# 计算机组成与系统结构 Computer Organization & System Architecture

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# **OpenMP**

- C extension: no new language to learn
- Multi-threaded, shared-memory parallelism
  - Compiler Directives, #pragma
  - Runtime Library Routines, #include <omp.h>

#### •#pragma

- Ignored by compilers unaware of OpenMP
- Same source for multiple architectures
  - E.g., same program for 1 & 16 cores
- Only works with shared memory

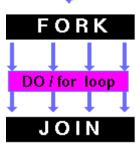
# OpenMP Parallel for pragma

```
#pragma omp parallel for
for (i=0; i<max; i++) zero[i] = 0;</pre>
```

 Master thread creates additional threads, each with a separate execution context

master

 All variables declared outside for loop are shared by default, except for loop index which is implicitly private per thread

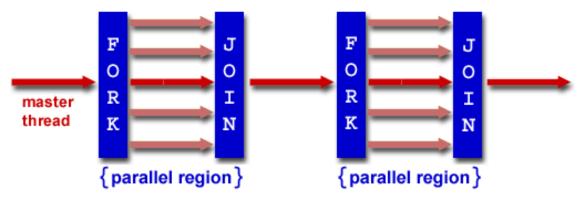


- Implicit "barrier" synchronization at end of for loop
- Divide index regions sequentially per thread
  - Thread 0 gets 0, 1, ..., (max/n)-1;
  - Thread 1 gets max/n, max/n+1, ..., 2\*(max/n)-1
  - Why?



# OpenMP Programming Model

Fork - Join Model:



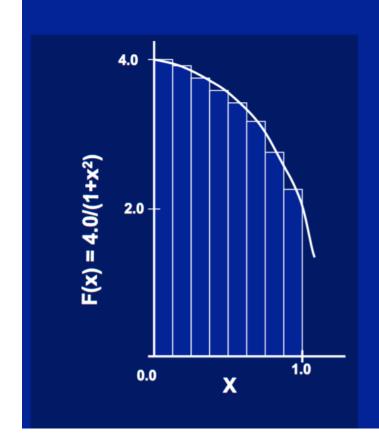
- OpenMP programs begin as single process (master thread)
  - Sequential execution
- When parallel region is encountered
  - Master thread "forks" into team of parallel threads
  - Executed simultaneously
  - At end of parallel region, parallel threads "join", leaving only master thread
- Process repeats for each parallel region
  - Amdahl's Law?

#### What Kind of Threads?

- OpenMP threads are operating system (software) threads
- OS will multiplex requested OpenMP threads onto available hardware threads
- Hopefully each gets a real hardware thread to run on, so no OS-level time-multiplexing
- But other tasks on machine compete for hardware threads!

# Example 2: Computing $\pi$

#### **Numerical Integration**



Mathematically, we know that:

$$\int_{0}^{1} \frac{4.0}{(1+x^2)} dx = \pi$$

We can approximate the integral as a sum of rectangles:

$$\sum_{i=0}^{N} F(x_i) \Delta x \approx \pi$$

Where each rectangle has width  $\Delta x$  and height  $F(x_i)$  at the middle of interval i.

http://openmp.org/mp-documents/omp-hands-on-SC08.pdf

#### Sequential $\pi$

```
#include <stdio.h>

void main () {
    const long num_steps = 10;
    double step = 1.0/((double)num_steps);
    double sum = 0.0;
    for (int i=0; i<num_steps; i++) {
        double x = (i+0.5) *step;
        sum += 4.0*step/(1.0+x*x);
    }
    printf ("pi = %6.12f\n", sum);
}</pre>
```

- pi = 3.142425985001
- Resembles  $\pi$ , but not very accurate
- Let's increase num\_steps and parallelize

# Parallelize (1) ...

```
#include <stdio.h>
void main () {
    const long num_steps = 10;
    double step = 1.0/((double)num_steps);
    double sum = 0.0;
#pragma parallel for
    for (int i=0; i<num_steps; i++) {</pre>
        double x = (i+0.5) *step;
        sum += 4.0*step/(1.0+x*x);

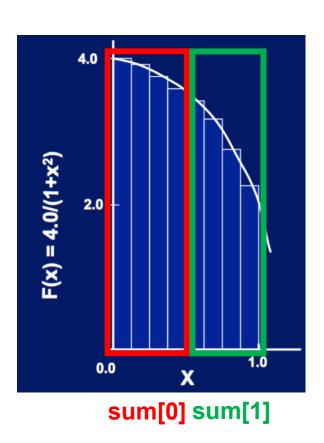
    Problem: each threads

    printf ("pi = %6.12f\n", sum);
                                        needs access to the
                                        shared variable sum
```

Code runs sequentially

. .

