# CSEP 524: Parallel Computation (week 3)

**Brad Chamberlain** 

Tuesdays 6:30 – 9:20

MGH 231



## **Shameless Plug**

 The Chapel team is looking to fill two internship positions this summer if someone you know is interested.

#### What We've Discussed

- Why parallelism matters
- A bunch of terminology
- Ways of measuring parallel performance
- How to create/join tasks in C+Pthreads and Chapel
- Block and Cyclic work distributions
- Hopefully you've seen speedup firsthand by now

#### What's Next?

- At a high level:
  - Discussion/Diagnosis of behavior in Assignment #1
  - Having tasks coordinate with one another

## **Discussion of Assignment #1**



## **Assignment #1 Discussion**

Q1: What kinds of parallel resources did you find?

- who has highest-core count desktop?
- what larger-scale systems are available to you?
- what parallel programming models did you identify?

We should soon have access to a UW CSE 8x4-core VM-based platform for the class to share

### **Assignment #1 Discussion**

Q4: What block distribution strategy did you use?

- e.g., when dividing 10 items by 4 tasks, did you use:
  - 3322
  - 3232
  - 2323
  - 3331
  - other?

## **Assignment #1 Discussion**

**Q5:** What were your predictions?

- random vs. ramp
- negation vs. factorial
- block vs. cyclic
- number of tasks
- What were the biggest surprises?

Did you see linear speedup?

## **Summary of Observations**

#### **Block Distribution**

#### Cyclic Distribution

	random	ramp		random	ramp
negation should be faster than factorial			negation		
factorial	should be faster than ramp because		factorial		

## **Parallel Programming is Hard**

(you may or may not agree with this sentiment yet, but it's true)

Keep track of your war stories this quarter

- for the purposes of classroom discussion
- because misery loves company



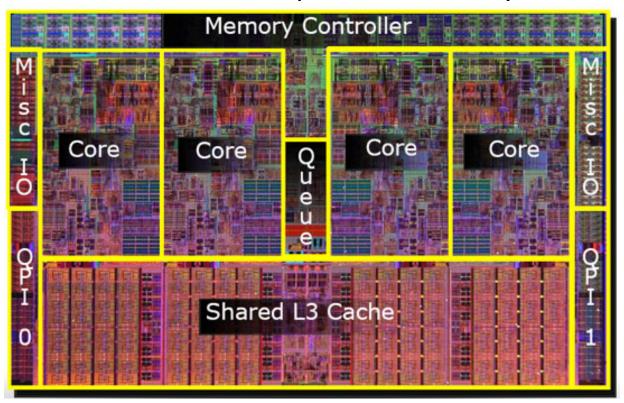
## **Two Performance Gotchas**



## Performance Gotcha #1: Memory

#### **Issue #1:** Competition for Memory Locations

 any time processors have non-shared caches there is the potential for them to compete for memory locations



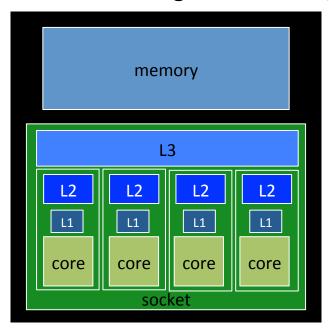


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## Performance Gotcha #1: Memory

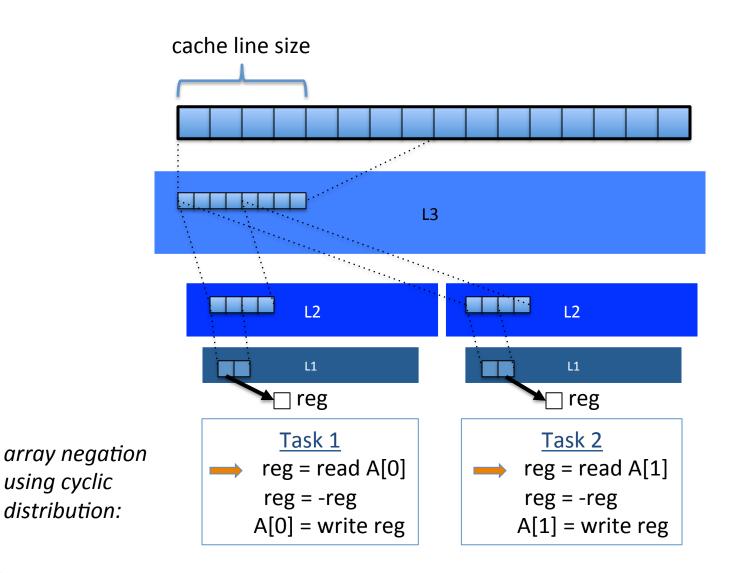
#### **Issue #1:** Competition for Memory Locations

- any time processors have non-shared caches there is the potential for them to compete for memory locations
  - read-only accesses should not be an issue
  - once a task/core starts writing to a location, competition may ensue





