Algorithms 2019: 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}, \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,

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

• Number of multiplications: n.

Evaluating Polynomials (cont.)

Algorithm Polynomial_Evaluation (\bar{a}, x) ; begin $P := a_n$;

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for i := 1 to n do

P := x * P + a_{n-i}
```

 \mathbf{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.)

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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
```

 \mathbf{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.

Celebrity (cont.)

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

Celebrity (cont.)

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

 \mathbf{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:

$$\begin{cases} T(1) = O(1) \\ T(n) = T(n-1) + O(n), n \ge 2 \end{cases}$$

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

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

• Merging two skylines every round:

$$\begin{cases} T(1) = O(1) \\ T(n) = 2T(\frac{n}{2}) + O(n), n \ge 2 \end{cases}$$

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

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



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.)

Algorithm Max_Consec_Subseq (X, n);

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\begin{aligned} & \text{begin} \\ & Global\_Max := 0; \\ & Suffix\_Max := 0; \\ & \text{for } i := 1 \text{ to } n \text{ do} \\ & \text{ if } x[i] + Suffix\_Max > Global\_Max \text{ then} \\ & Suffix\_Max := Suffix\_Max + x[i]; \\ & Global\_Max := Suffix\_Max \\ & \text{else if } x[i] + Suffix\_Max > 0 \text{ then} \\ & Suffix\_Max := Suffix\_Max + x[i] \\ & \text{else } Suffix\_Max := 0 \end{aligned}
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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 $0 \le k \le K$.

(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	-	Ι	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2 = 3$	0	-	0	I	-	Ι	-	-	-	-	-	-	-	-	-	-	-
$k_3 = 5$	0	-	0	0	-	0	-	I	I	-	Ι	-	-	-	-	-	-
$k_4 = 6$	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);

$$\begin{split} P[0,0].exist &:= true;\\ \textbf{for } k &:= 1 \textbf{ to } K \textbf{ do}\\ P[0,k].exist &:= false;\\ \textbf{for } i &:= 1 \textbf{ to } n \textbf{ do}\\ \textbf{for } k &:= 0 \textbf{ to } K \textbf{ do}\\ P[i,k].exist &:= false;\\ \textbf{ if } P[i-1,k].exist \textbf{ then}\\ P[i,k].exist &:= true;\\ P[i,k].belong &:= false\\ \textbf{ else if } k - S[i] &\geq 0 \textbf{ then}\\ \textbf{ if } P[i-1,k-S[i]].exist \textbf{ then}\\ P[i,k].exist &:= true;\\ P[i,k].belong &:= true;\\ P[i,k].belong &:= true \end{split}$$