Thread-Level Parallelism

15-213: Introduction to Computer Systems

26th Lecture, Nov. 30, 2010

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Today

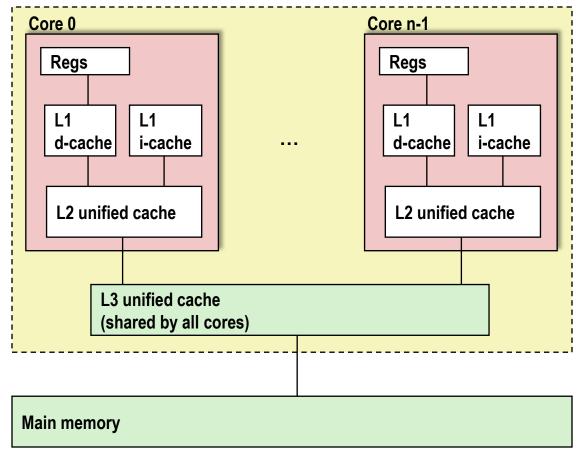
Parallel Computing Hardware

- Multicore
 - Multiple separate processors on single chip
- Hyperthreading
 - Replicated instruction execution hardware in each processor
- Maintaining cache consistency

Thread Level Parallelism

- Splitting program into independent tasks
 - Example: Parallel summation
 - Some performance artifacts
- Divide-and conquer parallelism
 - Example: Parallel quicksort

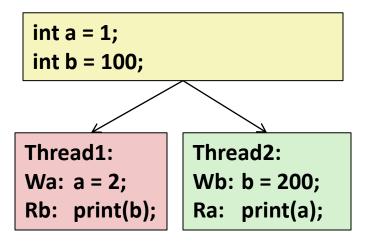
Multicore Processor



Intel Nehalem Processor

- E.g., Shark machines
- Multiple processors operating with coherent view of memory

Memory Consistency



Thread consistency constraints
Wa → Rb

Wb——→ Ra

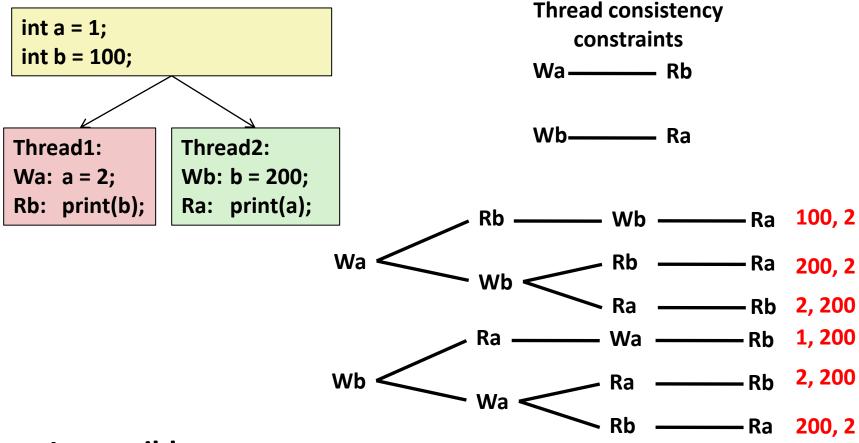
What are the possible values printed?

- Depends on memory consistency model
- Abstract model of how hardware handles concurrent accesses

Sequential consistency

- Overall effect consistent with each individual thread
- Otherwise, arbitrary interleaving