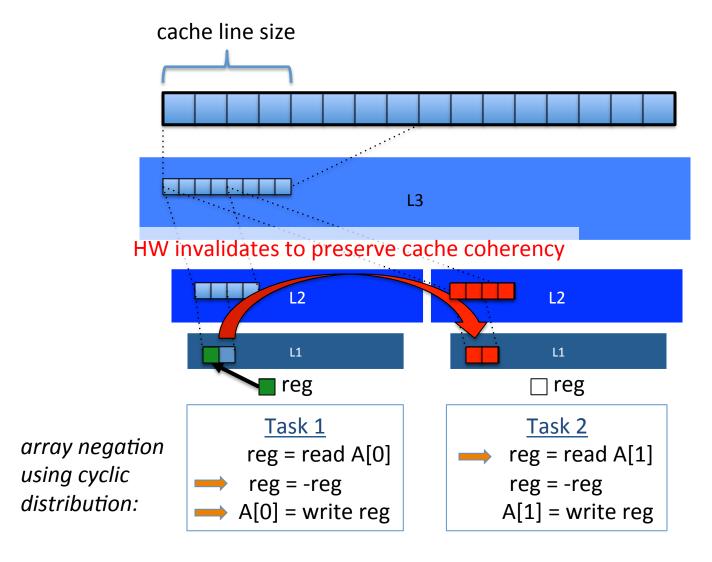
## **Example: Competition For Memory**





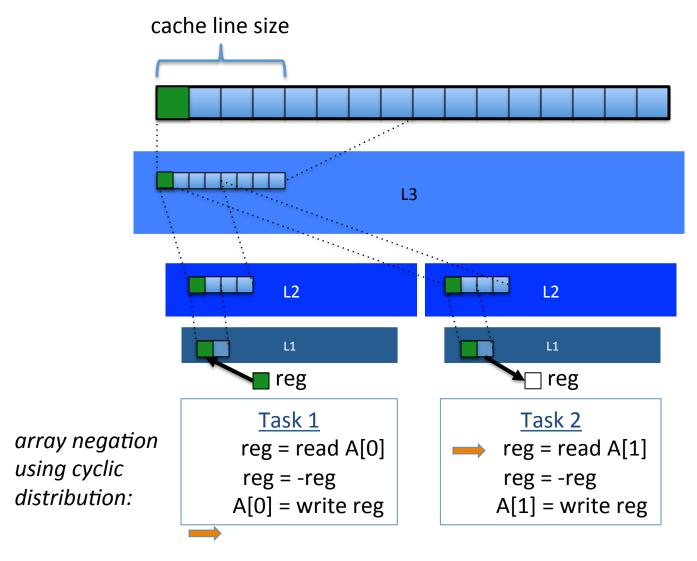
using cyclic

## **Example: Competition For Memory**





## **Example: Competition For Memory**





## **Definition: False Sharing**

False Sharing: When cache lines must be invalidated not because two tasks are accessing the same data, but because they're accessing data on the same cache line

- in reality, the data is truly independent, hence "false"
- the details of the granularity at which data is stored within HW is what causes the interdependence ("sharing")
- NOTE: On cache coherent architectures, this is a performance issue, not a correctness issue
- ("true sharing" might be considered when two tasks actually access the same shared variable/data)



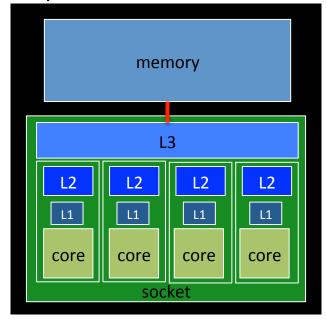
## **False Sharing Implications for Assignment #1**

- Writing to an array using a cyclic distribution can result in performance impacts due to false sharing
  - possible fixes:
    - have each task 0 start its cyclic iteration from a skewed position
      - e.g., have task t starts from element t + t\*n/p
      - but, results in more complex loop idioms due to need to wrap around
    - use padding/alignment pragmas to spread out array data
      - but, results in wasted space

## Performance Gotcha #1: Memory

#### **Issue #2:** *Memory is a bottleneck*

- typically, processors increase in speed faster than memory
- having multiple processors share memory doesn't help
  - there are only so many wires to access memory
  - cache coherence protocols also add overhead/complexity



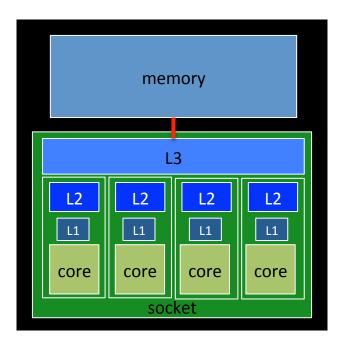


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## Performance Gotcha #1: Memory

#### **Issue #2:** *Memory is a bottleneck*

 algorithms with more computational intensity can better amortize these memory overheads





## **Definition: Computational Intensity**

#### **Computational Intensity:**



## **Definition: Computational Intensity**

**Computational Intensity:** How much computation is performed per memory access

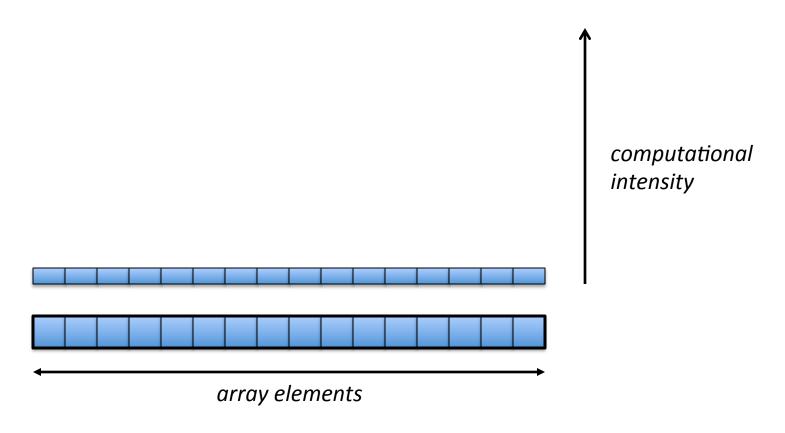
- high computational intensity: lots of OPS per load/store=> memory performance is less of an issue
- low computational intensity: few OPS per load/store=> memory performance is more of an issue



### Mem. Performance Implications for Assignment #1

- Computations with greater computational intensity should result in better speedup
  - e.g., factorial should speed up better than negation

