

第12章 并发编程

线程级并行 Thread-Level Parallelism

100076202: 计算机系统导论



任课教师:

宿红毅 张艳 黎有琦 颜珂

原作者:

Randal E. **Bryant and** David R. O'Hallaron



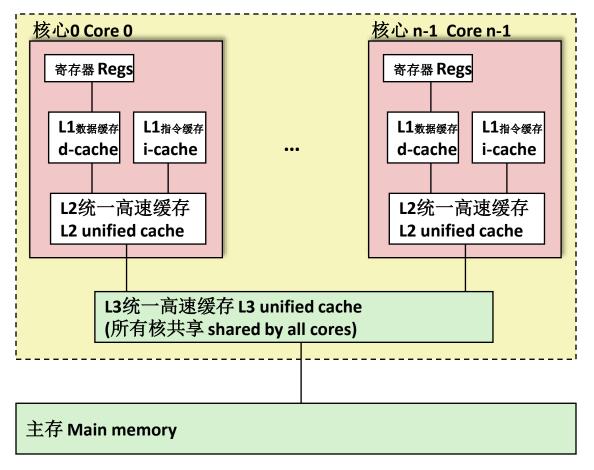
议题 Today

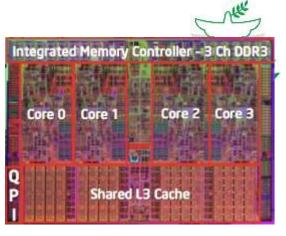


- 并行计算硬件 Parallel Computing Hardware
 - 多核 Multicore
 - 单个芯片上有多个独立处理器 Multiple separate processors on single chip
 - 超线程化 Hyperthreading
 - 在单核上高效执行多个线程 Efficient execution of multiple threads on single core
- 一致性模型 Consistency Models
 - 当多个线程读取和写入共享状态时会发生什么 What happens when multiple threads are reading & writing shared state
- 线程级并行 Thread-Level Parallelism
 - 将程序拆分为独立任务 Splitting program into independent tasks
 - 示例: 并行求和 Example: Parallel summation
 - 检查一些性能工件 Examine some performance artifacts
 - 分而治之 Divide-and conquer parallelism
 - 示例: 并行快速排序 Example: Parallel quicksort

典型的多核处理器

Typical Multicore Processor



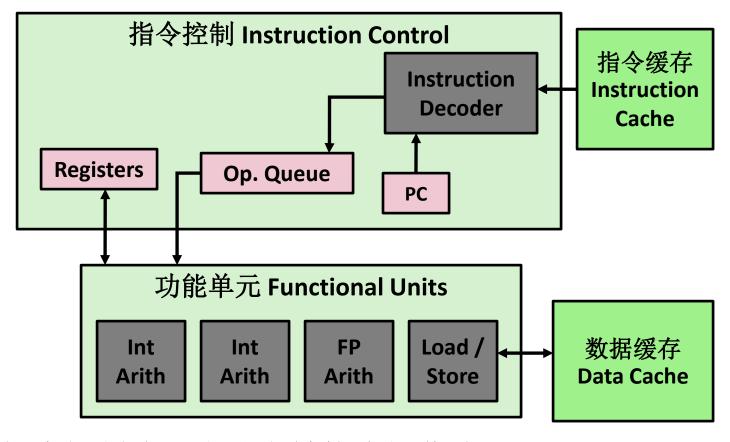


■ 多个处理器以一致的内存视图运行 Multiple processors operating with coherent view of memory

乱序处理器结构

Out-of-Order Processor Structure



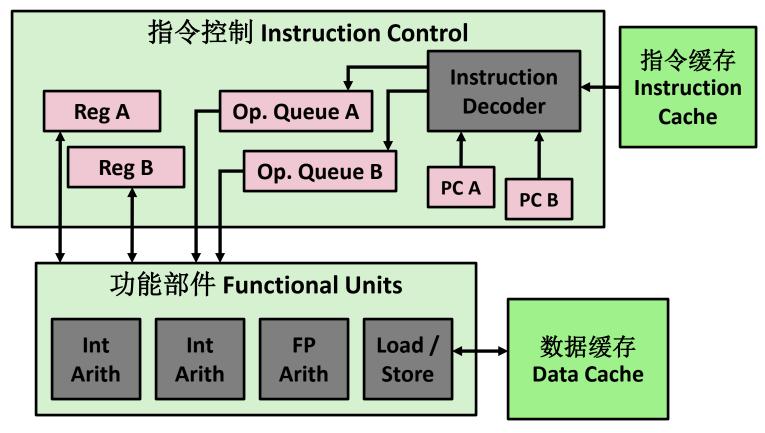


- 指令控制将程序动态转换为操作流 Instruction control dynamically converts program into stream of operations
- 操作映射到功能单元以并行方式执行 Operations mapped onto functional units to execute in parallel

超线程实现

J. Mark

Hyperthreading Implementation



- 复制指令控制以处理K个指令流 Replicate instruction control to process K instruction streams
- 所有寄存器有K份拷贝 K copies of all registers
- 共享功能单元 Share functional units



基准测试机 Benchmark Machine

- 从/proc/cpuinfo获取有关计算机的数据 Get data about machine from /proc/cpuinfo
- Shark机器 Shark Machines
 - Intel Xeon E5520 @ 2.27 GHz
 - Nehalem, ca. 2010
 - 8核 8 Cores
 - 每个核心可以执行2倍超线程 Each can do 2x hyperthreading

利用并行执行 Exploiting parallel execution

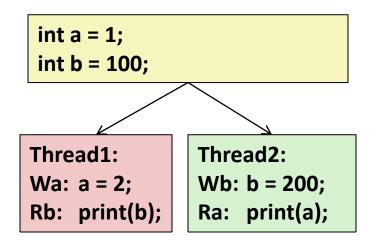
- 到目前为止,我们已经使用线程来处理I/O延迟 So far, we've used threads to deal with I/O delays
 - 例如每个客户端一个线程,以防止一个线程延迟另一个线程。e.g., one thread per client to prevent one from delaying another
- 多核CPU提供了另一个机会 Multi-core CPUs offer another opportunity
 - 在N个核心上并行执行的线程上扩展工作 Spread work over threads executing in parallel on N cores
 - 如果有许多独立任务,则自动发生 Happens automatically, if many independent tasks
 - 例如,运行许多应用程序或为许多客户端提供服务 e.g., running many applications or serving many clients
 - 还可以编写代码以加快一项大型任务的执行速度 Can also write code to make one big task go faster
 - 通过将其组织为多个并行子任务 by organizing it as multiple parallel sub-tasks

