenMP in C. Parallel Processin Parallel Processing Parallelism - Procedure Level Parallelism - Program (Process) Level Example - fork() 1 #define NUMPROCS 4 2 for (int i = 0; i < NUMPROCS; ++i) {</pre> A procedure is coupled with the execution process. pid\_t pid = fork(); A new process can be only a whole program. if (pid == 0) { procedure P: A new program is created by a system call, which creates a complete copy of itself compute(i, n): exit(0): including all variable and data at the moment of the call.  $PID x_{pid} = newprocess(P);$ } else f printf("Child %d created\n", pid); clang demo-fork.c && ./a.out Example - Creating a copy of the process by fork system call Child 2049 created  $killprocess(x_{pid});$ Process myid 0 start computing Child 2050 created 10 } 1 if (fork() == 0) { • P is a procedure and  $x_{pid}$  is a process identifier. printf("All processes created\n"); Process myid 1 start computing Process myid 2 start computing Child 2051 created /\* code executed by the child process \*/ 12 for (int i = 0; i < NUMPROCS; ++i) { Assignment of the procedure/function to the process at the declaration 3 } else { pid\_t pid = wait(&r); Child 2052 created Process myid 3 start computing /\* code executed by the parent process \*/ printf("Wait for pid %d return: %d\n", pid, r); PID  $x_{pid}$  process(P). All processes created • The process is created at the creation of the variable x. roid compute(int myid, int n) Process myid 1 finished Process myid 0 finished The process is terminated at the end of x or sooner. E.g., fork() in C Wait for pid 2050 return: 0 printf("Process myid %d start computing\n", myid); Process myid 3 finished E.g., Threads (pthreads) in C. 20 printf("Process myid %d finished\n", myid); Wait for pid 2049 return: 0 Wait for pid 2051 return: 21 } lec10/demo-fork. Wait for pid 2052 return: 0 Semaphore Semaphores Implementation 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 is a part of the program where exclusive access to the shared memory Semaphore is an integer variable with the following operations. (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 execution of the operation. If S > 0 then  $S \leftarrow S - 1$  (resources are available, in this case, acquire one). Wait(S): /\* Code of the critical section \*/
Signal(S); Otherwise suspend execution of the calling process (wait for S become S > 0). Machine instruction TestAndSet reads and stores a content of the addressed memory space and set the memory to a non-zero value. If there is a waiting process, awake it (let the process acquire one resource). Synchronization of the processes using semaphores. 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). Example of synchronization of processes. and access to the memory is not allowed for any other processor Semaphores can be used to control access to shared resource.  $\blacksquare$  S < 0 – shared resource is in use. The process asks for the access to the resources and /\* process p \*/ /\* process q \*/ waits for its release. InitSem(S.0) Signal(S); Wait(S): ... S > 0 - shared resource is available. The process releases the resource. exit(); The value of the semanhore can represent the number of available resources. Then we can acquire (or wait for) k resources – wait(k):  $S \leftarrow S - k$  for S > k, and also releases k resources – signal(k):  $S \leftarrow S + k$ . Process p waits for termination of the process q.





Pipeline-based communication Parallel Computing using GPU Parallel Computing using GPU Signal Generator and Visualization Massive parallelism using graphics cards Computational Power (2008) Let we have a signal generator sgen, which is a program that generate a sequence of What is the reported processor computational power? values to its stdout. Graphics (stream) processors. Image rendering performed pixel-by-pixel can be easily parallelized. ■ The visualization can be realized in another application tsignal\_viewer that reads 96 GigaFLOPs Cell 102 GigaFLOPs GeForce 8800 GTX 518 GigaFLOPs Graphics Processing Units (GPU) has similar (or even higher) degree of integration signal values from the stdin. with the main processors (CPU). Radeon HD 4670 480 GigaFLOPs ■ We can connect these two applications by ./sgen | ./tsignal\_viewer. Peak catalogue values They have huge number of parallel processors. Main processors : \$ ./sgen | wc E.g., GeForce GTX 1060 ~ 1280 cores. Phenom X4 9950 (@2.6 GHz) Core 2 Duo E8600 (@3.3 GHz) 21 GigaFLOPs 65772 65772 618529 • The computational power can also be used in another applications. 22 GigaFLOPs 1.031029 \$ ./sgen | ./tsignal\_viewer Cure 2 Quad QX9650 (@3.3 GHz) 35 GigaFLOPs 0.916540 Processing stream of data (SIMD instructions - processors). 0.748307 Cure 2 Quad QX9650 (@3.3 GHz) Core i7 970 (@3.2 GHz) 35 GigaFLOPs 0.549199 42 GigaFLOPs GPGPU - General Purpose computation on GPU. http://www.gpgpu.org 0.342897 0.149521 Test linpack 32-bit. · HW9B is an extension of the con- OpenCL (Open Computing Language) – GPGPU abstract interface. -0.016643 cept using user defined commu-Is the reported power really achievable? -0.147374 CUDA - Parallel programming interface for NVIDIA graphics cards. -0.147374 -0.239553 -0.293963 -0.314364 -0.306833 nication protocol between a sig-(float vs double) http://www.nvidia.com/object/cuda\_home.html nal generator and control applica-How about other indicators? 200 400 600 80 tion with visualization and multi-E.g., computational power / power consumption -0.279048 -0.239311 CSX700 has typical power consumption around 9W. -0.195373 Parallel Computing using GPU Parallel Computing using GPU Parallel Computing using GPI CUDA CUDA - Computational Model CUDA - Grid, Blocks, Threads, and Memory Access NVIDIA Compute Unified Device Architecture. Kernel (computation) is divided into blocks. Extension of the C to access to the parallel computational units of the GPU. Each block represent a parallel computation of the part of the result. Computation (kernel) is executed by the GPU. E.g., a part of the matrix multiplication. • Kernel is performed in parallel using available computational units. Each block consists of computational threads. Host - Main processor (process). Parallel computations are synchronization within the block. Device - GPU. Blocks are organized into the grid. Data must be in the memory accessible by the GPU. Scalability is realized by dividing the computation into blocks. Host memory → Device memory Blocks may not be necessarily computed in parallel. Based on the available number of ■ The result (of the computation) is stored in the GPU memory. Access time to the memory. Host memory ← Device memory Collisions for simultaneous access of several threads. Parallel Computing using GPU Parallel Computing using GPU Parallel Computing using GPU CUDA - Example - Matrix Multiplication 1/8 CUDA - Example - Matrix Multiplication 2/8 CUDA - Example - Matrix Multiplication 3/8 Naive implementation with transpose void simple\_multiply\_trans(const int n, NVIDIA CUDA SDK - Version 2.0, matrixMul. Naive implementation const float \*a, const float \*b, float \*c) Simple matrix multiplication. void simple\_multiply(const int n, float \* bT = create\_matrix(n);  $\mathbf{C} = \mathbf{A} \cdot \mathbf{B}$ const float \*A, const float \*B, float \*C) for (int i = 0; i < n; ++i) {
bT[i\*n + i] = b[i\*n + i]; ■ Matrices have identical dimensions  $n \times n$ . for (int j = i + 1; j < n; ++j) { where n is the multiple of the block size. for (int i = 0; i < n; ++i) {</pre> bT[i\*n + j] = b[j\*n + i]; bT[j\*n + i] = b[i\*n + j];for (int j = 0; j < n; ++j) { float prod = 0;