# Parallelize (2) ...



- Compute
  - sum[0] and sum[1]
- in parallel
- Compute
  - sum = sum[0] + sum[1]
- sequentially

#### Parallel $\pi$

```
#include <stdio.h>
#include <omp.h>
void main () {
    const int NUM_THREADS = 4;
    const long num steps = 10;
    double step = 1.0/((double)num_steps);
    double sum[NUM THREADS];
    for (int i=0; i<NUM_THREADS; i++) sum[i] = 0;</pre>
    omp_set_num_threads(NUM_THREADS);
#pragma omp parallel
        int id = omp_get_thread_num();
        for (int i=id; i<num_steps; i+=NUM_THREADS) {</pre>
            double x = (i+0.5) *step;
            sum[id] += 4.0*step/(1.0+x*x);
            printf("i =%3d, id =%3d\n", i, id);
    double pi = 0;
    for (int i=0; i<NUM_THREADS; i++) pi += sum[i];</pre>
    printf ("pi = %6.12f\n", pi);
```

#### **Trial Run**

```
i = 1, id = 1
i = 0, id = 0
i = 2, id = 2
i = 3, id = 3
i = 5, id = 1
i = 4, id = 0
i = 6, id = 2
i = 7, id = 3
i = 9, id = 1
i = 8, id = 0
pi = 3.142425985001
```

#### Parallel $\pi$

```
#include <stdio.h>
#include <omp.h>
                                              Scale up: num_steps = 10<sup>6</sup>
void main () {
    const int NUM THREADS = 4;
    const long num_steps = 1000000;
    double step = 1.0/((double)num_steps);
                                                       pi = 3.141592653590
    double sum[NUM THREADS];
    for (int i=0; i<NUM_THREADS; i++) sum[i] = 0;</pre>
    omp_set_num_threads(NUM_THREADS);
#pragma omp parallel
        int id = omp_get_thread_num();
        for (int i=id; i<num_steps; i+=NUM_THREADS) {</pre>
                                                       You verify how many
            double x = (i+0.5) *step;
            sum[id] += 4.0*step/(1.0+x*x);
                                                       digits are correct ...
            // printf("i =%3d, id =%3d\n", i, id);
    double pi = 0;
    for (int i=0; i<NUM_THREADS; i++) pi += sum[i];</pre>
    printf ("pi = %6.12f\n", pi);
```

# Can We Parallelize Computing sum?

```
#include <stdio.h>
#include <omp.h>
void main () {
    const int NUM THREADS = 1000;
    const long num_steps = 100000;
    double step = 1.0/((double)num_steps);
    double sum[NUM THREADS];
    for (int i=0; i<NUM_THREADS; i++) sum[i] = 0;</pre>
    double pi = 0;
    omp_set_num_threads(NUM_THREADS);
#pragma omp parallel
        int id = omp get thread num();
        for (int i=id; i<num_steps; i+=NUM_THREADS) {</pre>
            double x = (i+0.5) *step;
            sum[id] += 4.0*step/(1.0+x*x);
        pi += sum[id];
    printf ("pi = %6.12f\n", pi);
```

Always looking for ways to beat Amdahl's Law ...

# Summation inside parallel section

- Insignificant speedup in this example, but ...
- pi = 3.138450662641
- Wrong! And value changes between runs?!
- What's going on?

#### What Kind of Threads?

 What are the possible values of \*(x1) after executing this code by two concurrent threads?

