# Algorithms 2021: Design by Induction

(Based on [Manber 1989])

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# 1 Introduction

## Introduction

- It is not necessary to design the steps required to solve a problem from scratch.
- It is sufficient to guarantee the following:
  - 1. It is possible to solve one small instance or a few small instances of the problem. (base case)
  - 2. A solution to every problem/instance can be constructed from solutions to smaller problems/instances. (inductive step)

# 2 Evaluating Polynomials

# **Evaluating Polynomials**

**Problem 1.** Given a sequence of real numbers  $a_n$ ,  $a_{n-1}$ ,  $\cdots$ ,  $a_1$ ,  $a_0$ , and a real number x, compute the value of the polynomial

$$P_n(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0.$$

Motivation: different approaches to the inductive step may result in algorithms of very different time complexities.

#### Evaluating Polynomials (cont.)

- Let  $P_{n-1}(x) = a_{n-1}x^{n-1} + \dots + a_1x + a_0$ .
- Induction hypothesis (first attempt)

We know how to evaluate a polynomial represented by the input  $a_{n-1}, \dots, a_1, a_0$ , at the point x, i.e., we know how to compute  $P_{n-1}(x)$ .

- $P_n(x) = a_n x^n + P_{n-1}(x)$ .
- Number of multiplications:

$$n + (n-1) + \dots + 2 + 1 = \frac{n(n+1)}{2}$$
.

# Evaluating Polynomials (cont.)

- Induction hypothesis (second attempt)
  We know how to compute  $P_{n-1}(x)$ , and we know how to compute  $x^{n-1}$ .
- $P_n(x) = a_n x(x^{n-1}) + P_{n-1}(x)$ .
- Number of multiplications: 2n-1.

## Evaluating Polynomials (cont.)

- Let  $P'_{n-1}(x) = a_n x^{n-1} + a_{n-1} x^{n-2} + \dots + a_1$ .
- Induction hypothesis (final attempt)

We know how to evaluate a polynomial represented by the coefficients  $a_n$ ,  $a_{n-1}$ ,  $\cdots$ ,  $a_1$ , at the point x, i.e., we know how to compute  $P'_{n-1}(x)$ .

•  $P_n(x) = P'_n(x) = P'_{n-1}(x) \cdot x + a_0.$ 

#### Evaluating Polynomials (cont.)

• More generally,

$$\begin{cases} P'_0(x) = a_n \\ P'_i(x) = P'_{i-1}(x) \cdot x + a_{n-i}, \text{ for } 1 \le i \le n \end{cases}$$

• Number of multiplications: n.

#### Evaluating Polynomials (cont.)

Algorithm Polynomial\_Evaluation  $(\bar{a}, x)$ ; begin

$$P := a_n;$$
  
for  $i := 1$  to  $n$  do  
 $P := x * P + a_{n-i}$ 

end

This algorithm is known as *Horner's rule*.

# 3 Maximal Induced Subgraph

#### Maximal Induced Subgraph

**Problem 2.** Given an undirected graph G = (V, E) and an integer k, find an induced subgraph H = (U, F) of G of maximum size such that all vertices of H have degree  $\geq k$  (in H), or conclude that no such induced subgraph exists.

Design Idea: in the inductive step, we try to remove one vertex (that cannot possibly be part of the solution) to get a smaller instance.

#### Maximal Induced Subgraph (cont.)

• Recursive:

```
Algorithm Max_Ind_Subgraph (G, k); begin

if the degree of every vertex of G \ge k then

Max_Ind_Subgraph := G;

else let v be a vertex of G with degree < k;

Max_Ind_Subgraph := Max_Ind_Subgraph(G - v, k);

end

/* G - v denotes the graph obtained from G by removing vertex v and every edge incident to v. */

• Iterative:

Algorithm Max_Ind_Subgraph (G, k);

begin

while the degree of some vertex v of G < k do

G := G - v;

Max_Ind_Subgraph := G;

end
```

# 4 One-to-One Mapping

## One-to-One Mapping

**Problem 3.** Given a finite set A and a mapping f from A to itself, find a subset  $S \subseteq A$  with the maximum number of elements, such that (1) the function f maps every element of S to another element of S (i.e., f maps S into itself), and (2) no two elements of S are mapped to the same element (i.e., f is one-to-one when restricted to S).

