COMP 322: Parallel and Concurrent Programming

Lecture 18: Abstract vs. Real Performance

"Everything You Ever Wanted to Know About HJLib but Were Too Afraid to Ask"

Zoran Budimlić and Mack Joyner {zoran, mjoyner}@rice.edu

http://comp322.rice.edu



Functional Approach to Parallelism

- "Functional": futures, future tasks, streams, data-driven tasks and futures
- "Not-so functional": async tasks and finish scopes, tasks that modify shared memory
- Advantages to functional approach
 - Easier to reason about
 - Don't have to worry about data races
 - Leads to compact, elegant, easy to read code
 - Easy to scale to massively parallel (because you don't need to worry about data races)
- Disadvantages
 - May be hard to express exactly the computation graph you need (i.e. a finish scope with millions of tasks)
 - May be more expensive to execute (blocking future.get() vs. simply reading a shared memory location)
 - May need additional data structures (futures, data-driven futures) to express the computation
 - May need copying of data structures to avoid data races and mutation
 - Hard to scale to massively parallel (because of overheads)



Abstract vs. Real Performance

Abstract performance

- Focus on operation counts for WORK and CPL, regardless of actual execution time
- Ignore the nitty-gritty of task creation and execution overhead
- Same "performance" regardless of the machine
- Real performance
 - Lots of things happening "under the hood"
 - Operating system, runtime and hardware all have an impact
 - Process creation/execution vs. thread creation/execution vs. task creation/execution
 - Tasks could be blocked, waiting on some event
 - Complex matter, but important to at least have a general idea of the costs



Lab 4: Recursive Task Parallelism

```
private static double recursiveMaxParallel(final double[] inX, final int start, final int end)
        throws SuspendableException
    if (end - start = 2) {
        doWork(1);
        return 1/inX[end - 1] + 1/inX[start];
    } else {
        var bottom = future(() \rightarrow recursiveMaxParallel(inX, start, (end + start) / 2));
        var top = future(() \rightarrow recursiveMaxParallel(inX, (end+start) / 2, end));
        var bVal = bottom.get();
        var tVal = top.get();
        doWork(1);
        return bVal + tVal;
```



Lab 4: Recursive Task Parallelism

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        var top = future(() \rightarrow recursiveMaxParallel(inX, (end+start) / 2, end));
        var bVal = bottom.get();
        var tVal = top.get();
                                                                                           4 sec 459 ms
                                            Test Results
        doWork(1);
                                         edu.rice.comp322.Lab4CorrectnessTest
                                                                                           4 sec 459 ms
        return bVal + tVal;
                                               testReciprocalParallelism2Futures
                                                                                                 241 ms
                                               testReciprocalParallelism4Futures
                                                                                                 58 ms
                                               testReciprocalParallelism8Futures
                                                                                                 58 ms
                                               testReciprocalMaxParallelism
                                                                                            4 sec 102 ms
```



Cutoff Strategy for Recursive Task Parallelism

```
private static double recursiveMaxParallelCutoff(final double[] inX, final int start, final int end,
                                                     final int threshold) throws SuspendableException {
    if (end - start ≤ threshold) {
        double sum = 0.0;
        for(int i = start; i < end; i++) {</pre>
            doWork(1);
            sum = sum + 1 / inX[i];
        return sum;
    } else {
        var bottom = future(() \rightarrow recursive Max Parallel Cutoff(inX, start, (end + start) / 2, threshold));
        var top = future(() \rightarrow recursiveMaxParallelCutoff(inX, (end+start) / 2, end, threshold));
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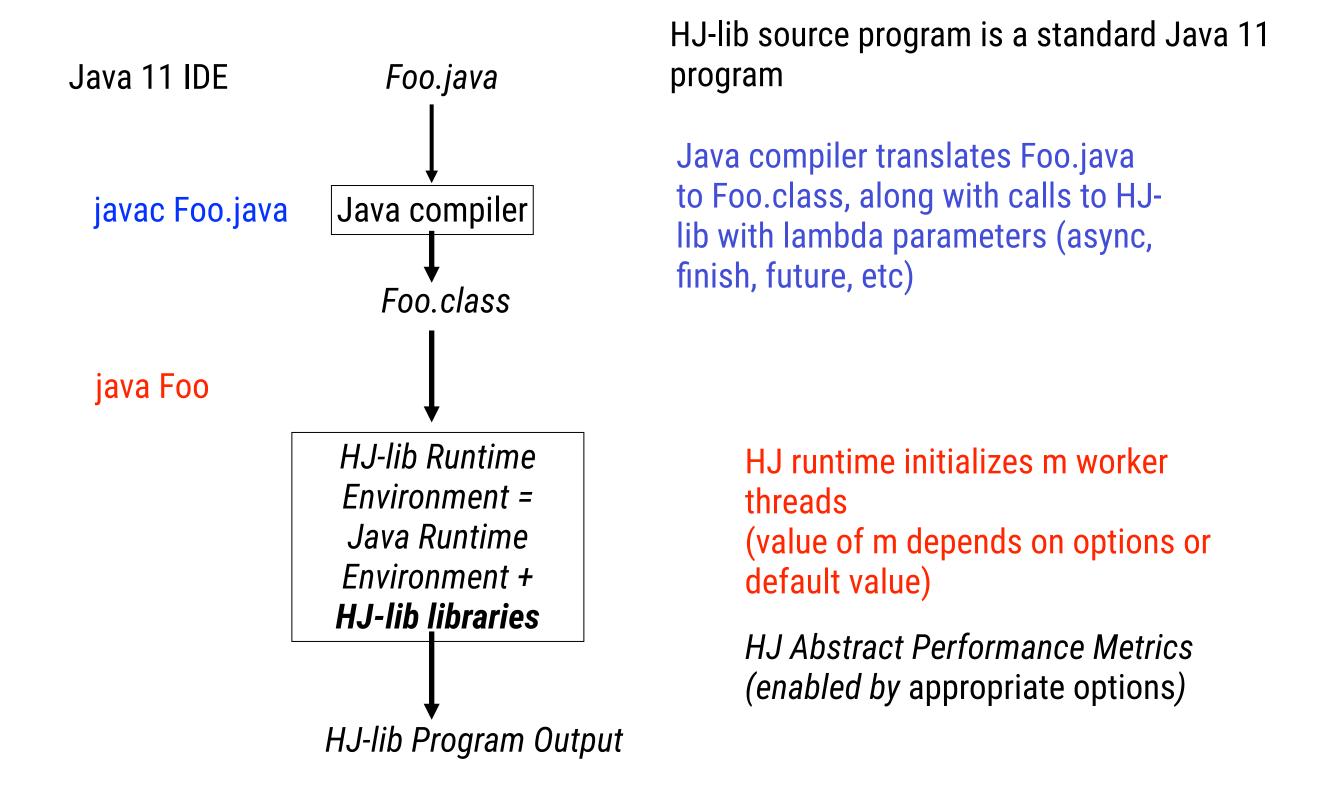