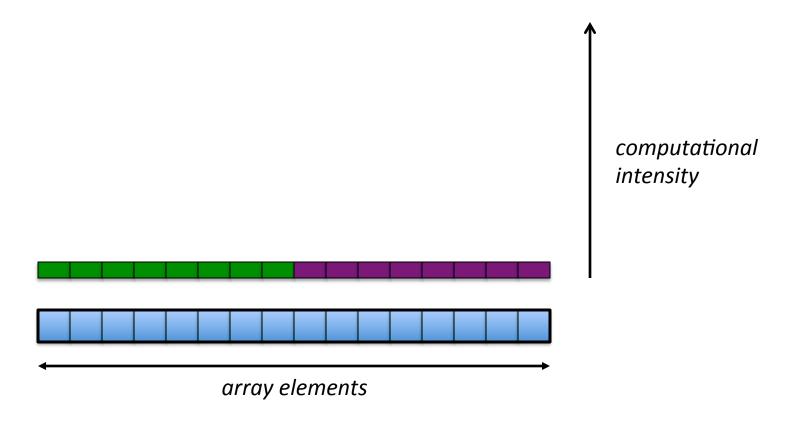
Negation + Ramp: Computational Intensity per Element





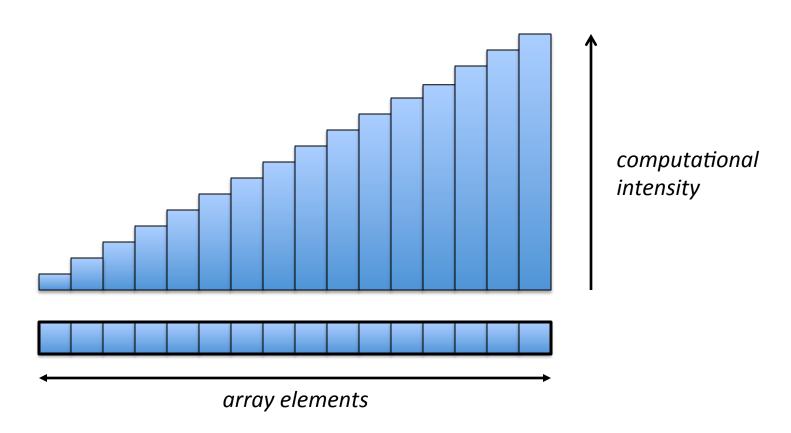
Negation + Ramp: Computational Intensity per Element

Block distribution: green and purple have ~the same work





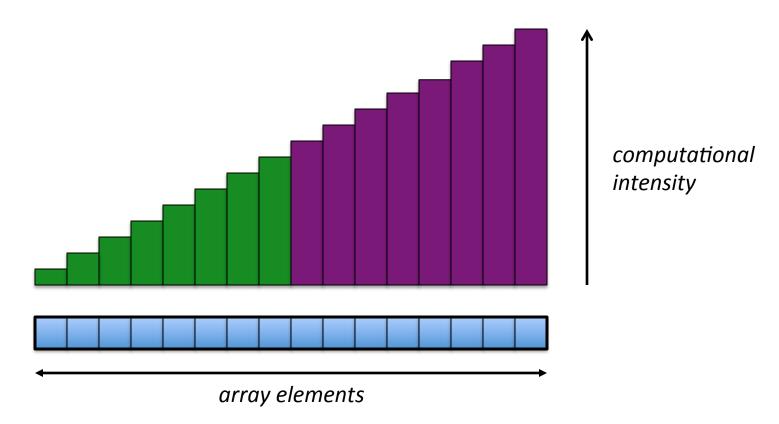
Factorial + Ramp: Computational Intensity per Element





Factorial + Ramp: Computational Intensity per Element

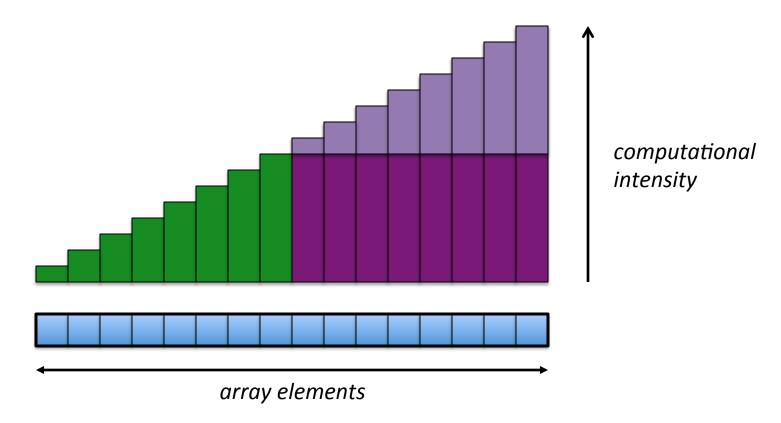
Block Distribution





Factorial + Ramp: Computational Intensity per Element

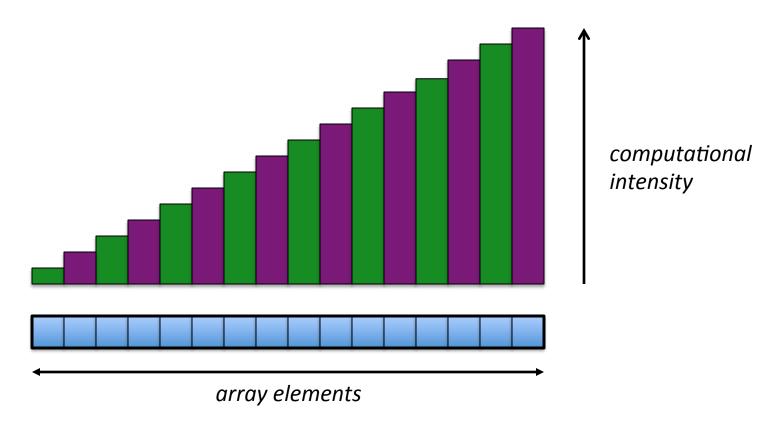
Block Distribution: Purple has ~3x as much work as green