Answer	*(x1)
RED	100 or 101
GREEN	101
ORANGE	101 or 102
	100 or 101 or 102

#### What's Going On?

```
#include <stdio.h>
#include <omp.h>
void main () {
    const int NUM THREADS = 1000;
    const long num_steps = 100000;
    double step = 1.0/((double)num_steps);
    double sum[NUM THREADS];
    for (int i=0; i<NUM_THREADS; i++) sum[i] = 0;</pre>
    double pi = 0;
    omp_set_num_threads(NUM_THREADS);
#pragma omp parallel
        int id = omp_get_thread_num();
        for (int i=id; i<num_steps; i+=NUM_THREADS) {</pre>
            double x = (i+0.5) *step;
            sum[id] += 4.0*step/(1.0+x*x);
        pi += sum[id];
    printf ("pi = %6.12f\n", pi);
```

- Operation is really
  pi = pi + sum[id]
- What if >1 threads reads current (same) value of pi, computes the sum, stores the result back to pi?
- Each processor reads same intermediate value of pi!
- Result depends on who gets there when
  - A "race" → result is <u>not</u>
     <u>deterministic</u>

#### OpenMP Reduction

```
double avg, sum=0.0, A[MAX]; int i;
#pragma omp parallel for private ( sum )
for (i = 0; i <= MAX ; i++)
    sum += A[i];
avg = sum/MAX; // bug</pre>
```

- Problem is that we really want sum over all threads!
- Reduction: specifies that 1 or more variables that are private to each thread are subject of reduction operation at end of parallel region:

#### reduction(operation:var) where

- Operation: operator to perform on the variables (var) at the end of the parallel region
- Var: One or more variables on which to perform scalar reduction.

```
double avg, sum=0.0, A[MAX]; int i;
#pragma omp for reduction(+ : sum)
for (i = 0; i <= MAX ; i++)
    sum += A[i];
avg = sum/MAX;</pre>
```

# Calculating π Original Version

```
#include <omp.h>
#define NUM THREADS 4
static long num steps = 100000; double step;
void main () {
  int i; double x, pi, sum[NUM THREADS];
  step = 1.0/(double) num steps;
  #pragma omp parallel private ( i, x )
    int id = omp get thread num();
    for (i=id, sum[id]=0.0; i<num steps; i=i+NUM THREADS)</pre>
      x = (i+0.5) *step;
      sum[id] += 4.0/(1.0+x*x);
  for(i=1; i<NUM THREADS; i++)</pre>
    sum[0] += sum[i]; pi = sum[0];
  printf ("pi = %6.12f\n", pi);
```

#### Version 2: parallel for, reduction

```
#include <omp.h>
#include <stdio.h>
/static long num steps = 100000;
double step;
void main ()
     int i; double x, pi, sum = 0.0;
     step = 1.0/(double) num_steps;
#pragma omp parallel for private(x) reduction(+:sum)
     for (i=1; i<= num steps; i++) {
          x = (i-0.5) *step;
          sum = sum + 4.0/(1.0+x*x);
     pi = sum;
   printf ("pi = %6.8f\n", pi);
```

# Synchronization

#### • Problem:

- Limit access to shared resource to 1 actor at a time
- E.g. only 1 person permitted to edit a file at a time
  - otherwise changes by several people get all mixed up

#### • Solution:



#### • Take turns:

- Only one person get's the microphone & talks at a time
- Also good practice for classrooms, btw ...

#### Data Races and Synchronization

- Two memory accesses form a *data race* if from different threads access same location, at least one is a write, and they occur one after another
- If there is a data race, result of program varies depending on chance (which thread first?)
- Avoid data races by synchronizing writing and reading to get deterministic behavior
- Synchronization done by user-level routines that rely on hardware synchronization instructions

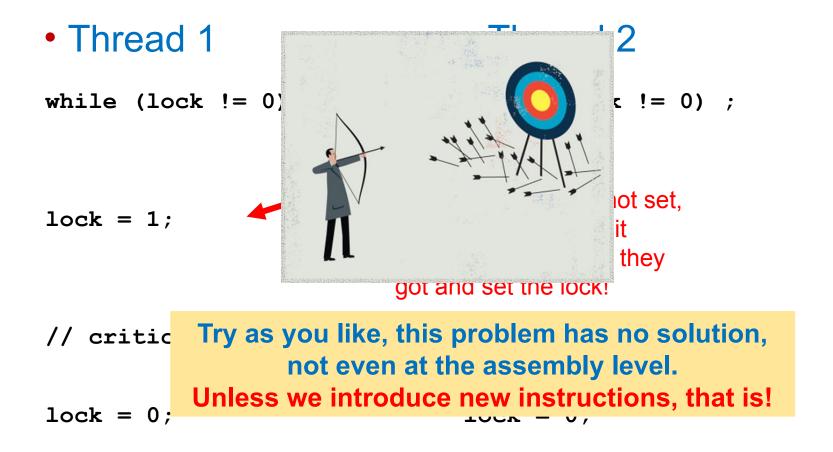
#### Locks

- Computers use locks to control access to shared resources
  - Serves purpose of microphone in example
  - Also referred to as "semaphore"
- Usually implemented with a variable
  - int lock;
    - 0 for unlocked
    - 1 for locked

