

# Algorithms 2017: 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  $a_n, a_{n-1}, \dots, a_1, a_0$ , 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  $P_{n-1}(x) = a_{n-1} x^{n-1} + \dots + a_1 x + a_0$ .
- **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  $P_{n-1}(x)$ .
- $P_n(x) = a_n x^n + P_{n-1}(x)$ .

### Evaluating Polynomials (cont.)

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

### 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  $a_n, a_{n-1}, \dots, a_1$ , 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,

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

### 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.

## 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.

## One-to-One Mapping (cont.)

```
Algorithm Mapping ( $f, n$ );
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.

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

### Celebrity (cont.)

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

### Celebrity (cont.)

```
 $wrong := false$ ;
 $k := 1$ ;
 $Know[candidate, candidate] := false$ ;
while not  $wrong$  and  $k \leq n$  do
```

```

if Know[candidate, k] then wrong := true;
if not Know[k, candidate] then
  if candidate ≠ k then wrong := true;
  k := k + 1;
if not wrong then celebrity := candidate
  else celebrity := 0;
end

```

## 6 The Skyline Problem

### The Skyline Problem

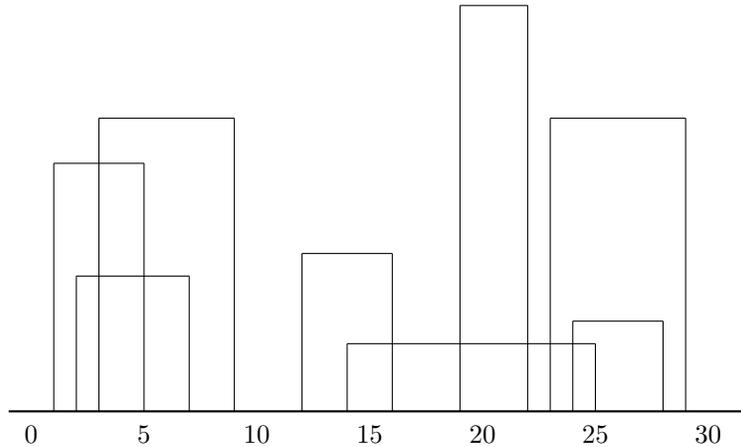
**Problem 5.** *Given the exact locations and shapes of several rectangular buildings in a city, draw the skyline (in two dimension) of these buildings, eliminating hidden lines.*

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

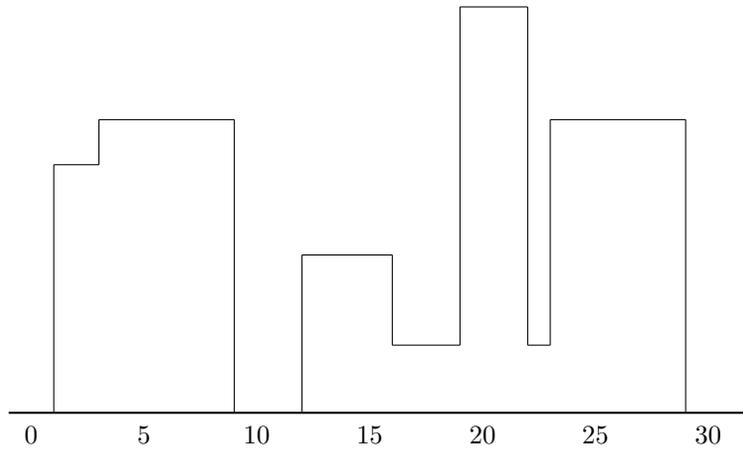
### Representation of a Skyline

(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).



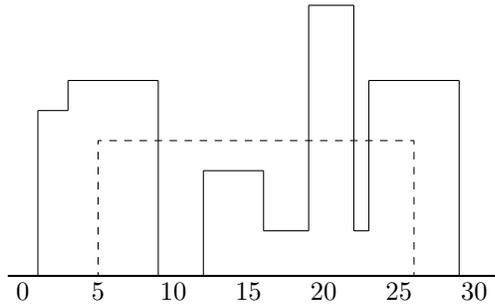
### Representation of a Skyline (cont.)

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



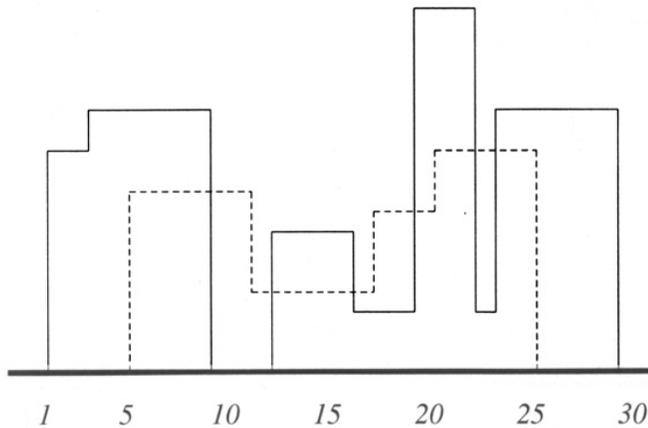
**Adding a Building**

- Add (5,9,26) to (1,11,3,13,9,0,12,7,16,3,19,18,22,3,23,13,29).



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

**Merging Two Skylines**



**Figure 5.7** Merging two skylines.

Source: [Manber 1989].

## 7 Balance Factors in Binary Trees

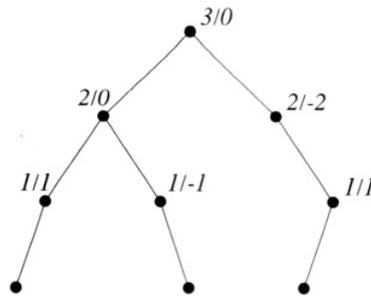
### Balance Factors in Binary Trees

**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 5.8** A binary tree. The numbers represent  $h/b$ , where  $h$  is the height and  $b$  is the balance factor.

Source: [Manber 1989].

### 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.

### Maximum Consecutive Subsequence (cont.)

**Algorithm** Max\_Consec\_Subseq ( $X, n$ );

**begin**

$Global\_Max := 0$ ;

$Suffix\_Max := 0$ ;

**for**  $i := 1$  **to**  $n$  **do**

**if**  $x[i] + Suffix\_Max > Global\_Max$  **then**

$Suffix\_Max := Suffix\_Max + x[i]$ ;

$Global\_Max := Suffix\_Max$

**else if**  $x[i] + Suffix\_Max > 0$  **then**

$Suffix\_Max := Suffix\_Max + x[i]$

**else**  $Suffix\_Max := 0$

**end**

## 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  $0 \leq k \leq K$ .

### 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
	O	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_1 = 2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2 = 3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3 = 5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4 = 6$	O	-	O	O	-	O	I	O	O	I	O	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.

## 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**  $k - S[i] \geq 0$  **then**

**if**  $P[i - 1, k - S[i]].exist$  **then**

$P[i, k].exist := true$ ;

$P[i, k].belong := true$