Cutoff Strategy for Recursive Task Parallelism

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        var bottom = future(() \rightarrow recursive Max Parallel Cutoff(inX, start, (end + start) / 2, threshold));
        var top = future(() \rightarrow recursiveMaxParallelCutoff(inX, (end+start) / 2, end, threshold));
        var bVal = bottom.get();
        var tVal = top.get();
                                                     Execution with threshold 64000 took 56 milliseconds.
        doWork(1);
                                                     Execution with threshold 128000 took 54 milliseconds.
        return bVal + tVal;
                                                     Execution with threshold 256000 took 4 milliseconds.
                                                     Execution with threshold 512000 took 3 milliseconds.
                                                     Execution with threshold 1024000 took 6 milliseconds.
                                                     Execution with threshold 2048000 took 10 milliseconds.
                                                     Execution with threshold 4096000 took 11 milliseconds.
```

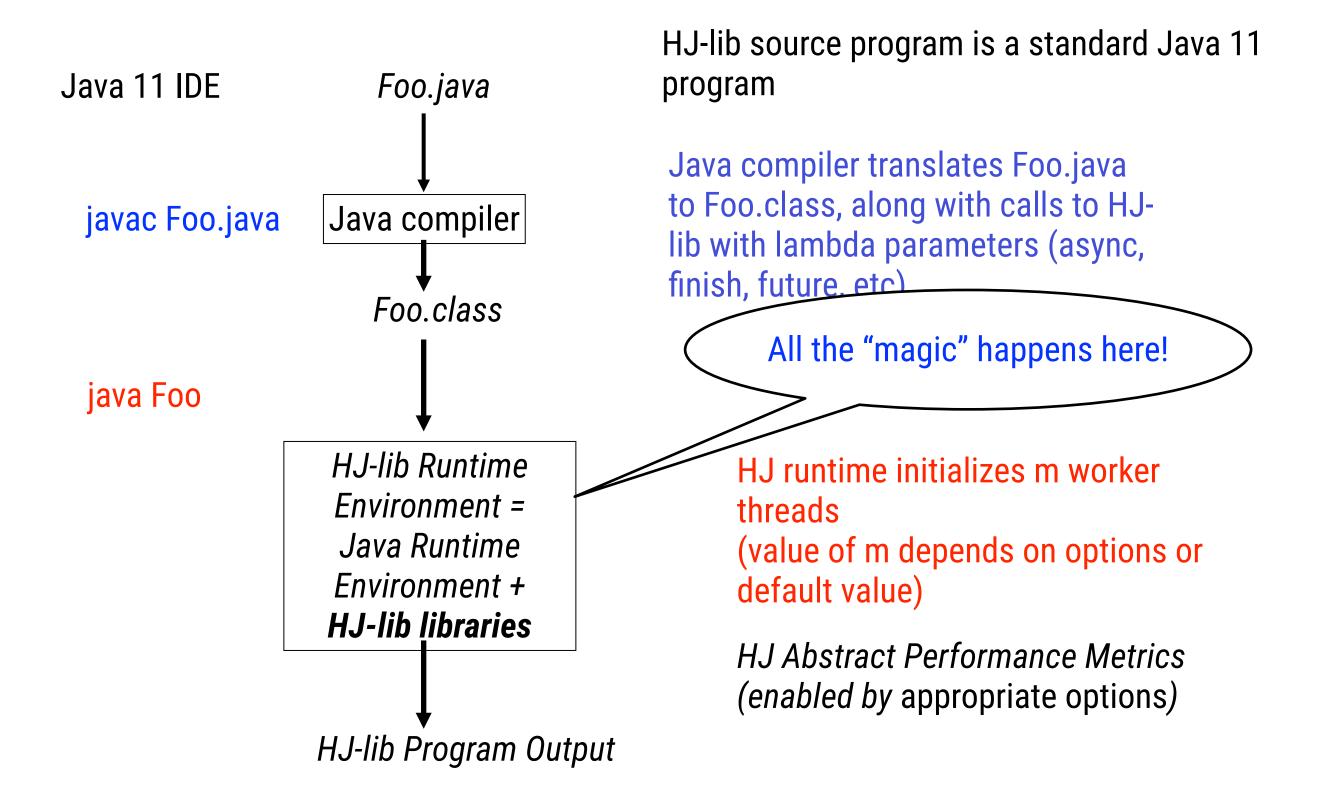


