

Longest increasing subsequence with dynamic programming (Brown CS  
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Another example of a problem that can be solved with Dynamic Programming is **longest increasing subsequence**: find the length of the longest increasing subsequence of a list

Approach: write down examples of what the answers would look like when we start with each element of the list.

Since the longest increasing sequence will have to start with one of those elements, the maximum such length will be our answer.

	A	B	C	D	E	F	G
1	3	7	4	2	5	is starting w/ 3:	3 the max
2							
3		7	4	2	5	is starting w/ 7:	1
4							
5			4	2	5	is starting w/ 4:	2
6							
7				2	5	is starting w/ 2:	2
8							
9					5	is starting w/ 5:	1
10							
11							

Note: why is 1 the answer for the LIS starting w/ 7? Because the sub-problem we are solving is to find the subsequence that starts with the element in question. This way of framing the problem will let us use the answers for the smaller sub-problems to answer the bigger sub-problems more efficiently!

How do we populate the table? One way to think about this is recursively: if we are considering the length of the longest subsequence starting with the element at index *start\_index*, and we consider every *possible\_next\_index* after start index, the answer for the longest increasing sequence starting with the element at *start\_index* will be (in pseudocode)

```
l_i_s(start_index, input_lst):
    if start_index == len(input_lst) - 1:
        return 1
    else:
        return 1 + max(l_i_s(possible_next_index, input_lst))
            # for all possible_next_index > start_index with
            # input_lst[possible_next_index] > start_index
```

So, the computation for index 3 of the list would look like:

$$l_i_s(3, \text{input\_lst}) = 1 + \max(l_i_s(4, \text{input\_lst})) \text{ (because } 4 > 3 \text{ and } \text{input\_lst}[4] > \text{input\_lst}[3] \text{ (} 5 > 2 \text{)}$$

$\downarrow$   
 $= 2$  (because  $\text{len}(\text{input\_lst})$  is 4)

The computation for index 1 of the list would look like:

$$l_i_s(1, \text{input\_lst}) = 1 + \max(\text{nothing}) \text{ (because there are no elements larger than } \text{input\_lst}[1] \text{ (} 7 \text{) in the rest of the list)}$$

$= 1$

The computation for index 0 of the list would look like:

$$l_i_s(0, \text{input\_lst}) = 1 + \max(l_i_s(1, \text{input\_lst}), l_i_s(2, \text{input\_lst}), l_i_s(4, \text{input\_lst})) = 3$$

$\downarrow$   
 $1 + \max(\text{nothing}) = 1$

$$1 + \max(l_i_s(4, \text{input\_lst})) = 2$$

$\downarrow$   
 $1$

redundant computations

We notice that that last computation has some redundancy when implemented recursively. Instead, we recognize that, if we store the result of these computations, from the largest starting index (smallest sub-list) to the smallest, then we could re-use them to do the computations for the smaller starting indices:

	A	B	C	D	E	F	G
1		3	7	4	2	5	
2		1	2	2	1		
3							
4		7	4	2	5		
5		1	2	2	1		
6							
7				4	2	5	
8				2	2	1	
9							
10				2	1		
11							
12				2	1		
13							
14							
15							
16							
17							
18							
19							
20							
21							
22							

Annotations on the table:

- Row 3: Handwritten note: "Analyze this".
- Row 4: Handwritten note: "then this".
- Row 5: Handwritten note: "then this".
- Row 6: Handwritten note: "do this next based on the computations done so far".
- Row 7: Handwritten note: "Start here".
- Row 8: Handwritten note: "1 + max(1, 2, 1)".
- Row 9: Handwritten note: "1 + max(nothin)".
- Row 10: Handwritten note: "1 + max(1)".
- Row 11: Handwritten note: "1 + max(1)".
- Row 12: Handwritten note: "1".

How do we translate this to python code? First, we set up the loop that allows us to fill in the computation table (longest\_seq\_so\_far) back-to-front:

*we are filling in this table*

```
def l_i_s(lst : "list[int]") -> int:
    """Produces the length of the LIS in lst"""
    longest_seq_so_far = [0] * len(lst)
    for start_index in range(len(lst) - 1, -1, -1):
        ...
        longest_seq_so_far[start_index] = ...
    return max(longest_seq_so_far)
```

*this loop  
occurs like this  
order of computation*

Then, we fill in the inner computation:

```
longest_seq_so_far = [0] * len(lst)
for start_index in range(len(lst) - 1, -1, -1):
    len_of_max_seq_from_start_index = 1
    for possible_next_index in range(start_index + 1, len(lst)):
        if lst[possible_next_index] > lst[start_index]:
            if longest_seq_so_far[possible_next_index] + 1 > len_of_max_seq_from_start_index:
                len_of_max_seq_from_start_index = longest_seq_so_far[possible_next_index] + 1
    longest_seq_so_far[start_index] = len_of_max_seq_from_start_index

return max(longest_seq_so_far)
```

