

Analysis of Algorithms

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CSCI 570

Lecture 5

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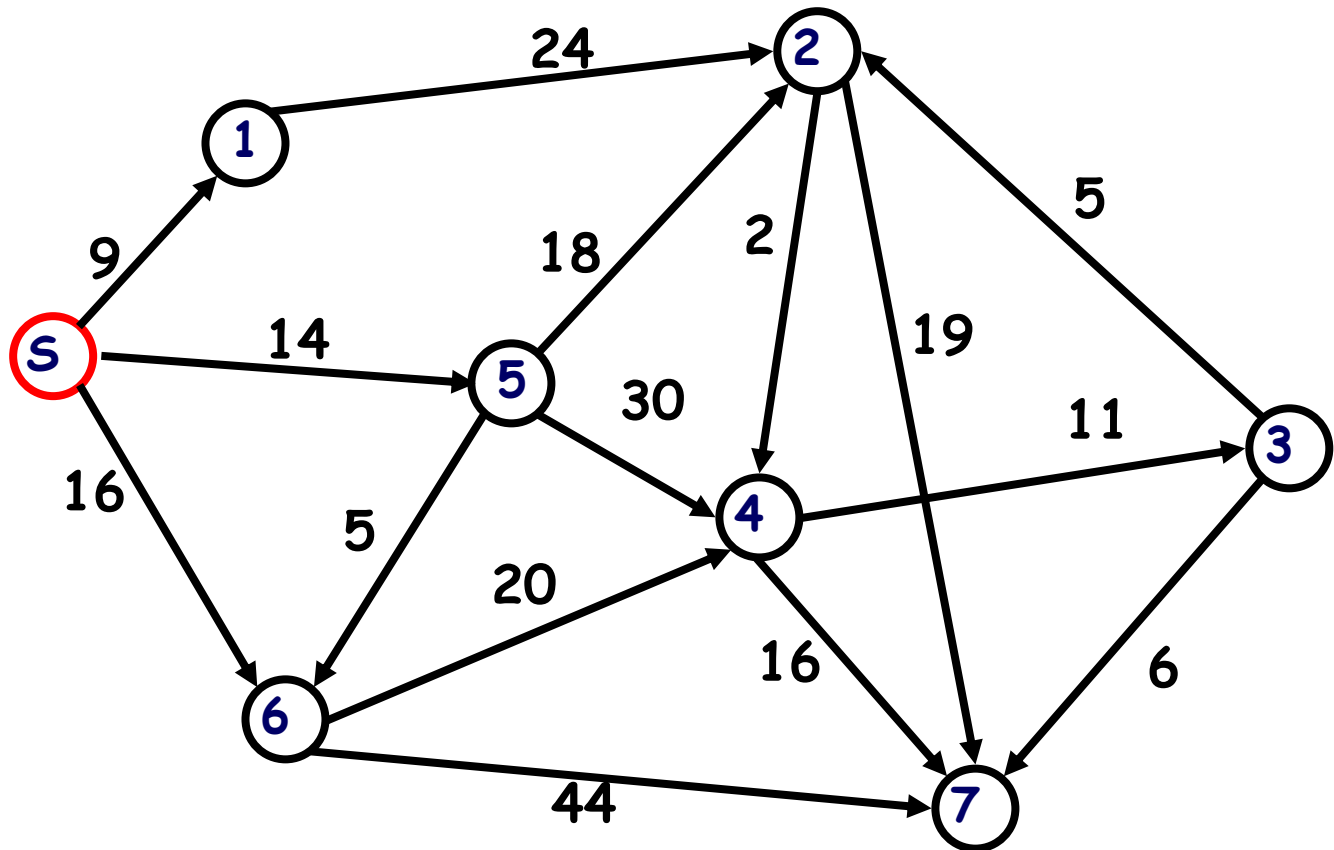
Fall 2023

Dijkstra's Algorithm Divide and Conquer Algorithms

Reading: chapters 4 & 5

The Shortest Path Problem

Given a positively weighted graph G with a source vertex s , find the shortest path from s to all other vertices in the graph.



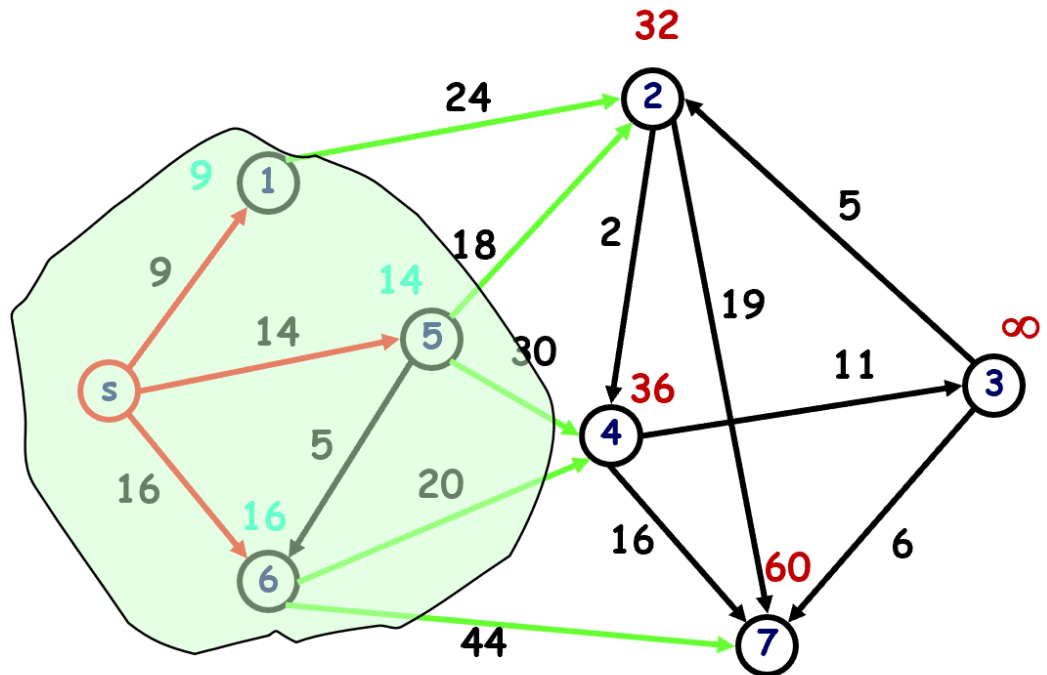
Greedy Approach

When algorithm proceeds all vertices are divided into two groups:
vertices whose shortest path from the source **s**