Factorial + Ramp: Computational Intensity per Element

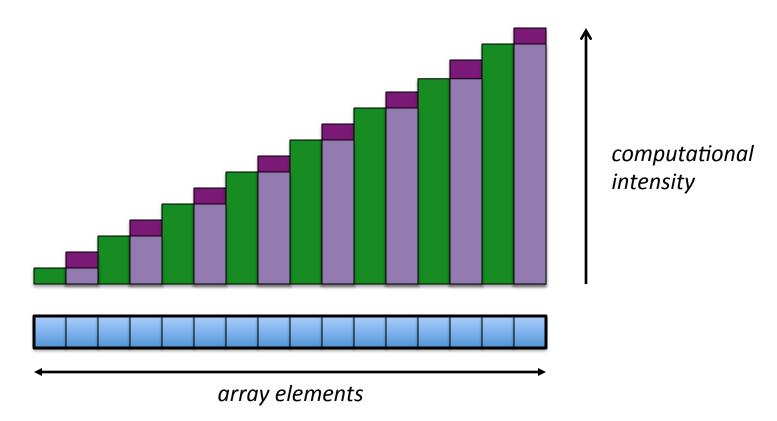
Cyclic Distribution





Factorial + Ramp: Computational Intensity per Element

Cyclic Distribution: Purple only has numItems/2 more work





#### Factorial + Random:

Block distribution: green has ~1.5x the work of purple

• (for the data set shown) computational intensity



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array elements

## **Load Balance Implications for Assignment #1**

- Block + factorial + ramp exhibits bad load balance
  - some tasks had significantly more work than others
  - cyclic/random input sets may result in better load balance
- Keep in mind that many algorithms must be written without knowing their input sets
  - i.e., can't think "aha, my input will be a ramp so ..."



## **Assignment #1 Debrief**

- Who saw execution time behaviors similar to what I just described?
  - what kinds of things did you "do right" to get this result?
  - what kinds of issues did others do differently to not see it?
  - or perhaps, rather, what did you stumble across then fix?
    - measuring aggregate performance of all threads, not wallclock time

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## **Assignment #1 Summary: Distributions**

#### **Block & Cyclic:**

- + give each task a similar number of work items
- + reasonably easy to compute

#### **Block:**

- + results in good spatial locality (touches adjacent elements)
- can expose sensitivities to work distribution
  - as in ramp+factorial

#### Cyclic:

- + less likely to be sensitive to work distribution
- can result in false sharing issues



# Time for a Break/Something Different?



## **Alternatives to Block and Cyclic**

- Other distributions can help address the drawbacks of block and cyclic:
  - Block-Cyclic distribution
  - Dynamic distributions
  - Algorithmically-aware distributions

## **Distribution #3: Block-Cyclic Distribution**

- As the name suggests, a hybrid of Block and Cyclic
  - deals blocks of items out cyclically
    - parameterized by block size, b
    - ideally, b should match or exceed cache line size
    - optimal choice of b often depends on algorithm, working set size, ...



#### – tradeoffs:

- + gives tasks chunks of work (good spatial locality; less false sharing)
- + like cyclic, results in probabilistically-oriented load balancing
- results in slightly more complicated loop nests

#### **Dynamic Distributions**

#### **Concept:**

- don't deal work out according to a fixed, a priori schedule
- instead, deal work out to tasks (or have them grab it) as they become idle

#### **Goal:**

no task gets stuck with more work than it can handle

#### **Challenge:**

- what granularity (granularities?) to deal out work?
  - if too large: tasks may get unlucky and stuck with too much work
  - if too small: too much effort coordinating, not enough computing



## **Algorithmically-Aware Distributions**

#### **Concept:**

- For some algorithms, there may be a way to scan the input data in order to compute a good distribution
  - e.g., dynamically sample the input data set to try and predict trends?
  - e.g., examine the placement of zeroes and non-zeroes in a sparse matrix?
  - e.g., compute a dependence graph for the computation and distribute it using a graph partitioning algorithm

#### **Goal:**

use algorithmic-centric knowledge to improve load balance

#### **Challenge:**

- Cost:Benefit ratio needs to be taken into account
  - since any overhead in computing a distribution is new work that wouldn't have been required in a serial version

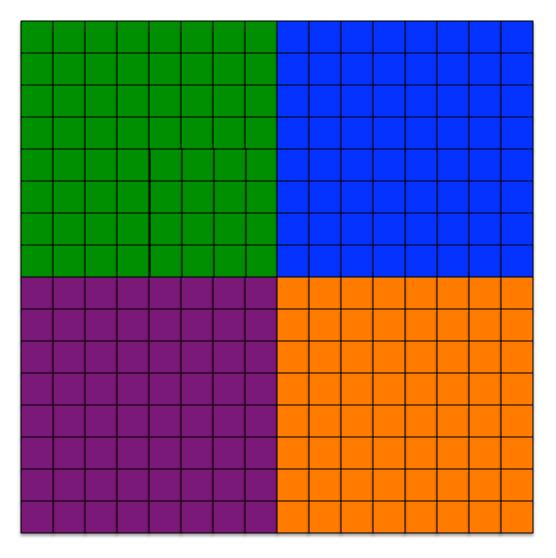


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#### **Multidimensional Distributions**