Sequential Consistency Example

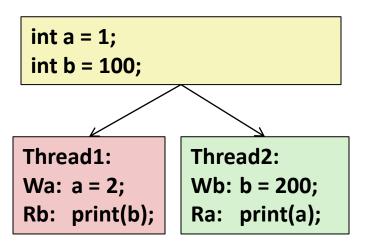


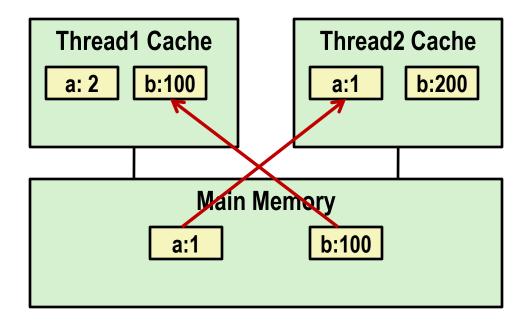
Impossible outputs

- 100, 1 and 1, 100
- Would require reaching both Ra and Rb before Wa and Wb

Non-Coherent Cache Scenario

Write-back caches, without coordination between them





print 1

print 100

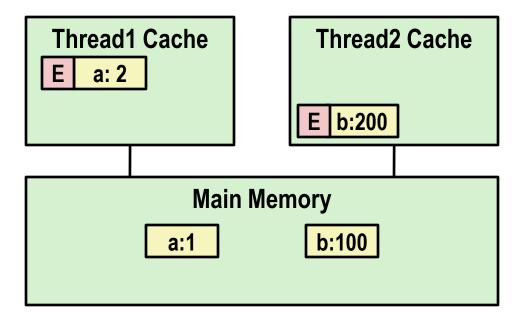
Snoopy Caches

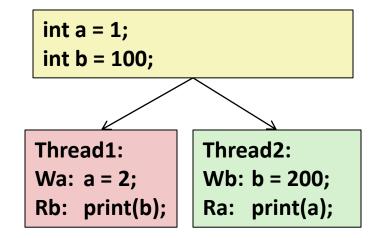
Tag each cache block with state

Invalid Cannot use value

Shared Readable copy

Exclusive Writeable copy





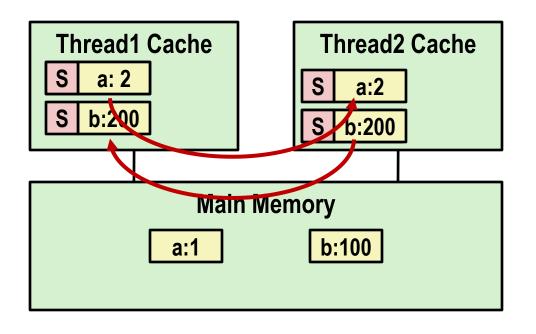
Snoopy Caches

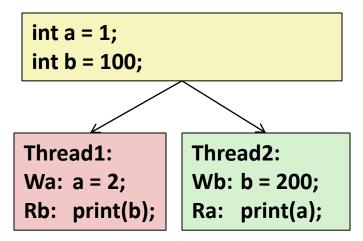
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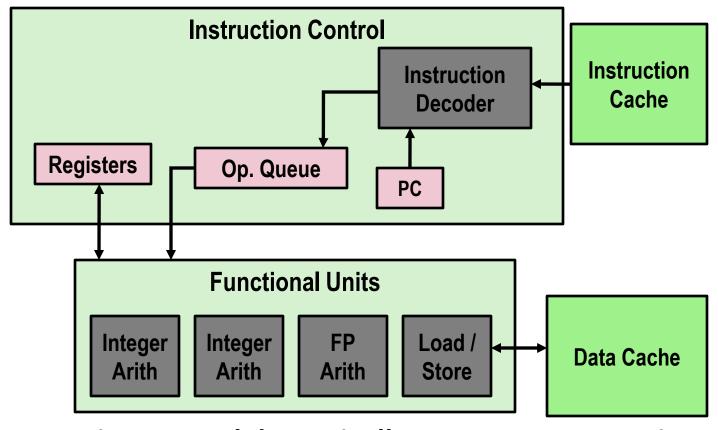


