


# Bellman Equations

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Prof. Subrahmanya Swamy

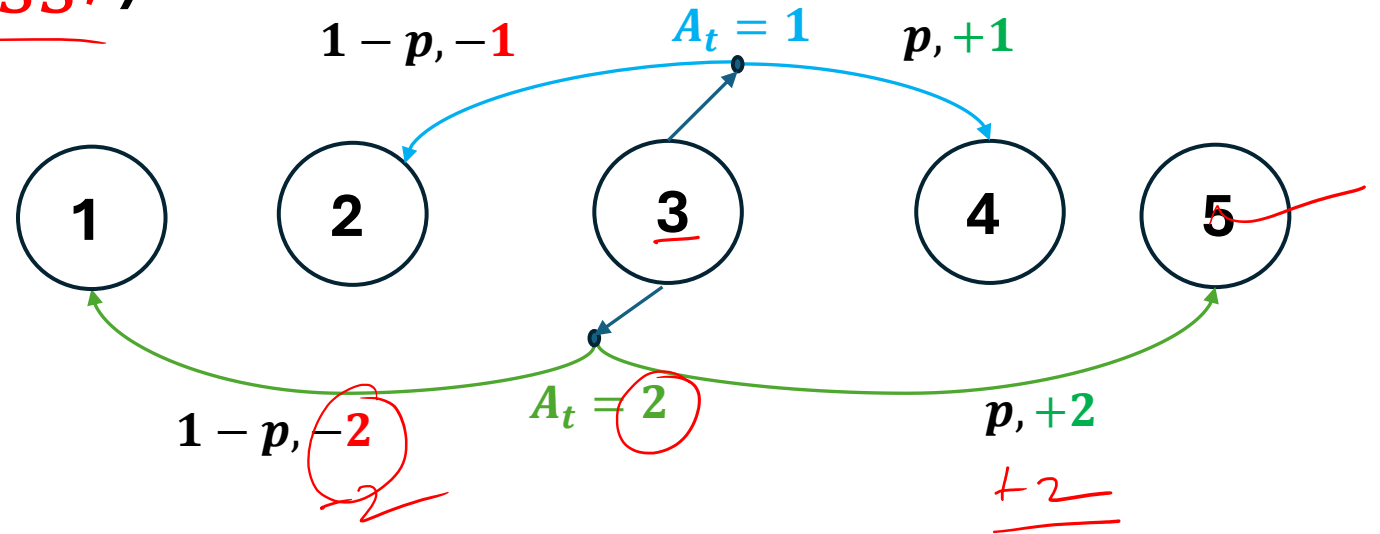
# Outline

- MDP Dynamics  $\underline{R_s^a}, \underline{P_{ss'}^a}$  ✓
  - Policy Dynamics  $R_s^\pi, P_{ss'}^\pi$  ✓
  - Value Function  $V_\pi(s)$  ✓
  - Action-Value Function  $Q_\pi(s, a)$  ✓
  - Bellman Equations
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# MDP Dynamics ( $R_s^a, P_{ss'}^a$ )

## Transition Probability

- $P_{ss'}^a = \mathbb{P}(S_{t+1} = s' \mid S_t = s, A_t = a)$
- Example:  $P_{3,5}^2 = p$  ✓



## Expected Reward

- $R_s^a = \mathbb{E}[R_{t+1} \mid S_t = s, A_t = a]$
- Example:

$$\begin{aligned}
 R_3^2 &= 2p - 2(1-p) \\
 &= 4p - 2 \\
 &= -1 \quad (\text{if } p = \frac{1}{4})
 \end{aligned}$$

# Policy Dynamics $(R_s^\pi, P_{ss'}^\pi)$

## Transition Probability

- $P_{ss'}^\pi = \mathbb{P}(S_{t+1} = s' \mid S_t = s, A_t \sim \pi)$
- $= \sum_a \pi(a \mid s) P_{ss'}^a$