Design Idea: similar to the previous problem; in the inductive step, we try to remove one element (that cannot possibly be part of the solution) to get a smaller instance.

An element that is not mapped to may be removed.

# One-to-One Mapping (cont.) Algorithm Mapping (f, n); begin

```
begin S := A; for j := 1 to n do c[j] := 0; for j := 1 to n do increment c[f[j]]; for j := 1 to n do
   if c[j] = 0 then put j in Queue;
   while Queue not empty do
    remove i from the top of Queue;
   S := S - \{i\}; decrement c[f[i]]; if c[f[i]] = 0 then put f[i] in Queue end
```

# 5 Celebrity

# Celebrity

**Problem 4.** Given an  $n \times n$  adjacency matrix, determine whether there exists an i (the "celebrity") such that all the entries in the i-th column (except for the ii-th entry) are 1, and all the entries in the i-th row (except for the ii-th entry) are 0.

```
Note: A celebrity corresponds to a sink of the directed graph.
```

```
Note: Every directed graph has at most one sink.
```

```
/* Proof by contradiction. */
```

Motivation: the trivial solution has a time complexity of  $O(n^2)$ . Can we do better, in O(n)?

To achieve O(n) time, we must reduce the problem size by at least one in constant time.

# Celebrity (cont.)

Basic idea: check whether i knows j.

In either case, one of the two may be eliminated.

/\* If i knows j, then i is not a celebrity. If i does not know j, then j is not a celebrity. \*/

The O(n) algorithm proceeds in two stages:

• Eliminate a node every round until only one is left.

/\* The node that remains is not necessarily a celebrity, as we have not checked whether it knows any previously deleted node or the other way around. \*/

• Check whether the remaining one is truly a celebrity.

else candidate := i;

## Celebrity (cont.)

```
Algorithm Celebrity (Know);
begin
i := 1;
j := 2;
next := 3;
while next \le n+1 do
if Know[i,j] then i := next
else j := next;
next := next + 1;
if i = n+1 then candidate := j
```

## Celebrity (cont.)

```
 wrong := false; \\ k := 1; \\ Know[candidate, candidate] := false; \\ \textbf{while } not \ wrong \ and \ k \leq n \ \textbf{do} \\ \textbf{if } Know[candidate, k] \ \textbf{then } wrong := true; \\ \textbf{if } not \ Know[k, candidate] \ \textbf{then} \\ \textbf{if } candidate \neq k \ \textbf{then } wrong := true; \\ k := k + 1; \\ \textbf{if } not \ wrong \ \textbf{then } celebrity := candidate \\ \textbf{else } celebrity := 0; \\ \textbf{end}
```

#### \_\_\_\_

# 6 The Skyline Problem

# The Skyline Problem

**Problem 5.** Given the exact locations and shapes of several rectangular buildings in a city, draw the skyline (in two dimension) of these buildings, eliminating hidden lines.

Motivation: different approaches to the inductive step may result in algorithms of very different time complexities.

Compare: adding buildings one by one to an existing skyline vs. merging two skylines of about the same size

## The Skyline Problem

• Adding one building at a time:

$$\left\{ \begin{array}{l} T(1) = O(1) \\ T(n) = T(n-1) + O(n), n \geq 2 \end{array} \right.$$

Time complexity:  $O(n^2)$ .