#### Synchronization with Locks

```
// wait for lock released
while (lock != 0);
// lock == 0 now (unlocked)
// set lock
lock = 1;
     // access shared resource ...
     // e.g. pi
     // sequential execution! (Amdahl ...)
// release lock
lock = 0;
```

# Lock Synchronization



#### Hardware Synchronization

- Solution:
  - Atomic read/write
  - Read & write in single instruction
    - No other access permitted between read and write
  - Note:
    - Must use shared memory (<u>multiprocessing</u>)
- Common implementations:
  - Atomic swap of register ↔ memory
  - Pair of instructions for "linked" read and write
    - write fails if memory location has been "tampered" with after linked read
- RISCV has variations of both, but for simplicity we will focus on the former

# RISCV Atomic Memory Operations (AMOs)

- AMOs atomically perform an operation on an operand in memory and set the destination register to the original memory value
- R-Type Instruction Format: Add, And, Or, Swap, Xor,
   Max, Max Unsigned, Min, Min Unsigned

	31	27	26	25	24	20	19	15 14	12	11	7 6	0
	funct5		aq	rl	rs2		rs1	fur	ict3	rd	opcode	
•	5		1	1	5		5		3	5	7	
	operation	n	orde	ring	src		addr	wie	$_{ m lth}$	dest	AMO	

- Load from address in rs1 to "t"
- •rd = "t", i.e., the value in memory
- Store at address in rs1 the calculation "t" <operation> rs2
- aq and rl insure in order execution

```
amoadd.w rd,rs2,(rs1):
    t = M[x[rs1]];
    x[rd] = t;
    M[x[rs1]] = t + x[rs2]
```

#### RISCV Critical Section

- Assume that the lock is in memory location stored in register a0
- The lock is "set" if it is 1; it is "free" if it is 0 (it's initial value)

# Lock Synchronization

#### **Broken Synchronization**

```
while (lock != 0);
lock = 1;

// critical section
lock = 0;
```

#### Fix (lock is at location (a0))

```
li t0, 1
Try amoswap.w.aq t1, t0, (a0)
    bnez t1, Try
Locked:

# critical section

Unlock:
    amoswap.w.rl x0, x0, (a0)
```

#### **OpenMP Locks**

```
#include <stdio.h>
#include <stdlib.h>
#include <omp.h>
int main(void) {
    omp_lock_t lock;
    omp_init_lock(&lock);
#pragma omp parallel
        int id = omp_get_thread_num();
        // parallel section
       // ...
        omp_set_lock(&lock);
        // start sequential section
        // ...
        printf("id = %d\n", id);
        // end sequential section
        omp_unset_lock(&lock);
        // parallel section
       // ...
    omp_destroy_lock(&lock);
```

#### Synchronization in OpenMP

- Typically are used in libraries of higher level parallel programming constructs
- E.g. OpenMP offers \$pragmas for common cases:
  - critical
  - atomic
  - barrier
  - ordered
- OpenMP offers many more features
  - See online documentation
  - Or tutorial at
    - http://openmp.org/mp-documents/omp-hands-on-SC08.pdf

#### OpenMP Critical Section

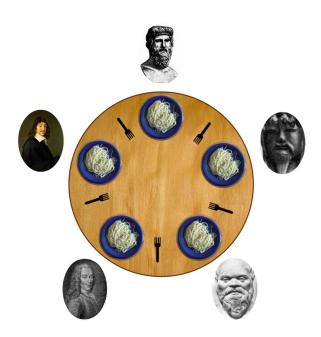
```
#include <stdio.h>
#include <omp.h>
void main () {
    const int NUM THREADS = 1000;
    const long num_steps = 100000;
    double step = 1.0/((double)num steps);
    double sum[NUM_THREADS];
    for (int i=0; i<NUM THREADS; i++) sum[i] = 0;</pre>
    double pi = 0:
    omp_set_num_threads(NUM_THREADS);
#pragma omp parallel
        int id = omp_get_thread_num();
        for (int i=id; i<num_steps; i+=NUM_THREADS) {</pre>
            double x = (i+0.5) *step;
            sum[id] += 4.0*step/(1.0+x*x);
#pragma omp critical
        pi += sum[id];
    printf ("pi = %6.12f\n", pi);
```

