

# Lecture 5: Dynamic Programming II

Scriber: Haoming Li

January 27, 2020

## 1 Designing a DP for Longest Increasing Subsequence (LIS)

Given a sequence of numbers, we want to find a strictly increasing *subsequence* of it that is also the longest. The numbers in the subsequence may not be consecutive in the original sequence. For example, given sequence  $a[] = \{4, 2, 5, 3, 9, 7, 8, 10, 6\}$ , its LIS is  $\{2, 5, 7, 8, 10\}$  or  $\{2, 3, 7, 8, 10\}$ , as they both have length 5.

### 1.1 A Failed Attempt

A natural subproblem is to have  $f[i]$  denote the length of the LIS of sequence  $a[1 \dots i]$ . A natural transition function is to consider whether the LIS of  $a[1 \dots i]$  should include  $a[i]$  or not, and take max of the two.

If  $a[i]$  is not included, then simply  $f[i] = f[i - 1]$ . If  $a[i]$  is included, however, we run into a problem: when the last element  $a[i]$  is in the sequence, we have the additional constraint that all other elements need to be smaller than  $a[i]$ . However, when we reference a previous subproblem  $f[j]$  where  $j < i$ , we do not know whether the solution for  $f[j]$  uses numbers strictly smaller than  $a[i]$ , hence our proposed transition function does not work.

### 1.2 Attempt 2

Consider the following subproblem definition: Let  $f[i]$  denote the length of the LIS of sequence  $a[1 \dots i]$  that ends at  $a[i]$ . (i.e. the subsequence must include  $a[i]$ )

The decision at  $f[i]$  is immediate, as we *have* to pick  $a[i]$  by definition. To compute  $f[i]$ , we can enumerate the number that is before  $a[i]$  in the sequence. This motivates our transition function:

$$f[i] = \max\{1, \max_{j < i, a[j] < a[i]} f[j] + 1\}$$

If the max evaluates to the first case then the subsequence is simply  $\{a[i]\}$ ; if it evaluates to the second case then the subsequence is  $\{\text{LIS ending at } a[j], a[i]\}$ .

For example, for the sequence mentioned above, we would fill out a DP table like below

$$a[] = \{4, 2, 5, 3, 9, 7, 8, 10, 6\}$$

$i$	1	2	3	4	5	6	7	8	9
$f[i]$	1	1	2	2	3	3	4	5	3

To complete our algorithm, we also need a base case that is  $f[0] = 0$ , and an output that is  $\max_{1 \leq i \leq n} f[i]$ .

### 1.2.1 Analyze Running Time

The running time of a DP, in general, is

$$\# \text{ states} \times \text{time for evaluating one transition function}$$

In the DP above, there are  $n$  states, and we take  $O(n)$  to evaluate one transition function. Hence the total running time is  $O(n^2)$

### 1.2.2 Proof of Correctness

We will use induction to prove that our DP computes the correct answer. Our inductive hypothesis, in general, is to assume that “smaller subproblems are computed correctly.”

- Base case:  $f[0] = 0$  is true by definition.
- Inductive hypothesis: assume that for every  $j < i$ ,  $f[j]$  is indeed the length of the LIS ending at  $a[j]$ .
- Induction step: Let  $b[]$  denote the LIS ending at  $a[i]$ .  $b[]$  is either of length 1 or of length greater than 1.
  - If  $b[]$  is of length 1, then it is considered by the first case of the transition function.
  - If  $b[]$  is of length greater than 1, let  $a[j]$  denote the second-to-last number in  $b[]$ . By definition  $j < i$  and  $a[j] < a[i]$ . By IH,  $f[j]$  is computed correctly. Hence  $f[i] = f[j] + 1$  is considered by the second case of the transition function.

Therefore,  $f[i]$  is also computed correctly.

- By induction,  $f[i]$  is computed correctly for all  $i \geq 0$ .