

Dynamic Programming 1: Pitfall of Recursion

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Today, we will start a series of lectures on **dynamic programming**, which is a technique for accelerating recursive algorithms.

Remark: Despite the word “programming”, the technique has nothing to do with programming languages.

Problem: Let A be an array of n positive integers.

Consider function

$$f(k) = \begin{cases} 0 & \text{if } k = 0 \\ \max_{i=1}^k (A[i] + f(k-i)) & \text{if } 1 \leq k \leq n \end{cases}$$

Goal: Compute $f(n)$.

Example: Consider the following array A :

i	1	2	3	4
$A[i]$	1	5	8	9

Then, $f(1) = 1$, $f(2) = 5$, $f(3) = 8$, and $f(4) = 10$.

Pitfall of Recursion

Consider the following recursive algorithm for computing $f(k)$.

$f(k)$

1. **if** $k = 0$ **then return** 0
2. $ans \leftarrow -\infty$
3. **for** $i \leftarrow 1$ to k **do**
4. $tmp \leftarrow A[i] + f(k - i)$
5. **if** $tmp > ans$ **then** $ans \leftarrow tmp$
6. **return** ans

Computing $f(n)$ with the above algorithm incurs running time $\Omega(2^n)$ (left as a regular exercise).

Pitfall of Recursion

$f(k)$

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Why is the algorithm so slow?

Answer: It re-computes $f(x)$ for the same x repeatedly!

How many times do we need to call $f(2)$ in computing $f(3)$, $f(4)$, $f(5)$, and $f(6)$, respectively?

Pitfall of recursion:

A recursive algorithm does considerable redundant work if the **same** subproblem is encountered over and over again.

Antidote: dynamic programming.

Principle of dynamic programming

Resolve subproblems according to a certain **order**. Remember the output of every subproblem to avoid re-computation.

Problem: Let A be an array of n positive integers.

$$f(k) = \begin{cases} 0 & \text{if } k = 0 \\ \max_{i=1}^k (A[i] + f(k-i)) & \text{if } 1 \leq k \leq n \end{cases}$$

Goal: Compute $f(n)$.

Order of subproblems: $f(1), \dots, f(n)$.

Resolve subproblem $f(1)$: $O(1)$ time

Resolve subproblem $f(2)$: $O(2)$ time, **given** $f(1)$.

...

Resolve subproblem $f(k)$: $O(k)$ time, **given** $f(1), \dots, f(k-1)$.

...

Resolve subproblem $f(n)$: $O(n)$ time, **given** $f(1), \dots, f(n-1)$.

In total: $O(n^2)$ time.

Pseudocode of our algorithm:

dyn-prog

1. initialize an array *ans* of size *n*
2. define special value $ans[0] \leftarrow 0$
3. **for** $k \leftarrow 1$ to n **do**
/* assuming $f(0), f(1), \dots, f(k-1)$ ready, compute $f(k)$ */
4. $ans[k] \leftarrow -\infty$
5. **for** $i \leftarrow 1$ to k **do**
6. $tmp \leftarrow A[i] + ans[k-i]$
7. **if** $tmp > ans[k]$ **then** $ans[k] \leftarrow tmp$

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