/\* 
$$T(n) = T(n-1) + O(n) = (T(n-2) + O(n-1)) + O(n) = \cdots = O(1) + O(2) + \cdots + O(n) = O(n^2).$$

• Merging two skylines every round:

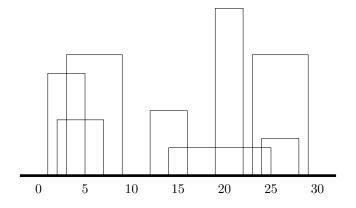
$$\left\{ \begin{array}{l} T(1)=O(1) \\ T(n)=2T(\frac{n}{2})+O(n), n\geq 2 \end{array} \right.$$

Time complexity:  $O(n \log n)$ .

/\* Apply the master theorem. Here, a = 2, b = 2, k = 1, and  $b^k = 2 = a$ . \*/

#### Representation of a Skyline

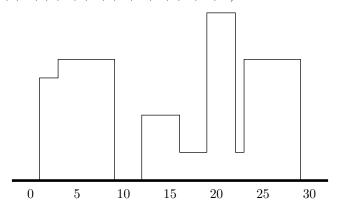
Input: 
$$(1,11,5)$$
,  $(2,6,7)$ ,  $(3,13,9)$ ,  $(12,7,16)$ ,  $(14,3,25)$ ,  $(19,18,22)$ ,  $(23,13,29)$ , and  $(24,4,28)$ .



Source: adapted from [Manber 1989, Figure 5.5(a)].

# Representation of a Skyline (cont.)

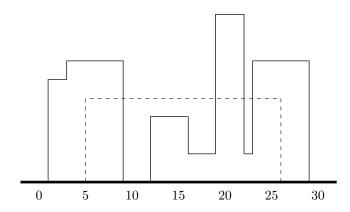
Representation: (1,11,3,13,9,0,12,7,16,3,19,18,22,3,23,13,29).



Source: adapted from [Manber 1989, Figure 5.5(b)].

# Adding a Building

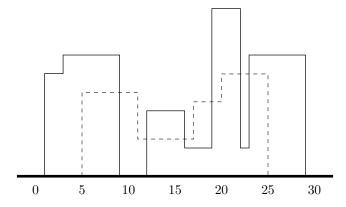
 $\bullet \ \mathrm{Add} \ (5, \boldsymbol{9}, 26) \ \mathrm{to} \ (1, \boldsymbol{11}, 3, \boldsymbol{13}, 9, \boldsymbol{0}, 12, \boldsymbol{7}, 16, \boldsymbol{3}, 19, \boldsymbol{18}, 22, \boldsymbol{3}, 23, \boldsymbol{13}, 29).$ 



Source: adapted from [Manber 1989, Figure 5.6].

• The skyline becomes (1,11,3,13,9,9,19,18,22,9,23,13,29).

# Merging Two Skylines



Source: adapted from [Manber 1989, Figure 5.7].

# 7 Balance Factors in Binary Trees

#### Balance Factors in Binary Trees

**Problem 6.** Given a binary tree T with n nodes, compute the balance factors of all nodes.

The balance factor of a node is defined as the difference between the height of the node's left subtree and the height of the node's right subtree.

Motivation: an example of why we must strengthen the hypothesis (and hence the problem to be solved).

# Balance Factors in Binary Trees (cont.)

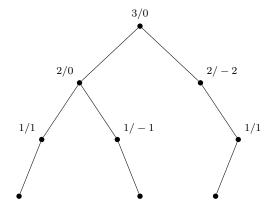


Figure: A binary tree. The numbers represent h/b, where h is the height and b is the balance factor. Source: redrawn from [Manber 1989, Figure 5.8].

# Balance Factors in Binary Trees (cont.)

## • Induction hypothesis

We know how to compute balance factors of all nodes in trees that have < n nodes.

## • Stronger induction hypothesis

We know how to compute balance factors and heights of all nodes in trees that have < n nodes.