HJ-lib Compilation and Execution Environment



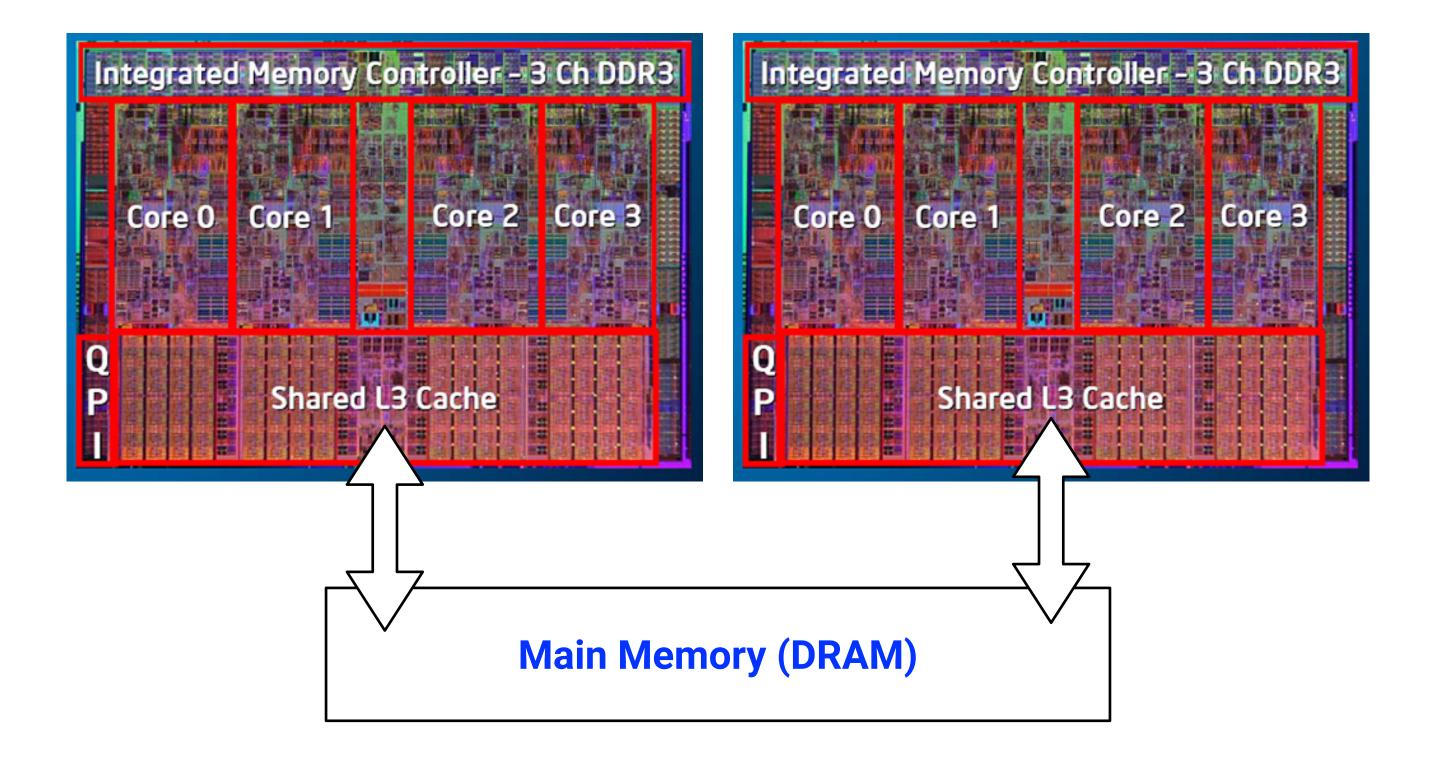


HJ-lib Compilation and Execution Environment





Looking under the hood - let's start with the hardware



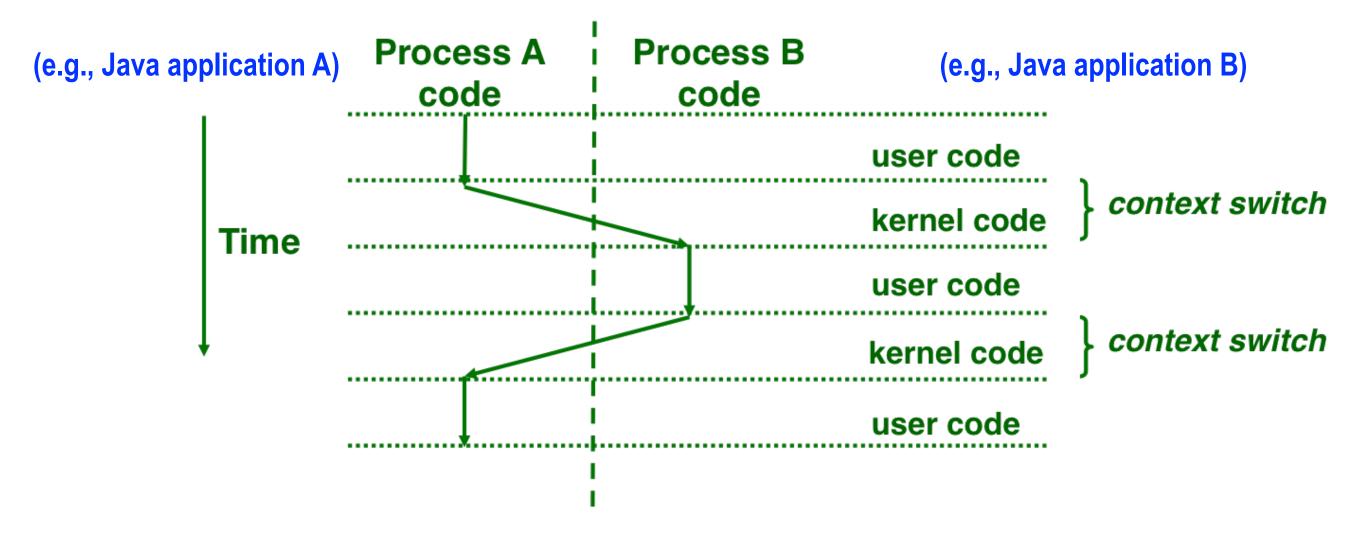


How does a process run on a single core?

Processes are managed by OS kernel

 Important: the kernel is not a separate process, but rather runs as part of some user process

Control flow passes from one process to another via a context switch



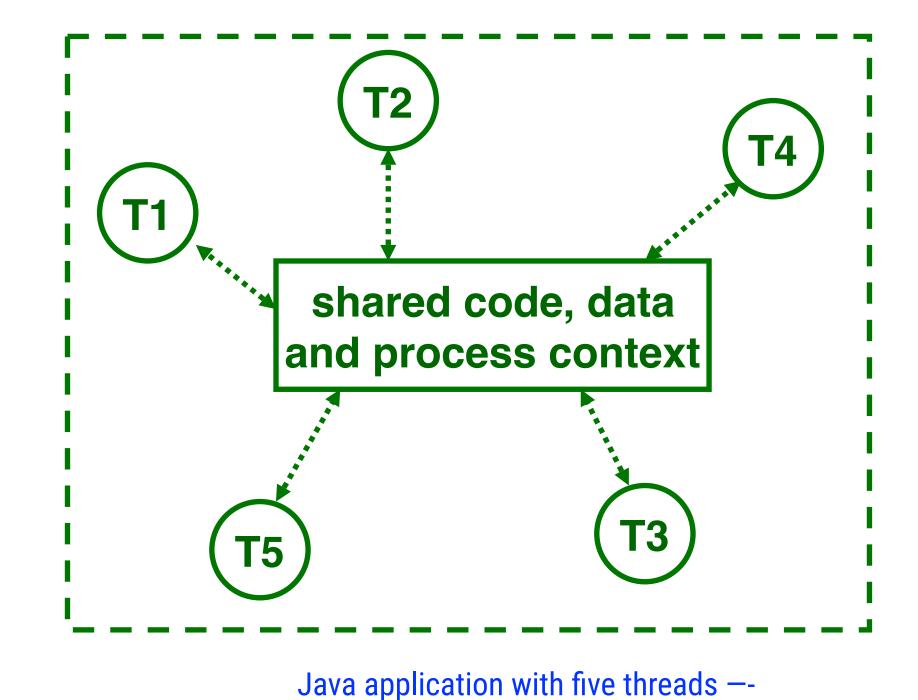
Context switches between two processes can be very expensive!

Source: COMP 321 lecture on Exceptional Control Flow (Alan Cox)



What happens when we execute a Java program

- A Java program executes in a single Java Virtual Machine (JVM) process with multiple threads
- Threads associated with a single process can share the same data
- Java main program starts with a single thread (T1), but can create additional threads (T2, T3, T4, T5) via library calls
- Java threads may execute concurrently on different cores, or may be context-switched on the same core

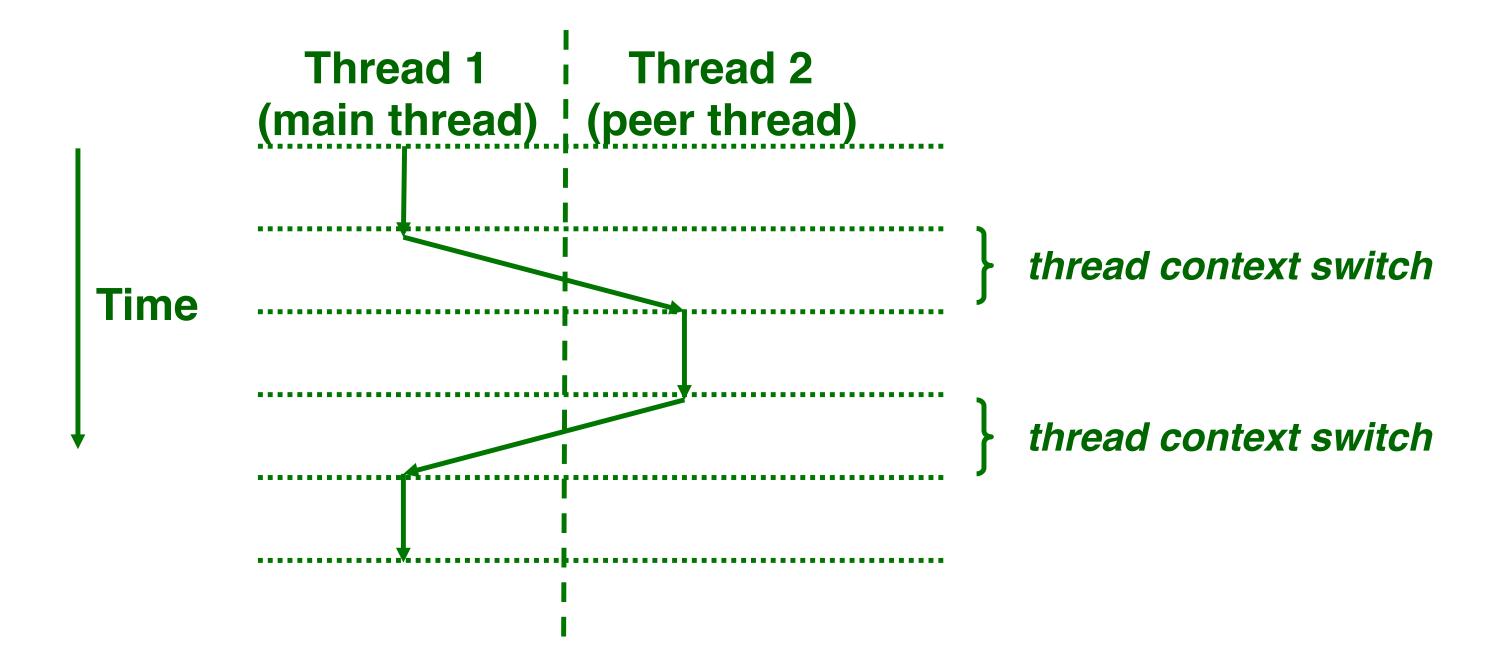


T1, T2, T3, T4, T5 — all of which can access a common set of shared objects

Figure source: COMP 321 lecture on Concurrency (Alan Cox)



