

# Markov Decision Process



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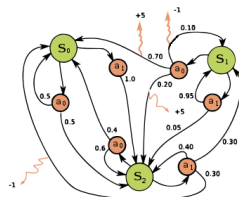
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## Markov Decision Processes

- describe environment in RL framework
- describe dynamical systems
- In optimal control problems MDPs are continuous

A Markov Decision Process (MDP) is a tuple  $\langle \mathcal{S}, \mathcal{A}, \mathcal{P}, \mathcal{R}, \gamma \rangle$

- $\mathcal{S}$ : The set of states.
- $\mathcal{A}$ : The set of actions.
- $\mathcal{P}$ : The set of transition probability.
- $\mathcal{R}$ : The set of immediate rewards associated with the state-action pairs.
- $0 \leq \gamma \leq 1$ : Discount factor.



Modified version of @ [https://en.wikipedia.org/wiki/Markov\\_decision\\_process](https://en.wikipedia.org/wiki/Markov_decision_process)

**States:** Describe internal status of MDP

**Actions:** Possible choices to make in each state of MDP

The state and action space can be finite or infinite and it is extremely important!

**Transitions probability:**  $\mathcal{P}$  is the set of transition probability with  $n_a$  matrices each of dimension  $n_s \times n_s$  where  $s, s'$  entry reads

$$[\mathcal{P}^a]_{ss'} = p[s_{t+1} = s' | s_t = s, a_t = a] \quad (1)$$

## Reward:

$$r_t = r(s, a) \quad (2)$$

Total reward:

$$R(T) = \sum_{t=1}^T \gamma^t r_t \quad (3)$$

Average reward:

$$R(T) = \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T r_t \quad (4)$$

# Do you care about future as much as now (and past)?

- $\gamma \rightarrow 0$ : We only care about the current reward not what we'll receive in future
- $\gamma \rightarrow 1$ : We care about all rewards equally

$$\mathcal{S} = \{s_0, s_1, s_2\},$$

$$\mathcal{A} = \{a_0, a_1\},$$

$$\mathcal{P}^{a_0} = \begin{bmatrix} 0.5 & 0 & 0.5 \\ 0.7 & 0.1 & 0.2 \\ 0.4 & 0 & 0.6 \end{bmatrix},$$

$$\mathcal{P}^{a_1} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0.95 & 0.05 \\ 0.3 & 0.3 & 0.4 \end{bmatrix}.$$

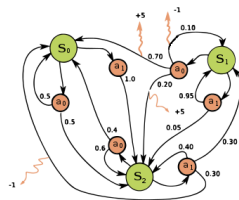


Photo Credit: © [https://en.wikipedia.org/wiki/Markov\\_decision\\_process](https://en.wikipedia.org/wiki/Markov_decision_process)



- Policy: The agent's decision
  - Deterministic policy  $a = \pi(s)$
  - stochastic policy  $\pi(a|s) = P[a_t = a | s_t = s]$
- Value function: how good the agent does in a state

$$V(s) = \mathbf{E} \left[ r_t + \gamma r_{t+1} + \gamma^2 r_{t+2} + \dots | s_t = s \right]$$

- Model: The agent's interpretation of the environment

Not all components are necessary!

# Email your questions to

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