print 2

print 200

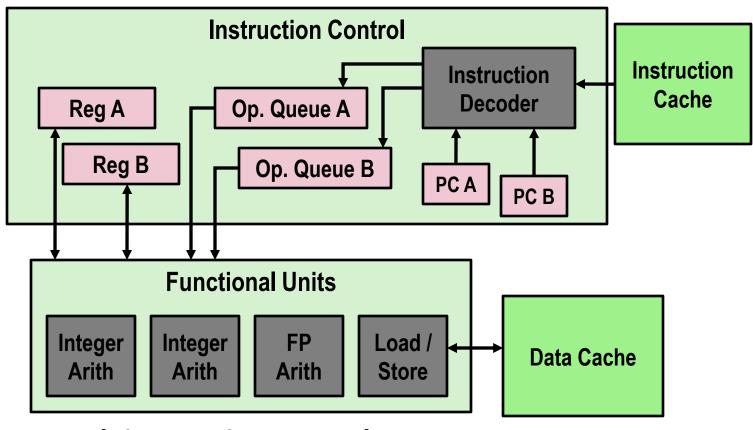
- When cache sees request for one of its E-tagged blocks
 - Supply value from cache
 - Set tag to S

Out-of-Order Processor Structure



- Instruction control dynamically converts program into stream of operations
- Operations mapped onto functional units to execute in parallel

Hyperthreading



- Replicate enough instruction control to process K instruction streams
- K copies of all registers
- Share functional units

Summary: Creating Parallel Machines

Multicore

- Separate instruction logic and functional units
- Some shared, some private caches
- Must implement cache coherency

Hyperthreading

- Also called "simultaneous multithreading"
- Separate program state
- Shared functional units & caches
- No special control needed for coherency

Combining

- Shark machines: 8 cores, each with 2-way hyperthreading
- Theoretical speedup of 16X
 - Never achieved in our benchmarks

Summation Example

- Sum numbers 0, ..., N-1
 - Should add up to (N-1)*N/2
- Partition into K ranges
 - LN/K values each
 - Accumulate leftover values serially
- Method #1: All threads update single global variable
 - 1A: No synchronization
 - 1B: Synchronize with pthread semaphore
 - 1C: Synchronize with pthread mutex
 - "Binary" semaphore. Only values 0 & 1

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];
```

Accumulating in Single Global Variable: Operation

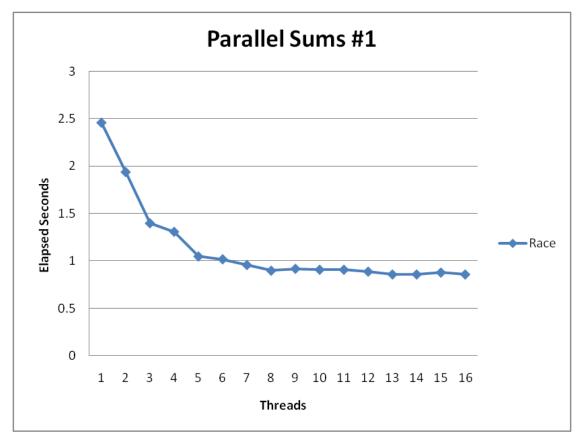
```
nelems per thread = nelems / nthreads;
/* Set global value */
global sum = 0;
/* Create threads and wait for them to finish */
for (i = 0; i < nthreads; i++) {
   myid[i] = i;
   Pthread create(&tid[i], NULL, thread fun, &myid[i]);
for (i = 0; i < nthreads; i++)
   Pthread join(tid[i], NULL);
result = global sum;
/* Add leftover elements */
for (e = nthreads * nelems per thread; e < nelems; e++)</pre>
   result += e;
```

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>
```

Unsynchronized Performance



- $N = 2^{30}$
- Best speedup = 2.86X
- Gets wrong answer when > 1 thread!