利用并行执行 Exploiting parallel execution

- Shark机器可以同时执行16个线程 Shark machines can execute 16 threads at once
 - 8核心,每个带2路超线程 8 cores, each with 2-way hyperthreading
 - 理论上16倍加速比 Theoretical speedup of 16X
 - 在我们的基准测试中从未达到 never achieved in our benchmarks

内存一致性 Memory Consistency





线程一致性约束 Thread consistency constraints Wa——→ Rb

Wb → Ra

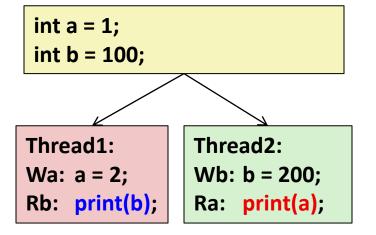
- 打印的可能值是什么? What are the possible values printed?
 - 取决于内存一致性模型 Depends on memory consistency model
 - 硬件如何处理并发访问的抽象模型 Abstract model of how hardware handles concurrent accesses

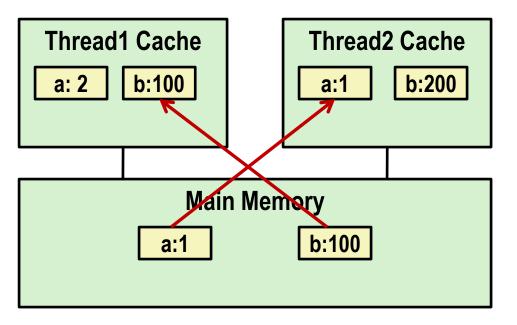
非一致性高速缓存方案

- The state of the

Non-Coherent Cache Scenario

■ 写回高速缓存,线程间没有 协作 Write-back caches, without coordination between them





print 1

print 100

稍后,a:2和b:200被写回主存储器 At later points, a:2 and b:200 are written back to main memory

Snoopy缓存 Snoopy Caches

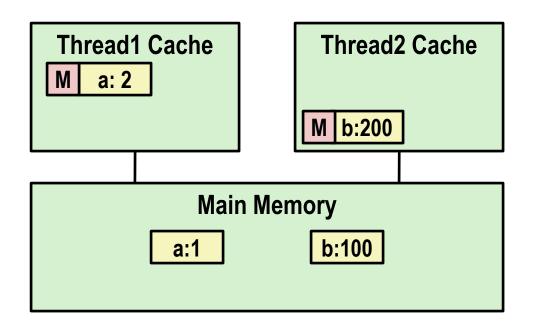


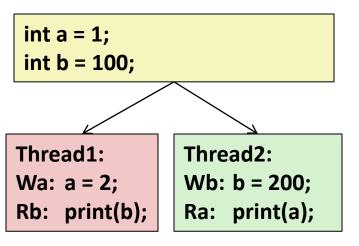
■ 用状态标记每个缓存块 Tag each cache block with state

无效 Invalid 不能使用其值 Cannot use value

共享 Shared 可读拷贝 Readable copy

修改 Modified 可写拷贝 Writeable copy





Snoopy缓存 **Snoopy Caches**

无效 Invalid

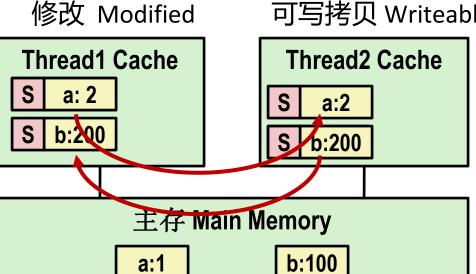
共享 Shared

- 用状态标记每个缓存块
- Tag each cache block with state

不能使用其值 Cannot use value

可读拷贝 Readable copy ■ 当缓存看到对其M标记

可写拷贝 Writeable copy



块之一的请求时 When cache sees request for one of its M-tagged

print 2

int a = 1;

Thread1:

Wa: a = 2;

Rb: print(b);

blocks

int b = 100;

print 200

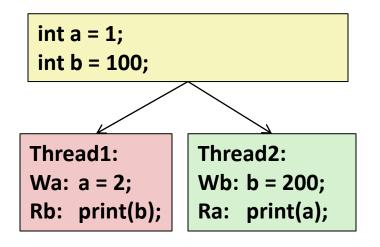
- 从缓存提供值(注 意:内存中的值可 能已过时) Supply value from cache (Note: value in memory may be stale)
- 将标记设置为S Set tag to S

Wb: b = 200;

Ra: print(a);

内存一致性 Memory Consistency

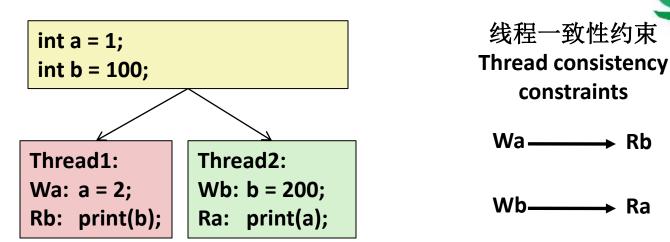




线程一致性约束
Thread consistency
constraints
Wa——→ Rb
Wb——→ Ra

- 打印的可能值是什么? What are the possible values printed?
 - 取决于内存一致性模型 Depends on memory consistency model
 - 硬件如何处理并发访问的抽象模型 Abstract model of how hardware handles concurrent accesses

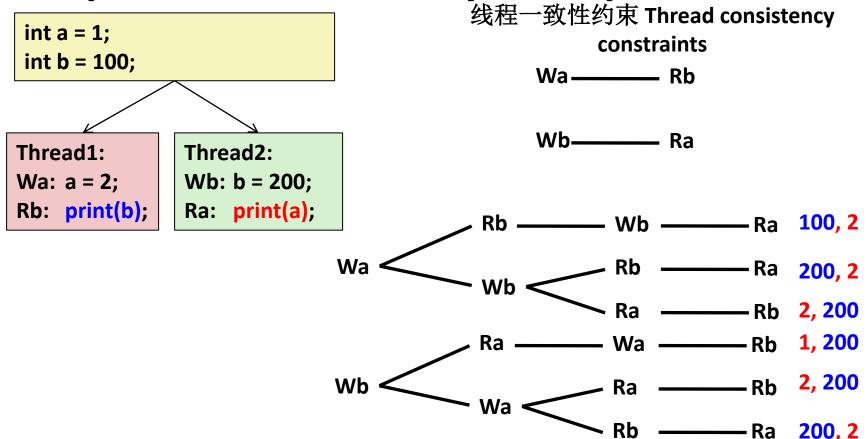
内存一致性 Memory Consistency



- 打印的可能值是什么? What are the possible values printed?
 - 取决于内存一致性模型 Depends on memory consistency model
 - 硬件如何处理并发访问的抽象模型 Abstract model of how hardware handles concurrent accesses
- 顺序一致性 Sequential consistency
 - 就好像一次只有一个操作一样,其顺序与每个线程内的操作顺序一致 As if only one operation at a time, in an order consistent with the order of operations within each thread
 - 因此,总体效果与每个单独的线程一致,但允许任意交错 Thus, overall effect consistent with each individual thread but otherwise allows an arbitrary interleaving

