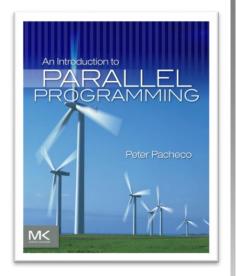


## An Introduction to Parallel Programming Peter Pacheco



#### Chapter 1

Why Parallel Computing?



### Roadmap

- Why we need ever-increasing performance.
- Why we're building parallel systems.
- Why we need to write parallel programs.
- How do we write parallel programs?
- What we'll be doing.
- Concurrent, parallel, distributed!



### **Changing times**

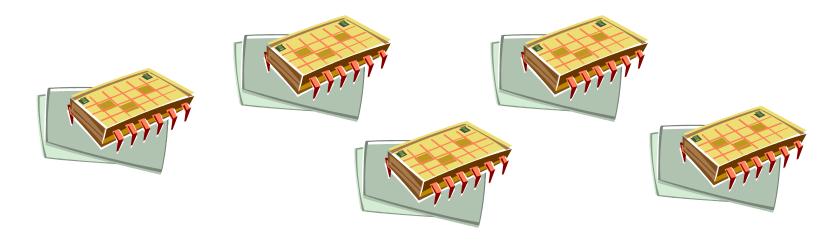
■ From 1986 – 2002, microprocessors were speeding like a rocket, increasing in performance an average of 50% per year.

Since then, it's dropped to about 20% increase per year.



## **An intelligent solution**

 Instead of designing and building faster microprocessors, put <u>multiple</u> processors on a single integrated circuit.





## Now it's up to the programmers

- Adding more processors doesn't help much if programmers aren't aware of them...
- or don't know how to use them.

 Serial programs don't benefit from this approach (in most cases).



## Why we need ever-increasing performance

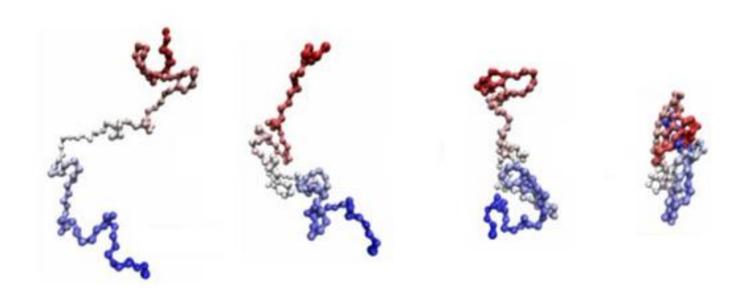
- Computational power is increasing, but so are our computation problems and needs.
- Problems we never dreamed of have been solved because of past increases, such as decoding the human genome.
- More complex problems are still waiting to be solved.

## **Climate modeling**





## **Protein folding**





## **Drug discovery**





## **Energy research**







## **Data analysis**





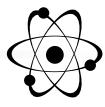
# Why we're building parallel systems

 Up to now, performance increases have been attributable to increasing density of transistors.

But there are inherent problems.



## A little physics lesson

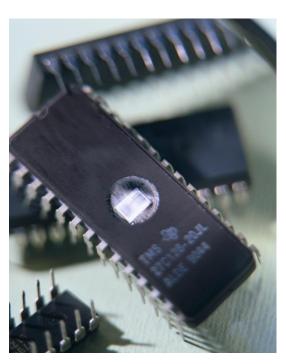


- Smaller transistors = faster processors.
- Faster processors = increased power consumption.
- Increased power consumption = increased heat.
- Increased heat = unreliable processors.



#### **Solution**

- Move away from single-core systems to multicore processors.
- "core" = central processing unit (CPU)



Introducing parallelism!!!



## Why we need to write parallel programs

- Running multiple instances of a serial program often isn't very useful.
- Think of running multiple instances of your favorite game.

What you really want is for it to run faster.



## Approaches to the serial problem

Rewrite serial programs so that they're parallel.

- Write translation programs that automatically convert serial programs into parallel programs.
  - This is very difficult to do.
  - Success has been limited.



### **More problems**

- Some coding constructs can be recognized by an automatic program generator, and converted to a parallel construct.
- However, it's likely that the result will be a very inefficient program.
- Sometimes the best parallel solution is to step back and devise an entirely new algorithm.