Thread Function: Semaphore / Mutex

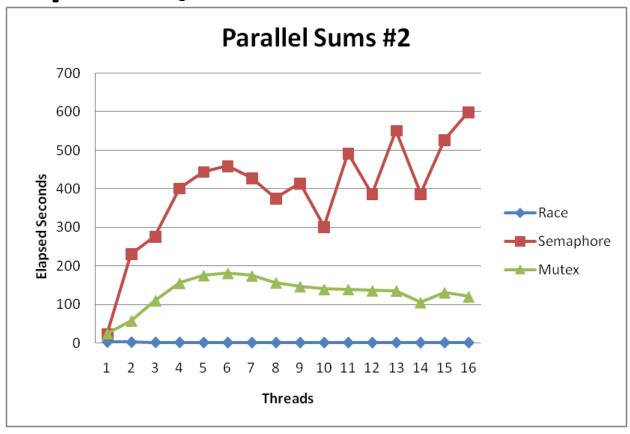
Semaphore

```
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;
```

Mutex

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

Semaphore / Mutex Performance



- Terrible Performance
 - 2.5 seconds → ~10 minutes
- Mutex 3X faster than semaphore
- Clearly, neither is successful

Separate Accumulation

- Method #2: Each thread accumulates into separate variable
 - 2A: Accumulate in contiguous array elements
 - 2B: Accumulate in spaced-apart array elements
 - 2C: Accumulate in registers

```
/* Partial sum computed by each thread */
data_t psum[MAXTHREADS*MAXSPACING];
/* Spacing between accumulators */
size_t spacing = 1;
```

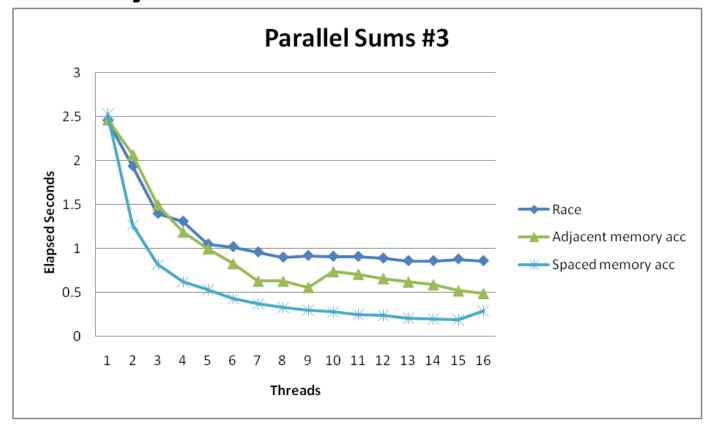
Separate Accumulation: Operation

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

Thread Function: Memory Accumulation

```
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++) {
       psum[index] += i;
    return NULL;
```

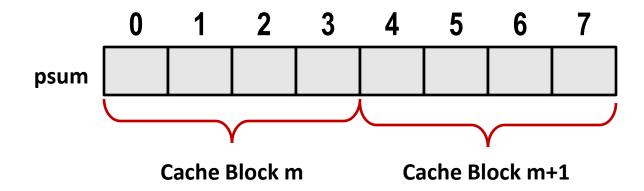
Memory Accumulation Performance



Clear threading advantage

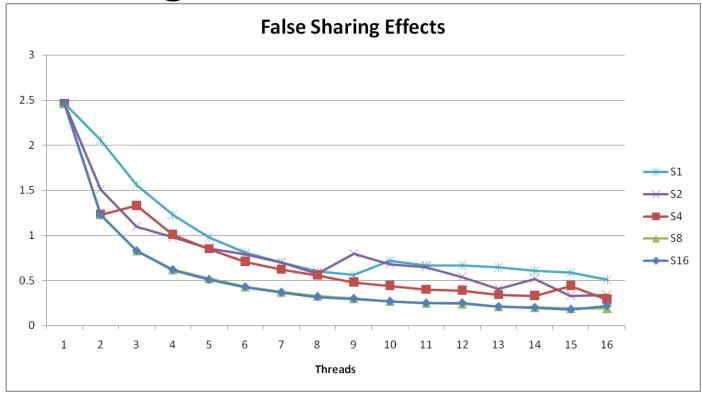
- Adjacent speedup: 5 X
- Spaced-apart speedup: 13.3 X (Only observed speedup > 8)
- Why does spacing the accumulators apart matter?