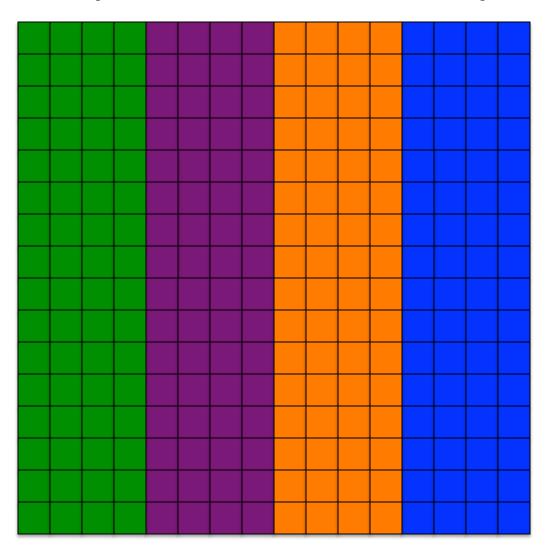
- So far, we've looked solely at 1D distributions
- Distributions can also be multidimensional
  - one option is to apply a 1D distribution per dimension

# 2D Block x Block (distributed to 2x2 tasks)



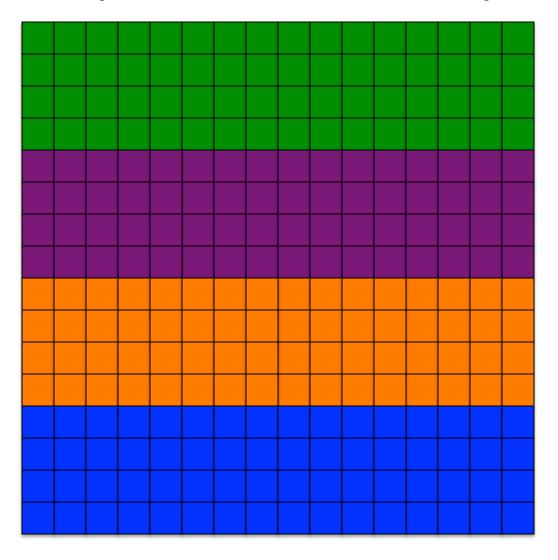


# 2D Block x Block (distributed to 1x4 tasks)

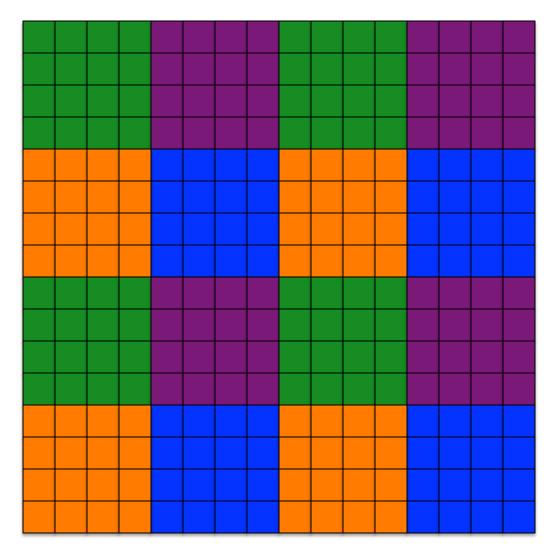




# 2D Block x Block (distributed to 4x1 tasks)



# 2D Block-Cyclic x Block-Cyclic (distributed to 2x2 tasks)



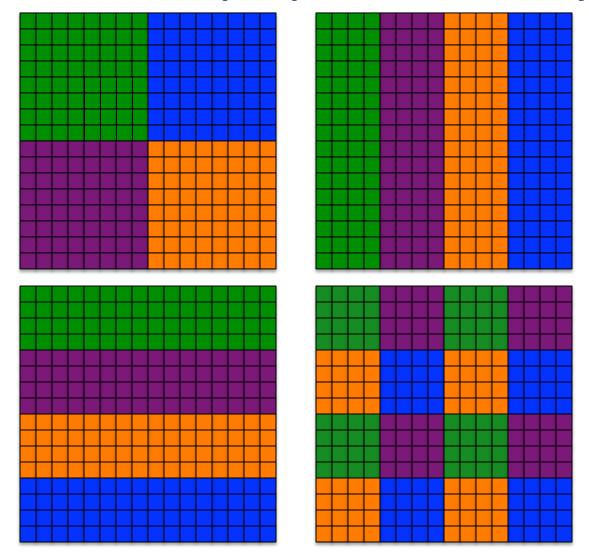


#### ...and so on and so forth

- Cyclic x Cyclic
- Block x Cyclic
- Cyclic x Block
- Block-Cyclic x Block-Cyclic with different block sizes
- Block-Cyclic x Block
- Block x Block-Cyclic
- etc.



# Q: In a Shared-Memory setting, which would you use from the perspective of memory?



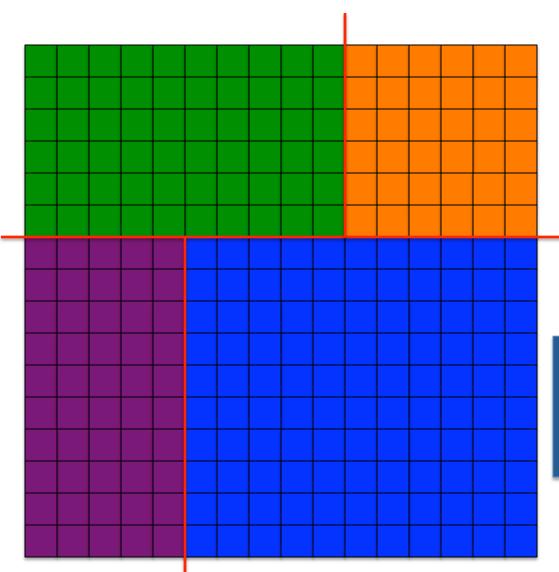


#### **Multidimensional Distributions**

- So far, we've primarily looked at 1D distributions
- Distributions can also be multidimensional
  - one option is to apply a 1D distribution per dimension
  - another is to distribute the items holistically

