

# Parallel Programming

## Recitation Session 5

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# Executive Summary

- Data partitioning with parallel matrix multiplication
- How to manage threads

# Outline

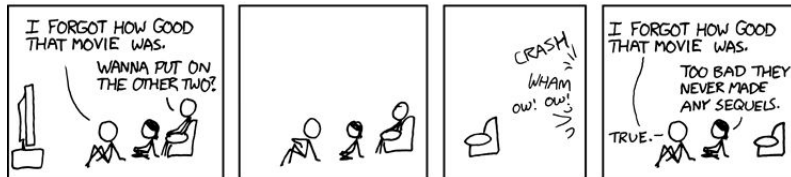
## 1 Matrix Multiplication

## 2 Thread Pools

# Parallel Matrix Multiplication

## Data partitioning based on

- Output matrix **C**
- Input matrix **A** and input matrix **B**



Source: <http://www.xkcd.com>

# Output Partitioning (2 Threads)

```
// Thread 0
for (i=0; i<N/2; i++) {
    for (j=0; j<N; j++) {
        for (k=0; k<N; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}

// Thread 1
for (i=N/2; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=0; k<N; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}
```

# Input Partitioning: Error in Last Week's Slides

```
// Thread 0
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=0; k<N/2; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}

// Thread 1
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=N/2; k<N; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}
```

# Input Partitioning: Error in Last Week's Slides

```
// Thread 0
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=0; k<N/2; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}

// Thread 1
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=N/2; k<N; k++) {
            c[i][j] += a[i][k] * b[k][j];
        }
    }
}
```

**Error:** Possible  
race condition  
when writing  
`c[i][j]`

# Input Partitioning: Locking

```
Object[][] lock; // Lock "matrix"
// Thread 0
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        synchronized(lock[i][j]) {
            for (k=0; k<N/2; k++) {
                c[i][j] += a[i][k] * b[k][j];
            }
        }
    }
}
// Thread 1
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        synchronized(lock[i][j]) {
            for (k=N/2; k<N; k++) {
                c[i][j] += a[i][k] * b[k][j];
            }
        }
    }
}
```



# Overhead?

- A complete row is locked
- Actual lock contention will be moderate to low
- In practice the slow-down – with respect to output partitioning – is moderate (only a few percent)

# Input Partitioning: Fine-Grain Locking

```

Object[][] lock; // Lock "matrix"
// Thread 0
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=0; k<N/2; k++) {
            synchronized(lock[i][j]) {
                c[i][j] += a[i][k] * b[k][j];
            }
        }
    }
}
// Thread 1
for (i=0; i<N; i++) {
    for (j=0; j<N; j++) {
        for (k=N/2; k<N; k++) {
            synchronized(lock[i][j]) {
                c[i][j] += a[i][k] * b[k][j];
            }
        }
    }
}

```

Significant overhead:

About 3 times slower than  
coarse-grain locking

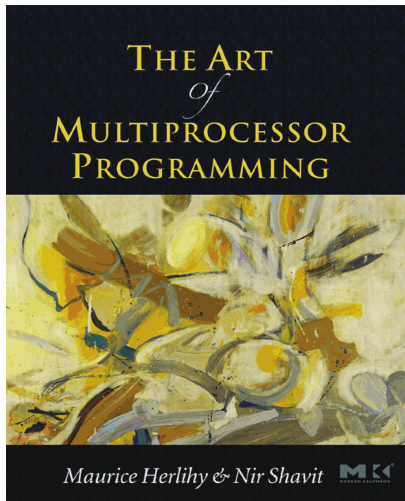
# Outline

1 Matrix Multiplication

**2 Thread Pools**

# The Art of Multiprocessor Programming

Source of the following material about Thread Pools: “The Art of Multiprocessor Programming” by Maurice Herlihy and Nir Shavit



# Thread Overhead

- Threads require resources
  - Memory for stacks
  - Setup, teardown
- Scheduler overhead
- Worse for short-lived threads

# Thread Pools

- More sensible to keep a pool of long-lived threads
- Threads assigned short-lived tasks
  - Runs the task
  - Rejoins pool
  - Waits for next assignment

# Thread Pool = Abstraction

- Insulate programmer from platform
  - Big machine, big pool
  - And vice-versa
- Portable code
  - Runs well on any platform
  - No need to mix algorithm/platform concerns

# ExecutorService Interface

In `java.util.concurrent`:

- Task = Runnable object
  - If no result value expected
  - Calls `run()` method.
- Task = `Callable<T>` object
  - If result value of type T expected
  - Calls `T call()` method.



# Future<T>

```
Callable<T> task = ...;  
...  
Future<T> future = executor.submit(task);  
...  
T value = future.get();
```

- Submitting a `Callable<T>` task returns a `Future<T>` object
- The `Future`'s `get()` method blocks until the value is available

# Future<?>

```
Runnable task = ...;  
...  
Future<?> future = executor.submit(task);  
...  
future.get();
```

- Submitting a Runnable task returns a Future<?> object
- The Future's get() method blocks until the computation is complete

# Note

- Executor Service submissions are purely advisory in nature
- The executor
  - Is free to ignore any such advice
  - And could execute tasks sequentially ...

# Fibonacci

$$F(n) := \begin{cases} 1, & n = 0 \\ 1, & n = 1 \\ F(n-1) + F(n-2), & n > 1 \end{cases}$$

- Potential parallelism
- Dependencies

# Disclaimer

- This Fibonacci implementation is very inefficient
  - So don't deploy it!
- But illustrates our point
  - How to deal with dependencies

# Multithreaded Fibonacci

```
class FibTask implements Callable<Integer> {
    static ExecutorService exec =
        Executors.newCachedThreadPool();
    int arg;
    public FibTask(int n) {
        arg = n;
    }
    public Integer call() {
        if (arg > 2) {
            Future<Integer> left =
                exec.submit(new FibTask(arg-1));
            Future<Integer> right =
                exec.submit(new FibTask(arg-2));
            return left.get() + right.get();
        } else {
            return 1;
        }
    }
}
```

# Summary

## Enjoy your vacation!