False Sharing



- Coherency maintained on cache blocks
- To update psum[i], thread i must have exclusive access
 - Threads sharing common cache block will keep fighting each other for access to block

False Sharing Performance

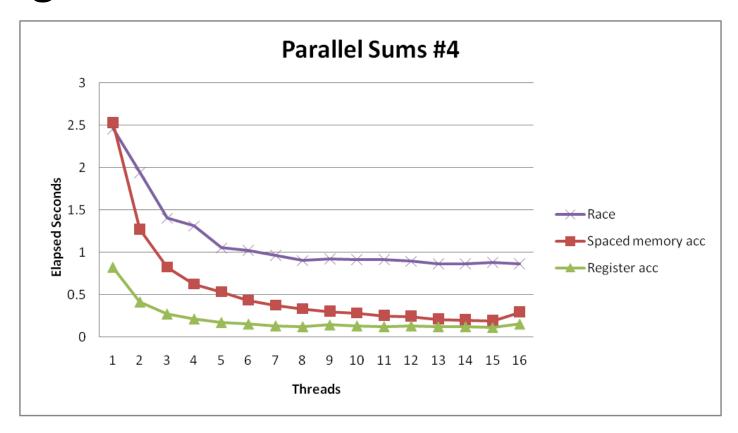


- Best spaced-apart performance 2.8 X better than best adjacent
- Demonstrates cache block size = 64
 - 8-byte values
 - No benefit increasing spacing beyond 8

Thread Function: Register Accumulation

```
void *sum_local(void *vargp)
{
    int myid = *((int *)vargp);
    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++) {
        sum += i;
    }
    psum[index] = sum;    return NULL;
}</pre>
```

Register Accumulation Performance

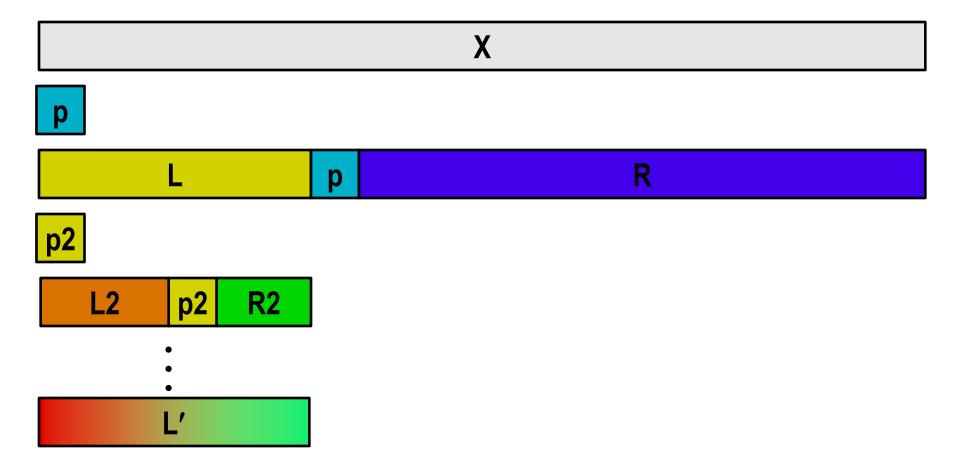


- Clear threading advantage
 - Speedup = 7.5 X
- 2X better than fastest memory accumulation