顺序一致性示例

Sequential Consistency Example



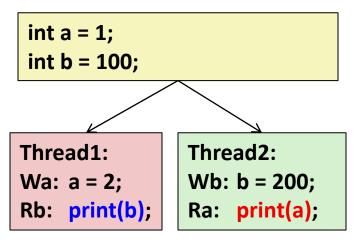
- 不可能输出 Impossible outputs
 - 100, 1 and 1, 100
 - 需要在Wa或Wb之前达到Ra和Rb Would require reaching both Ra and Rb before either Wa or Wb

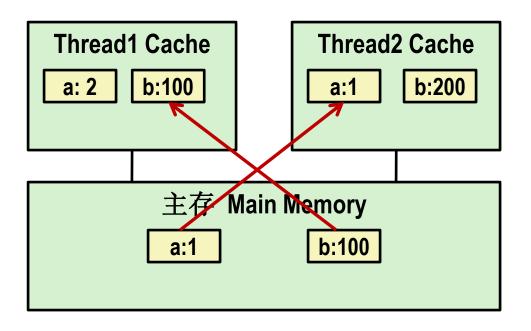
非一致性缓存方案

- Aller

Non-Coherent Cache Scenario

■ 写回缓存,线程间没有协作 Write-back caches, without coordination between them





print 1

print 100

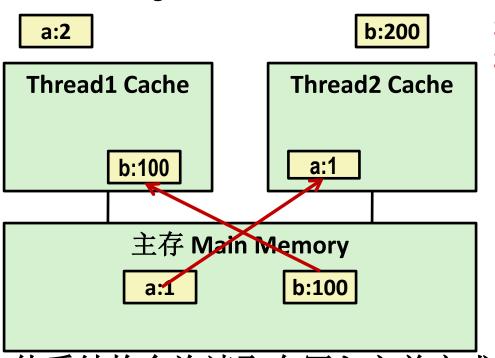
顺序一致性? 否! Sequentially consistent? No!

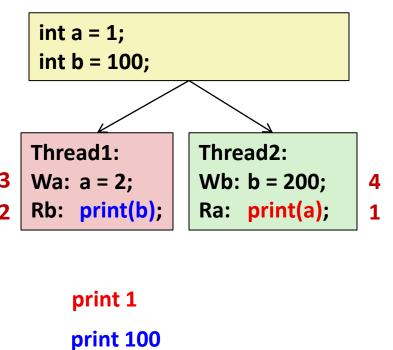
非顺序一致性方案

Non-Sequentially Consistent Scenario



■ 一致性缓存,但由于操作重新排序 而违反了线程一致性约束 Coherent caches, but thread consistency constraints violated due to operation reordering



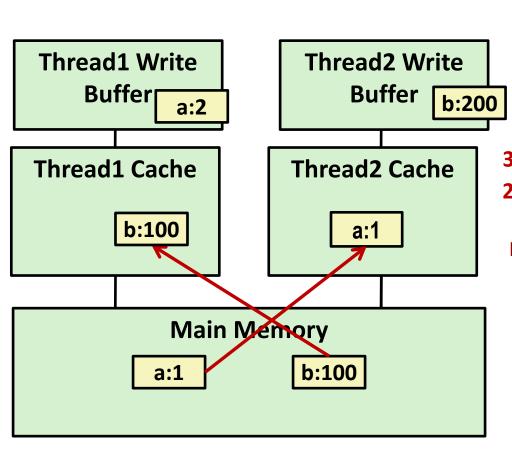


体系结构允许读取在写入之前完成,因为单个线程访问不同的内存位置 Architecture lets reads finish before writes because single thread accesses different memory locations

非顺序一致性方案

- ARK

Non-Sequentially Consistent Scenario



```
int a = 1;
int b = 100;

Thread1:
    Wa: a = 2;
    Rb: print(b);
    Ra: print(a);
```

■ 为什么重新排序?写入需要很长时间。缓冲区写入,让读取继续。*指令级并行性* Why Reordered? Writes take long time. Buffer write, let read go ahead.

Instruction-level parallelism

■ 修复: 在Wa&Rb和Wb&Ra之间添加SFENCE指令 Fix: Add SFENCE instructions between Wa & Rb and Wb & Ra



内存模型 Memory Models

- 顺序一致: Sequentially Consistent:
 - 每个线程以正确的顺序执行,任意交错 Each thread executes in proper order, any interleaving
- 为了确保,需要 To ensure, requires
 - 正确的缓存/内存行为 Proper cache/memory behavior
 - 适当的线程内排序约束 Proper intra-thread ordering constraints
- 线程排序约束 Thread ordering constraints
 - 使用同步确保程序没有数据竞争 Use synchronization to ensure the program is free of data races

议题 Today



- 并行计算硬件 Parallel Computing Hardware
 - 多核 Multicore
 - 单芯片上有多个独立的处理器 Multiple separate processors on single chip
 - 超线程化 Hyperthreading
 - 在单核上高效执行多个线程 Efficient execution of multiple threads on single core
- 一致性模型 Consistency Models
 - 当多个线程在读/写共享状态时会发生什么情况 What happens when multiple threads are reading & writing shared state
- 线程级并行 Thread-Level Parallelism
 - 将程序分成独立的任务 Splitting program into independent tasks
 - 例如: 并行求和 Example: Parallel summation
 - 检查一些性能小工件 Examine some performance artifacts
 - 分而治之 Divide-and conquer parallelism
 - 例如:并行快速排序 Example: Parallel quicksort

求和示例 Summation Example



- 求数字0, ..., N-1的和 Sum numbers 0, ..., N-1
 - 应该加起来得到(N-1)*N/2 Should add up to (N-1)*N/2
- 分区成K个区域 Partition into K ranges
 - 每个区域有LN/K」个值 LN/K」values each
 - t个线程每个处理一个区域 Each of the t threads processes 1 range
 - 连续累加剩余值 Accumulate leftover values serially
- 方法#1: 所有线程更新单个全局变量 Method #1: All threads update single global variable
 - 1A: 无同步 1A: No synchronization
 - 1B: 用pthread信号量同步 1B: Synchronize with pthread semaphore
 - 1C: 用pthread互斥锁同步 1C: Synchronize with pthread mutex
 - "二元"信号量,仅取值0和1 "Binary" semaphore. Only values 0 & 1