Thread-level Context Switching on the same processor core

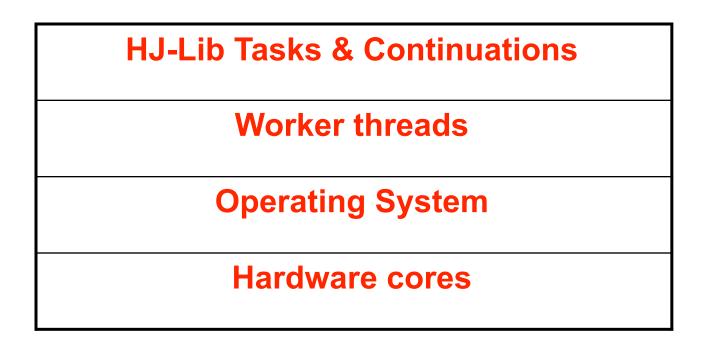


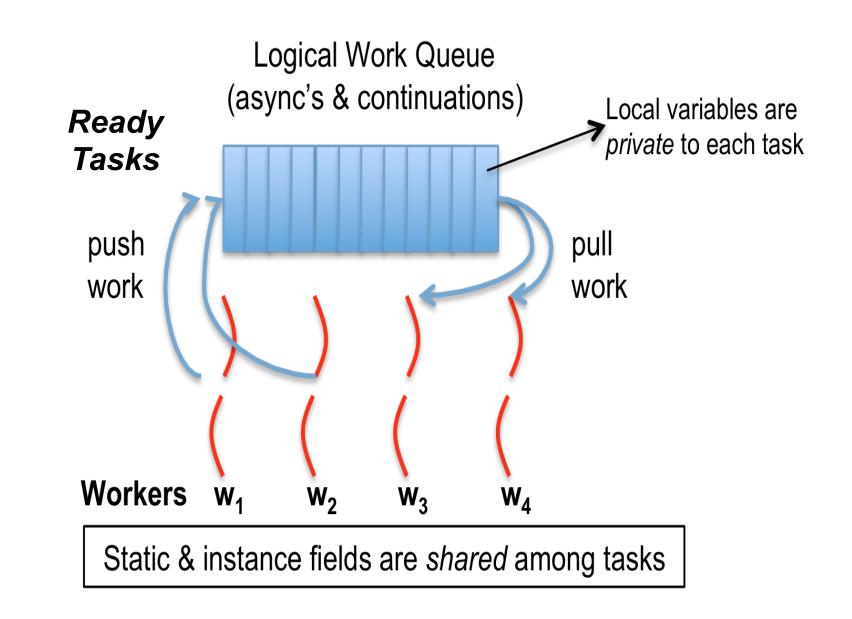
- Thread context switch is cheaper than a process context switch, but is still expensive (just not "very" expensive!)
- It would be ideal to just execute one thread per core (or hardware thread context) to avoid context switches

Figure source: COMP 321 lecture on Concurrency (Alan Cox)



Now, what happens is a task-parallel Java program (e.g., HJ-lib, Java Fork/Join, etc.)





- HJ-lib runtime creates a small number of worker threads, typically one per core
- Workers push new tasks and "continuations" into a logical work queue
- Workers pull task/continuation work items from logical work queue when they are idle (remember greedy scheduling?)



Task-Parallel Model: Checkout Counter Analogy



Think of each checkout counter as a processor core

Image sources: http://www.deviantart.com/art/Randomness-20-178737664, http://www.wholefoodsmarket.com/blog/whole-story/new-haight-ashbury-store



Task-Parallel Model: Checkout Counter Analogy



- Think of each checkout counter as a processor core
- And of customers as tasks

