

Parallel Algorithms (vs. serial algorithms)

Random-access machine model used for serial algorithms
Many models for parallel algorithms, no general agreement.

Model used: Dynamic Multithreading

- appropriate for multicore machines, shared memory programming
- not appropriate for distributed memory programs

Ex:

```

      Fib (n)
A  [ if n < 2
    [ then return n
B  [ x ← spawn Fib (n-1)
    [ y ← spawn Fib (n-2)
empty [ sync
C  [ return (x+y)

```

spawn, sync, or return terminate
current thread
threads

A: includes $n-1$ computation
- upto spawn

B: - includes $n-2$ computation
- upto spawn

empty: ignored now

C: - includes $x+y$ computation
- upto return, after sync

spawn: - subroutine can execute at same time as parent

sync: - wait until all children are done

Description of logical parallelism, not actual (does not describe # processors)

A scheduler determines how to map dynamically unfolding execution onto processors.

Serial instruction stream: when in a loop, chain of subsequent instructions.

Logical serial instruction stream is actually not executed sequentially by a processor \rightarrow instruction-level parallelism

not a focus here. The focus is on logical parallelism.

Parallel instruction stream:

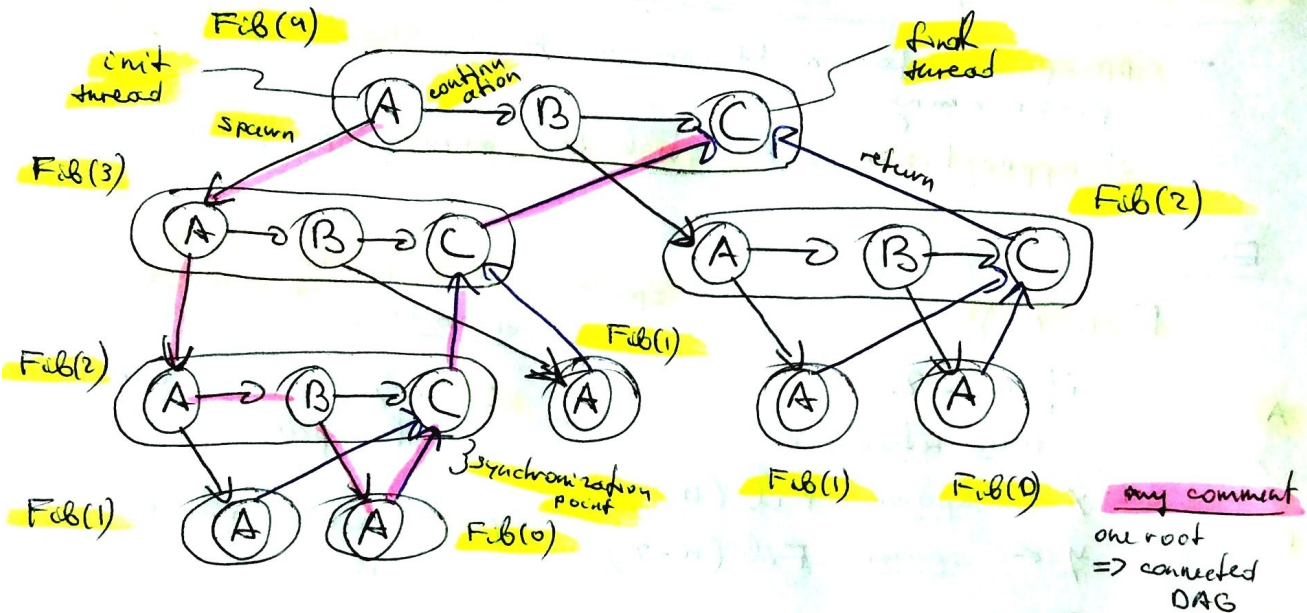
DAG

Multi-threaded computation

Parallel instruction stream = DAG

vertices are threads: maximal sequence of instructions not containing parallel control (spawn, sync, return)

edges: spawn, return, continuation



Performance measures

T_p = running time on p processors

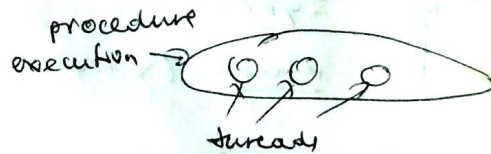
T_1 = work = serial time (just like getting rid of spawn, sync)

T_{∞} = critical path length = longest path in DAG

Ex: Fib(4) $T_1 = 17$ (assume each thread is unit time)

$T_{\infty} = 8$ (longest path in DAG threads that must be sequentially executed)

if unit-time threads



my comment
longest path across topo-
logically sorted DAG
after threads can be parallel

6.046

Lecture 22

this model does not take
communication into account

(2)

Lower bound on T_p

$$T_p \geq \frac{T_1}{P}$$

- P proc can do $\leq P$
work in 1 step

- if ~~contradiction~~,

$$T_p < \frac{T_1}{P},$$

Processor can do $> P$
work in 1 step.

my comment

Suppose $T_p < \frac{T_1}{P}$,
then \exists a task that
is not executed.
contradiction to
 T_p def.

$$T_p \geq T_\infty$$

- P processors can't do more work than
 \approx processors

Speedup

$$T_1/T_p = \text{speedup of } p \text{ processors}$$

$T_1/T_p = \Theta(P) \Rightarrow$ linear speedup each processor contributed
within a constant factor is measure of full
support

$T_1/T_p = P \Rightarrow$ perfect linear speedup

$T_1/T_p > P \Rightarrow$ super-linear speedup

NOT Possible in this model ($T_p < \frac{T_1}{P}$)
contr.