累积在单个全局变量中: 声明

Accumulating in Single Global Variable: Declarations

```
typedef unsigned long data t;
/* Single accumulator */
volatile data t global sum;
```

累积在单个全局变量中: 声明

Accumulating in Single Global Variable: Declarations

```
typedef unsigned long data t;
/* Single accumulator */
volatile data t global sum;
/* Mutex & semaphore for global sum */
sem t semaphore;
pthread mutex t mutex;
```

累积在单个全局变量中: 声明

Accumulating in Single Global Variable: Declarations

```
typedef unsigned long data t;
/* Single accumulator */
volatile data t global sum;
/* Mutex & semaphore for global sum */
sem t semaphore;
pthread mutex t mutex;
/* Number of elements summed by each thread */
size t nelems per thread;
/* Keep track of thread IDs */
pthread t tid[MAXTHREADS];
/* Identify each thread */
int myid[MAXTHREADS];
```

累积在单个全局变量中:操作

Accumulating in Single Global Variable: Operation

```
nelems per thread = nelems / nthreads;
/* Set global value */
                                                      线程例程
global sum = 0;
                                 线程ID Thread ID
                                                    Thread routine
/* Create threads and wait for them to finish *
for (i = 0; i < nthreads; 1++) {</pre>
   myid[i] = i;
   Pthread create(&tid[i], NULL, thread fun, &myid[i]);
for (i = 0; i < nthreads; i++)</pre>
                                             线程参数Thread argum
   Pthread join(tid[i], NULL);
                                                     (void *p)
result = global sum;
/* Add leftover elements */
for (e = nthreads * nelems per thread; e < nelems; e++)</pre>
    result += e;
```

线程函数: 无同步

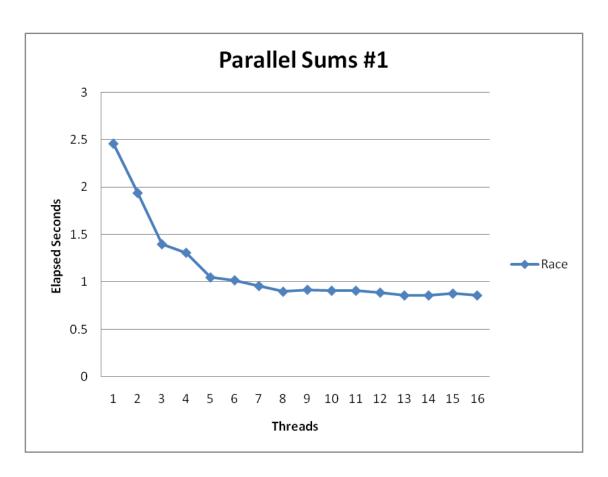


Thread Function: No Synchronization

```
void *sum_race(void *vargp)
{
    int myid = *((int *)vargp);
    size_t start = myid * nelems_per_thread;
    size_t end = start + nelems_per_thread;
    size_t i;

    for (i = start; i < end; i++) {
        global_sum += i;
    }
    return NULL;
}</pre>
```

无同步的性能 Unsynchronized Performance



- $N = 2^{30}$
- 最佳的加速比 Best speedup = 2.86X
- 当大于1个线程时得到错误的答案 Gets wrong answer when > 1 thread! 为何? Why?

线程函数: 信号量/互斥锁

S. Mark

Thread Function: Semaphore / Mutex

信号量 Semaphore

```
void *sum sem(void *varqp)
    int myid = *((int *)varqp);
    size t start = myid * nelems per thread;
    size t end = start + nelems per thread;
    size t i;
    for (i = start; i < end; i++) {
       sem wait(&semaphore);
       global sum += i;
       sem post(&semaphore);
    return NULL;
```

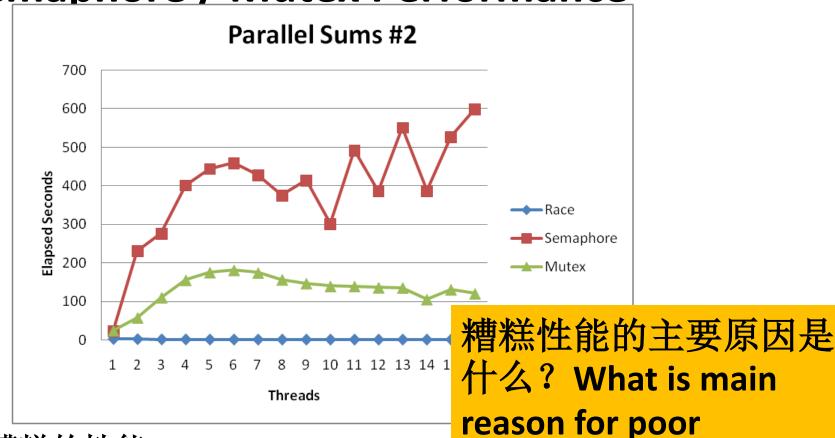
互斥锁 Mutex

```
pthread_mutex_lock(&mutex);
global_sum += i;
pthread_mutex_unlock(&mutex);
```

信号量/互斥锁性能



Semaphore / Mutex Performance



- 糟糕的性能 Terrible Performance
 - 2.5 seconds 秒 → ~10 minutes 分钟
- 互斥锁比信号量快3倍 Mutex 3X faster than semaphore

performance?

■ 很明显,这些方法都不成功 Clearly, neither is successful

单独累积 Separate Accumulation

- 方法#2:每个线程累积到单独的变量中 Method #2: Each thread accumulates into separate variable
 - 2A: 在相邻数组元素中累加 2A: Accumulate in contiguous array elements
 - 2B: 在间隔开的数组元素中累加 2B: Accumulate in spaced-apart array elements
 - 2C: 在寄存器中累加 2C: Accumulate in registers

```
/* Partial sum computed by each thread */
data_t psum[MAXTHREADS*MAXSPACING];

/* Spacing between accumulators */
size_t spacing = 1;
```

单独累积:操作

THE THE PERSON OF THE PERSON O

Separate Accumulation: Operation

```
nelems per thread = nelems / nthreads;
/* Create threads and wait for them to finish */
for (i = 0; i < nthreads; i++) {</pre>
   myid[i] = i;
   psum[i*spacing] = 0;
   Pthread create(&tid[i], NULL, thread fun, &myid[i]);
for (i = 0; i < nthreads; i++)</pre>
   Pthread join(tid[i], NULL);
result = 0;
/* Add up the partial sums computed by each thread */
for (i = 0; i < nthreads; i++)</pre>
   result += psum[i*spacing];
/* Add leftover elements */
for (e = nthreads * nelems per thread; e < nelems; e++)</pre>
    result += e;
```

