

CSE 417: Algorithms and Computational Complexity

6: Dynamic Programming, III Longest Increasing Subseq.

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Three Steps to Dynamic Programming

- Formulate the answer as a recurrence relation or recursive algorithm
- Show that number of different parameters in the recursive algorithm is "small" (e.g., bounded by a low-degree polynomial)
- Specify an order of evaluation for the recurrence so that already have the partial results ready when you need them.

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Longest Increasing Subsequence

- Given a sequence of integers s_1, \dots, s_n find a subsequence $s_{i_1} < s_{i_2} < \dots < s_{i_k}$ with $i_1 < \dots < i_k$ so that k is as large as possible.
- e.g. Given 9,5,2,8,7,3,1,6,4 as input,
 - possible increasing subsequence is 5,7
 - better is 2,3,6 or 2,3,4 (either or which would be a correct output to our problem)

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Find recursive algorithm

- Solve sub-problem on s_1, \dots, s_{n-1} and then try to extend using s_n
- Two cases:
 - s_n is not used
 - answer is the same answer as on s_1, \dots, s_{n-1}
 - s_n is used
 - answer is s_n preceded by the longest increasing subsequence in s_1, \dots, s_{n-1} that ends in a number smaller than s_n

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Refined recursive idea (stronger notion of subproblem)

- Suppose that we knew for each $i < n$ the longest increasing subsequence in s_1, \dots, s_n that ends in s_i .
 - $i = n-1$ is just the $n-1$ size sub-problem we tried before.
- Now to compute value for $i = n$ find
 - s_n preceded by the maximum over all $i < n$ such that $s_i < s_n$ of the longest increasing subsequence ending in s_i
- First find the best **length** rather than trying to actually compute the sequence itself.

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Recurrence

- Let $L[i]$ = length of longest increasing subsequence in s_1, \dots, s_n that ends in s_i .
- $L[j] = 1 + \max\{L[i] : i < j \text{ and } s_i < s_j\}$
(where max of an empty set is 0)
- Length of longest increasing subsequence:
 - $\max\{L[i] : 1 \leq i \leq n\}$

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Computing the actual sequence

- For each j , we computed

$$L[j] = 1 + \max\{L[i] : i < j \text{ and } s_i < s_j\}$$
 (where max of an empty set is 0)
- Also maintain $P[j]$ the value of the i that achieved that max
 - this will be the index of the predecessor of s_j in a longest increasing subsequence that ends in s_j
 - by following the $P[j]$ values we can reconstruct the whole sequence in linear time.

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Longest Increasing Subsequence Algorithm

- for $j=1$ to n do
 - $L[j] \leftarrow 1$
 - $P[j] \leftarrow 0$
 - for $i=1$ to $j-1$ do
 - if $(s_i < s_j \ \& \ L[i]+1 > L[j])$ then
 - $P[j] \leftarrow i$
 - $L[j] \leftarrow L[i]+1$
 - endfor
- Now find j such that $L[j]$ is largest and walk backwards through $P[j]$ pointers to find the sequence

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Example

i	1	2	3	4	5	6	7	8	9
s_i									
l_i									
p_i									

9

Example

i	1	2	3	4	5	6	7	8	9
s_i	90	50	20	80	70	30	10	60	40
l_i									
p_i									

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