In other models possible
(e.g. caching effects)

Max possible speedup, given T_1, T_∞ , is $T_1/T_\infty = \text{parallelism}$

= average amount of
work that can be done in
parallel along each step
of critical path.

adding more
processors does not
improve speedup

$$= \bar{P}$$

Scheduling

Map computation to P processors

Done by runtime system (scheduler algorithm)
typically language runtime system

On-line schedulers are complex (randomized schedulers with guarantees!!!)
Illustrate ideas using off-line scheduler.

Greedy scheduler (P processors) ~~states~~

in DAG: cannot execute a node until nodes preceding it are executed

- Do as much as possible on every step.

→ do not guess if something is worth delaying

- Complete step: $\geq P$ threads ready to run.

execute any P threads. May be not optimal.

There maybe a particular thread, if executed now, enables more parallelism later.

- Incomplete step: $< P$ threads ready to run.

Execute all of them.

!! Scheduling optimally a DAG on P processors is NP-complete.

Theorem (Graham, Brent):

A greedy scheduler executes any computation with work T_1 and critical path length T_∞ in time

$$T_P \leq \frac{T_1}{P} + T_\infty \leq 2OPT \quad T_P \geq \frac{T_1}{P} \Rightarrow 2OPT$$

on a computer with P processors.

$$T_P \geq T_\infty$$

2-competitive.

$$OPT \geq$$

$$\max\left(\frac{T_1}{P}, T_\infty\right)$$

my comment

$$\frac{T_1}{P} \geq T_\infty \Rightarrow \frac{T_1}{P} + T_\infty \leq 2 \frac{T_1}{P} \leq 2OPT$$

6.046

Lecture 22

my comment

min # available

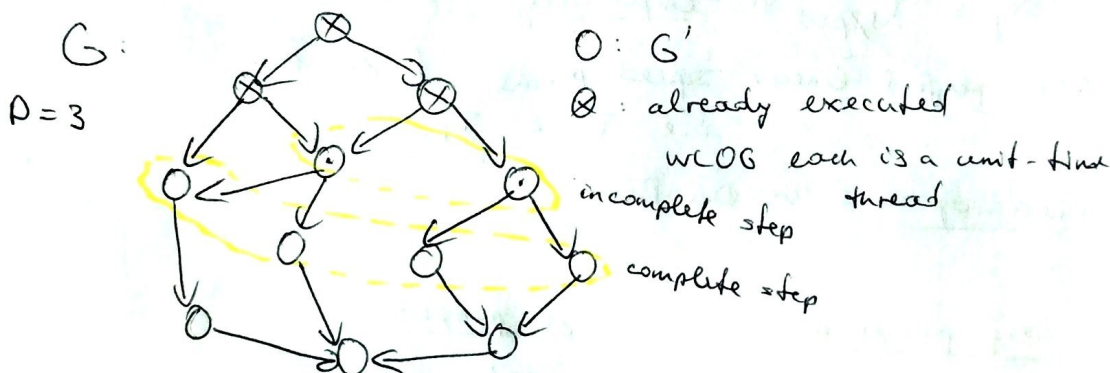
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Proof: # complete steps $\leq \frac{T_1}{P}$

since otherwise more than T_1 work would be done.

← assume p threads each step then at most $\frac{T_1}{p}$ ~~steps~~

Consider an incomplete step, and let G' be subgraph of G that remains to be executed.



Threads \otimes with in-degree 0 in G' are ready to be executed

The critical path length of G' is reduced by 1

\Rightarrow # incomplete steps $\leq T_\infty$

~~very~~

$$\Rightarrow T_P \leq \frac{T_1}{P} + T_\infty$$

⏟

foundational theorem of Scheduling!

Corollary: $\frac{T_1}{T_P} = \Theta(P)$

Linear speedup when $P = O(\bar{P})$

with greedy scheduler

← bound of P grows as T_1 and T_∞ grow

$$\bar{P} = \frac{T_1}{T_\infty} \Rightarrow P = O\left(\frac{T_1}{T_\infty}\right) \Rightarrow T_\infty = O\left(\frac{T_1}{P}\right)$$

$$\text{Thus } T_P \leq \frac{T_1}{P} + O\left(\frac{T_1}{P}\right) = O\left(\frac{T_1}{P}\right)$$



✓

critical path must include one of \otimes nodes and each of them ~~may~~ cannot be reached if not included as start node in the critical path of G' (a DAG) may not be connected but each "root" is included because the step is incomplete

when running on fewer processor than \bar{P} , can get speedup.

Cilk

Randomized online scheduler

$$\mathbb{E}[T_p] = T_1/p + O(T_\infty) \text{ provably}$$

$$T_p \approx T_1/p + T_\infty \text{ empirically}$$

Near-perfect linear speedup if $P \ll \bar{P}$
i.e. $T_\infty \ll T_1/p$

Chess programs vs. DeepBlue

Orig. program

$$T_{32} = 65 \text{ sec}$$

$$T_1 = 2048$$

$$T_\infty = 1$$

Opt. program

$$T'_{32} = 40 \text{ sec}$$

Reject.

$$T'_1 = 1024$$

$$T'_\infty = 8$$

$$T_{32} = T_1/32 + T_\infty = 65$$

$$T'_{32} = T'_1/32 + T'_\infty = 40$$

Extrapolate on a larger machine

$$T_{512} = T_1/512 + T_\infty = 5$$

$$T'_{512} = T'_1/512 + T'_\infty = 10$$