线程函数: 内存累积

Thread Function: Memory Accumulation

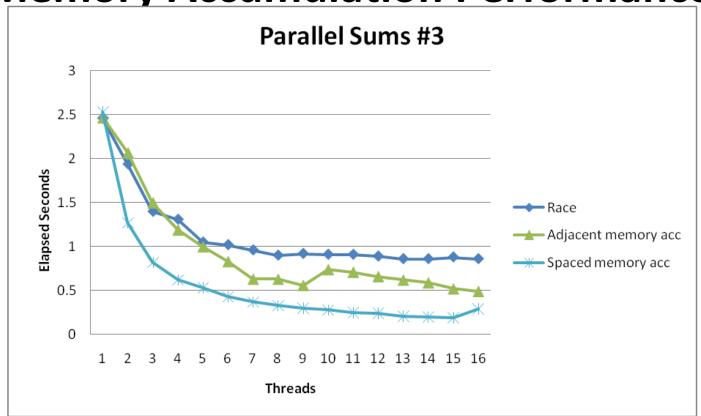
互斥锁在哪? Where is the mutex?

```
void *sum global(void *vargp)
    int myid = *((int *)varqp);
    size t start = myid * nelems per thread;
    size t end = start + nelems per thread;
    size t i;
    size t index = myid*spacing;
    psum[index] = 0;
    for (i = start; i < end; i++) {</pre>
       psum[index] += i;
    return NULL;
```

内存累积性能



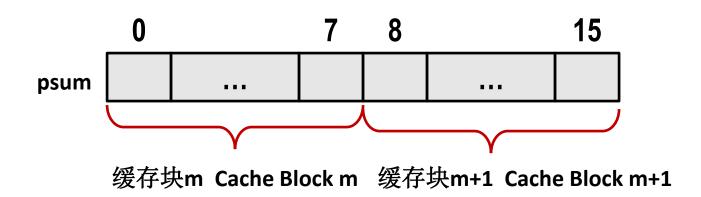
Memory Accumulation Performance



- 单独线程累积的优势 Clear threading advantage
 - 连续累积加速比: Adjacent speedup: 5 X
 - 间隔累积加速比: Spaced-apart speedup: 13.3 X (仅观察到加速比大于8 Only observed speedup > 8)
- 为什么进行间隔开累加性能更佳? Why does spacing the accumulators apart matter?



虚假共享 False Sharing

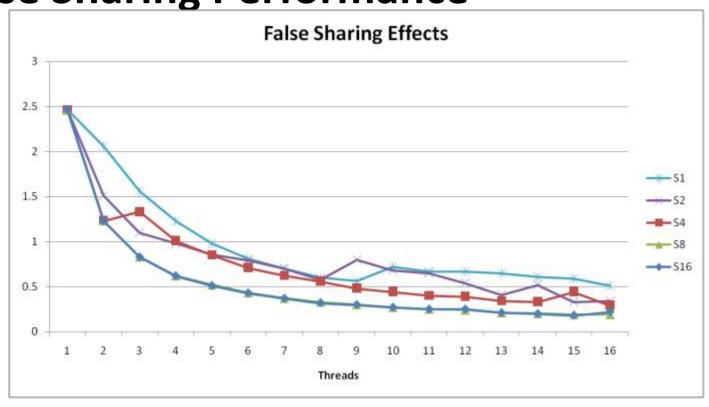


- 缓存块上保持一致性 Coherence maintained on cache blocks
- 要更新psum[i],线程i必须具有独占访问权限 To update psum[i], thread i must have exclusive access
 - 共享公共缓存块的线程将继续为访问块而相互争斗 Threads sharing common cache block will keep fighting each other for access to block

虚假共享的性能

False Sharing Performance





- 最佳间隔性能比最佳相邻性能高2.8倍 Best spaced-apart performance 2.8 X better than best adjacent
- 演示缓存块大小为64 Demonstrates cache block size = 64
 - 8字节值 8-byte values
 - 将间隔增加到8以上没有性能改善 No benefit increasing spacing beyond 8

线程函数: 寄存器累积

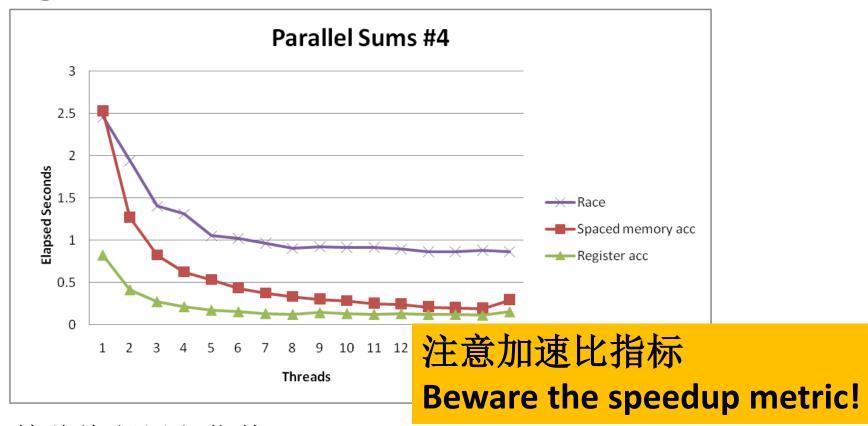
Thread Function: Register Accumulation

```
void *sum local(void *vargp)
    int myid = *((int *)varqp);
    size t start = myid * nelems per thread;
    size t end = start + nelems per thread;
    size t i;
    size t index = myid*spacing;
    data t sum = 0;
    for (i = start; i < end; i++) {</pre>
       sum += i;
    psum[index] = sum;
    return NULL;
```

寄存器累积性能

THE THE PERSON NAMED IN TH

Register Accumulation Performance



- 单独线程累积优势 Clear threading advantage
 - 加速比/Speedup = 7.5 X
- 比最快的内存累积好2倍 2X better than fastest memory accumulation



经验教训 Lessons learned

- 共享内存可能开销很高 Sharing memory can be expensive
 - 关注真实共享 Pay attention to true sharing
 - 注意虚假共享 Pay attention to false sharing
- 尽可能使用寄存器 Use registers whenever possible
 - (记住cachelab Remember cachelab)
 - 尽可能使用本地缓存 Use local cache whenever possible
- 处理剩余的数据 Deal with leftovers
- 在检查性能时,与最佳顺序实现进行比较 When examining performance, compare to best possible sequential implementation

