

Informatics 2 – Introduction to Algorithms and Data Structures

Tutorial 7 - Dynamic Programming

1. Consider the weighted directed graph $G = (V, E)$ of Figure 1. Run the Bellman-Ford algorithm to compute the value of $M[i, v]$ for every node $x \in V$. Recall that $M[i, v]$ is the cost of the minimum-cost $v \sim t$ path that uses at most i edges.

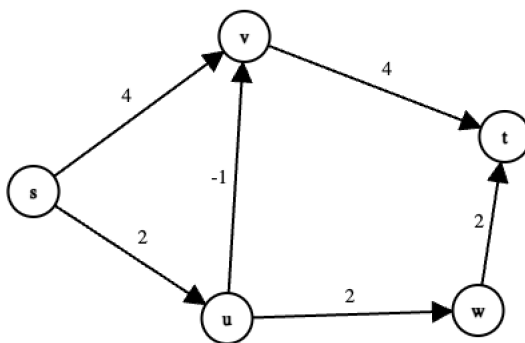


Figure 1: A directed graph with edge costs indicated. Algorithms Illuminated Example 18.2.6.

SOLUTION: We run the algorithm referring to the pseudocode presented in Lecture 20. In the initialisation step, we have $M[0, t] = 0$ and $M[0, x] = \infty$ for every other node $x \in V \setminus \{t\}$. Then we enter the nested for loops.

- Let $i = 1$. We consider the nodes one by one and compute $M[1, x]$ for each one of them. For this, we use the following recurrence relation

$$M[1, x] = \min\{M[0, x] + \min_{y \in N(x)} c_{xy} + M[0, y]\},$$

where $N(x)$ is the set of nodes that are reachable from x via a single edge. In our calculation, we observe that $M[0, x] = \infty$ for every node $x \in V \setminus \{t\}$. Therefore the nodes for which $M[1, x] \neq \infty$ (besides t) will be nodes v and w , which can reach t via a single edge. In other words, we have

$$M[1, s] = M[1, u] = \infty, M[1, t] = 0$$

For $M[1, v]$ we calculate the minimum of the expression as $M[1, v] = c_{vt} + M[0, t] = 4$. Similarly, for $M[1, w]$ we calculate $M[1, w] = c_{wt} + M[0, t] = 2$.

- Let $i = 2$. We consider the nodes one by one and compute $M[2, x]$ for each one of them. For this, we use the following recurrence relation

$$M[2, x] = \min\{M[1, x] + \min_{y \in N(x)} c_{xy} + M[1, y]\}.$$

We calculate

$$M[2, s] = \min\{M[1, s] + \min_{y \in N(s)} c_{sy} + M[1, y]\} = c_{sv} + M[1, v] = 4 + 4 = 8$$

$$M[2, v] = \min\{M[1, v] + \min_{y \in N(v)} c_{vy} + M[1, y]\} = M[1, v] = 4$$

$$M[2, u] = \min\{M[1, u] + \min_{y \in N(u)} c_{uy} + M[1, y]\} = c_{uv} + M[1, v] = -1 + 4 = 3$$

$$M[2, w] = \min\{M[1, w] + \min_{y \in N(w)} c_{wy} + M[1, y]\} = M[1, w] = 2$$

$$M[2, t] = 0$$

- Let $i = 3$. We consider the nodes one by one and compute $M[3, x]$ for each one of them. For this, we use the following recurrence relation

$$M[3, x] = \min\{M[2, x] + \min_{y \in N(x)} c_{xy} + M[2, y]\}.$$

Similarly to above we calculate

$$M[3, s] = \min\{M[2, s] + \min_{y \in N(s)} c_{sy} + M[2, y]\} = c_{su} + M[2, u] = 2 + 3 = 5$$

$$M[3, v] = \min\{M[2, v] + \min_{y \in N(v)} c_{vy} + M[2, y]\} = M[2, v] = 4$$

$$M[3, u] = \min\{M[2, u] + \min_{y \in N(u)} c_{uy} + M[2, y]\} = M[2, u] = 3$$

$$M[3, w] = \min\{M[2, w] + \min_{y \in N(w)} c_{wy} + M[2, y]\} = M[2, w] = 2$$

$$M[3, t] = 0$$

- Let $i = 4$. We consider the nodes one by one and compute $M[4, x]$ for each one of them. For this, we use the following recurrence relation

$$M[4, x] = \min\{M[3, x] + \min_{y \in N(x)} c_{xy} + M[3, y]\}.$$

It is not hard to see that in this case we will have $M[4, x] = M[3, x]$ for all $x \in V$. In the end, the 2D array M looks as follows.

