

# CS-E3190 Principles of Algorithmic Techniques

## 03. Dynamic Programming – Graded Exercise

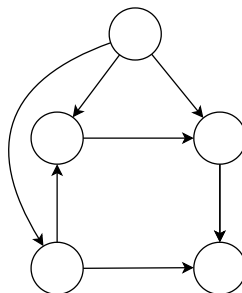
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Please read the following **rules** very carefully.

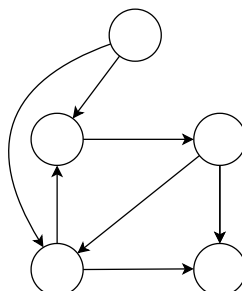
- Do not consciously search for the solution on the internet.
- You are allowed to discuss the problems with your classmates but you should **write the solutions yourself**.
- Be aware that **if plagiarism is suspected**, you could be asked to have an interview with teaching staff.
- Each week the second exercise is an **individual exercise**, and the teaching staff will not give hints or help with them. You are allowed to ask for hints for the first exercise.
- In order to ease grading, we want the solution of each problem and subproblem to start on a **new page**. If this requirement is not met, **points will be reduced**.

### 1. Basics of dynamic programming.

- (a) (3p.) Consider the recursive algorithm that was found last week for computing binomial coefficients.
- i. Draw the recursion tree for  $\text{Binom}(4, 2)$ . Draw a DAG based on the recursion tree. Each function call with specific parameters can only appear once in the DAG.
  - ii. Design a dynamic algorithm using memoization to improve the recursive algorithm.
  - iii. Analyze the runtime and the memory complexity of your algorithm.
- (b) (2p.) For full points you need to justify your answers.
- i. Inspect the following graph. Is it a DAG?



- ii. How about the graph below?



2. **Individual exercise: Knapsack.** Consider the following problem called Knapsack problem. Given a knapsack with capacity  $C \in \mathbb{N}$ , and a set of items  $I = \{1, 2, \dots, n\}$  such that item  $i$  has value  $v_i \in \mathbb{N}$  and weight  $w_i \in \mathbb{N}$ , select a subset  $S \subseteq I$  of the items to pack that maximises total value without exceeding the weight capacity  $C$ . Denote the total value of the subset  $S$  by  $V(S) = \sum_{i \in S} v_i$  and the weight by  $W(S) = \sum_{i \in S} w_i$ . If  $S$  is empty both of the sums are 0. The optimal value of the Knapsack problem can be expressed as

$$OPT = \max_{S \subseteq I} \{V(S) \text{ subject to } W(S) \leq C\}$$

*Sub-problems.* Let  $I_k = \{1, \dots, k\}$  for  $k \in \{1, \dots, n\}$ , and  $I_0 = \emptyset$ . Then  $(I_k, w)$  defines a sub-problem in which we maximise the value of a selection  $S_k \subseteq I_k$  subject to  $W(S_k) \leq w$ . Note that  $V(0, w) = 0, \forall w$ , because there are no items to pack. Denote the optimum value of the subproblem  $(I_k, w)$  by

$$V(k, w) = \max_{S_k \subseteq I_k} \{V(S_k) \text{ subject to } W(S_k) \leq w\}.$$

It follows from the definition that  $V(n, C) = OPT$ . Moreover  $V(k+1, w) \geq V(k, w), \forall k \in \{0, 1, \dots, n-1\}, \forall w \geq 0$ , since having more items to choose from can only improve the value.

- (a) (2p.) The optimal value  $V(k, w) = V(S_k^*)$  satisfies

$$V(S_k^*) = \begin{cases} V(k-1, w) & \text{if } k \notin S_k^* \\ V(k-1, w - w_k) + v_k & \text{if } k \in S_k^*. \end{cases} \quad (1)$$

Using this fact, prove that

$$V(k, w) = \begin{cases} V(k-1, w) & \text{if } w_k > w \\ \max\{V(k-1, w), V(k-1, w - w_k) + v_k\} & \text{if } w_k \leq w \end{cases} \quad (2)$$

*Hint: What decisions regarding  $S \subseteq I$  do the various terms represent? If proving the claim is difficult, try filling in a table of  $V(k, w)$ -values on a toy instance.*

- (b) (1p.) Consider the following pseudocode for the recursive algorithm solving the problem. Draw the recursion tree for the problem for the following items and capacity 10.

item	value	weight
1	5	3
2	1	2
3	7	6
4	5	5

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**Algorithm 1:** Knapsack( $I, (v_k, w_k)_{k \in I}, C$ )

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 $n \leftarrow |I|$ 
if  $I = \emptyset$  or  $C = 0$  then
  | return 0
end
else
  | if  $w_n > C$  then
    | /* Remove the last item, it can't fit. */
    | return Knapsack( $I \setminus \{n\}, C$ )
  | else
    | return max{Knapsack( $I \setminus \{n\}, C$ ), Knapsack( $I \setminus \{n\}, C - w_n$ ) +  $v_n$ }
  | end
end
return  $V[n, C]$ 
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- (c) (2p.) Based on the recursive algorithm, design a  $O(nC)$  time dynamic algorithm using  $O(C)$  memory. (*Hint: Because the weights are assumed to be integers, we can iterate through all the subproblems  $(I_k, w)$ ,  $k = 0, \dots, n$  and  $w = 0, \dots, C$ .)*)