更重要的示例:排序

The state of the s

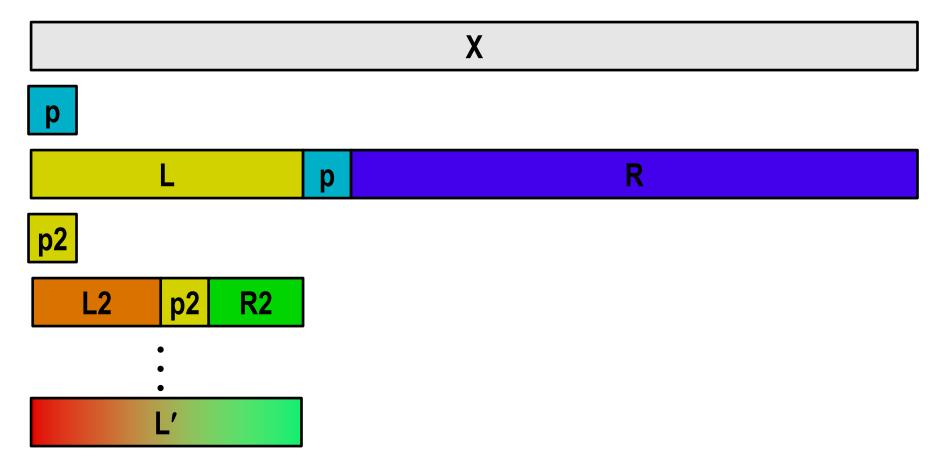
A More Substantial Example: Sort

- N个随机数集合排序 Sort set of N random numbers
- 多种可能的算法 Multiple possible algorithms
 - 使用并行版本的快速排序 Use parallel version of quicksort
- 对X集合进行顺序快速排序 Sequential quicksort of set of values X
 - 从X选择"中心点"p Choose "pivot" p from X
 - 重新排列X Rearrange X into
 - 左边集合: 值小于等于p L: Values ≤ p
 - 右边集合: 值大于等于p R: Values ≥ p
 - 对左边集合进行递归排序得到L' Recursively sort L to get L'
 - 对右边集合进行递归排序得到R′ Recursively sort R to get R′
 - 返回 Return L':p:R'

顺序快速排序可视化



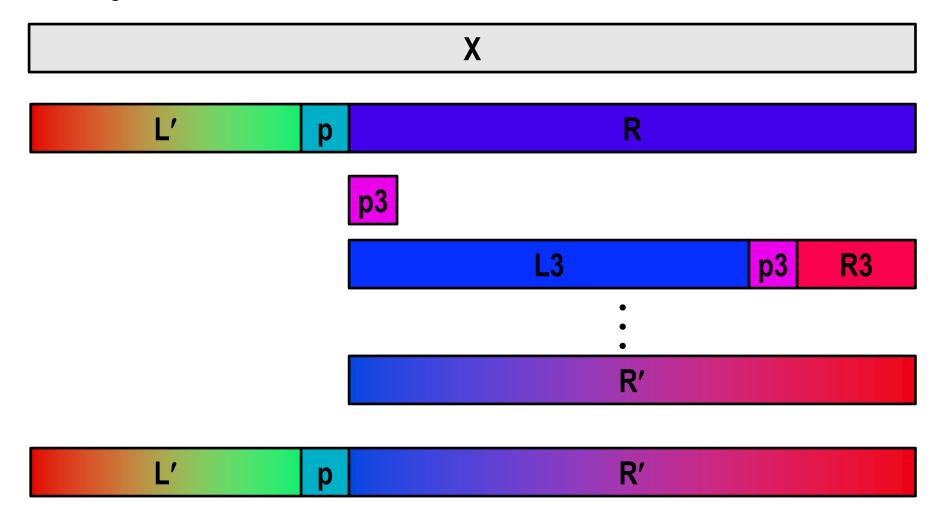
Sequential Quicksort Visualized



顺序快速排序可视化



Sequential Quicksort Visualized



顺序快速排序代码

New York

Sequential Quicksort Code

```
void qsort serial(data t *base, size t nele) {
  if (nele <= 1)</pre>
    return;
  if (nele == 2) {
    if (base[0] > base[1])
      swap(base, base+1);
    return;
  }
  /* Partition returns index of pivot */
  size t m = partition(base, nele);
  if (m > 1)
   qsort serial(base, m);
  if (nele-1 > m+1)
    qsort serial(base+m+1, nele-m-1);
```

- 从base开始对nele个元素排序 Sort nele elements starting at base
 - 如果有多于一个元素,则递归排序L或R Recursively sort L or R if has more than one element

并行快速排序 Parallel Quicksort



- 集合X的并行快速排序 Parallel quicksort of set of values X
 - 如果N小于等于Nthresh, 执行顺序快速排序 If N ≤ Nthresh, do sequential quicksort
 - 否则 Else
 - 从X选择"中心点"p Choose "pivot" p from X
 - 重新排列X Rearrange X into

– 左集合:值小于等于p L: Values ≤ p

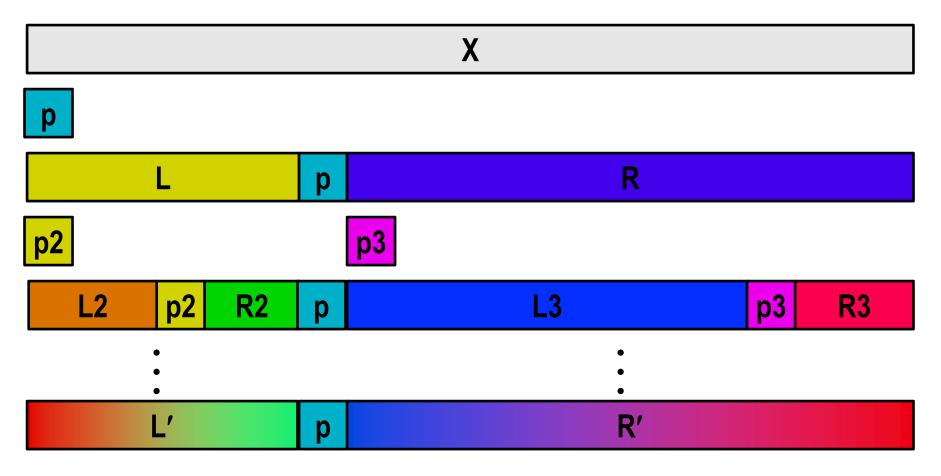
– 右集合:值大于等于p R: Values ≥ p

- 递归生成单独的线程 Recursively spawn separate threads
 - 排序L以获得L' Sort L to get L'
 - 排序R以获得R' Sort R to get R'
- 返回 Return L':p:R'