for (int i = 0; i < n; ++i) {
 for (int j = 0; j < n; ++j) {
 float tmp = 0;
 for (int k = 0; k < n; ++k) {</pre>

c[i\*n + j] = tmp;

free(hT).

21 22

tmp += a[i\*n + k] \* bT[j\*n + k];

for (int k = 0; k < n; ++k) {

C[i \* n + j] = prod;

10

11

12 }

13 }

prod += A[i \* n + k] \* B[k \* n + j];

naive implementation in C (3× for loop),

CUDA implementation.

Hardware

naive implementation in C with matrix transpose.

CPU - Intel Core 2 Duo @ 3 GHz, 4 GB RAM,

■ GPU - NVIDIA G84 (GeForce 8600 GT), 512 MB RAM.

Parallel Computing using GPU Parallel Computing using GPU CUDA - Example - Matrix Multiplication 4/8 CUDA - Example - Matrix Multiplication 5/8 CUDA - Implementation - main function void cuda\_multiply(const int n, CUDA – computation strategy const float \*hostA, const float \*hostB, float \*hostC) Divide matrices into blocks. const int size = n \* n \* sizeof(float);
float \*devA, \*devB, \*devC; ■ Each block computes a single sub-matrix C<sub>sub</sub>. cudaMalloc((void\*\*)&devA, size); cudaMalloc((void\*\*)&devB, size);
cudaMalloc((void\*\*)&devC, size); ■ Each thread of the individual blocks computes a single element of  $C_{sub}$ . cudaMemcpy(devA, hostA, size, cudaMemcpyHostToDevice);
cudaMemcpy(devB, hostB, size, cudaMemcpyHostToDevice); dim3 threads(BLOCK\_SIZE, BLOCK\_SIZE); // BLOCK\_SIZE == 16 dim3 grid(n / threads.x, n /threads.y); // Call kernel function matrixMul matrixMul<<<grid, threads>>>(n, devA, devB, devC); cudaMemcpy(hostC, devC, size, cudaMemcpyDeviceToHost); cudaFree(devA); 23 cudaFree(devR) cudaFree(devC); 24 Parallel Computing using GPU Parallel Computing using GPU CUDA – Example – Matrix Multiplication 7/8 CUDA - Example - Matrix Multiplication 8/8 Computational time (in milliseconds) CUDA source codes. 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. nvcc --cuda cuda\_func.cu -o cuda\_func.cu.cc CUDA CUDA Naive Transp. Naive Transp. 3. Compilation of the cuda\_func.cu.cc file using standard compiler. 122 208 11 11 82 1104 6360 1628 235 304 308 35 33 1264 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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CUDA - Example - Matrix Multiplication 6/8

CUDA implementation - kernel function

1 __global__ void matrixMultint n, float* A, float* B, float* C) {
2    int bx = blockIdx.x; int by = blockIdx.y;
3    int tx = threadIdx.x; int ty = threadIdx.y;
4    int aBegin = n * BLOCK_SIZE* by; //beginning of sub-matrix in the block
5    int aBnd = aBegin + n - 1; //end of sub-matrix in the block
6    float Caub = 0;
7    for (int a = aBegin, b = BLOCK_SIZE * bx;
a <= aEnd;
a <= aEnd;
10    a *= BLOCK_SIZE b *= BLOCK_SIZE * bx;
a <= aEnd;
11    a *= BLOCK_SIZE b *= BLOCK_SIZE * bx;
12    a <= begin + b = BLOCK_SIZE * bx;
13    a *= begin + b = BLOCK_SIZE * bx;
14    Ast[v][tx] = A[a + n * vy + tx]; // each thread reads a single element
15    Bs[ty][tx] = B[b + n * vy + tx]; // each thread reads a single element
16    Scub + a As[ty][tx] = B[b] * (a *= b) * (b *= b) * (
```

BAB36PRGA - Přednáška 10: Parallel Programmin

C[c + n \* ty + tx] = Csub; // write the results to memory

Parallel Computing using GPU