



# Towards a better Parallelism

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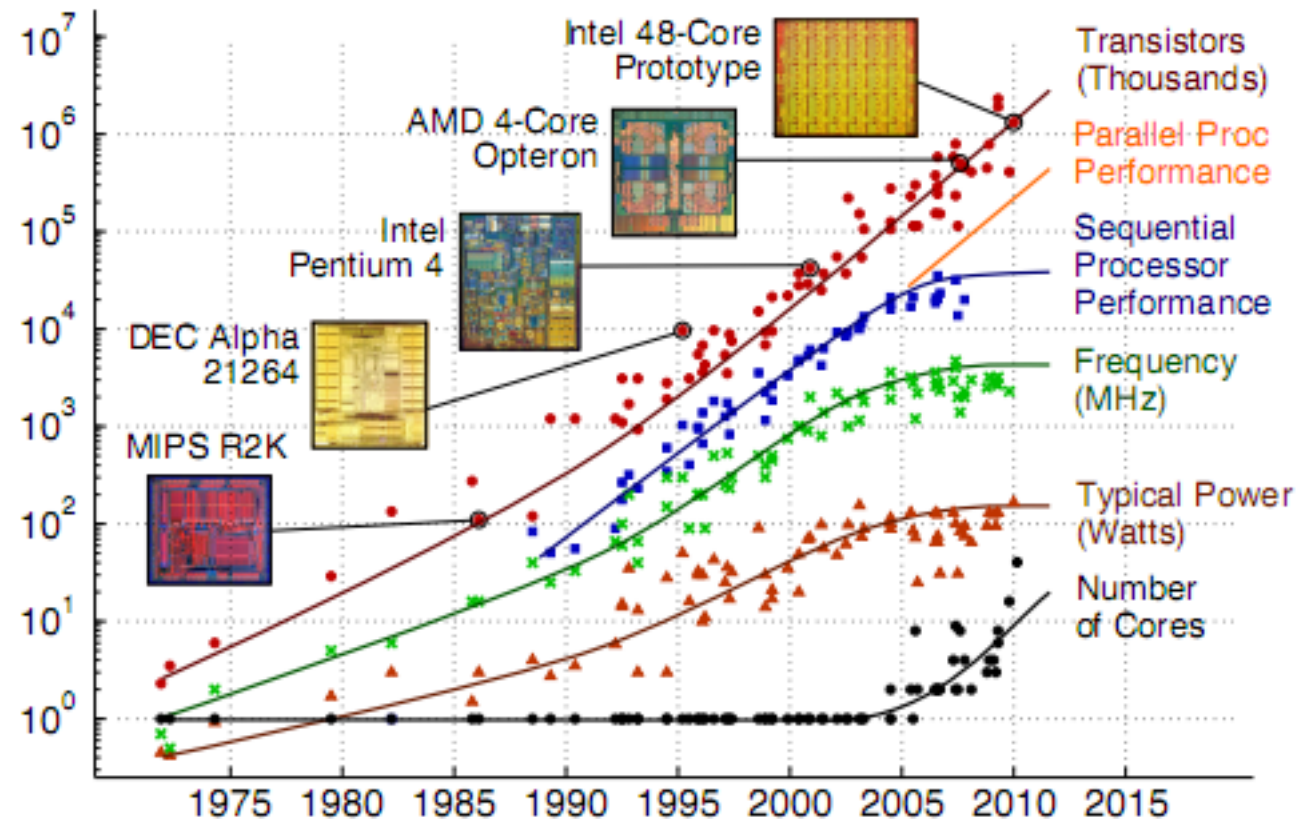
# Safe Harbor Statement

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# Agenda of the day ...

- 1 ➤ Future of Hardware
- 2 ➤ Software followed Hardware
- 3 ➤ Productive Parallelism
- 4 ➤ Performance Consideration
- 5 ➤ A glimpse of Parallel Streams
- 6 ➤ Where we are heading