### **Example**

- Compute n values and add them together.
- Serial solution:

```
sum = 0;
for (i = 0; i < n; i++) {
    x = Compute_next_value(. . .);
    sum += x;
}</pre>
```



- We have p cores, p much smaller than n.
- Each core performs a partial sum of approximately n/p values.

```
my_sum = 0;
my_first_i = . . . ;
my_last_i = . . . ;
for (my_i = my_first_i; my_i < my_last_i; my_i++) {
    my_x = Compute_next_value( . . .);
    my_sum += my_x;
}</pre>
Each core uses it's own private varial
```

Each core uses it's own private variables and executes this block of code independently of the other cores.

After each core completes execution of the code, is a private variable my\_sum contains the sum of the values computed by its calls to Compute\_next\_value.

Ex., 8 cores, n = 24, then the calls to Compute\_next\_value return:

1,4,3, 9,2,8, 5,1,1, 5,2,7, 2,5,0, 4,1,8, 6,5,1, 2,3,9



 Once all the cores are done computing their private my\_sum, they form a global sum by sending results to a designated "master" core which adds the final result.



```
if (I'm the master core) {
    sum = my_x;
    for each core other than myself {
        receive value from core;
        sum += value;
    }
} else {
    send my_x to the master;
}
```



Core	0	1	2	3	4	5	6	7
my_sum	8	19	7	15	7	13	12	14

#### Global sum

$$8 + 19 + 7 + 15 + 7 + 13 + 12 + 14 = 95$$

Core	0	1	2	3	4	5	6	7
my_sum	95	19	7	15	7	13	12	14



**But wait!** 

There's a much better way to compute the global sum.





## Better parallel algorithm

- Don't make the master core do all the work.
- Share it among the other cores.
- Pair the cores so that core 0 adds its result with core 1's result.
- Core 2 adds its result with core 3's result, etc.
- Work with odd and even numbered pairs of cores.



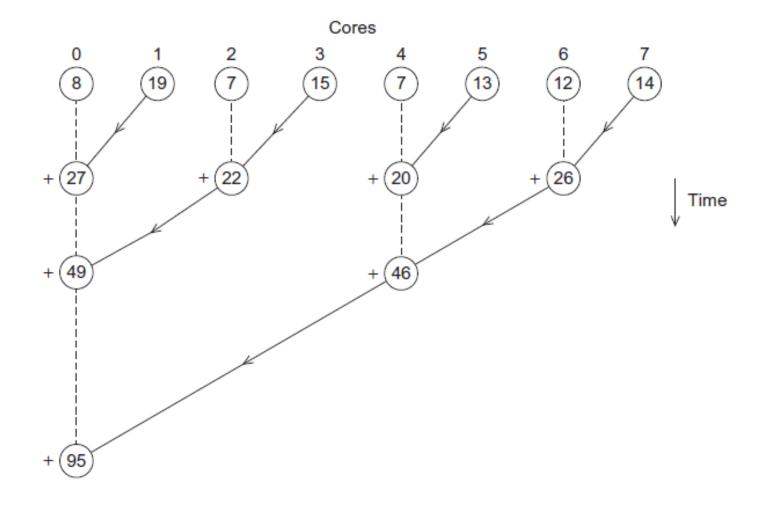
## Better parallel algorithm (cont.)

- Repeat the process now with only the evenly ranked cores.
- Core 0 adds result from core 2.
- Core 4 adds the result from core 6, etc.

Now cores divisible by 4 repeat the process, and so forth, until core 0 has the final result.



## Multiple cores forming a global sum





### **Analysis**

In the first example, the master core performs 7 receives and 7 additions.

In the second example, the master core performs 3 receives and 3 additions.

The improvement is more than a factor of 2!



## **Analysis (cont.)**

- The difference is more dramatic with a larger number of cores.
- If we have 1000 cores:
  - The first example would require the master to perform 999 receives and 999 additions.
  - The second example would only require 10 receives and 10 additions.
- That's an improvement of almost a factor of 100!



# How do we write parallel programs?

- Task parallelism
  - Partition various tasks carried out solving the problem among the cores.
- Data parallelism
  - Partition the data used in solving the problem among the cores.
  - Each core carries out similar operations on it's part of the data.