Image sources: http://www.deviantart.com/art/Randomness-20-178737664, http://www.wholefoodsmarket.com/blog/whole-story/new-haight-ashbury-store



All is well until a task blocks ...



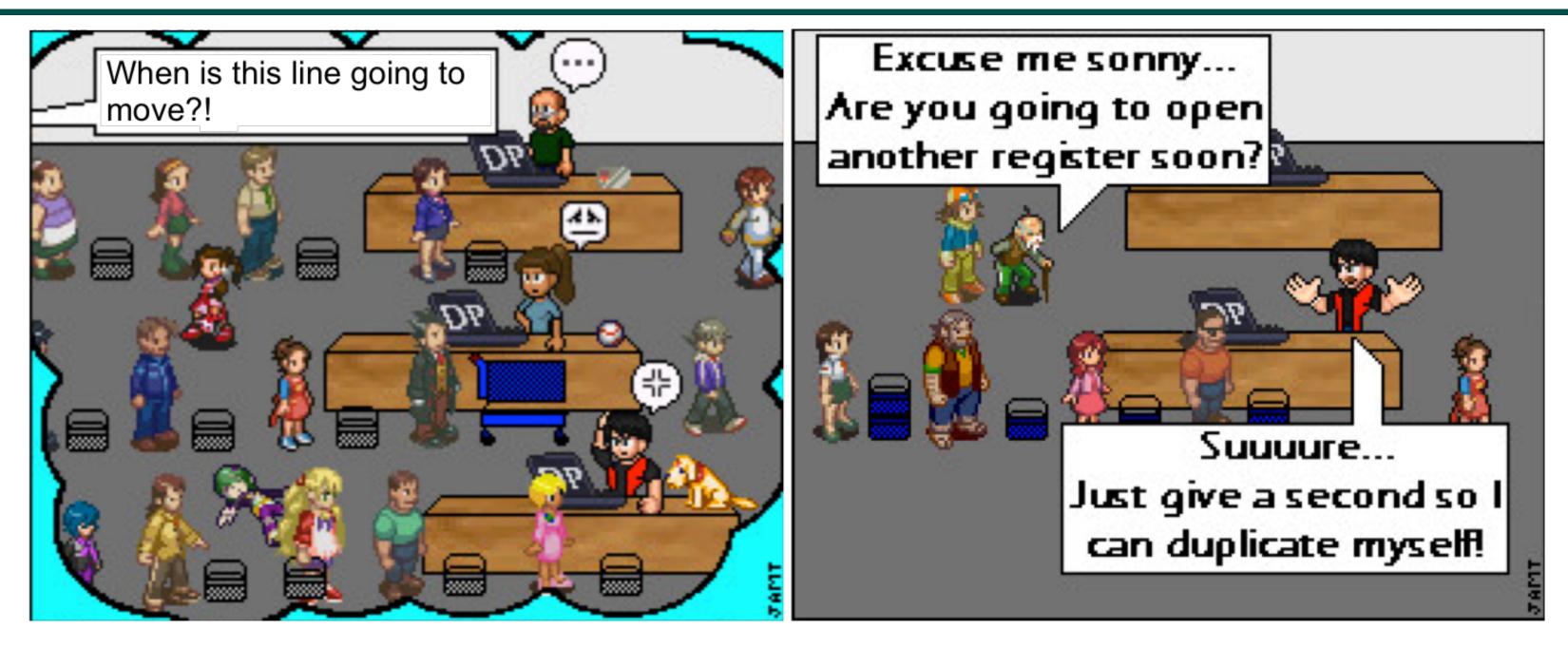


- A blocked task/customer can hold up the entire line
- What happens if each checkout counter has a blocked customer?

source: http://viper-x27.deviantart.com/art/Checkout-Lane-Guest-Comic-161795346



Approach 1: Create more worker threads (as in HJ-Lib's Blocking Runtime)

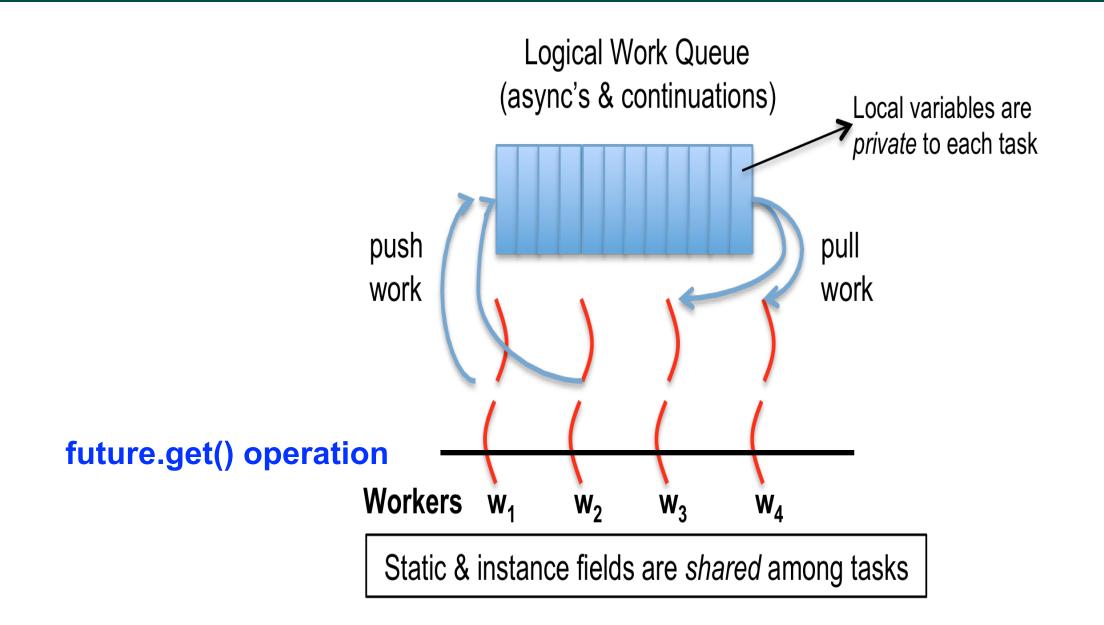


- Creating too many worker threads can exhaust system resources (OutOfMemoryError)
- Leads to context-switch overheads when blocked worker threads get unblocked

source: http://www.deviantart.com/art/Randomness-5-90424754



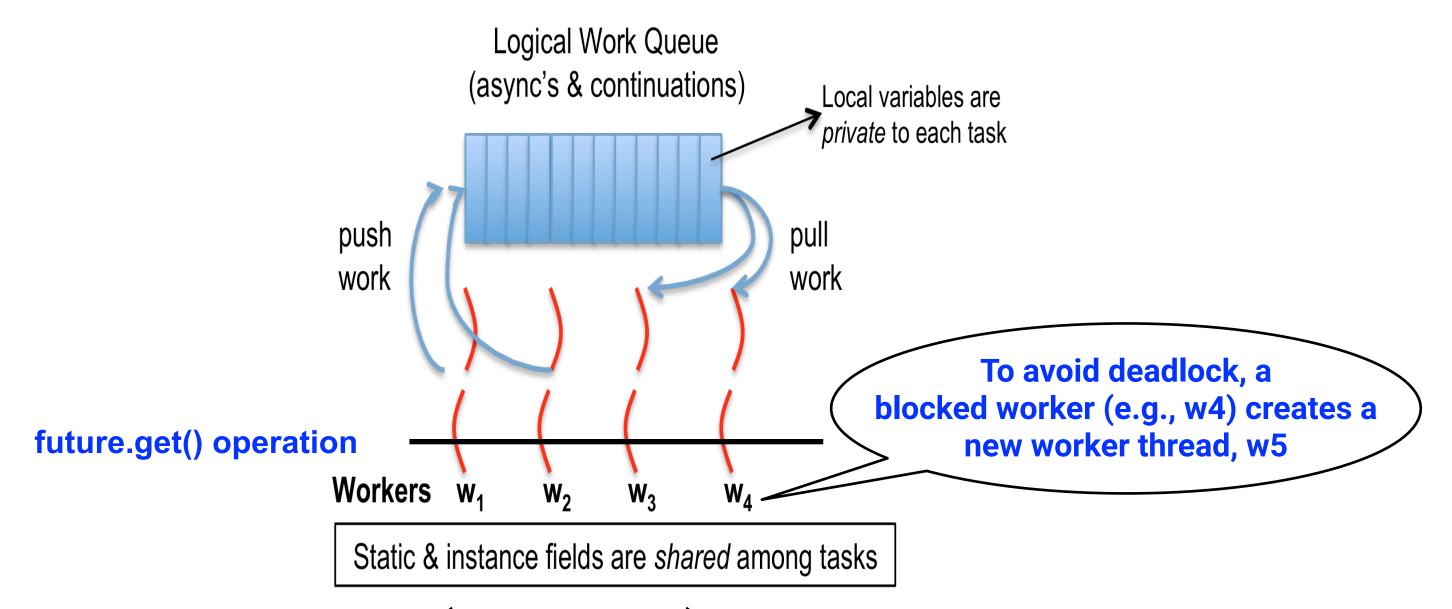
Blocking Runtime (contd)



- Assume that there are five tasks (A1 ... A5)
- Q: What happens if four tasks (say, A1 ... A4) executing on workers w1 ... w4 all wait
 on the same future that's computed by A5?



Blocking Runtime (contd)



- Assume that there are five tasks (A1 ... A5)
- Q: What happens if four tasks (say, A1 ... A4) executing on workers w1 ... w4 all wait on the same future that's computed by A5?
- A: Deadlock! (All four tasks will wait for task A5 to compute the future.)