并行快速排序可视化

The state of the s

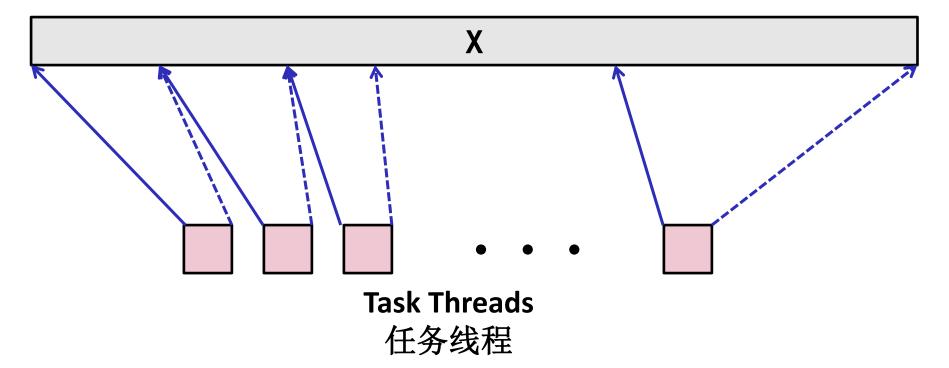
Parallel Quicksort Visualized



线程结构: 排序任务

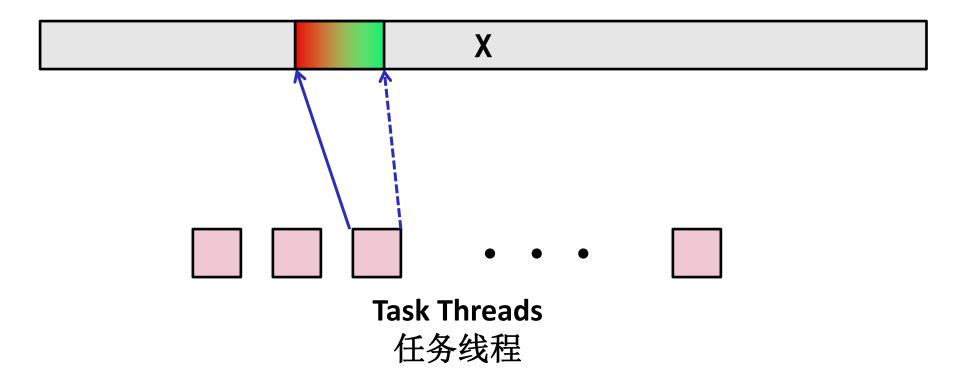
New York

Thread Structure: Sorting Tasks



- 任务: 排序子范围数据 Task: Sort subrange of data
 - 指定为: Specify as:
 - base: 起始地址 **base**: Starting address
 - nele: 子范围中的元素数 **nele**: Number of elements in subrange
- 作为单独线程运行 Run as separate thread

小排序任务操作 Small Sort Task Operation

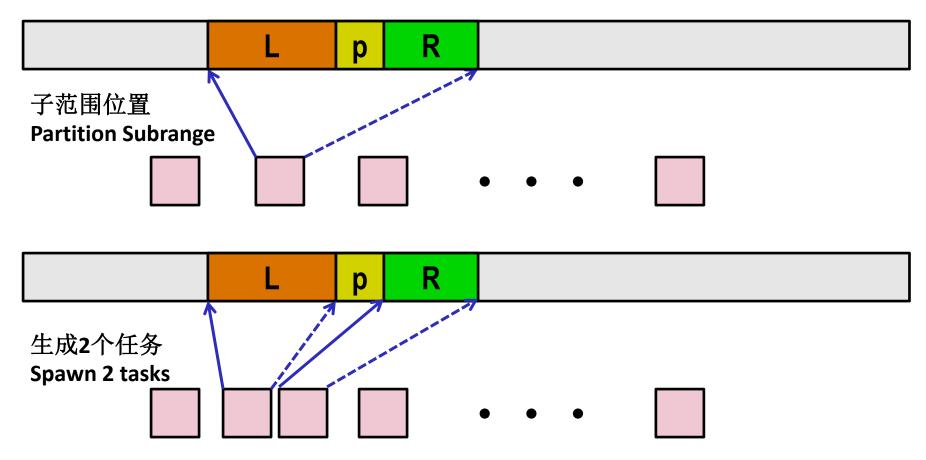


■ 排序子范围数据使用串行快速排序 Sort subrange using serial quicksort

大排序任务操作

- ARK

Large Sort Task Operation



顶层函数(简化)

The state of the s

Top-Level Function (Simplified)

```
void tqsort(data_t *base, size_t nele) {
   init_task(nele);
   global_base = base;
   global_end = global_base + nele - 1;
   task_queue_ptr tq = new_task_queue();
   tqsort_helper(base, nele, tq);
   join_tasks(tq);
   free_task_queue(tq);
}
```

- 初始化数据结构 Sets up data structures
- 调用递归排序例程 Calls recursive sort routine
- 保持加入线程,直到没有剩余 Keeps joining threads until none left
- 释放数据结构 Frees data structures

递归排序例程(简化)

- Aller

Recursive sort routine (Simplified)

- 小分区:按顺序排序 Small partition: Sort serially
- 大分区: 生成新的排序任务 Large partition: Spawn new sort task

排序任务线程(简化)

Sort task thread (Simplified)



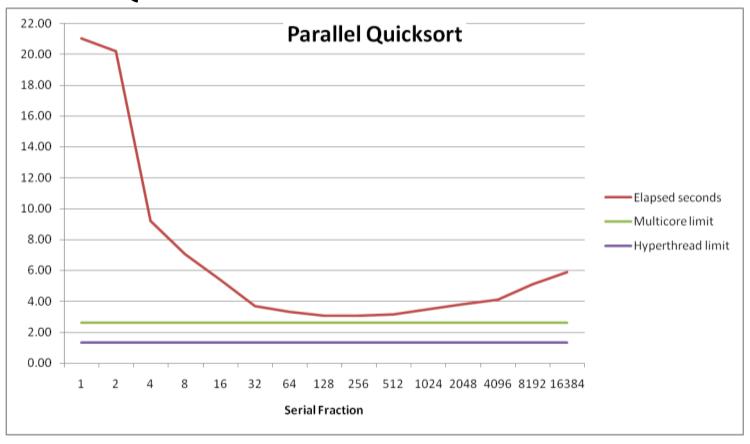
```
/* Thread routine for many-threaded quicksort */
static void *sort thread(void *vargp) {
    sort task t *t = (sort task t *) vargp;
   data t *base = t->base;
    size t nele = t->nele;
    task queue ptr tq = t->tq;
    free (varqp);
    size t m = partition(base, nele);
    if (m > 1)
        tqsort helper(base, m, tq);
    if (nele-1 > m+1)
        tqsort helper(base+m+1, nele-m-1, tq);
    return NULL;
```