- is known
- is NOT discovered yet

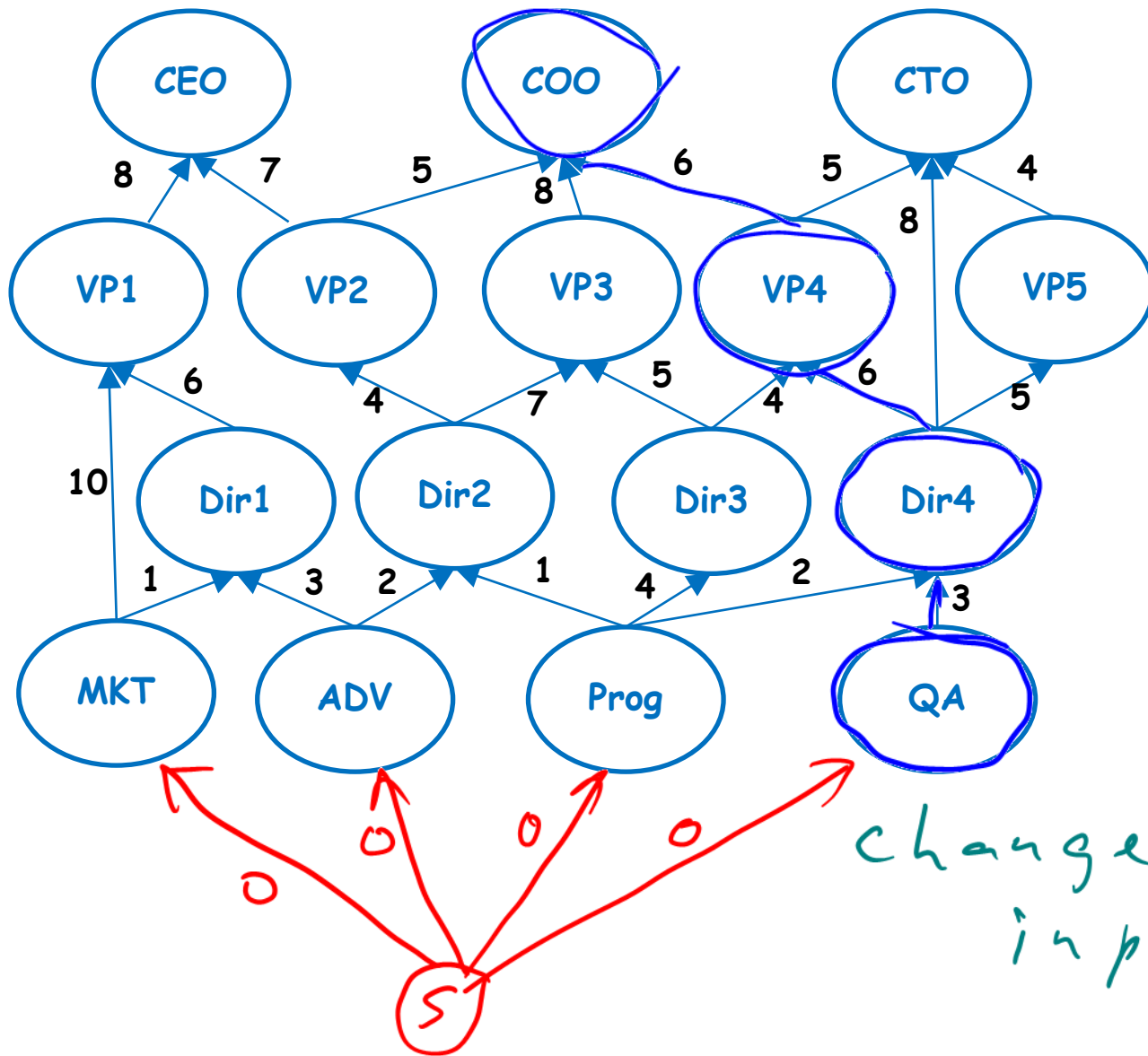
Move vertices one at a time from the undiscovered set of vertices to the known set of the shortest distances, based on the shortest distance from the source.