- Blocking Runtime's solution to avoid deadlock: keep task blocked on worker thread, and create a new worker thread when task blocks

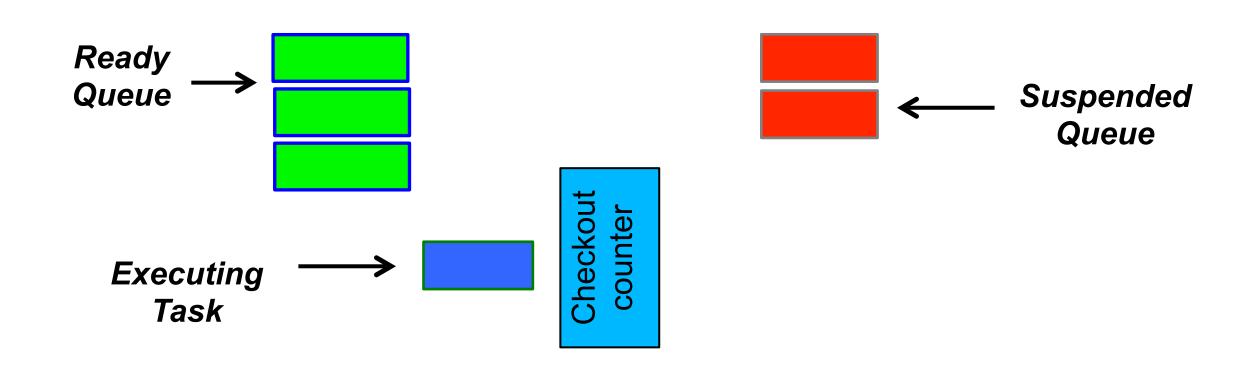


Blocking Runtime (contd)

- Examples of blocking operations
 - End of finish
 - Future get
 - Barrier next
- Approach: Block underlying worker thread when task performs a blocking operation, and launch an additional worker thread
- Too many blocking operations can result in exceptions and/or poor performance, e.g.,
 - java.lang.IllegalStateException: Error in executing blocked code! [89 blocked threads]
- Maximum number of worker threads can be configured if needed
 - HjSystemProperty.maxThreads.set(100);



Approach 2: Suspend task continuations at blocking points (as in HJ-Lib's Cooperative Runtime)



- Upon a blocking operation, the currently executing tasks suspends itself and yields control back to the worker
- Task's continuation is stored in the suspended queue and added back into the ready queue when it is unblocked
- Pro: No overhead of creating additional worker threads
- Con: Need to create continuations (enabled by -javaagent option)

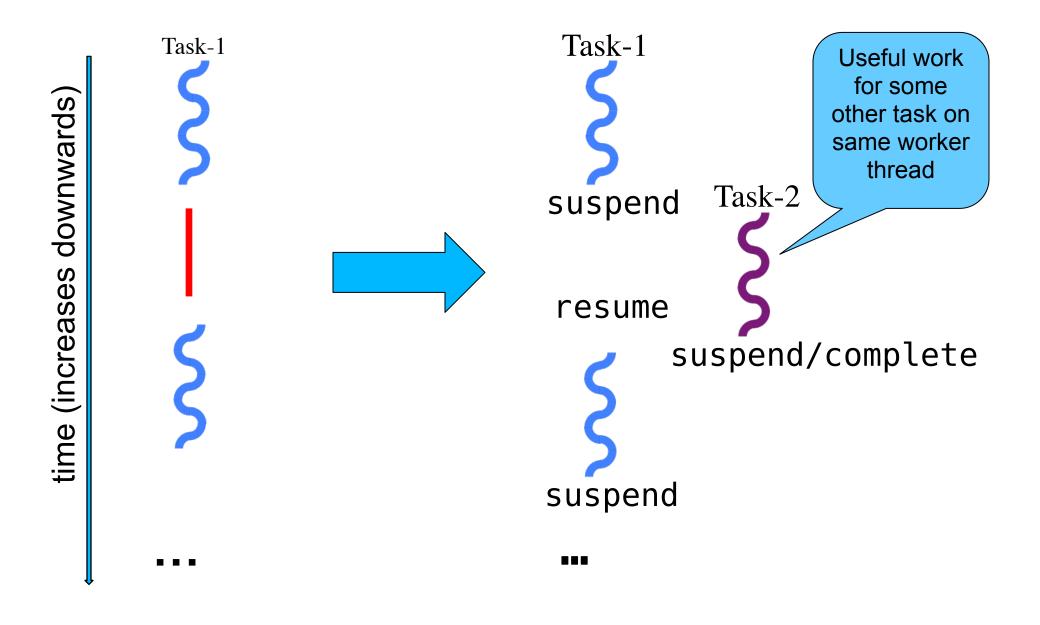


Continuations