| | | | | | |
|-----|----------|----------|---|---|---|
| | 0 | 1 | 2 | 3 | 4 |
| s | ∞ | ∞ | 8 | 5 | 5 |
| u | ∞ | ∞ | 3 | 3 | 3 |
| v | ∞ | 4 | 4 | 4 | 4 |
| w | ∞ | 2 | 2 | 2 | 2 |
| t | 0 | 0 | 0 | 0 | 0 |

2. Assume that we wanted to use the Bellman-Ford algorithm to find the cost of the minimum-cost paths from a node s to all the nodes $x \in V$ in the graph G . Think about how to modify the algorithm to achieve this and run the modified algorithm on the graph of Figure 1 to compute the costs of all the minimum-cost paths from s to the nodes in V .

Solution: The modification that we need to make is that now $M[i, x]$ will denote the cost of the minimum-cost path from node s to node $x \in V$ that uses at most i edges. In our initialisation step, we will have $M[0, s] = 0$ and $M[0, x] = \infty$ for every node $x \in V \setminus \{s\}$. Our recurrence relation will still be

$$M[i, x] = \min\{M[i-1, x] + \min_{y \in N^-(x)} c_{yx} + M[i-1, y]\},$$

where now $N^-(x)$ denotes the set of nodes y for which there is an edge $(y, x) \in E$.

If we run the algorithm on the graph of Figure 1, we get the following 2D array M :

| | | | | | |
|-----|----------|----------|---|---|---|
| | 0 | 1 | 2 | 3 | 4 |
| s | 0 | 0 | 0 | 0 | 0 |
| u | ∞ | 2 | 2 | 2 | 2 |
| v | ∞ | 4 | 1 | 1 | 1 |
| w | ∞ | ∞ | 4 | 4 | 4 |
| t | ∞ | ∞ | 8 | 5 | 5 |

3. Consider the knapsack problem given by the following table, with capacity $W = 7$.

| Item | Value | Weight |
|------|-------|--------|
| 1 | 1 | 1 |
| 2 | 2 | 3 |
| 3 | 3 | 2 |
| 4 | 4 | 5 |
| 5 | 5 | 5 |

Use the dynamic programming algorithm presented in the lectures to compute the value of the optimal solution.

Solution: Recall that we will use a 2D array for which the entry $M[i, w]$ will correspond to the optimal solution using the first i intervals and capacity w . For the capacities w , we will consider all of the integers that are at most $W = 7$, i.e., $\{0, 1, 2, 3, 4, 5, 6, 7\}$.

Our algorithm first sets $M[0, w] = 0$ for all $w \in \{1, \dots, 7\}$. This results in our partially filled 2D array looking like:

Then we run the first outer loop, for $i = 1$, and calculate the value of $M[1, w]$ for every $w \in \{1, \dots, 7\}$. For that, we use the recurrence relation:

$$M[1, w] = \max\{M[0, w], v_i + M[0, w - w_i]\},$$

if $w_i > w$ (i.e., the item fits), otherwise we set $M[1, w] = M[0, w]$. Let's run this iteration explicitly:

| | | | | | | | | |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| 5 | | | | | | | | |
| 4 | | | | | | | | |
| 3 | | | | | | | | |
| 2 | | | | | | | | |
| 1 | | | | | | | | |
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |

$M[1, 0] = M[0, 0]$ since $w_1 > 0$. When considering $w = 1$, the item now fits, as $w_1 = 1$. Therefore we have to check the recurrence relation. We have that $M[0, 1] = M[0, 0] = 0$, so the maximum is given by the second term and is equal to $v_i = 1$. Therefore we have $M[1, 1] = 1$. Similarly, for $M[1, 2]$ we have that $M[0, 2] = M[0, 1] = 0$, and the maximum is given again by the second term. We again have $M[1, 2] = 1$. Similarly, we calculate $M[1, x] = 1$ for all $x \in \{1, \dots, 7\}$. Using this, we can populate our $2D$ array as follows:

| | | | | | | | | |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| 5 | | | | | | | | |
| 4 | | | | | | | | |
| 3 | | | | | | | | |
| 2 | | | | | | | | |
| 1 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |

Now let's consider the next iteration, when $i = 2$. We calculate the value of $M[2, w]$ for every $w \in \{1, \dots, 7\}$. For that, we use the recurrence relation:

$$M[2, w] = \max\{M[1, w], v_i + M[1, w - w_i]\},$$

if $w_i > w$ (i.e., the item fits), otherwise we set $M[2, w] = M[1, w]$. Let's also run this iteration explicitly: For $w = 0$, $w = 1$ or $w = 2$, we see that $3 = w_2 > w$, so for those cases we will have $M[2, x] = M[1, x]$ for $x \in \{0, 1, 2\}$. This means that $M[2, 0] = 0$ and $M[2, 1] = M[2, 2] = 1$. For $w = 3$, we now have $w_2 = w$, and we can use the recurrence relation. We have that $M[1, 3] = 1$ and $v_2 + M[1, 0] = 2 + 0$, so we will have $M[2, 3] = 2$. For $w = 4$, we have $M[1, 4] = 1$ and $v_2 + M[1, 1] = 2 + 1 = 3$, so we have $M[2, 4] = 3$. Continuing like this, we can compute $M[2, x] = 3$ for all $x \in \{5, 6, 7\}$. Using this, we can populate our $2D$ array as follows:

| | | | | | | | | |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| 5 | | | | | | | | |
| 4 | | | | | | | | |
| 3 | | | | | | | | |
| 2 | 0 | 1 | 1 | 2 | 3 | 3 | 3 | 3 |
| 1 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |

Continuing like this, in the end we complete our $2D$ array:

| | | | | | | | | |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| 5 | 0 | 1 | 3 | 4 | 4 | 5 | 6 | 8 |
| 4 | 0 | 1 | 3 | 4 | 4 | 5 | 6 | 7 |
| 3 | 0 | 1 | 3 | 4 | 4 | 5 | 6 | 6 |
| 2 | 0 | 1 | 1 | 2 | 3 | 3 | 3 | 3 |
| 1 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |

4. Recall the following simple context-free grammar for arithmetic expressions from Lecture 21. The start symbol is Exp .

$$\begin{aligned} \text{Exp} &\rightarrow \text{Var} \mid \text{Num} \mid (\text{Exp}) \\ \text{Exp} &\rightarrow \text{Exp} + \text{Exp} \\ \text{Exp} &\rightarrow \text{Exp} * \text{Exp} \\ \text{Var} &\rightarrow x \mid y \mid z \\ \text{Num} &\rightarrow 0 \mid \dots \mid 9 \end{aligned}$$

- (a) How many syntax trees are there for each of the following three strings? Draw them all.

$$3 + x * y \qquad 3 + (x * y) \qquad z + 10$$

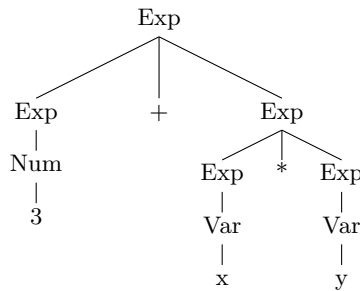
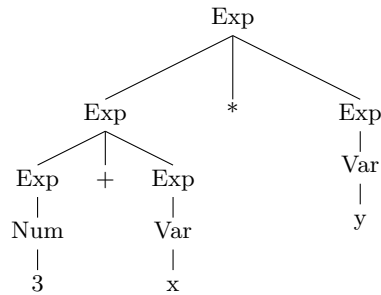
- (b) Design a new context-free grammar that generates exactly the same language as the one above, but with the property that it is *unambiguous*: every string in the language should have exactly one syntax tree. Informally, your grammar should enforce the familiar convention that $*$ takes precedence over $+$. You will find it helpful to introduce some additional non-terminal symbols.

[Hint: First try to do this for the grammar with the rule for $\text{Exp} * \text{Exp}$ omitted. To ensure that a string like $3 + 4 + 5$ has only one tree, you might want to draw inspiration from the grammar for comma-separated lists in Lecture 21. Then try to adapt your grammar to cater for $*$, building in the precedence rule.]

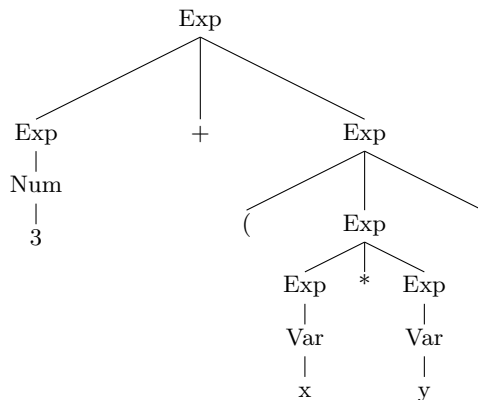
- (c) For the grammar you have designed in part (b), draw the *unique* syntax tree for any of the strings from part (a) that had more than one syntax tree with respect to the original grammar.