solution tree = { s, 1, 5, 6 }
heap = { 2, 3, 4, 7 }



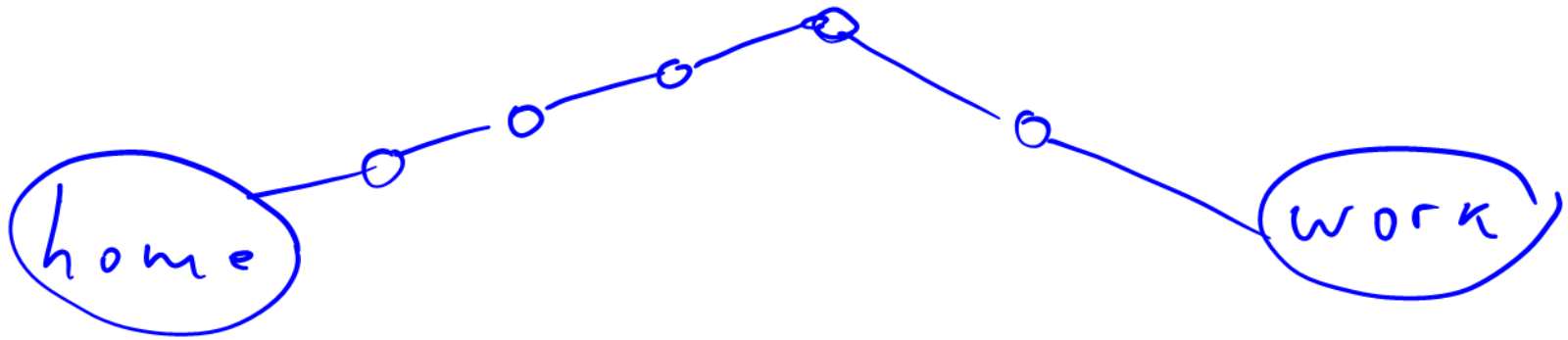
Discussion Problem 1

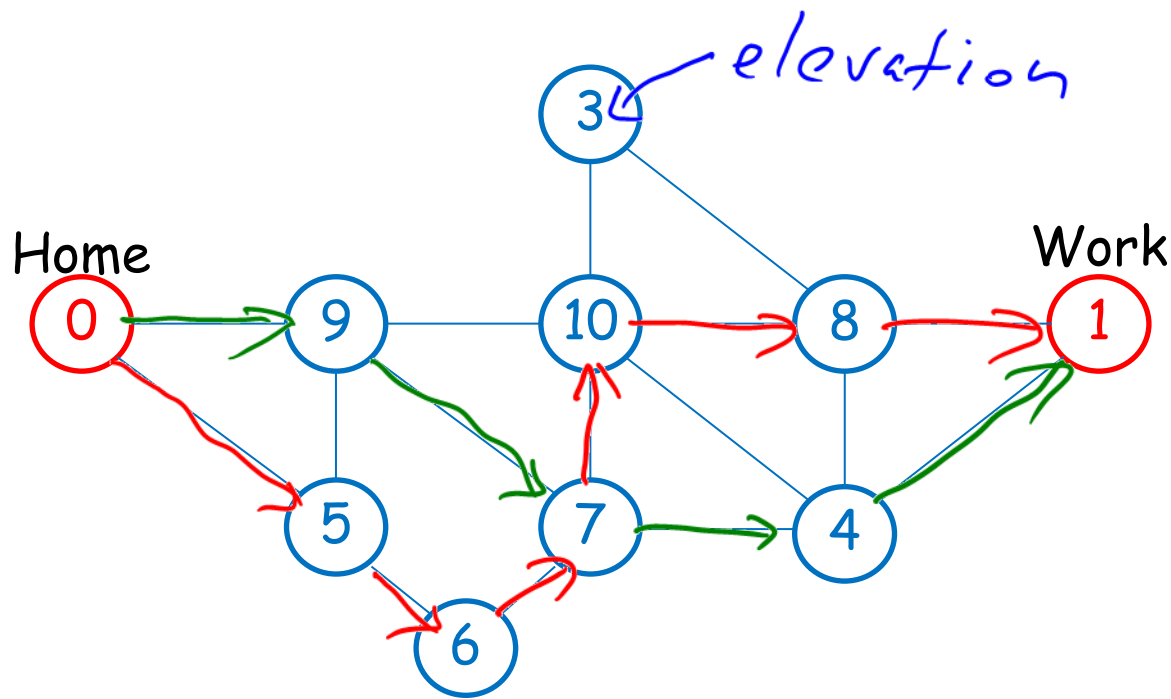
You are given a graph representing the several career paths available in industry. Each node represents a position and there is an edge from node v to node u if and only if v is a pre-requisite for u . Top positions are the ones which are not pre-requisites for any positions. Start positions are the ones which have no pre-requisites. The cost of an edge (v,u) is the effort required to go from one position v to position u . Salma wants to start a career and achieve a top position with minimum effort. Using the given graph can you provide an algorithm with the same run time complexity as Dijkstra's algorithm?



Discussion Problem 2

Hardy decides to start running to work in San Francisco to get in shape. He prefers a route to work that goes first entirely uphill and then entirely downhill. To guide his run, he prints out a detailed map of the roads between home and work. Each road segment has a positive length, and each intersection has a distinct elevation. elevation. Assuming that every road segment is either fully uphill or fully downhill, give an efficient algorithm to find the shortest path that meets Hardy's specifications.



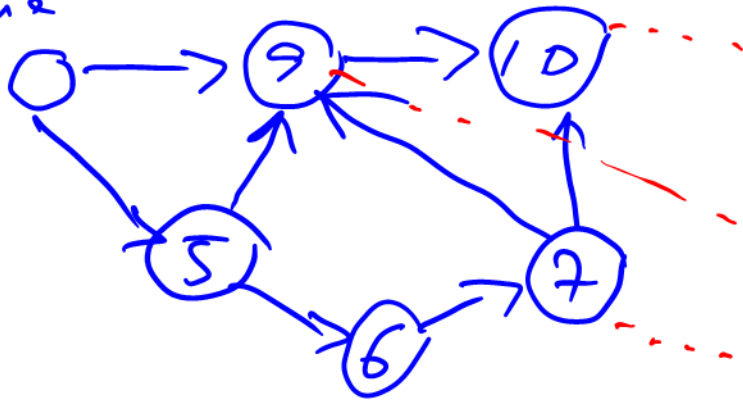


change the input

- ① create a new graph of uphill vertices starting from home
- ② create a downhill graph from Work