#### The Trouble with Locks ...

- ... is **dead-locks**
- Consider 2 cooks sharing a kitchen
  - Each cooks a meal that requires salt and pepper (locks)
  - Cook 1 grabs salt
  - Cook 2 grabs pepper
  - Cook 1 notices s/he needs pepper
    - it's not there, so s/he waits
  - Cook 2 realizes s/he needs salt
    - it's not there, so s/he waits
- A not so common cause of cook starvation
  - But deadlocks are possible in parallel programs
  - Very difficult to debug
    - malloc/free is easy ...

#### Deadlock

- Deadlock: a system state in which no progress is possible
- Dining Philosopher's Problem:
  - Think until the left fork is available; when it is, pick it up
  - Think until the right fork is available; when it is, pick it up
  - When both forks are held, eat for a fixed amount of time
  - Then, put the right fork down
  - Then, put the left fork down
  - Repeat from the beginning
- Solution?



# **OpenMP Timing**

Elapsed wall clock time:

```
double omp_get_wtime(void);
```

- Returns elapsed wall clock time in seconds
- Time is measured per thread, no guarantee can be made that two distinct threads measure the same time
- Time is measured from "some time in the past," so subtract results
  of two calls to omp get wtime to get elapsed time

# Matrix Multiply in OpenMP

```
// C[M][N] = A[M][P] \times B[P][N]
start time = omp get wtime();
#pragma omp parallel for private(tmp, j, k)
  for (i=0; i<M; i++) {
                                     Outer loop spread across N threads;
    for (j=0; j<N; j++) { ←
                                     inner loops inside a single thread
      tmp = 0.0;
      for (k=0; k<P; k++) {
         /* C(i,j) = sum(over k) A(i,k) * B(k,j)*/
         tmp += A[i][k] * B[k][j];
      C[i][j] = tmp;
run_time = omp_get_wtime() - start_time;
```

# Matrix Multiply in Open MP

- More performance optimizations available:
  - Higher compiler optimization (-O2, -O3) to reduce number of instructions executed
  - Cache blocking to improve memory performance
  - Using SIMD AVX instructions to raise floating point computation rate (DLP)

#### And, in Conclusion, ...

- Sequential software execution speed is limited
- Parallel processing is the only path to higher performance
  - SIMD: instruction level parallelism
    - Implemented in all high performance CPUs today (x86, ARM, ...)
    - Partially supported by compilers
  - MIMD: thread level parallelism
    - Multicore processors
    - Supported by Operating Systems (OS)
    - Requires programmer intervention to exploit at single program level
      - E.g. OpenMP
  - SIMD & MIMD for maximum performance
- Synchronization
  - Requires hardware support: specialized assembly instructions
  - Typically use higher-level support
  - Beware of deadlocks

# Peer Instruction: Why Multicore?

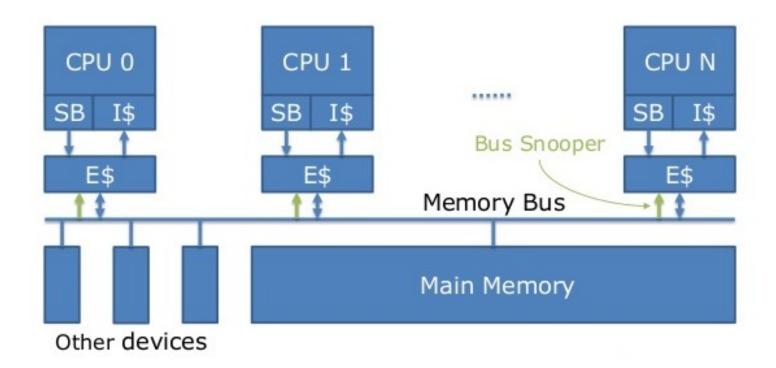
The switch in ~ 2005 from one processor per chip to multiple processors per chip happened because:

- I. The "power wall" meant that no longer get speed via higher clock rates and higher power per chip
- II. There was no other performance option but replacing one inefficient processor with multiple efficient processors
- III. OpenMP was a breakthrough in ~2000 that made parallel programming easy

RED	I only	
GREEN	II only	
ORANGE	I & II only	
YELLOW	I, II, & III onl	y

# (Chip) Multicore Multiprocessor

- SMP: (Shared Memory) Symmetric Multiprocessor
  - Two or more identical CPUs/Cores
  - Single shared coherent memory

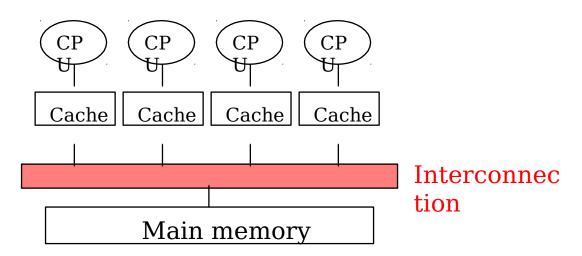


# Multiprocessor Key Questions

- Q1 How do they share data?
- Q2 How do they coordinate?
- Q3 How many processors can be supported?

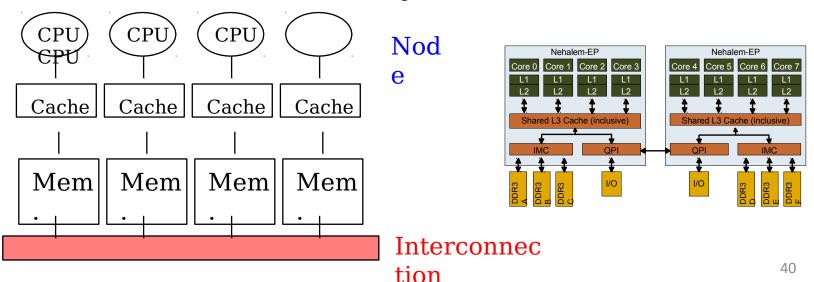
#### **Taxonomy of Parallel Computers**

- According to physical organization of processors and memory:
  - Physically centralized memory, uniform memory access (UMA)
    - All memory is allocated at same distance from all processors
    - Also called symmetric multiprocessors (SMP)
    - Memory bandwidth is fixed and must accommodate all processors → does not scale to large number of processors
    - Used in CMPs today (single-socket ones)



#### **Taxonomy of Parallel Computers**

- According to physical organization of processors and memory:
  - Physically distributed memory, non-uniform memory access (NUMA)
    - A portion of memory is allocated with each processor (node)
    - Accessing local memory is much faster than remote memory
    - If most accesses are to local memory than overall memory bandwidth increases linearly with the number of processors
    - Used in multi-socket CMPs E.g Intel Nehalem



#### **Taxonomy of Parallel Computers**

- According to memory communication model
  - Shared address or shared memory
    - Processes in different processors can use the same virtual address space
    - Any processor can directly access memory in another processor node
    - Communication is done through shared memory variables
    - Explicit synchronization with locks and critical sections
    - Arguably easier to program??
  - Distributed address or message passing
    - Processes in different processors use different virtual address spaces
    - Each processor can only directly access memory in its own node
    - Communication is done through explicit messages
    - Synchronization is implicit in the messages
    - Arguably harder to program??
    - Some standard message passing libraries (e.g., MPI)

# Shared Memory vs. Message Passing

Shared memory

```
Producer (p1)

flag = 0;

a = 10;

flag = 1;

Consumer (p2)

flag = 0;

while (!flag) {}

x = a * y;
```

Message passing

```
Producer (p1)

...

a = 10;

send(p2, a, label);

Consumer (p2)

...

receive(p1, b, label);

x = b * y;
```

# Shared Memory Multiprocessor (SMP)

- Q1 Single address space shared by all processors/cores
- Q2 Processors coordinate/communicate through shared variables in memory (via loads and stores)
  - Use of shared data must be coordinated via synchronization primitives (locks) that allow access to data to only one processor at a time
- All multicore computers today are SMP

#### Multiprocessor Caches

- Memory is a performance bottleneck even with one processor
- Use caches to reduce bandwidth demands on main memory
- Each core has a local private cache holding data it has accessed recently
- Only cache misses have to access the shared common memory