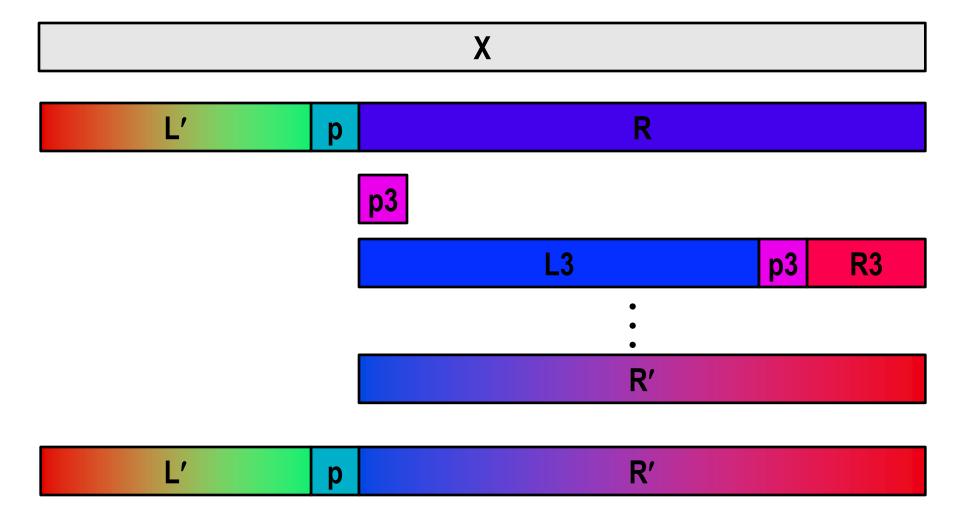
A More Interesting Example

- Sort set of N random numbers
- Multiple possible algorithms
 - Use parallel version of quicksort
- Sequential quicksort of set of values X
 - Choose "pivot" p from X
 - Rearrange X into
 - L: Values ≤ p
 - R: Values $\geq p$
 - Recursively sort L to get L'
 - Recursively sort R to get R'
 - Return L' : p : R'

Sequential Quicksort Visualized



Sequential Quicksort Visualized



Sequential Quicksort Code

```
void qsort serial(data t *base, size t nele) {
  if (nele <= 1)
    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);
```

Sort nele elements starting at base

Recursively sort L or R if has more than one element

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Parallel Quicksort

Parallel quicksort of set of values X

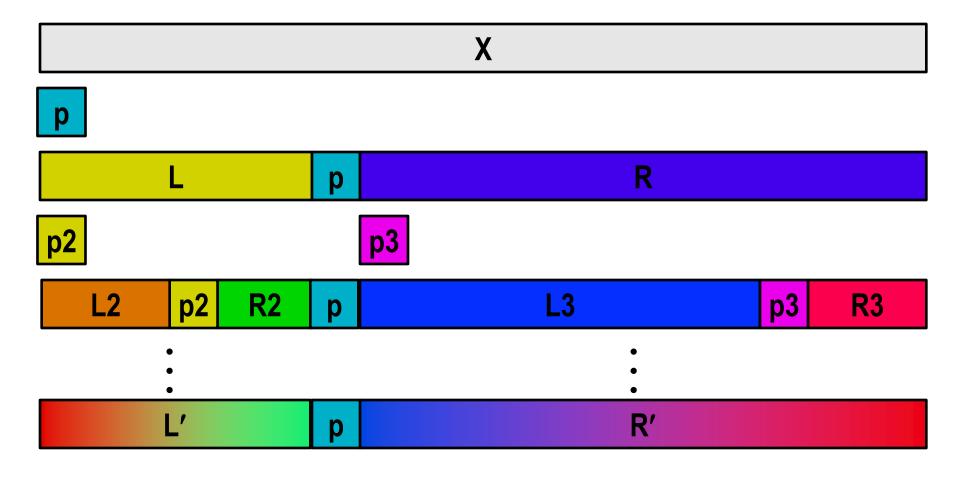
- If N ≤ Nthresh, do sequential quicksort
- Else
 - Choose "pivot" p from X
 - Rearrange X into
 - L: Values \leq p
 - R: Values ≥ p
 - Recursively spawn separate threads
 - Sort L to get L'
 - Sort R to get R'
 - Return L': p: R'