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#### **Holistic Distribution: Recursive Bisection**



Note: Can't be expressed as the conflation of two 1D distributions



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#### **Multidimensional Distributions**

- So far, we've primarily looked at 1D distributions
- Distributions can also be multidimensional
  - one option is to apply a 1D distribution per dimension
  - another is to distribute the items holistically
- Or, even unstructured (e.g., distribute a graph)
  - a topic for another day



## **Measuring Load Imbalance**

• In assignment #1, we used the following pattern to measure the overall execution time of the code:

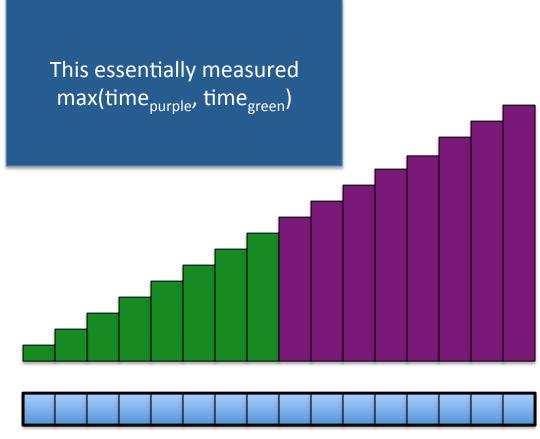
start timer

create tasks

do work

join tasks

check timer





## **Measuring Load Imbalance**

Imagine instead, pushing the timing into the loop:

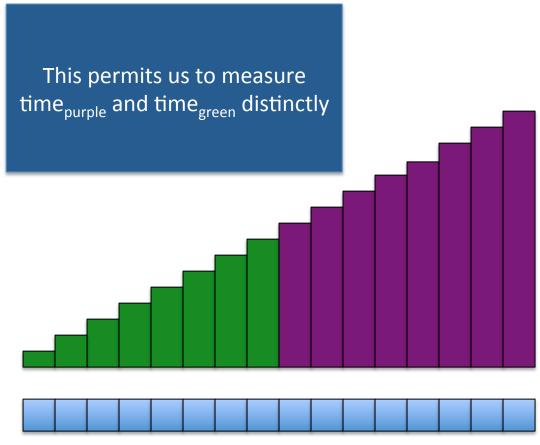
create tasks

start timer

do work

check timer

join tasks





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#### **Measuring Load Imbalance**

 Now, we can compute statistics on a task-by-task basis:

```
var totTime, maxTime = 0.0;
var minTime = max(real);
                                  What's the bug in this code?
coforall tid in 0..#numTasks {
  start timer
    do work
  const myTime = check timer
  totTime += myTime;
  if myTime < minTime then minTime = myTime;</pre>
  if myTime > maxTime then maxTime = myTime;
const avgTime = totTime / numTasks;
```



- The previous slide contains a classic bug
  - Code that looks innocuous is actually problematic
  - Cause: reading parallel code as though it were sequential

```
coforall tid in 0..#numTasks {
    ...
    totTime += myTime;
    ...
}
Task 1
```

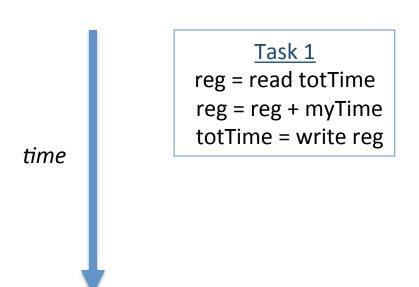
reg = read totTime
reg = reg + myTime
totTime = write reg

#### Task 2

reg = read totTime
reg = reg + myTime
totTime = write reg



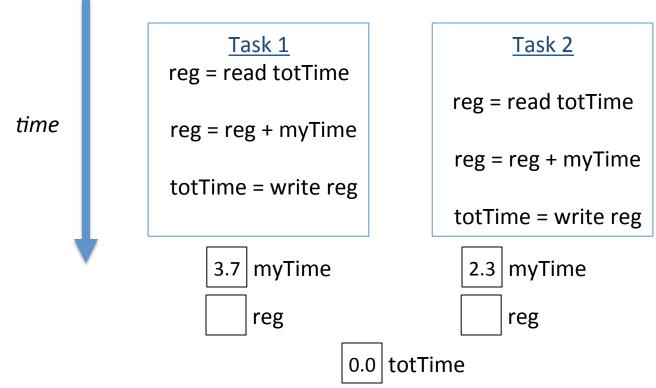
- Whether or not this bug exhibits itself depends on the scheduling of the tasks
  - the following schedule would be fine:



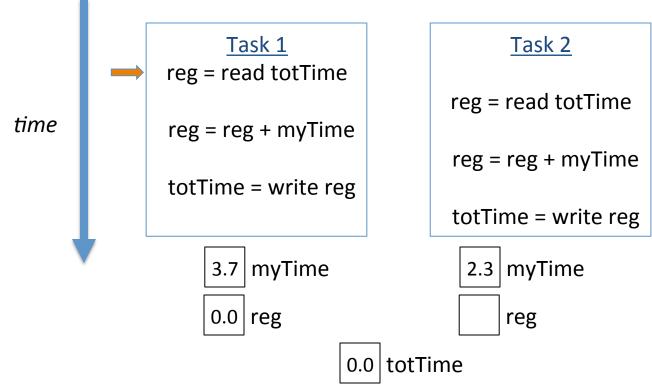
#### Task 2

reg = read totTime
reg = reg + myTime
totTime = write reg

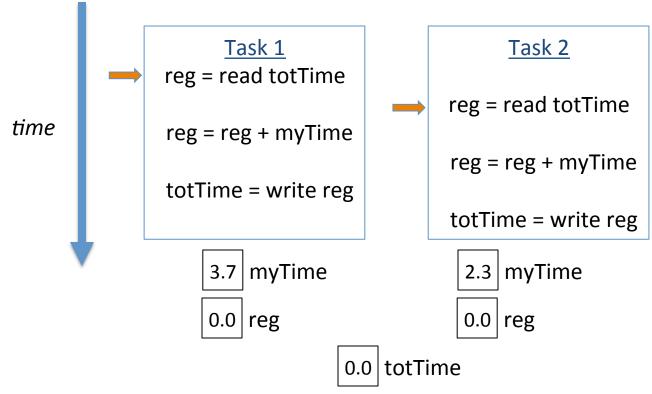
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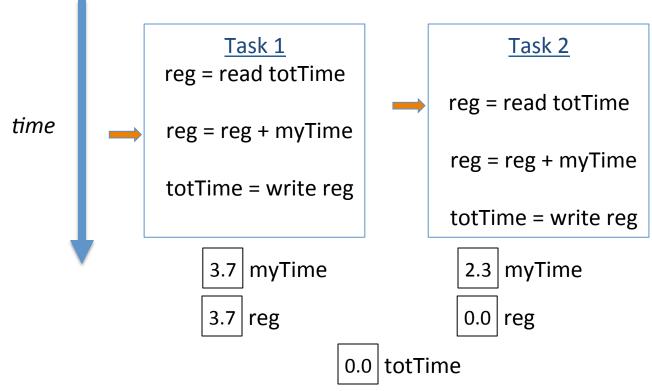


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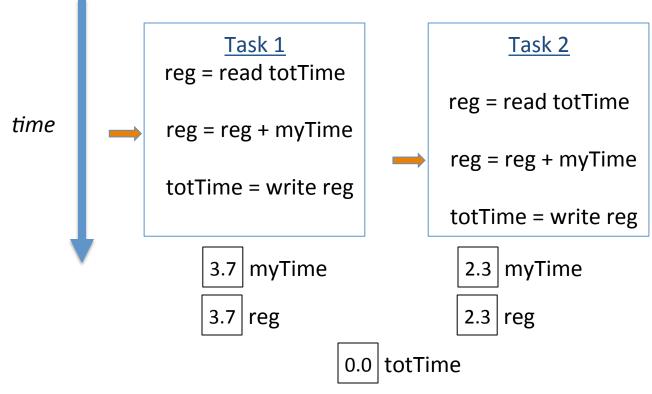


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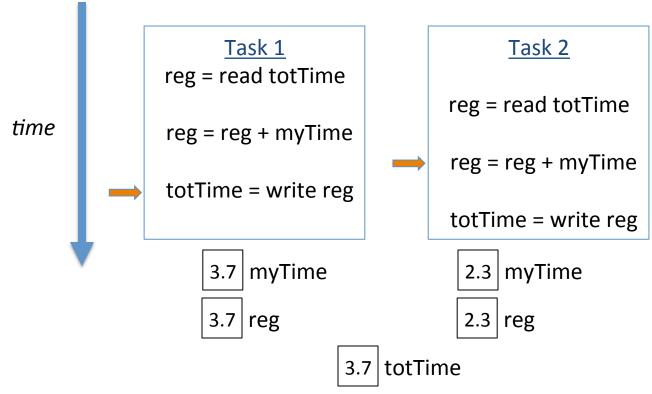


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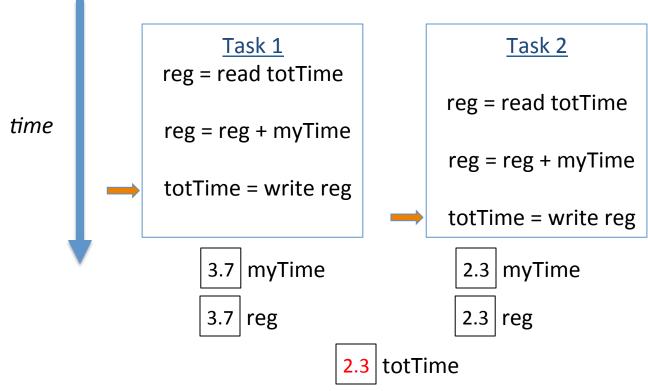


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  - the following schedule is problematic:





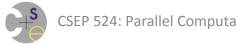
- Whether or not this bug exhibits itself depends on the scheduling of the tasks
  - the following schedule is problematic:



#### Bug of the week: RRWW (Read-Read-Write-Write)

- Due to interleaving, uncoordinated reads and writes to shared state may cause values to be lost
- The fix is to coordinate such accesses to shared state
  - in this case, totTime, minTime, maxTime
  - e.g., could protect each/all of them by a lock

```
coforall tid in 0..#numTasks
...
   grab totTime lock
   totTime += myTime;
   release totTime lock
```



# **Glossary: Synchronization**

## **Synchronization:**



# **Glossary: Synchronization**

**Synchronization:** Coordination between tasks

## **Synchronization Mechanisms in Pthreads**

- 1) mutex: "mutual exclusion" essentially a lock
  - operations:
    - init, destroy: create and destroy them
    - lock, unlock: grab and release the lock
    - trylock: attempt to grab the lock, but don't block if you can't



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# **Synchronization Mechanisms in Pthreads**