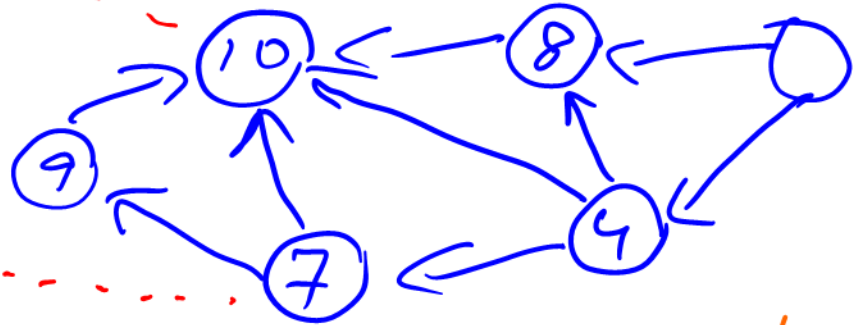
Uphill

Home



~~Downhill~~
Uphill

Work



Run Dijkstra's
from Home

Run Dijkstra's
from Work

③ find common vertices

④ check all common vertices
and those distances

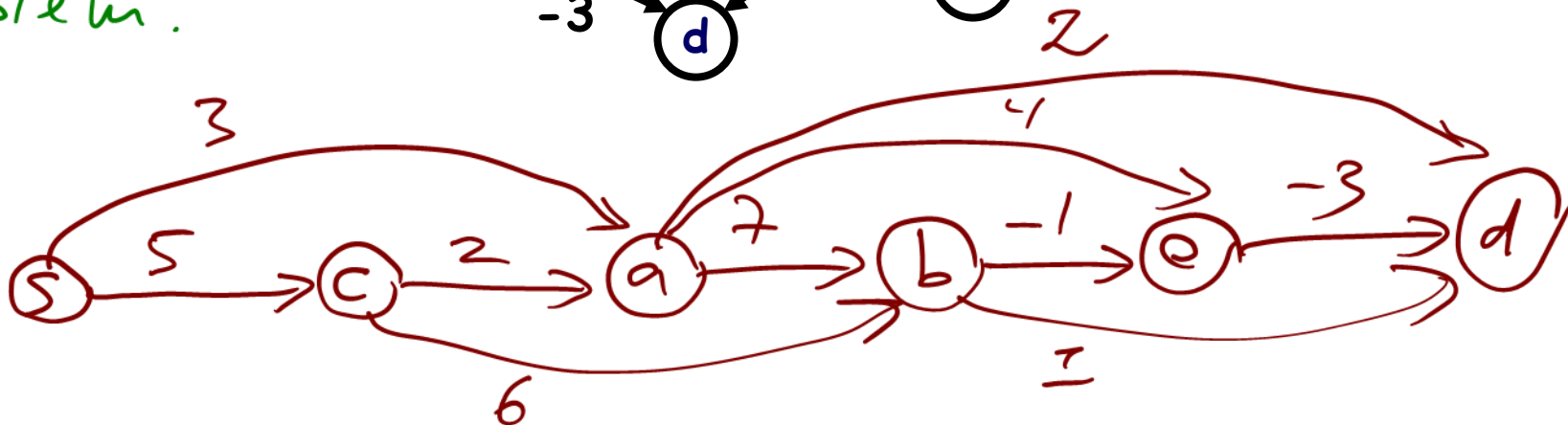
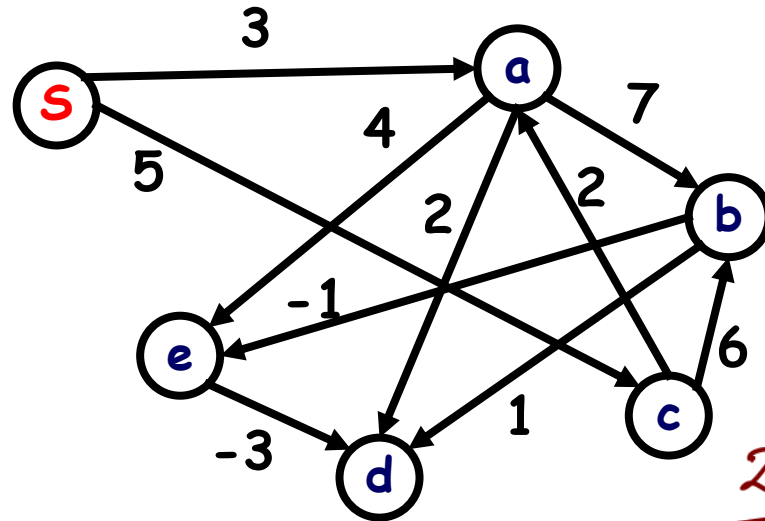
Discussion Problem 3

no Dijkstra

Design a linear time algorithm to find shortest distances in a DAG.

a) traversal/
b) topological sort

how topological
sort helps us
to solve this
problem?



s	a	b	c	d	e	
0	3	$\overline{5+6}$ 10	5	—	—	s
				3+2	3+4	s, c
	3	10	5	5	7	s, c, a
	3	10	5	4	7	s, c, a, b
						s, c, a, b, e

Runtime: topological sort + $O(E)$

Discussion Problem 4

Why doesn't Dijkstra's greedy algorithm work on graphs with negative weights?

Run Dijkstra

$d(s) = 0$, $d(c) = 3$

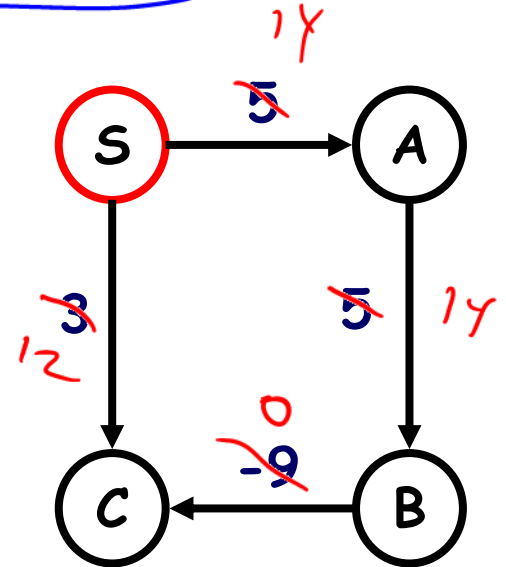
$d(A) = 5$, $d(B) = 10$

How do you fix Dijkstra?

a) re-weight the graph

Run Dijkstra: $s - c = 12$

$s - A - B - C$



break

Divide and Conquer Algorithms



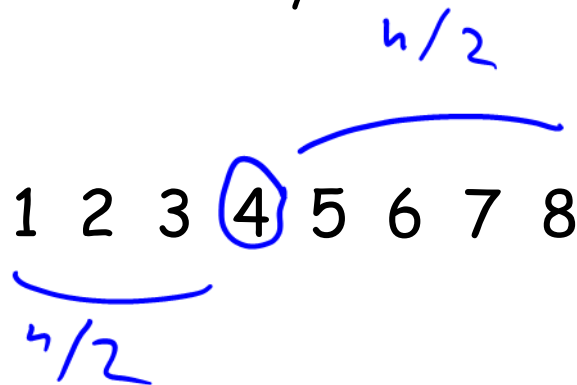
A divide-and-conquer algorithm consists of

- dividing a problem into smaller subproblems
- solving (recursively) each subproblem
- then combining solutions to subproblems to get solution to original problem

Binary Search

Given a sorted array of size n :

- compare the search item with the middle
- if it's less, search in the lower half
- if it's greater, search in the upper half
- if it's equal or the entire array has been searched, terminate.