- A continuation can be a point immediately following a blocking operation, such as an end-finish, future get(), barrier/phaser next(), etc.
- Continuations are also referred to as task-switching points
 - Program points at which a worker may switch execution between different tasks (depends on scheduling policy)

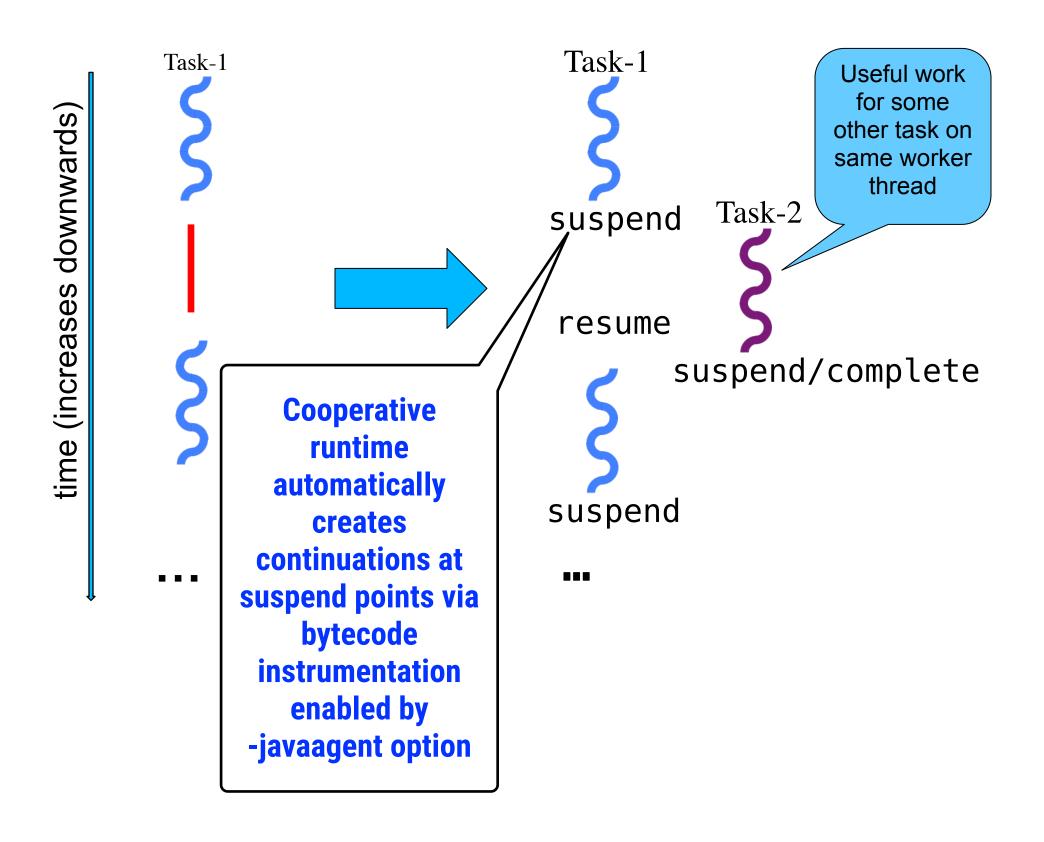


Cooperative Scheduling (view from a single worker)



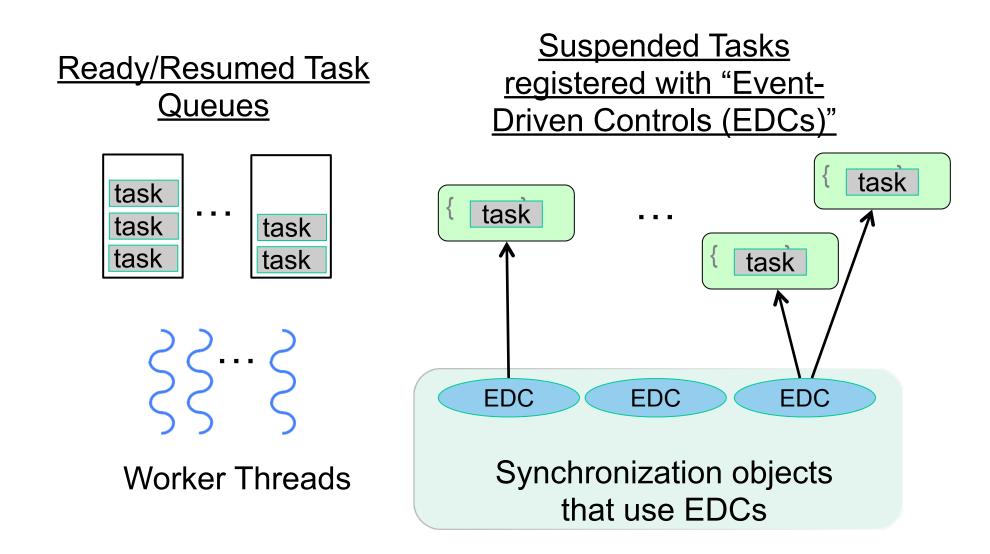


Cooperative Scheduling (view from a single worker)





HJ-lib's Cooperative Runtime (contd)



Any operation that contributes to unblocking a task can be viewed as an event e.g., task termination in finish, return from a future, signal on barrier, put on a data-driven-future, ...



Why are Data-Driven Tasks (DDTs) more efficient than Futures?

- Consumer task blocks on get() for each future that it reads, whereas asyncAwait does not start
 execution until all Data-Driven Futures (DDFs) are available
 - An "asyncAwait" call does not block the worker, unlike a future.get()
 - No need to create a continuation for asyncAwait; a data-driven task is directly placed on the Suspended queue by default
- Therefore, DDTs can be executed on a Blocking Runtime without the need to create additional worker threads, or on a Cooperative Runtime without the need to create continuations



