Question Recall the dynamic program for longest increasing subsequence (LIS) for an input sequence of n numbers a1,a2,...,an. If L(i) is the length of the LIS that ends in and includes a_i , then L(i) = 1 + max{L(j) : j < i and $a_i < a_i$ }

1. Give pseudocode that turns this formula for L(i) into an algorithm for finding the length of the LIS of the original sequence. Use the ideas of dynamic programming.

Solution

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\begin{split} L(1) &= 1 \\ k &= 0 \\ \text{for } i &= 2, ..., n \\ \text{a.} \quad L(i) &= 1 \\ & \text{for } j &= 1, ..., i \\ & \text{if } a_j < a_i \\ & L(i) &= \max\{L(i), 1 + L(j)\} \\ & k &= \max\{k, L(i)\} \end{split} return k
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2. What is the running time of your algorithm in terms of n?

solution $\Theta(n^2)$; being clever, you can implement this algorithm in $O(n \log n)$ time.

- 3. Prove that the formula $L(i) = 1 + max\{ L(j) : j < i \text{ and } a_j < a_i \}$ correctly computes the LIS that ends in and includes a_i
- 4. What is the longest increasing subsequence of the following input sequence?

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0 8 4 12 2 10 6 14 1 9 5 13 3 11 7 15
```

Solution 0, 2, 6, 9, 13, 15 or 0, 4, 6, 9, 11, 15 or ?

Practice questions

1. Modify the dynamic program for the knapsack problem to find a set of items of maximum value whose total weight is exactly the capacity of the knapsack. Note that for a given set of items, there may not be a subset of items whose total weight is exactly the capacity of the knapsack; in this case, your algorithm should correctly say there is no solution.

You should:

- (a) Define the dynamic programming table.
- (b) Give a recursive formula for an entry in the dynamic programming table.
- (c) Describe in words how to fill the dynamic programming table.
- (d) Give pseudocode for the final algorithm including how to find and return the items in the knapsack.

Solution

(a) aT (i,w) is the highest value knapsack using a subset of the items 1,2,3,..., i with total weight exactly equal to w. Let the entry $-\infty$ denote an infeasible knapsack (ie. there is no subset of the items 1,2,3,..., i with weight exactly w.

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(b) T(i,0) = 0, \forall i=0,...,n

T(0,w) = -\infty, \forall w = 1,...,W

T(i,w) = \max\{T(i-1,w),T(i-1,w-w_i) + v_i \text{ if } w_i \leq w\}
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- (c) Fill in the table either column-wise or row-wise.
- (d) Input n items with item i's weight w_i and value vi and the capacity of the knapsack W. initialize p and T as n+1 by W+1 arrays

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for i=0 ...n

T(i,0) = 0

for w=1 ...W

T(0,w) = -\infty

for i=1 ...n

for w=1 ...W

T(i,w) = T(i-1,w)

p(i,w) = w \text{ if } w_i \le w

if T(i-1,w-w_i) + vi > T(i,w)

T(i,w) = T(i-1,w-w_i) + vi

p(i,w) = w-w_i
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The final value of the knapsack is T(n,W). If this is $-\infty$, there is no solution. Otherwise, the set of items in the best knapsack can be placed in a linked list O by:

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i=n, w=W
initialize O as an empty linked list
while i>0
if p(i,w)=w-w_i % take item i add i to O w=w-w_i
i=i-1
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2. Give an O(nt) dynamic programming algorithm for the following task:

Input: A list of n positive integers a₁,a₂,...,a_n and a positive integer K. Question: Does some subset of the a_i's add up to K? (You can use each a_i at most once.)

You should:

- (a) Define the dynamic programming table.
- (b) Give a recursive formula for an entry in the dynamic programming table.
- (c) Describe in words how to fill the dynamic programming table.
- (d) Give pseudocode for the final algorithm.
- (e) Give the running time of your algorithm.

Solution

- (a) Let T (i,k) = 1 if there is a subset of the integers $a_1, a_2, ..., a_i$ that adds to k and o otherwise.
- (b) T(i,0)=1 for all i T(0,k)=0 for all k>0 $T(i,k) = max\{T(i-1,k),T(i-1,k-a_i)(if a_i \le k)\}$
- (c) Column-wise or row-wise.
- (d) initialize T as an $n+1 \times K + 1$ array for i=0 ...n T(i,0) = 1 for k=1 ...K T(0,k) = 0 for i=1 ...N for k=1 ...N T(i,k) = T(i-1,k) if $a_i \le k$ if $T(i-1,k-a_i) = 1$, T(i,k) = 1

If T(n,K) = 0, then there is no subset of the a_i 's that add up to K. If T(n,K) = 1, then there is a subset of the a_i 's that add up to K.

(e) The running time is O(nK). Note that this is not polynomial time.