### **Professor P**

15 questions300 exams



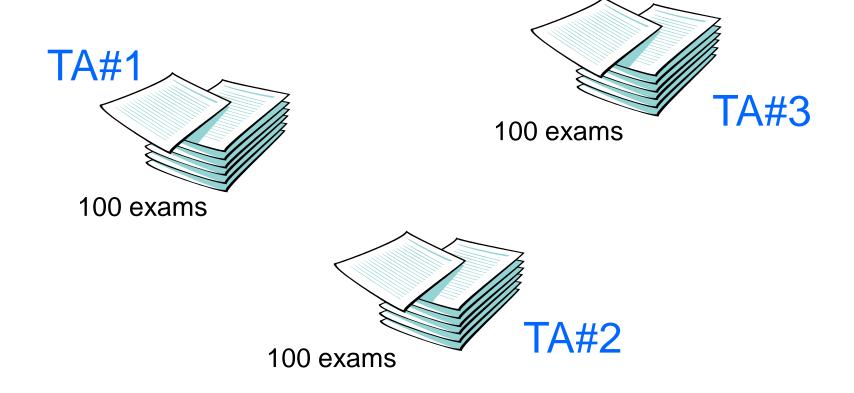


## Professor P's grading assistants





# Division of work – data parallelism





## Division of work – task parallelism

**TA#1** 



Questions 1 - 5



**TA#3** 

Questions 11 - 15



Questions 6 - 10

**TA#2** 



## Division of work – data parallelism

```
sum = 0;
for (i = 0; i < n; i++) {
    x = Compute_next_value(. . .);
    sum += x;
}</pre>
```



## Division of work – task parallelism

```
if (I'm the master core) {
    sum = my_x;
    for each core other than myself {
        receive value from core;
        sum += value;
    }
} else {
        send my_x to the master;
        1) Receiving
}
```



#### Coordination

- Cores usually need to coordinate their work.
- Communication one or more cores send their current partial sums to another core.
- Load balancing share the work evenly among the cores so that one is not heavily loaded.
- Synchronization because each core works at its own pace, make sure cores do not get too far ahead of the rest.



## What we'll be doing

- Learning to write programs that are explicitly parallel.
- Using the C language.
- Using three different extensions to C.
  - Message-Passing Interface (MPI)
  - Posix Threads (Pthreads)
  - OpenMP

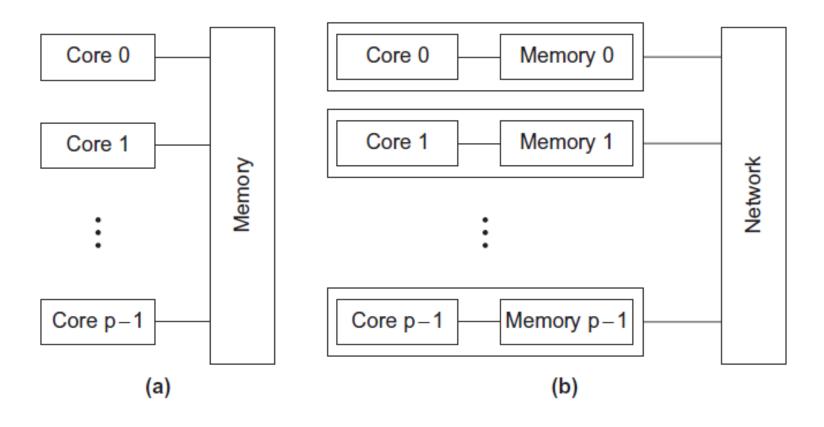


### Type of parallel systems

- Shared-memory
  - The cores can share access to the computer's memory.
  - Coordinate the cores by having them examine and update shared memory locations.
- Distributed-memory
  - Each core has its own, private memory.
  - The cores must communicate explicitly by sending messages across a network.



## Type of parallel systems



Shared-memory

Distributed-memory



## **Terminology**

- Concurrent computing a program is one in which multiple tasks can be in progress at any instant.
- Parallel computing a program is one in which multiple tasks <u>cooperate closely</u> to solve a problem
- Distributed computing a program may need to cooperate with other programs to solve a problem.



## **Concluding Remarks (1)**

- The laws of physics have brought us to the doorstep of multicore technology.
- Serial programs typically don't benefit from multiple cores.
- Automatic parallel program generation from serial program code isn't the most efficient approach to get high performance from multicore computers.



## **Concluding Remarks (2)**

- Learning to write parallel programs involves learning how to coordinate the cores.
- Parallel programs are usually very complex and therefore, require sound program techniques and development.