Degree of parallelism

- Top-level partition: none
- Second-level partition: 2X

• •••

Parallel Quicksort Visualized



Parallel Quicksort Data Structures

```
/* Structure that defines sorting task */
typedef struct {
   data_t *base;
   size_t nele;
   pthread_t tid;
} sort_task_t;

volatile int ntasks = 0;
volatile int ctasks = 0;
sort_task_t **tasks = NULL;
sem_t tmutex;
```

Data associated with each sorting task

base: Array start

nele: Number of elements

tid: Thread ID

Generate list of tasks

Must protect by mutex

Parallel Quicksort Initialization

```
static void init_task(size_t nele) {
  ctasks = 64;
  tasks = (sort_task_t **) Calloc(ctasks, sizeof(sort_task_t *));
  ntasks = 0;
  Sem_init(&tmutex, 0, 1);
  nele_max_serial = nele / serial_fraction;
}
```

- Task queue dynamically allocated
- Set Nthresh = N/F:
 - N Total number of elements
 - F Serial fraction
 - Fraction of total size at which shift to sequential quicksort

Parallel Quicksort: Accessing Task Queue

```
static sort task t *new task(data t *base, size t nele) {
 P(&tmutex);
  if (ntasks == ctasks) {
   ctasks *= 2;
   tasks = (sort task t **)
          Realloc(tasks, ctasks * sizeof(sort task t *));
  int idx = ntasks++;
  sort task t *t = (sort task t *) Malloc(sizeof(sort task t));
  tasks[idx] = t;
 V(&tmutex);
 t->base = base;
 t->nele = nele;
 t->tid = (pthread t) 0;
 return t;
```

- Dynamically expand by doubling queue length
 - Generate task structure dynamically (consumed when reap thread)
- Must protect all accesses to queue & ntasks by mutex

Parallel Quicksort: Top-Level Function

```
void tqsort(data_t *base, size_t nele) {
  int i;
  init_task(nele);
  tqsort_helper(base, nele);
  for (i = 0; i < get_ntasks(); i++) {
    P(&tmutex);
    sort_task_t *t = tasks[i];
    V(&tmutex);
    Pthread_join(t->tid, NULL);
    free((void *) t);
}
```

- Actual sorting done by tqsort_helper
- Must reap all of the spawned threads
 - All accesses to task queue & ntasks guarded by mutex

Parallel Quicksort: Recursive function

```
void tqsort_helper(data_t *base, size_t nele) {
  if (nele <= nele_max_serial) {
    /* Use sequential sort */
    qsort_serial(base, nele);
    return;
  }
  sort_task_t *t = new_task(base, nele);
  Pthread_create(&t->tid, NULL, sort_thread, (void *) t);
}
```

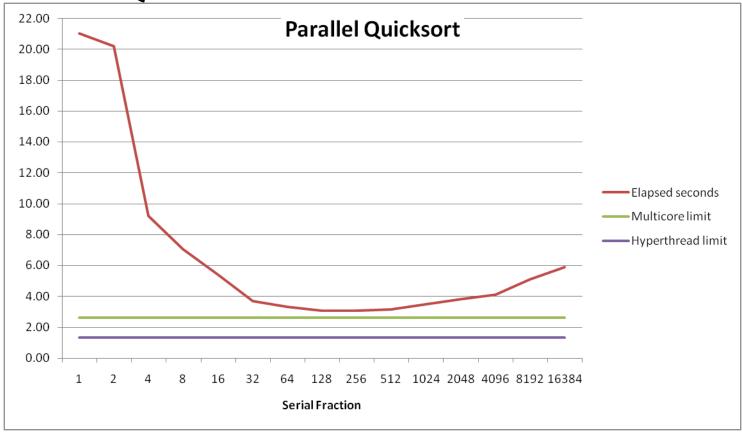
- If below Nthresh, call sequential quicksort
- Otherwise create sorting task