SOLUTION:

1. (a) $3 + x * y$ has two trees:



$3 + (x * y)$ has just one tree:



$z + 10$ has no trees. This is not a sentence of the language: our grammar doesn't cater for multi-digit numerals like 10.

- (b) First for the grammar with the clause for $*$ omitted: The key observation is that a general expression is a list of one or more 'simple expressions', separated by $+$. Drawing inspiration from the comma list example, the following grammar does the trick:

$$\begin{aligned} \text{Exp} &\rightarrow \text{SimpleExp PlusList} \\ \text{SimpleExp} &\rightarrow \text{Var} \mid \text{Num} \mid (\text{Exp}) \\ \text{PlusList} &\rightarrow \epsilon \mid + \text{SimpleExp PlusList} \end{aligned}$$

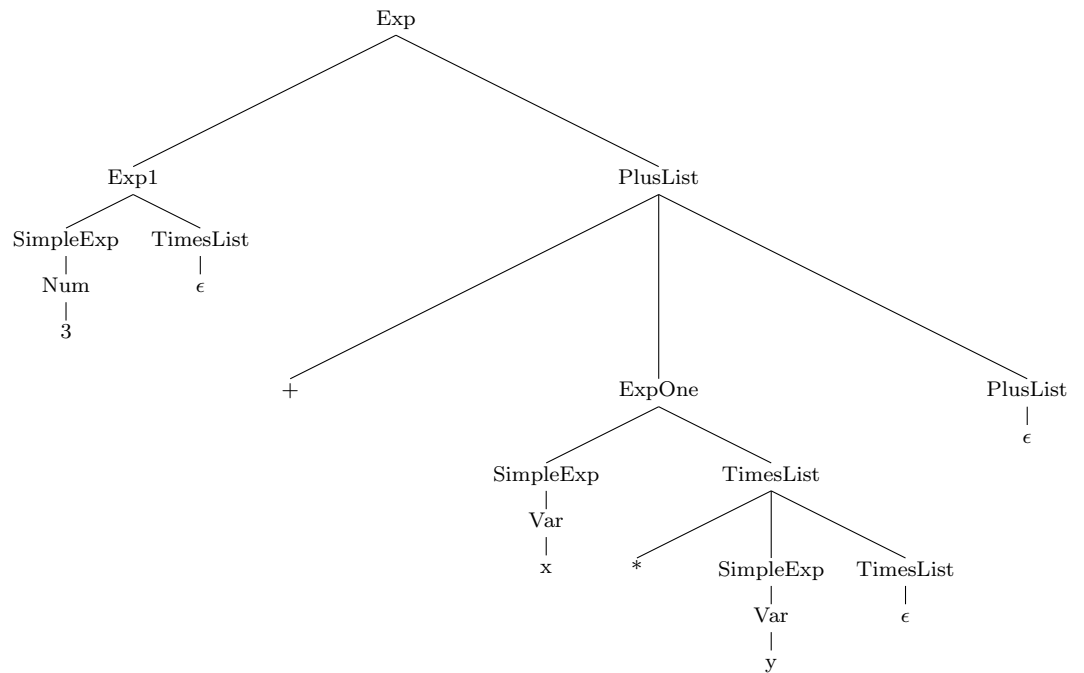
(with the same rules for Var and Num as before). Intuitively, we here distinguish two 'levels' of expressions, corresponding to Exp and SimpleExp . To cater for $*$ as

well, and to enforce the precedence rule, we can extend this idea to allow three levels:

$$\begin{aligned} \text{Exp} &\rightarrow \text{Exp1 PlusList} \\ \text{Exp1} &\rightarrow \text{SimpleExp TimesList} \\ \text{SimpleExp} &\rightarrow \text{Var} \mid \text{Num} \mid (\text{Exp}) \\ \text{TimesList} &\rightarrow \epsilon \mid * \text{SimpleExp TimesList} \\ \text{PlusList} &\rightarrow \epsilon \mid + \text{Exp1 PlusList} \end{aligned}$$

(plus the usual `Var` and `Num` rules). Other solutions are possible, but the above grammar turns out to be particularly well-adapted to ‘left-to-right parsing’.)

(c) The unique syntax tree for $3 + x * y$ is now:



Here we include explicit ϵ 's for clarity, though of course they contribute nothing to the string in question.