linear	Binary
n	n
$n-1$	$n/2$
$n-2$	$n/4$
	$O(\log n)$

Mergesort

divides an unsorted list into two equal or nearly equal sub lists

sorts each of the sub lists by calling itself recursively, and then

merges the two sub lists together to form a sorted list

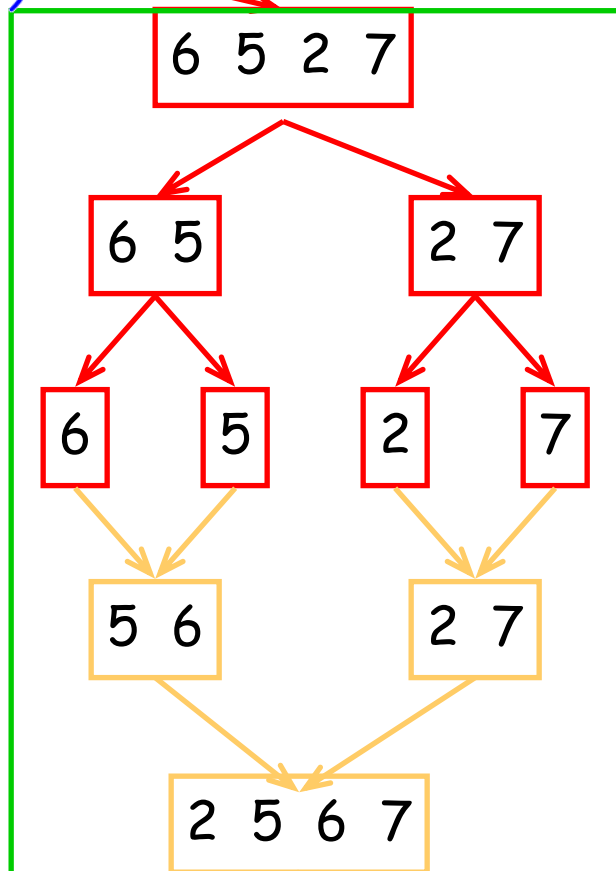
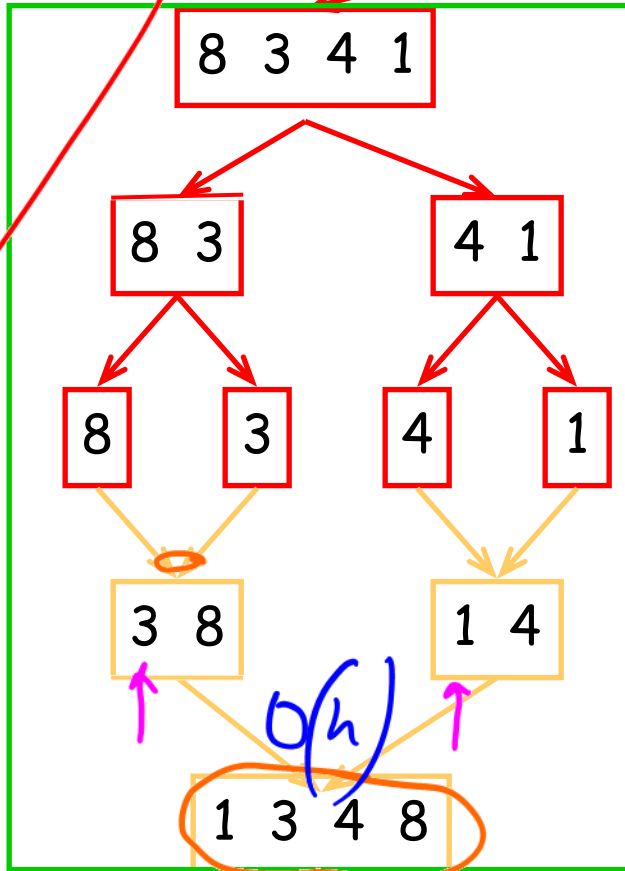
$$T(n) = 2 \cdot T\left(\frac{n}{2}\right) + O(n) + O(1) \text{ split}$$

8 3 4 1 6 5 2 7

$O(1)$

dividing

merging



1 2 3 4 5 6 7 8

D&C Recurrences

Suppose $T(n)$ is the number of steps in the worst case needed to solve the problem of size n .

We define the runtime complexity $T(n)$ by a recurrence equation.

Binary Search: $T(n) = T\left(\frac{n}{2}\right) + O(1)$

MergeSort: $T(n) = 2T\left(\frac{n}{2}\right) + O(n)$

D&C Recurrences

Suppose $T(n)$ is the number of steps in the worst case needed to solve the problem of size n .

Let us divide a problem into $a \geq 1$ subproblems, each of which is of the input size n/b where $b > 1$.

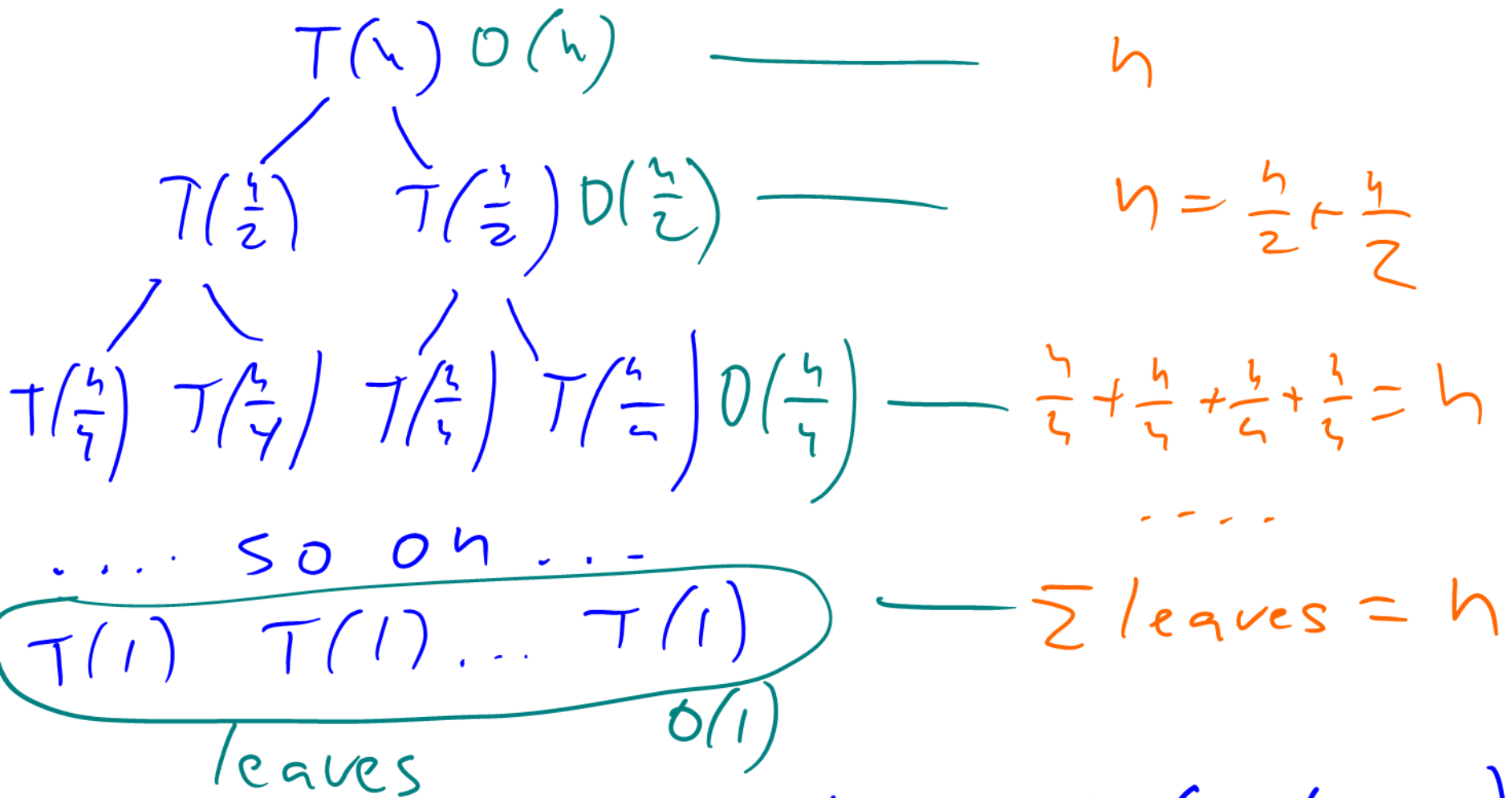
The total complexity $T(n)$ is obtained by

