# DM510: Threads and Concurrency

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#### Disclaimer

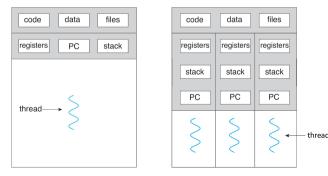
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### Today's lecture

• Chapter 4 of course book

# Multi-threaded process

- Last lecture: several processes (running simultaneously) can coorporate on tasks
- Instead of several processes, often several threads within one process are used
- Reasons of using several threads/processes: continuing during a blocking call or CPU intensive task (especially for responsiveness), parallelism (multicore processor)

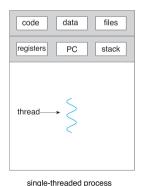


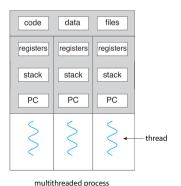
single-threaded process

multithreaded process

## Multi-threaded process

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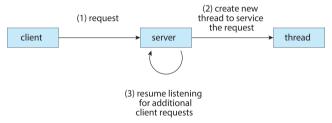


#### Advantages of threads

- Lower resource overhead (especially memory)
- Faster thread creation and thread switches
- Communication via shared memory simpler (address space is shared)

### Examples

• A web server responding to HTTP requests:



• A database needs to search a large table for a specific entry. It scans the first quarter on one thread (running on CPU core 1), the second quarter on another thread (running on CPU core 2), etc.

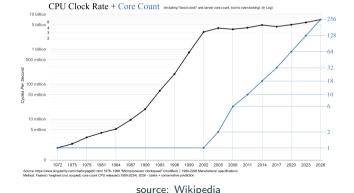
# Parallelism

### Background on parallelism

- Speed at which processors execute instructions (in million cycles per second: MHz) is no longer increasing due to physical limits
- Performance improvements are now mainly due to parallelism (more CPU cores)

### Challenges

- Dividing activities
- Balancing
- Data splitting
- Data dependency
- Testing and debugging



### Concurrency and parallelism

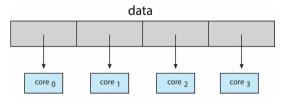
Note that **concurrency** (simultaneous progress on different tasks) is also possible on single core (via preemptive scheduler):

True **parallelism** on several cores:

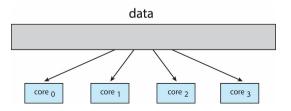
core 1 
$$T_1$$
  $T_3$   $T_1$   $T_3$   $T_1$  ...  
core 2  $T_2$   $T_4$   $T_2$   $T_4$   $T_2$  ...  
time

### Forms of parallelism

Data parallelism: distribute data to cores performing the same task on each batch



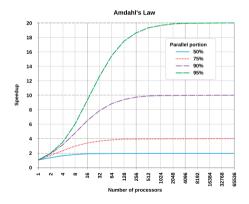
Task parallelism: distribute different tasks to cores, working on same data



### Amdahl's law

Possible speedup due to parallelism (even with many cores) is bounded. It depends highly on how sequential the program is

Theoretical speedup Let S be the ratio of sequential operations to all operations. Then speedup  $\leq \frac{1}{S + \frac{1-S}{1-S}}$ Optimistic estimate that does not account for communication overhead



Source: Wikipedia

# Implementation Details

# Types of threads

- Typically, libraries handle internals of thread implementation
- **pthreads** (POSIX threads) is a widespread API specification implemented in many libraries, especially on UNIX systems, see demo/course resources

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- Conceptionally, two implementations of threads can be distinguished:

#### Kernel threads

The kernel creates, schedules, terminates threads like processes, except resources (e.g. address space) are shared. Thus:

- threads may run on different cores
- other threads can continue when one is in a blocking system call

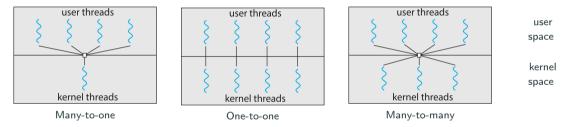
#### **User threads**

Threads implemented in user mode. Kernel possibly not aware of threads.