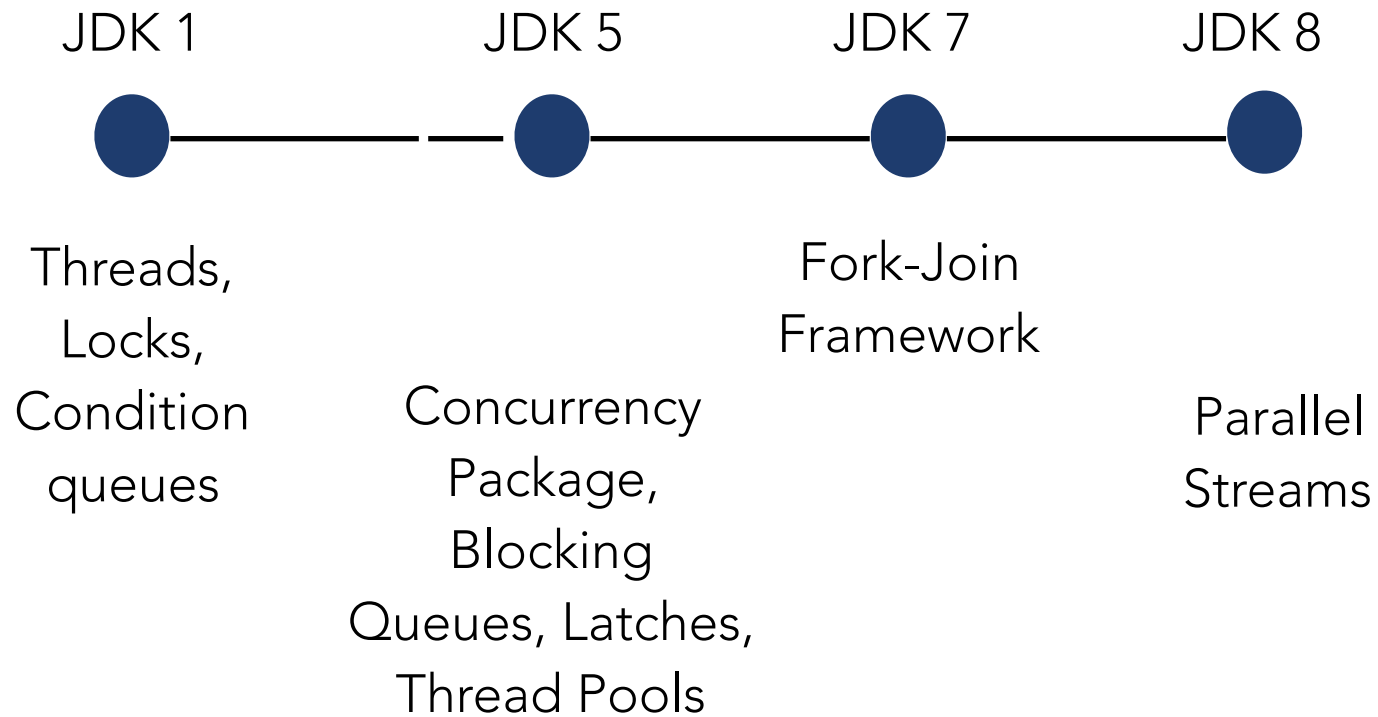
# Where future is trending ...



Data partially collected by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond

Prepared by C. Batten - School of Electrical and Computer Engineering - Cornell University - 2005 - retrieved Dec 12 2012 - <http://www.csl.cornell.edu/courses/ece5950/handouts/ece5950-overview.pdf>

# Software followed Hardware



# What is Parallelism ...

- Breaking the tasks into sub-tasks and get results faster.
- Faster and only faster (with more resources)
- Golden rule
  - Analyze -> Implement -> Test -> Repeat
- If results are not better (?), drop the idea.



Reference: Internet Movie Firearms Database ([http://www.imfdb.org/wiki/Matrix,\\_The](http://www.imfdb.org/wiki/Matrix,_The))

# Any overhead ?

- Parallel computation always involves “more” task than its alternative sequential computation.
- Always a slow startup.

DECOMPOSE THE PROBLEM  
INTO TASK

LAUNCH TASK, MANAGE TASK,  
WAIT FOR COMPLETION

COMBINE RESULTS

+

SOLVE THE PROBLEM

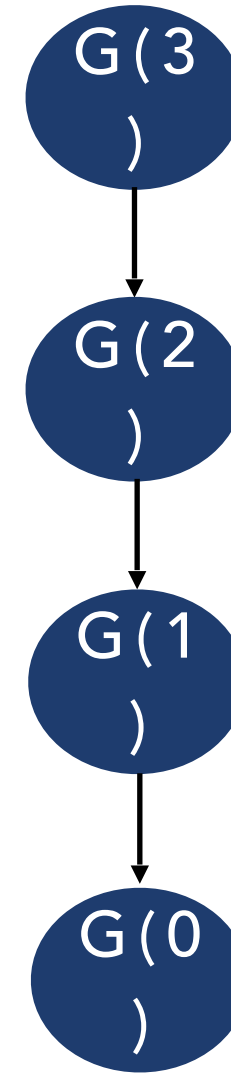


# Data Flow Dependency

- Consider a problem and look at the data flow dependency.
- If it's a real iterative problem - Bad idea to think of parallelism.