$$T(n) = a \cdot T(n/b) + f(n)$$

→ 0

Here $f(n)$ is a complexity of combining subproblem solutions (including complexity of *dividing* step).

Mergesort: tree of recursive calls

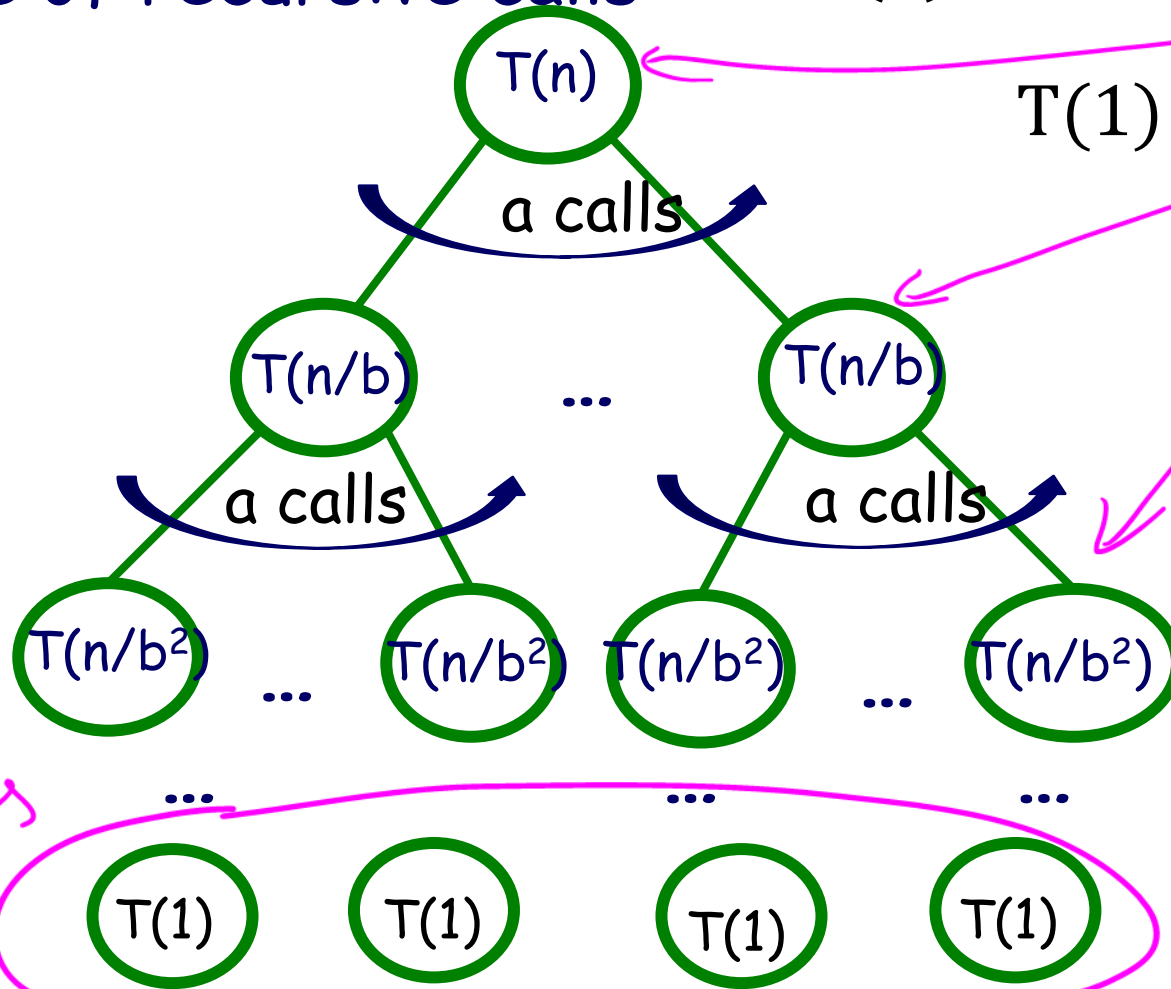


Total work: $n \cdot \text{height} = O(n \log n)$

Tree of recursive calls

$$T(n) = a \cdot T(n/b) + f(n)$$

$$T(1) = O(1)$$



leaves

how many leaves?

height

$$\log_b n$$

Counting leaves

Let $h = \text{height}$

$$T(n) = a \cdot T(n/b) + f(n)$$

$$\begin{aligned} \# \text{ of leaves} &= a^h \\ a^h &= a^{\log_b n} = \left(a^{\log_a n} \right)^{1/\log_a b} = n^{\log_b a} \end{aligned}$$