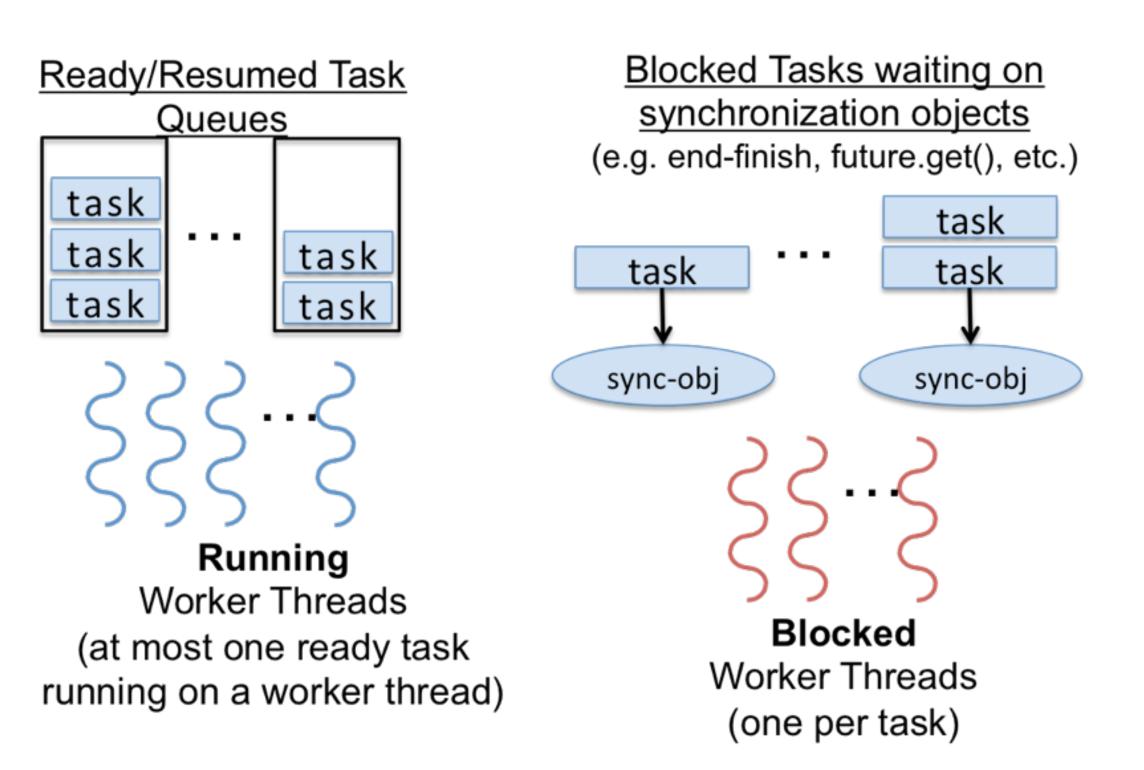
Abstract vs Real Performance in HJ-Lib

Abstract Performance

 Abstract metrics focus on operation counts for WORK and CPL, regardless of actual execution time

Real Performance

 HJlib uses ForkJoinPool implementation of Java Executor interface with Blocking or Cooperative Runtime (default)





Summary

- Functional approach is great, but sometimes can lead to performance issues
- Knowing what is happening "under the covers" can help you design better performing algorithms
- Cutoff strategy is a great way to balance parallelism and overhead for recursive task parallelism
- Depending on the runtime, your task parallel program may have some tasks that could block the whole CPU thread
- Processes are more expensive than threads, threads are more expensive than tasks
- In order to deliver performance, most runtimes assume they have a full control of OS threads
 - Don't mix Java parallel Streams with HJLib constructs
 - Don't mix Java threads with HJLib tasks and/or Java parallel Streams
 - An HJ runtime instance inside of its own Java thread is usually OK
 - A Java parallel Stream computation inside an HJ task is usually OK