- 获取任务参数 Get task parameters
- 执行分区步骤 Perform partitioning step
- 在每个分区上调用递归排序例程(如果部分大小大于1) Call recursive sort routine on each partition (if size of part > 1)

并行快速排序性能

Parallel Quicksort Performance



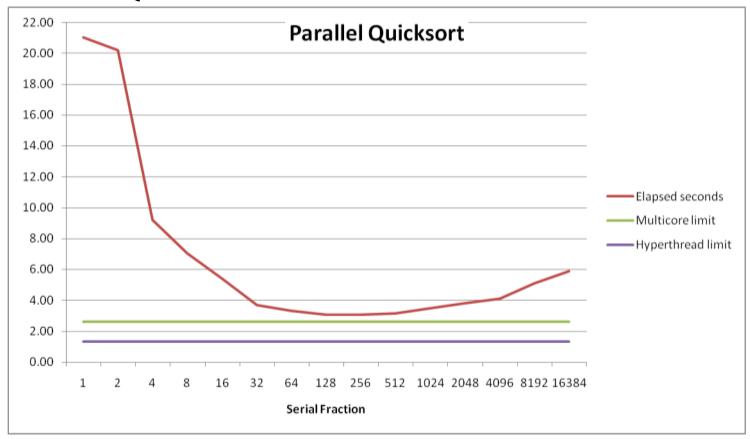


- 串行比例: 进行串行排序的输入比例 Serial fraction: Fraction of input at which do serial sort
- 排序128M随机值 Sort 2²⁷ (134,217,728) random values
- 最佳加速比 Best speedup = 6.84X

并行快速排序性能

Parallel Quicksort Performance





- 在广泛的串行比例范围内表现良好 Good performance over wide range of fraction values
 - F太小: 并行度不够 F too small: Not enough parallelism
 - F太大: 线程开销太高 F too large: Thread overhead too high

阿姆达尔定律(旅行模拟)

Amdahl's Law (Travel Analogy)



- 从PIT直飞LHR Flying jet non-stop from PIT -> LHR: 7.5 Hours
- 或者,老式SST方式: Or, old fashioned SST way:
 - Fly jet from PIT -> JFK: 1.5 Hours
 - Fly SST from JFK -> LHR: 3.5 Hours

5 Hours

1.5x

- 或者,使用FTL Or, Using FTL:
 - Fly jet from PIT -> JFK: 1.5 Hours
 - Fly FTL from JFK -> LHR: .01 Hours

1.51 Hours

~5x

- 最好的加速比是5倍,即使是FTL,因为必须到达纽约 Best possible speed up is 5X, even with FTL because have to get to New York.
- PIT: 匹兹堡 LHR: 伦敦 JFK: 纽约
- SST: 超音速客机 FTL: 超光速

New York

阿姆达尔定律 Amdahl's Law

- 总体问题 Overall problem
 - T所需的总顺序执行时间 T Total sequential time required
 - p可加速的总比例 p Fraction of total that can be sped up (0 ≤ p ≤ 1)
 - k加速系数 k Speedup factor
- 最终性能 Resulting Performance
 - $T_k = pT/k + (1-p)T$
 - 可以加速的部分速度快k倍 Portion which can be sped up runs k times faster
 - 无法加速的部分保持不变 Portion which cannot be sped up stays the same
 - 最大可能加速比 Maximum possible speedup
 - $k = \infty$
 - $T_{\infty} = (1-p)T$

阿姆达尔定律(旅行模拟)

Amdahl's Law (Travel Analogy)



- 从PIT直飞LHR Flying jet non-stop from PIT -> LHR: **7.5 Hours**
- 或者,老式SST方式: Or, old fashioned SST way:
 - Fly jet from PIT -> JFK: 1.5 Hours
 - Fly SST from JFK -> LHR: 3.5 Hours

5 Hours

1.5x

- 或者,使用FTL Or, Using FTL:
 - Fly jet from PIT -> JFK: 1.5 Hours
 - Fly FTL from JFK -> LHR: .01 Hours

1.51 Hours

~5x

- 最好的加速比是5倍,即使是FTL,因为必须到达纽约 Best possible speed up is 5X, even with FTL because have to get to New York.
 - T=7.5, p=6/7.5=.8, k= $\infty \implies T_{\infty} = (1-p)T=1.5$ max speed-up =5x

最大加速比

阿姆达尔定律的示例

Amdahl's Law Example



- 总体问题 Overall problem
 - T = 10 Total time required 所需总时间
 - p = 0.9 Fraction of total which can be sped up 可加速的总比例
 - k = 9 Speedup factor 加速系数
- 最终性能 Resulting Performance
 - T₉ = 0.9 * 10/9 + 0.1 * 10 = 1.0 + 1.0 = 2.0 (5倍加速比 a 5x speedup)
- 最大可能加速比 Maximum possible speedup
 - T_∞ = 0.1 * 10.0 = 1.0 (10倍加速比 a 10x speedup)
 - 拥有无限的并行计算资源! With infinite parallel computing resources!
 - 极限加速比显示**算法**极限 Limit speedup shows **algorithmic** limitation

阿姆达尔定律和并行快速排序

THE THE PERSON OF THE PERSON O

Amdahl's Law & Parallel Quicksort

- 顺序程序瓶颈 Sequential bottleneck
 - 顶层分区: 无加速 Top-level partition: No speedup
 - 第二级:小于等于2倍加速比 Second level: ≤ 2X speedup
 - 第k级: 小于等于2^{k-1}加速比 kth level: ≤ 2^{k-1}X speedup
- 启示 Implications
 - 小规模并行的良好性能 Good performance for small-scale parallelism
 - 需要并行化分区步骤以获得大规模并行性 Would need to parallelize partitioning step to get large-scale parallelism
 - 基于规则抽样的并行排序 Parallel Sorting by Regular Sampling
 - "并行与分布式计算" H. Shi & J. Schaeffer, J. Parallel & Distributed Computing, 1992

经验教训 Lessons Learned

- 必须具有并行化策略 Must have parallelization strategy
 - 划分为K个独立部分 Partition into K independent parts
 - 分而治之 Divide-and-conquer
- 内部循环必须无同步 Inner loops must be synchronization free
 - 同步操作非常耗时 Synchronization operations very expensive
- 当心硬件瑕疵 Watch out for hardware artifacts
 - 需要了解处理器和内存结构 Need to understand processor & memory structure
 - 共享和虚假共享全局数据 Sharing and false sharing of global data
- 当心阿姆达尔定律 Beware of Amdahl's Law
 - 串行代码可能成为瓶颈 Serial code can become bottleneck
- 你能行! You can do it!
 - 实现适度的并行性并不困难 Achieving modest levels of parallelism is not difficult
 - 建立实验框架并测试多种策略 Set up experimental framework and test multiple strategies