Parallel Quicksort: Sorting Task Function

```
static void *sort_thread(void *vargp) {
   sort_task_t *t = (sort_task_t *) vargp;
   data_t *base = t->base;
   size_t nele = t->nele;
   size_t m = partition(base, nele);
   if (m > 1)
      tqsort_helper(base, m);
   if (nele-1 > m+1)
      tqsort_helper(base+m+1, nele-m-1);
   return NULL;
}
```

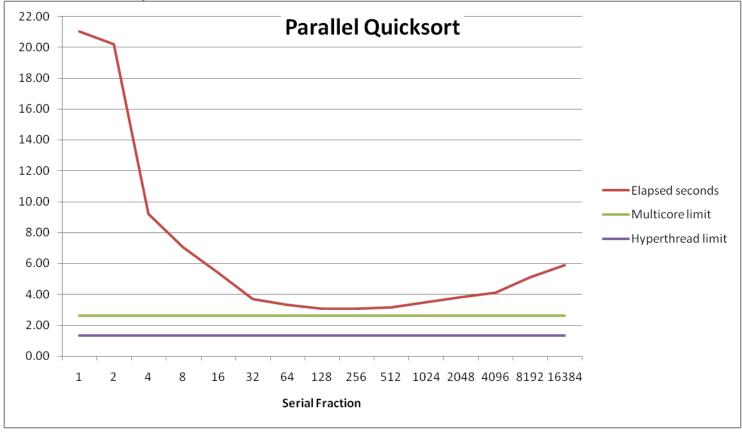
Same idea as sequential quicksort

Parallel Quicksort Performance



- Sort 2³⁷ (134,217,728) random values
- Best speedup = 6.84X

Parallel Quicksort Performance



Good performance over wide range of fraction values

- F too small: Not enough parallelism
- F too large: Thread overhead + run out of thread memory

Implementation Subtleties

Task set data structure

Array of structs

```
sort_task_t *tasks;
```

- new_task returns pointer or integer index
- Array of pointers to structs

```
sort_task_t **tasks;
```

new_task dynamically allocates struct and returns pointer

Reaping threads

Can we be sure the program won't terminate prematurely?

Amdahl's Law

Overall problem

- T Total time required
- p Fraction of total that can be sped up $(0 \le p \le 1)$
- k Speedup factor

Resulting Performance

- $T_k = pT/k + (1-p)T$
 - Portion which can be sped up runs k times faster
 - Portion which cannot be sped up stays the same
- Maximum possible speedup
 - $k = \infty$
 - $T_{\infty} = (1-p)T$

Amdahl's Law Example

Overall problem

- T = 10 Total time required
- p = 0.9 Fraction of total which can be sped up
- k = 9 Speedup factor

Resulting Performance

- $T_9 = 0.9 * 10/9 + 0.1 * 10 = 1.0 + 1.0 = 2.0$
- Maximum possible speedup
 - $T_{\infty} = 0.1 * 10.0 = 1.0$

Amdahl's Law & Parallel Quicksort

Sequential bottleneck

- Top-level partition: No speedup
- Second level: ≤ 2X speedup
- k^{th} level: $\leq 2^{k-1}X$ speedup

Implications

- Good performance for small-scale parallelism
- Would need to parallelize partitioning step to get large-scale parallelism
 - Parallel Sorting by Regular Sampling
 - H. Shi & J. Schaeffer, J. Parallel & Distributed Computing, 1992

Lessons Learned

- Must have strategy
 - Partition into K independent parts
 - Divide-and-conquer
- Inner loops must be synchronization free
 - Synchronization operations very expensive
- Watch out for hardware artifacts
 - Sharing and false sharing of global data
- You can do it!
 - Achieving modest levels of parallelism is not difficult