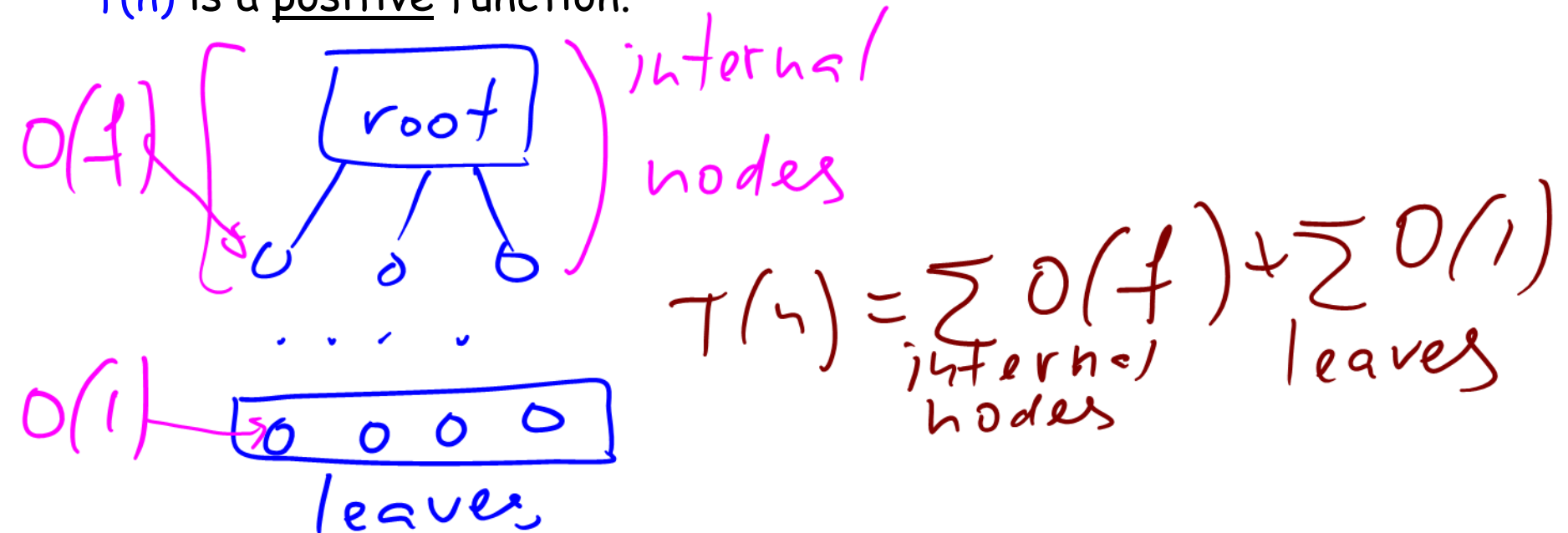
$$\log_b n = \frac{\log_a n}{\log_a b}$$

The Master Theorem

The master method provides a straightforward ("cookbook") method for solving recurrences of the form

$$T(n) = a \cdot T(n/b) + f(n)$$

where $a \geq 1$ and $b > 1$ are constants (not necessarily integers) and $f(n)$ is a positive function.



The Master Theorem

$$T(n) = a \cdot T(n/b) + f(n),$$

$a \geq 1$ and $b > 1$

any

Let $c = \log_b a$.

≥ 0

Case 1: (only leaves)

if $f(n) = O(n^{c-\epsilon})$, then $T(n) = \Theta(n^c)$ for some $\epsilon > 0$.

Case 2: (all nodes)

Merge sort

if $f(n) = \Theta(n^c \log^k n)$, $k \geq 0$, then $T(n) = \Theta(n^c \log^{k+1} n)$

extra log

Case 3: (only internal nodes)

if $f(n) = \Omega(n^{c+\epsilon})$, then $T(n) = \Theta(f(n))$ for some $\epsilon > 0$.

Discussion Problem 5

Solve the recurrence by the Master Theorem:

$$T(n) = 16 T(n/4) + 5n^3$$

$$a = 16$$

$$b = 4$$

$$c = \log_b a = 2 \rightarrow \# \text{ of leaves } O(n^2)$$

$$f(n) = O(n^3)$$

$$T(n) = a \cdot T(n/b) + f(n)$$

Case 1: if $f(n) = O(n^{c-\epsilon})$, then $T(n) = \Theta(n^c)$

Case 2: if $f(n) = \Theta(n^c \log^k n)$, then $T(n) = \Theta(n^c \log^{k+1} n)$

Case 3: if $f(n) = \Omega(n^{c+\epsilon})$, then $T(n) = \Theta(f(n))$

where $c = \log_b a$.

$$T(n) = \Theta(n^3)$$

Discussion Problem 6

$$C = \log_b a$$

Solve the recurrence by the Master Theorem:

case 1

$$1. A(n) = 3 A(n/3) + 15$$

case 3

$$2. B(n) = 4 B(n/2) + n^3$$

case 2

$$3. C(n) = 4 C(n/2) + n^2$$

$$4. D(n) = 4 D(n/2) + n$$

case 2, $K=1$

$$5. E(n) = 4 E(n/2) + n^2 \log n$$

leaves	$f(n)$	$T(n)$
$O(n)$	$O(1)$	$\Theta(n)$
$O(n^2)$	$O(n^3)$	$\Theta(n^3)$
$O(n^2)$	$O(n^2)$	$\Theta(n^2 \log_4 n)$
		$\Theta(n^2)$
		$\Theta(n^2 \log^2 n)$
$O(n^2)$	$O(n^2 \log_4 n)$	$\Theta(n^2 \log^2 n)$

Break

Integer Multiplication

Given two n -digit integers a and b , compute $a \times b$.

Brute force solution: $O(n^2)$ bit operations.

$$154517766 = \underbrace{15451}_{n/2} \cdot 10^4 + \underbrace{7766}_{n/2}$$

n bits

$$a \times b = (x_1 \cdot 10^{n/2} + x_0) \cdot (y_1 \cdot 10^{n/2} + y_0) =$$
$$= x_1 y_1 \cdot 10^n + (x_1 y_0 + x_0 y_1) 10^{n/2} + x_0 y_0$$

$O(1)$ $O(n)$

$$\begin{array}{r} 1234 \\ \times 1111 \\ \hline 1234 \\ 1234 \\ 1234 \\ 1234 \\ \hline 1370974 \end{array}$$

$$T(n) = 4 \cdot T\left(\frac{n}{2}\right) + O(1) + O(n)$$

$$T(n) = O(n^2)$$

196/

Karatsuba's algorithm

Divide-and-conquer algorithm. Split each integer in two parts and consider their product:

$$a \times b = (x_1 \cdot 10^{n/2} + x_0) \cdot (y_1 \cdot 10^{n/2} + y_0)$$

$$x_1 y_0 + x_0 y_1 = (x_0 + x_1)(y_0 + y_1) - \underbrace{x_0 y_0 - x_1 y_1}_{\text{hash table}}$$

$$a \times b = x_1 y_1 10^n + \left[\begin{matrix} \text{hash table} \end{matrix} \right] 10^{\frac{n}{2}} + x_0 y_0$$

$$T(n) = 3 \cdot T\left(\frac{n}{2}\right) + O(n), \quad T(n) = \Theta\left(n^{\log 3}\right) \\ = \Theta\left(n^{1.58}\right)$$

