

# Parallel computing

- "History": From 1980's to ~2004, processor performance increased mainly due to frequency scaling (increased clock rate).

[Codes would run faster and faster without changes]

$$\text{Runtime} = \frac{\text{Instructions}}{\text{program}} \times \frac{\text{cycles}}{\text{instruction}} \times \frac{\text{time}}{\text{cycle}}$$

$\frac{1}{f_{\text{proc}}}$

Power consumption in chip:

$$P \propto f$$

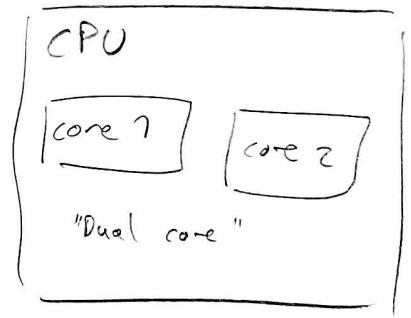
so increased  $f \rightarrow$  increased  $P$  (as expected!)

$$\left[ f_{\text{proc}} = \frac{\text{cycles}}{\text{second}} \right]$$

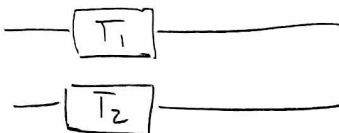
$\Rightarrow$  Problems w/ overheating etc.

From ~2004; performance increase mainly from shift to parallel comp., and in part multicore processors

- Challenge: Requires changes on software side!  
Need to distribute tasks or data across threads or processes!



[Old-school parallelization:  
Give each student in a class their own ~~eg~~ eg to solve :)]



## • Two main approaches

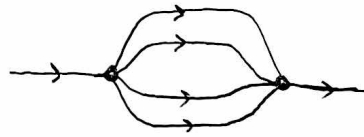
### 1) • Shared memory

#### • Threading

#### • Single computer/node

#### • Example : OpenMP , $\langle \text{thread} \rangle$

#### • Single process , can switch between one and multiple threads



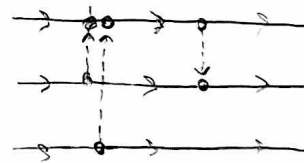
### 2) • Distributed memory

#### • Message passing

#### • can be used on single computer/nodes or between computers/nodes

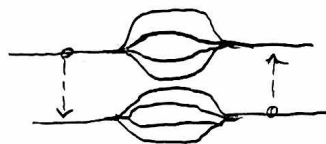
#### • Example : MPI

#### • Multiple indep. processes



## • These approaches can be combined :

- Multiple processes , each spawning multiple threads (sharing that process' memory)



[ Mention  
GAMBIT ]

## • Parallelization comes with some overhead , from spawning threads , passing messages , etc.

Only useful if  $\Delta t_{\text{task}} > \Delta t_{\text{overhead}}$  . [ Also : comes with substantial room for mistakes and bugs ... ]

## ◦ Parallelization (cont. from last time)

### ◦ Recap:

- Two approaches: Shared memory (e.g. OpenMP)  
Distributed memory (e.g. MPI)

### ◦ [Go through OpenMP examples in the Git repo]

### ◦ Different approaches to parallelize project 4:

Alt 1) Parallelize loop over temperatures (simplest!)

Alt 2) For each temp., use parallelization to run multiple MCMC chains (multiple "walkers")  
Can either:

- Increase number of threads and decrease cycles per thread

Each walker/chain needs burn-in

or:

⇒ same accuracy, shorter time

- Increase number of threads while keeping number of cycles per thread fixed

⇒ higher accuracy (more MC samples), same time

Alt 3) For each temperature and each MC cycle, parallelize the "sweep" over the spin matrix

- Most complicated! (Don't do this...)
- Most overhead

- How to define speedup from parallelization

$$\text{Speedup} = \frac{\text{time with single thread/process}}{\text{time with } n \text{ threads/processes}} = \frac{T_1}{T_n}$$

- Ideal case:  $n$  threads  $\Leftrightarrow T_n = \frac{T_1}{n} \Leftrightarrow$  speedup factor is  $n$

- In most cases we will not get ideal speedup
- In rare cases we can get better than ideal speedup (e.g. through changes in memory access)

- Keep in mind: A complicated algorithm with less than ideal speedup from parallelization can still be a better choice than a simple algorithm with better (ideal?) parallelization speedup!

- Example: Find the maximum of a complicated, high-dim function through

1) random sampling (ideal speedup, "embarrassingly parallelizable") or grid scan

2) sophisticated optimization algorithm, e.g. differential evolution

[show example from paper]

(needs communication and synchronization  $\rightarrow$  less than ideal speedup)

◦ Upper bound on speedup :

- A task takes time  $T_1$  on single thread/process

→ Fraction of time spent in perfectly parallelizable code :  $f$

→ Non-parallelizable fraction :  $1-f$

◦ Single thread/process :

$$T_1 = (1-f)T_1 + fT_1$$

◦ On  $n$  threads/processes :

$$T_n = (1-f)T_1 + f \frac{T_1}{n}$$

◦ Speedup :

$$\frac{T_1}{T_n} = \frac{T_1}{(1-f)T_1 + f \frac{T_1}{n}} = \frac{1}{(1-f) + \frac{f}{n}}$$

$$\boxed{\lim_{n \rightarrow \infty} \frac{T_1}{T_n} = \frac{1}{1-f}}$$

Amdahl's law

Example: If 99% of a task is parallelizable ( $f=0.99$ )  
the maximum possible speedup factor  
is  $\frac{1}{1-0.99} = \frac{1}{0.01} = 100$

Example 2 :  $f = 0.80 \Rightarrow$  max speedup is 5