# 8 Maximum Consecutive Subsequence

## Maximum Consecutive Subsequence

**Problem 7.** Given a sequence  $x_1, x_2, \dots, x_n$  of real numbers (not necessarily positive) find a subsequence  $x_i, x_{i+1}, \dots, x_j$  (of consecutive elements) such that the sum of the numbers in it is maximum over all subsequences of consecutive elements.

Example: In the sequence (2, -3, 1.5, -1, 3, -2, -3, 3), the maximum subsequence is (1.5, -1, 3).

Motivation: another example of strengthening the hypothesis.

## Maximum Consecutive Subsequence (cont.)

# • Induction hypothesis

We know how to find the maximum subsequence in sequences of size < n.

#### • Stronger induction hypothesis

We know how to find, in sequences of size < n, the maximum subsequence overall and the maximum subsequence that is a suffix.

Reasoning: the maximum subsequence of problem size n is obtained either

- directly from the maximum subsequence of problem size n-1 or
- from appending the *n*-th element to the maximum suffix of problem size n-1.

# Maximum Consecutive Subsequence (cont.)

```
\label{eq:Algorithm_Max_Consec_Subseq} \begin{aligned} & \textbf{Algorithm Max\_Consec\_Subseq} \ (X,n); \\ & \textbf{begin} \\ & \textit{Global\_Max} := 0; \\ & \textit{Suffix\_Max} := 0; \\ & \textbf{for} \ i := 1 \ \textbf{to} \ n \ \textbf{do} \\ & \textbf{if} \ x[i] + \textit{Suffix\_Max} > \textit{Global\_Max} \ \textbf{then} \\ & \textit{Suffix\_Max} := \textit{Suffix\_Max} + x[i]; \\ & \textit{Global\_Max} := \textit{Suffix\_Max} \\ & \textbf{else} \ \textbf{if} \ x[i] + \textit{Suffix\_Max} > 0 \ \textbf{then} \\ & \textit{Suffix\_Max} := \textit{Suffix\_Max} + x[i] \\ & \textbf{else} \ \textit{Suffix\_Max} := 0 \end{aligned}
```

# 9 The Knapsack Problem

#### The Knapsack Problem

**Problem 8.** Given an integer K and n items of different sizes such that the i-th item has an integer size  $k_i$ , find a subset of the items whose sizes sum to exactly K, or determine that no such subset exists.

Design Idea: use strong induction so that solutions to all smaller instances may be used.

## The Knapsack Problem (cont.)

- Let P(n, K) denote the problem where n is the number of items and K is the size of the knapsack.
- Induction hypothesis

We know how to solve P(n-1, K).

• Stronger induction hypothesis

We know how to solve P(n-1, k), for all  $0 \le k \le K$ .

Reasoning: P(n,K) has a solution if either

- -P(n-1,K) has a solution or
- $-P(n-1,K-k_n)$  does, provided  $K-k_n \ge 0$ .

#### The Knapsack Problem (cont.)

An example of the table constructed for the knapsack problem:

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
	0	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_1 = 2$	0	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2 = 3$	0	-	0	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3 = 5$	0	-	0	0	-	0	-	I	I	-	I	-	-	-	-	-	-
$k_4 = 6$	0	-	0	0	-	0	I	0	0	I	0	I	-	I	I	-	I

"I": a solution containing this item has been found.

"O": a solution without this item has been found.

"-": no solution has yet been found.

Source: adapted from [Manber 1989, Figure 5.11].

# The Knapsack Problem (cont.)

```
Algorithm Knapsack (S,K); P[0,0].exist := true; for k := 1 to K do P[0,k].exist := false; for i := 1 to n do for k := 0 to K do P[i,k].exist := false; if P[i-1,k].exist then P[i,k].exist := true; P[i,k].belong := false else if k-S[i] \ge 0 then if P[i-1,k-S[i]].exist then P[i,k].exist := true; P[i,k].belong := true; P[i,k].belong := true
```