Discussion Problem 7

Consider another divide and conquer algorithm for integer multiplication. The key idea is to divide a large integer into 3 parts (rather than 2) of size approximately $n/3$ and then multiply those parts. What would be the runtime complexity of this multiplication?

$$154517766 = 154 \cdot 10^6 + 517 \cdot 10^3 + 766$$

$$T(n) = 9T\left(\frac{n}{3}\right) + O(n), \quad T(n) = \Theta(n^2)$$

$\leftarrow \log_3 9$

5 multiplications

$$T(n) = 5T\left(\frac{n}{3}\right) + O(n), \quad T(n) = \Theta(n^{\log_3 5})$$

Matrix Multiplication

$$\begin{matrix} & A & B & & C = A \times B \\ & & & O(n) & \\ n & \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} & \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix} & = & \begin{pmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{pmatrix} \end{matrix}$$

CPU — integers

GPU — matrix

TPU — tensor, 3D matrix

$$\text{Runtime: } O(n \cdot n^2) = O(n^3)$$

Matrix Multiplication

The usual rules of matrix multiplication holds for
block matrices

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} = \begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{pmatrix}$$

8 multiplications

$$A + B = \begin{matrix} \boxed{} \\ \boxed{} \end{matrix} + \begin{matrix} \boxed{} \\ \boxed{} \end{matrix}$$

D&C Algorithm

Let $n = 2^k$ and $M(A,B)$ denote the matrix product

1. if A is 1×1 matrix, return $a_{11} * b_{11}$.

2. write $A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$ $B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$

where A_{ij} and B_{ij} are $n/2 \times n/2$ matrices.

3. Compute $C_{ij} = M(A_{i1}, B_{1j}) + M(A_{i2}, B_{2j})$

4. Return

$$\begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

Runtime: $T(n) = 8 \cdot T\left(\frac{n}{2}\right) + O(n^2)$ *additions*
 $T(n) = O(n^3)$

1968

Strassen's Algorithm

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix} = \begin{pmatrix} s_1 + s_2 - s_4 + s_6 & s_4 - s_5 \\ s_6 + s_7 & s_2 - s_3 + s_5 - s_7 \end{pmatrix}$$

$$s_1 = (a_{12} - a_{22})(b_{21} + b_{22})$$

$$s_2 = (a_{11} + a_{22})(b_{11} + b_{22})$$

$$s_3 = (a_{11} - a_{21})(b_{11} + b_{12})$$

$$s_4 = (a_{11} + a_{12})b_{22}$$

$$s_5 = a_{11}(b_{12} - b_{22})$$

$$s_6 = a_{22}(b_{21} - b_{11})$$

$$s_7 = (a_{21} + a_{22})b_{11}$$

It takes 7 multiplications

$$T(n) = 7 \cdot T\left(\frac{n}{2}\right) + O(n^2)$$

$$T(n) = \Theta(n^{\log_2 7})$$

2.8

$$s_6 + s_7 = a_{22}b_{21} - \cancel{a_{22}b_{11}} + \cancel{a_{21}b_{11}} + \cancel{a_{22}b_{11}}$$

Fast Matrix Multiplication

1969, Strassen $O(n^{2.808})$.

1978, Pan $O(n^{2.796})$

1979, Bini $O(n^{2.78})$

1981, Schonhage $O(n^{2.548})$

1981, Pan $O(n^{2.522})$

1982, Romani $O(n^{2.517})$

1982, Coppersmith and Winograd $O(n^{2.496})$

1986, Strassen $O(n^{2.479})$

1989, Coppersmith and Winograd $O(n^{2.376})$

2010, Stothers $O(n^{2.374})$

2011, Williams $O(n^{2.3728642})$

2014, Le Gall $O(n^{2.3728639})$

Finding the Maximum Subsequence Sum

Given an array $A[0, \dots, n-1]$ of integers, design a D&C algorithm that finds a subarray $A[i, \dots, j]$ such that

$$A[i] + A[i + 1] + \dots + A[j]$$

is the maximum.

For example,

$$A = \{3, -4, \underbrace{5, -2, -2, 6, -3, 5}_{\text{sum}}, -3, 2\}$$

Output: $\{5, -2, -2, 6, -3, 5\}$

$$\text{Sum} = 5 - 2 - 2 + 6 - 3 + 5 = 9$$

Finding the Maximum Subsequence Sum (MSS)

3, -4, 5, -2, -2, 6, -3, 5, -3, 2
A₁ A₂

$[l_1, r_1, \max_1] = \text{MSS}(A_1)$ recursive

$[l_2, r_2, \max_2] = \text{MSS}(A_2)$ recursive

$[l_3, r_3, \max_3] = \text{SPAN}(A_1, A_2)$ iterative

return $\text{MAX}(\max_1, \max_2, \max_3)$

Finding the Maximum Subsequence Sum (MSS)

←
3, -4, 5, -2, -2, 6, -3, 5, -3, 2

Implementation of span A_1 A_2

② must be a part of span

⑥ must be a part of span

Compute partial sums

0, -3, ①, -4, -2 | 6, 3, ⑧, 5, 7
 r_3 r_3

Runtime: $T(n) = 2 \cdot T\left(\frac{n}{2}\right) + \Theta(n)$
 $T(n) = \Theta(n \log n)$

Review

For each of the following recurrences, give an expression for the runtime $T(n)$ if the recurrence can be solved with the Master Theorem. Otherwise, indicate that the Master Theorem does not apply.

1. $T(n) = 16 T(n/4) + 5n^3 + \log n$

$T(n) = \Theta(n^3)$

2. $T(n) = 4 T(n/2) + n^2 \log n + n^2$

$T(n) = \Theta(n^2 \log^2 n)$

3. $T(n) = 4 T(n/8) - n^2$ negative

$T(n) = NA$

4. $T(n) = 2^n T(n/2) + n$

$T(n) = NA$

5. $T(n) = 0.2 T(n/2) + n \log n$

$T(n) = NA$

≥ 1

Review

Design a new Mergesort algorithm in which instead of splitting the input array in half we split it in **the ratio 1:3**.

Write down the recurrence relation for the number of comparisons.
What is the runtime complexity of this algorithm?

$$T(n) = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + O(n)$$

$$T(n) = \Theta(n \log n)$$