$$G(N) = F(G(N-1)), \text{ IF } N > 0$$

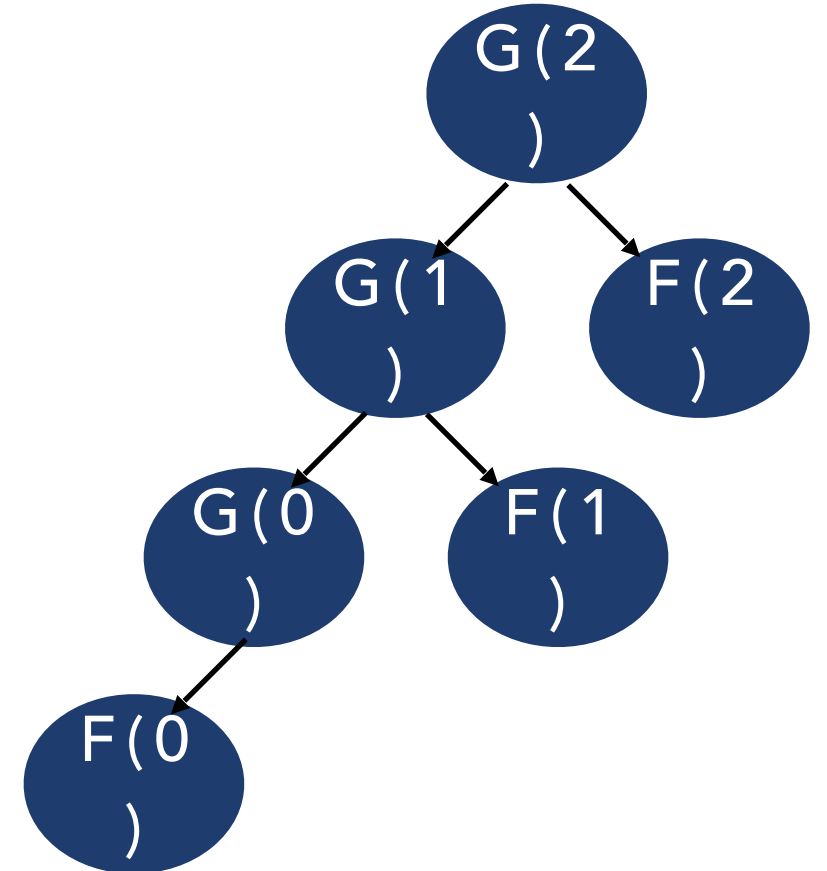
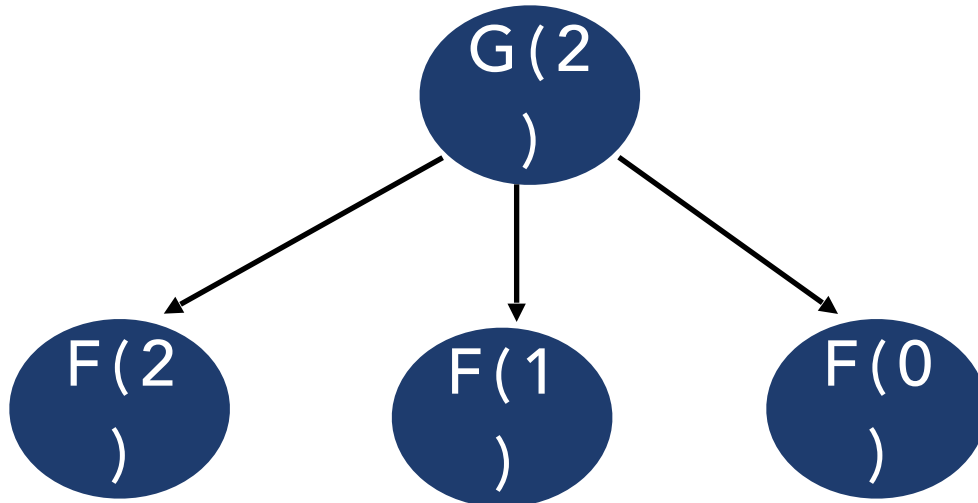
$$G(0) = F(0)$$



# Parallelism can be exploited

$$G(N) = F(N) + G(N-1), \text{ IF } N > 0$$

$$G(0) = F(0)$$



# Let's reach out to the parallel computation

```
int sum (int[] arr) {  
    int sum = 0;  
    for (int i : arr)  
        sum = sum + i;  
    return sum;  
}
```

Sequential approach. Accumulator Pattern.



