

## Mario Exercise

Design a Super Mario Bros MDP

- Reward function  $R$
- Discount factor  $\gamma$

Your states are: eat mushroom, collect coins, die, game over

Compute discounted return for:

- Eat mushroom at  $t = 10$
- Collect coins at  $t = 11, 12$
- Die to bowser at  $t = 20$
- Game over screen at  $t = 21 \dots \infty$
- $r = 0$  for other timesteps

## Answer

We have the states:

$$S = \{\text{mushroom, coin, die, game over, } 0\}$$

You should define some scalar reward for each, this is up to you to decide, but you probably want game over to be zero so your return is finite

$$R(\text{mushroom}) = 2, R(\text{coin}) = 1, R(\text{die}) = -5, R(\text{game over}) = 0$$

You can choose any  $\gamma$  between 0 and 1

$$\gamma = 0.9$$

Then, the return is

$$\begin{aligned} G &= \gamma^{10} R_{\text{mushroom}} + \gamma^{11} R_{\text{coin}} + \gamma^{12} R_{\text{coin}} + \gamma^{20} R_{\text{die}} \\ G &= 0.9^{10} \cdot 2 + 0.9^{11} \cdot 1 + 0.9^{12} \cdot 1 + 0.9^{20} \cdot -5 \\ G &\approx 0.686 \end{aligned}$$

## Markov Process Exercise

Design an Markov process about a problem you care about

- 4 states
- State transition function  $\text{Tr} = \Pr(s_{t+1} \mid s_t)$  for all  $s_t, s_{t+1} \in S$
- Create a terminal state
- Given a starting state  $s_0$ , what will your state distribution be for  $s_2$ ?

$$\Pr(s_n \mid s_0) = \sum_{s_1, s_2, \dots, s_{n-1} \in S} \prod_{t=0}^{n-1} \Pr(s_{t+1} \mid s_t)$$

### Answer

**States:**

$$S = \{S_a, S_b, S_c, S_d\}$$

$S_d$  is a terminal state

**Transition Function (Tr):**

You can come up with whatever state transition function you want, as long as each section sums to one

**From  $s_a$ :**

$$\Pr(s_a \mid s_a) = 0.5$$

$$\Pr(s_b \mid s_a) = 0.3$$

$$\Pr(s_c \mid s_a) = 0.1$$

$$\Pr(s_d \mid s_a) = 0.1$$

**From  $s_b$ :**

$$\Pr(s_a \mid s_b) = 0.2$$

$$\Pr(s_b \mid s_b) = 0.6$$

$$\Pr(s_c \mid s_b) = 0.1$$

$$\Pr(s_d \mid s_b) = 0.1$$

**From  $s_c$ :**

$$\Pr(s_a \mid s_c) = 0.1$$

$$\Pr(s_b \mid s_c) = 0.1$$

$$\Pr(s_c \mid s_c) = 0.7$$

$$\Pr(s_d \mid s_c) = 0.1$$

**From  $s_d$  (terminal):**

$$\Pr(s_a \mid s_d) = 0$$

$$\Pr(s_b \mid s_d) = 0$$

$$\Pr(s_c \mid s_d) = 0$$

$$\Pr(s_d \mid s_d) = 1.0$$

### Roll the Process Forward

Compute  $\Pr(s_2 \mid s_0 = s_a)$  by summing over all paths through intermediate states  $s_1$ . You can either compute all conditional probabilities at each timestep, or you can just enumerate all possible paths from  $s_0$  to  $s_2$ . In this case, I choose the latter.

**For  $s_2 = s_a$ :**

$$\begin{aligned} & \Pr(s_a \rightarrow s_a \rightarrow s_a) + \Pr(s_a \rightarrow s_b \rightarrow s_a) + \Pr(s_a \rightarrow s_c \rightarrow s_a) + \Pr(s_a \rightarrow s_d \rightarrow s_a) \\ &= (0.5 * 0.5) + (0.3 * 0.2) + (0.1 * 0.1) + (0.1 * 0) \\ &= 0.25 + 0.06 + 0.01 + 0 = 0.32 \end{aligned}$$

**For  $s_2 = s_b$ :**

$$\begin{aligned} & \Pr(s_a \rightarrow s_a \rightarrow s_b) + \Pr(s_a \rightarrow s_b \rightarrow s_b) + \Pr(s_a \rightarrow s_c \rightarrow s_b) + \Pr(s_a \rightarrow s_d \rightarrow s_b) \\ &= (0.5 * 0.3) + (0.3 * 0.6) + (0.1 * 0.1) + (0.1 * 0) \\ &= 0.15 + 0.18 + 0.01 + 0 = 0.34 \end{aligned}$$

**For  $s_2 = s_c$ :**

$$\begin{aligned} & \Pr(s_a \rightarrow s_a \rightarrow s_c) + \Pr(s_a \rightarrow s_b \rightarrow s_c) + \Pr(s_a \rightarrow s_c \rightarrow s_c) + \Pr(s_a \rightarrow s_d \rightarrow s_c) \\ &= (0.5 * 0.1) + (0.3 * 0.1) + (0.1 * 0.7) + (0.1 * 0) \\ &= 0.05 + 0.03 + 0.07 + 0 = 0.15 \end{aligned}$$

**For  $s_2 = s_d$ :**

$$\begin{aligned} & \Pr(s_a \rightarrow s_a \rightarrow s_d) + \Pr(s_a \rightarrow s_b \rightarrow s_d) + \Pr(s_a \rightarrow s_c \rightarrow s_d) + \Pr(s_a \rightarrow s_d \rightarrow s_d) \\ &= (0.5 * 0.1) + (0.3 * 0.1) + (0.1 * 0.1) + (0.1 * 1.0) \\ &= 0.05 + 0.03 + 0.01 + 0.10 = 0.19 \end{aligned}$$

**Final Distribution for  $s_2$ :**

Make sure this sums to one

$$\Pr(s_2 = s_a \mid s_0 = s_a) = 0.32$$

$$\Pr(s_2 = s_b \mid s_0 = s_a) = 0.34$$

$$\Pr(s_2 = s_c \mid s_0 = s_a) = 0.15$$

$$\Pr(s_2 = s_d \mid s_0 = s_a) = 0.19$$