*this corresponds  
to the highlighted  
cells above  
) this keeps  
the max  
updated*

We need the inner-most if statement to compute the max, by going through each possible\_next\_index and examining whether it will help us find the maximum possible subsequence starting with start\_index. We could have equivalently written the code using list comprehensions:

```

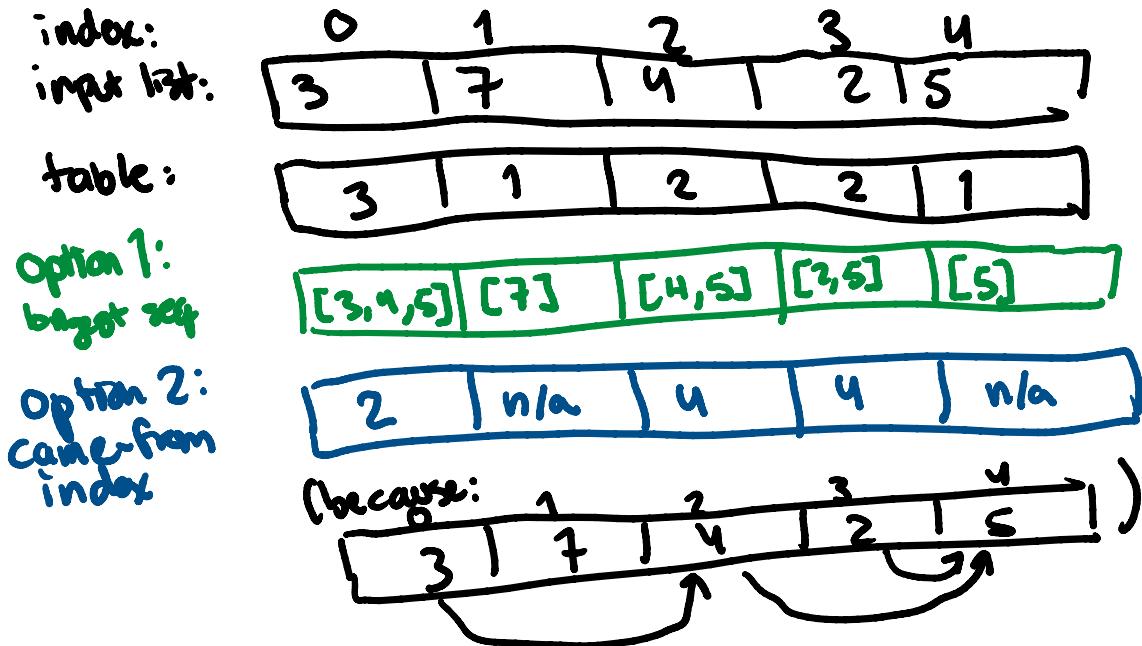
longest_seq_so_far = [0] * len(lst)
for start_index in range(len(lst) - 1, -1, -1):
    possible_next_indices = [next_index for next_index in range(start_index + 1, len(lst)) \
        if lst[next_index] > lst[start_index]]
    len_of_next_index_sequences = [longest_seq_so_far[next_index] for next_index in possible_next_indices]
    longest_seq_so_far[start_index] = 1 + max(len_of_next_index_sequences, default = 0)

return max(longest_seq_so_far)

```

this corresponds to  
the highlighted cells  
→ this gets the computed  
table vals at those  
indices

Instead of returning the length of the longest sequence, how would we return the sequence itself? There are two possible approaches we might take: keeping track of the sequence itself, or keeping track of the indices that got us to the sequence.



How would we insert this into our existing code?

```

longest_seq_so_far = [0] * len(lst)
for start_index in range(len(lst) - 1, -1, -1):
    len_of_max_seq_from_start_index = 1
    for possible_next_index in range(start_index + 1, len(lst)):
        if lst[possible_next_index] > lst[start_index]:
            if longest_seq_so_far[possible_next_index] + 1 > len_of_max_seq_from_start_index:
                len_of_max_seq_from_start_index = longest_seq_so_far[possible_next_index] + 1
    longest_seq_so_far[start_index] = len_of_max_seq_from_start_index

```

initialize array here  
(e.g. came\_from = [-1] \* len(lst))  
update new value here:  
(e.g. came\_from[start\_index] = possible-next-index)

for came\_from, loop back through, just like we did for graphs.

Compute argmax (index that got us the max)  
loop through came\_from until we get to -1,  
adding on the value of the list at each index