

## Tutorial 7: solution sketches

1. The Bellman optimality equations are as follows:

$$\begin{aligned}
 x_6 &= 1 \\
 x_5 &= \max\{x_1, x_2\} \\
 x_4 &= x_4 \\
 x_3 &= \max\{x_2, x_4\} \\
 x_2 &= 2x_1/5 + x_4/5 + 2x_6/5 \\
 x_1 &= x_2/6 + x_4/6 + x_5/6 + x_6/2
 \end{aligned}$$

Our goal is to compute the unique minimal (least non-negative) solution vector  $p^* = (p_1^*, \dots, p_6^*)$ , which gives the optimal probabilities of reaching  $s_6$  starting from each state. It is clear that  $p_4^* = 0$ , so that at  $s_3$  the node  $s_2$  is always chosen, giving  $p_3^* = p_2^*$ . Also, since  $p_6^* = 1$ , it only remains to solve for  $p_1^*, p_2^*$ , and  $p_5^*$ . From the optimality conditions we see that the equations governing these are as follows:

$$\begin{aligned}
 p_5^* &= \max\{p_1^*, p_2^*\} \\
 p_2^* &= 2p_1^*/5 + 2/5 \\
 p_1^* &= p_2^*/6 + p_5^*/6 + 1/2
 \end{aligned}$$

We can (as shown on the lecture slides) solve this by computing the unique optimal solution for the following linear programming problem:

**Minimize:**  $x_1 + x_2 + x_5$

**Subject to:**

$$x_5 \geq x_1$$

$$x_5 \geq x_2$$

$$x_2 \geq (2/5)x_1 + (2/5)$$

$$x_1 \geq (1/6)x_2 + (1/6)x_5 + (1/2)$$

In this small example, we can also avoid using linear programming, by enumerating all possible cases of the max equation. There are two cases to consider: (i)  $\max\{p_1^*, p_2^*\} = p_2^*$  and (ii)  $\max\{p_1^*, p_2^*\} = p_1^*$ . In both of these cases we know the value of  $p_5^*$ , so we can calculate the rest.

In case (i), the equations reduce to

$$\begin{aligned} p_2^* &= 2p_1^*/5 + 2/5 \\ p_1^* &= p_2^*/3 + 1/2 \end{aligned}$$

These can be solved to get  $p_1^* = 19/26$  and  $p_2^* = 18/26$ . This contradicts our assumption that  $p_2^* = \max\{p_1^*, p_2^*\}$ .

In case (ii), the equations reduce to

$$\begin{aligned} p_2^* &= 2p_1^*/5 + 2/5 \\ p_1^* &= p_1^*/6 + p_6^*/6 + 1/2 \end{aligned}$$

which gives us  $p_1^* = 17/23$  and  $p_2^* = 16/23$ . This gives us the full solution to the original problem:  $p^* = (p_1^*, \dots, p_6^*) = (17/23, 16/23, 16/23, 0, 17/23, 1)$ . Player 1s optimal strategy is to choose  $s_2$  when at node  $s_3$ , and to choose  $s_1$  when at node  $s_5$ .

2. As we are working with a congestion game, we can find a pure Nash Equilibrium by starting at any pure strategy profile, and iteratively improving it until we can't. To get a concrete starting point, let's say all players take the route  $s \rightarrow v_3 \rightarrow t$ . Then we can do iterative improvements for example<sup>1</sup> as follows:

- (i) Player 1 switches to  $s \rightarrow v_2 \rightarrow v_1 \rightarrow t$
- (ii) Player 2 switches to  $s \rightarrow v_2 \rightarrow v_1 \rightarrow t$
- (iii) Player 3 switches to  $s \rightarrow v_1 \rightarrow t$ .
- (iv) Player 2 switches to  $s \rightarrow v_1 \rightarrow t$

At (iv) no further improvements can be made, so we reached the following NE:

Player 1:  $s \rightarrow v_2 \rightarrow v_1 \rightarrow t$   
 Player 2:  $s \rightarrow v_1 \rightarrow t$   
 Player 3:  $s \rightarrow v_1 \rightarrow t$

Note that in the above sequence we weren't done at stage (iii), even though every player had switched once. Other starting points will take

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<sup>1</sup>at many stages there's more than one option on who improves and how

through other sequences of steps, and they might end up in a different NE, although it turns out that in this game all pure Nash equilibria send two players via the route  $s \rightarrow v_1 \rightarrow t$  and one via  $s \rightarrow v_2 \rightarrow v_1 \rightarrow t$ , differing only in which player chooses the path  $s \rightarrow v_2 \rightarrow v_1 \rightarrow t$ .