---

```
int sum (int[] arr) {  
    int sum = 0;  
    INT MID = arr.length/2;  
    CONCURRENT {  
        { for (int i=0; i< MID; i++)  
            sum = sum + arr[i]; }  
        { for (int i=MID; i<arr.length; i++)  
            sum = sum + arr[i]; }  
    }  
    return sum;  
}
```

//need to atomic

//need to atomic

# Cont...

- The last solution is bad. Because we are spending time to safeguard the data.
- Right approach (in the order) :-
  - Don't share
  - Don't mutate
  - Coordinate access

# Cont...

```
int sum (int[] arr) {  
    int LEFT = 0, RIGHT = 0;  
    int MID = arr.length/2;  
    CONCURRENT {  
        { for (int i=0; i< MID; i++)  
            ATOMIC { LEFT = LEFT + arr[i]; } }  
        { FOR (int i=MID;i<arr.length; i++)  
            ATOMIC { RIGHT = RIGHT + arr[i]; } }  
    }  
    return LEFT + RIGHT;  
}
```

Fairly a good solution

Generic solution is required.

Leverage all the cores, split till sequential gives the better result.

# Pseudo-Code for generic solution

*// PSEUDOCODE*

```
Result solve(Problem problem) {  
    if (problem.size < SEQUENTIAL_THRESHOLD)  
        return solveSequentially(problem);  
    else {  
        Result left, right;  
        INVOKE-IN-PARALLEL {  
            left = solve(extractLeftHalf(problem));  
            right = solve(extractRightHalf(problem));  
        }  
        return combine(left, right);  
    }  
}
```

# Introduction to Fork-Join Framework

- JDK7 introduced Fork And Join Framework.
- Works on Divide and Conquer
- Implements work-stealing algorithm.

```
if (my portion of the work is small enough)
    do the work directly

else
    split my work into two pieces
    invoke the two pieces and wait for the results
```

# Introduction of ParallelStreams in Java8

- Build on top of FnJ Framework.
- FnJ was not so easy to use.
- You just have to say “.parallel()”

Arrays

```
.stream(a)  
.parallel()  
.filter(s -> s > 5000000)  
.sum();
```

Magical, but unfortunately not...



# Parallel Stream consideration

# Performance Considerations

- Splitting of Source
  - source should be splittable
  - cost of computation for split
  - evenness, predictability of split
- Some source are awesome to split and some are not.
- Arrays (Good), LinkedList (Not good), Tree (so-so)
- Not only splitting, combining result should be considered.

# Minimum Data to get benefit (Run Size: 500K)

THREAD/ THRESHOLD	500K	50K	5K	500	50
2 thread	1.0	1.07	1.02	0.82	0.2
4 thread	0.88	3.02	3.2	2.22	0.43
8 thread	1.0	5.29	5.73	4.53	2.03
32 thread	0.98	10.46	17.21	15.34	6.49

# The NQ model

- Simple model for parallel performance
  - N: No of elements
  - Q: amount of work per item
  - $N \times Q > 10,000$  -> chance for parallelism
  - Important for say “minimum data”

# Locality Issues

- Locality is dangerous.
- CPU should be busy in good work
  - Cache misses are not good.
- `Streams.of(int[].sum())` vs `Streams.of(Integer[].sum())`

Speed Up	N=1K	N=10K	N=1M
int	1x	6.2x	7.9x
Integer	-4.9x	1.5x	3.5x

# Order insensitive operations

- Least parallel friendly.
- Always remember the bottle-neck principles.

```
OptionalLong m = Arrays.stream(a).parallel() // killer
    .filter(s -> s > 5000000)
    .filter(x -> x == 5465759)
    .findFirst();
```

# Must visit reference

- Java Concurrency in Practice (Book) - Brian Goetz
- [Concurrency to Parallelism](#)
- [Introduction to Streams Library](#)
- [Java Fork-Join Framework - Doug Lea](#)