# 2) condition variables: a "waiting room" for some condition to become true

- operations:
  - init, destroy: create and destroy them
  - wait: wait for a condition to become true
  - **signal/broadcast:** signal to one/multiple thread(s) that it is
- rationale: avoid spinning on some test in user code
  - e.g., "wait for this variable to take on some nonzero value"
  - such spinning is typically not a wise use of resources
  - instead: let the thread library manage who should wake up when

## **Condition Variables: Fiddly Details**

#### There are some details that complicate condition vars:

- mutex argument: must be managed properly
- spurious wakeups: verifying that the condition is still true once you've awoken from a wait()

#### See Ch. 6 of the text and/or this tutorial for details:

https://computing.llnl.gov/tutorials/pthreads/#ConditionVariables



## Using Mutexes to fix RRWW bugs

```
pthread mutex t totTimeMutex;
pthread mutex init(&totTimeMutex, NULL);
create tasks
  pthread mutex lock(&totTimeMutex);
    totTime += myTime;
  pthread mutex unlock(&totTimeMutex);
join tasks
pthread mutex destroy(&totTimeMutex);
```



# Using Mutexes to fix RRWW bugs

The result is that there are only two legal orderings of the totTime updates:

#### task 1 grabs the mutex first

#### Task 1

mutex lock
reg = read totTime
reg = reg + myTime
totTime = write reg
mutex unlock

#### Task 2

mutex lock

(...blocks...)

reg = read totTime
reg = reg + myTime
totTime = write reg
mutex unlock

#### time

#### task 2 grabs the mutex first

#### Task 1

mutex lock

(...blocks...)

reg = read totTime
reg = reg + myTime
totTime = write reg
mutex unlock

#### Task 2

mutex lock reg = read totTime reg = reg + myTime totTime = write reg mutex unlock



# **Synchronization Mechanisms in Chapel**

- 1) synchronization variables
- 2) single-assignment variables

#### Synchronization Variables



Syntax

```
sync-type:
sync type
```

- Semantics
  - Stores full/empty state along with normal value
  - Defaults to full if initialized, empty otherwise
  - Default read blocks until full, leaves empty
  - Default write blocks until empty, leaves full
- Examples: Critical sections and futures

```
var lock$: sync bool;
lock$ = true;
critical();
var lockval = lock$;
```

```
var future$: sync real;

begin future$ = compute();
computeSomethingElse();
useComputedResults(future$);
```





#### Example: Bounded Buffer Producer/Consumer

```
var buff$: [0..#buffersize] sync real;
cobegin {
  producer();
  consumer();
proc producer() {
  var i = 0;
  for ... {
    i = (i+1) % buffersize;
   buff$[i] = ...; // blocks until empty, leaves full
proc consumer() {
  var i = 0;
  while ... {
    i= (i+1) % buffersize;
    ...buff$[i]...; // blocks until full, leaves empty
```



#### Single-Assignment Variables



Syntax

```
single-type:
single type
```

- Semantics
  - Similar to sync variables, but stays full once written
- Example: Multiple Consumers of a future

```
var future$: single real;

begin future$ = compute();
computeSomethingElse(future$);
computeSomethingElse(future$);
```



#### Synchronization Type Methods



- block until full, leave empty, return value readFE():t readFF():t block until full, leave full, return value return value (non-blocking) readXX():t block until *empty*, set value to  $\forall$ , leave *full* writeEF(v:t) wait until *full*, set value to v, leave *full* writeFF(v:t) writeXF(v:t) set value to  $\vee$ , leave *full* (non-blocking) reset value, leave *empty* (non-blocking) • reset() return true if full else false (non-blocking) • isFull: bool
- Defaults: read: readFE, write: writeEF





#### Single Type Methods



- readFE():t block until full, leave empty, return value
- readFF():t block until full, leave full, return value
- readXX():t return value (non-blocking)
- writeEF(v:t) block until empty, set value to v, leave full
- writeFF (v:t) wait until full, set value to v, leave full
- writeXF(v:t) set value to v, leave full (non-blocking)
- reset value, leave empty (non-blocking)
- isFull: bool return true if full else false (non-blocking)
- Defaults: read: readFF, write: writeEF



#### Using Sync vars to fix RRWW bugs

```
var totTime$: sync real = 0.0; // starts full

coforall tid in 0..#numTasks {
    ...
    totTime$ += myTime; // readFE followed by writeEF
    ...
}
```



## Summary: Pthreads vs. Chapel Synchronization

#### Pthreads mutex & condition variables:

- + arguably a reasonable backbone for synchronization
  - based on the endurance of Pthreads
  - use of these concepts in other languages/contexts
- arguably result in complex code for common patterns

#### **Chapel sync/single variables:**

- + data-centric synchronization: expressing synchronization in terms of the data being accessed
- arguably a little artificial/confusing when used as a mutex
  - e.g., see unused boolean value in previous critical section example

Both approaches also have some common liabilities (stay tuned)



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# **Diagnosing Deadlock/Livelock in Chapel**

- If you suspect you have a deadlock problem...
  - re-execute your program using –b/--blockreport
    - adds a certain amount of overhead, but beats deadlocking!
  - if deadlock is detected, the program will...
    - terminate
    - do its best to tell you where the tasks were
- If you suspect you have a livelock problem...
  - re-execute your program using –t/--taskreport
    - again, adds a certain amount of overhead
  - upon hitting Ctrl-C/sending SIGINT, the program will...
    - terminate and do its best to tell you where the tasks are



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# This week's assignment

- extend the single-producer/single-consumer bounded buffer pattern shown in lecture to support multiple producers and consumers
  - in Chapel (to get practice with sync/single variables)
  - in Pthreads (to get practice with mutex/condition variables)
- write a dynamic load balancing distribution in Chapel OR Pthreads
  - apply to ramp + factorial case
- some written questions