- Limited: process uses single core, blocking call suspends all threads
- Low on resources, applicable also to e.g. embedded systems without OS or with very limited OS

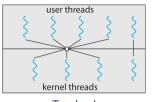
### User-kernel thread mapping

• In principle one could map user threads (of one process) to a smaller number of kernel threads in different ways:



- one-to-one model is by far the most popular and default option
- Possible motivation for not using one-to-one is that kernel threads are more resource heavy than user threads and we may not have perfect control over scheduling of kernel threads

# Light-weight processes

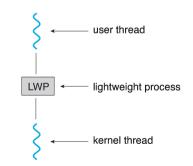


Two-level

- A variant of many-to-many is **two-level** model that allows user process to carefully decide how user threads are allocated to kernel threads
- This provides fine-grained control, but also requires a sophisticated kernel interface, usually achieved by lightweight processes (LWP)

### Lightweight process

- A LWP is a wrapper for a kernel thread
- threads are allocated to LWPs by user process
- Kernel informs process via **upcall** of: user thread enters blocking call (so that LWP can be used for other user thread) or thread is no longer blocked



### **Cancelling threads**

- We could cancel a threads execution at whatever point it currently is (asynchronous cancellation), but this might leave process in an inconsistent state
- Safer alternative: thread specifies points where it can be cancelled (deferred cancellation). Then it continues to run until it reaches such a point

# Implicit Multi-Threading

# Motivation of implicit multi-threading

Optimizing performance and ensuring correctness in multithreaded programs can be difficult:

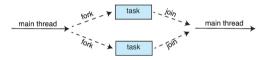
- how many kernel threads should be used? Depends on CPU architecture
- How to pass data between different threads? *Requires synchronization, which causes overhead*
- In general, writing optimized code for all architectures may be difficult.

At the same time, many similar programs face these challenges.

There are various libraries and frameworks, which take a computational task and parallelize it almost automatically

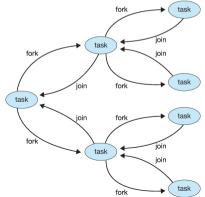
# Fork-join model

• Structure: when task is large, split into subtasks, do both in parallel (fork) and wait for results (join)



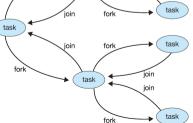
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```
import java.util.concurrent.*;
public class SumTask extends RecursiveTask<Integer>
  static final int THRESHOLD = 1000:
  private int begin:
  private int end:
  private int[] array;
  public SumTask(int begin, int end, int[] array) {
     this.begin = begin;
     this.end = end:
     this.array = array;
  protected Integer compute()
     if (end - begin < THRESHOLD)
       int sum = 0:
       for (int i = begin; i \le end; i++)
          sum += array[i];
       return sum:
     else -
       int mid = (begin + end) / 2:
       SumTask leftTask = new SumTask(begin, mid, array);
       SumTask rightTask = new SumTask(mid + 1, end, arrav);
       leftTask.fork():
       rightTask.fork();
       return rightTask.join() + leftTask.join();
```

### OpenMP

- Library for parallelizing C/C++ programs
- Parallelization of for-loops requires only minimal changes: adding "#pragma omp parallel for" before the loop
- See demo/course resources

# Other parallelization

### Single-instruction-multiple-data (SIMD)

- Modern processors come with special registers that can hold vectors, e.g. 4 integers (of 4 bytes each)
- SIMD instructions can perform parallel operations on vectors.
- They perform data-parallelism
- Note: this is not threading or an operating system aspect

#### General-purpose graphics processors (GPGPU)

- Graphic processors come with very many, sometimes 1000s, of (very simple) cores
- These traditionally perform computer graphics tasks, but are being used increasingly for other purposes (e.g. linear algebra) as well
- Accessed as a device via driver
- Not easy to program