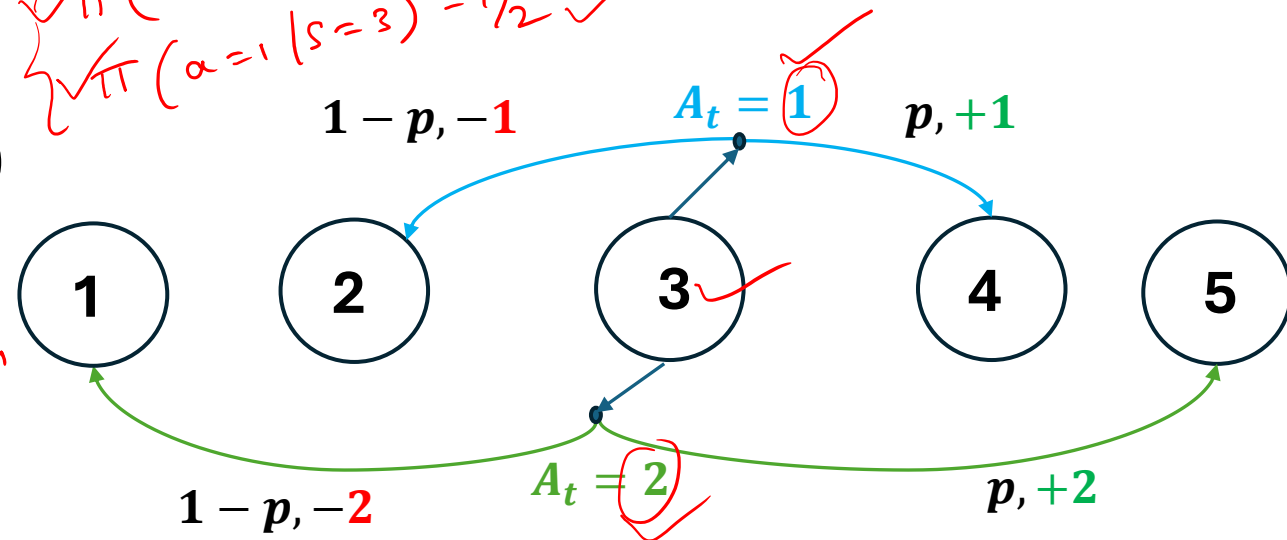
$$P_{ss'}^\pi = \frac{1}{2} \cdot P_{ss'}^2 + \frac{1}{2} P_{ss'}^1$$

$$\begin{aligned} \checkmark \frac{1}{11} (a=2 \mid s=3) &= \frac{1}{2} \\ \checkmark \pi(a=1 \mid s=3) &= \frac{1}{2} \checkmark \end{aligned}$$

## Expected Reward

- $R_s^\pi = \mathbb{E}[R_{t+1} \mid S_t = s, A_t \sim \pi]$
- $= \sum_a \pi(a \mid s) R_{ss'}^a$

$$\frac{R_3^\pi}{1}$$

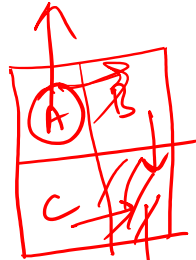


# Value Function ( $V_{\pi}(s)$ )

Bellman Expectation Eq.

The expected return for following policy  $\pi$  starting from state  $s$

$$\begin{aligned} \pi(\text{Left}|s) &= \frac{1}{4} \\ \pi(\text{R}|s) &= \frac{1}{4} \\ \pi(\text{D}|s) &= \frac{1}{4} \\ \pi(\text{U}|s) &= \frac{1}{4} \end{aligned}$$



↑↑↑↑↑  
A A A A A ... B G

$$V_{\pi}(s) := \mathbb{E}_{\pi} [G_t | S_t = s]$$

$$\begin{aligned} &\rightarrow P(A, \text{Right}, -1, B, \text{Down}, -1, G) \rightarrow -2 \\ &\rightarrow A, \text{UP}, -1, A, \text{Down}, -1, C, \text{Right}, -1, G \rightarrow -3 \end{aligned}$$

$$\pi(\text{Right}) \times \pi(\text{Down}) = \frac{1}{4} \times \frac{1}{4} = \frac{1}{16}$$

$$-2 \times \frac{1}{16} - 3 \times \left(\frac{1}{4}\right)^3 =$$

# Action-Value Function ( $Q_\pi(s, a)$ )

The expected return for taking action  $a$  in current state  $s$  and then following policy  $\pi$  from the next state

$$Q_\pi(\underline{s}, a) := \mathbb{E}_\pi [G_t \mid S_t = s, A_t = a]$$

$V_\pi(s)$   $Q_\pi(s, a)$

# Relating $Q_\pi$ and $V_\pi$

$$\underline{V_\pi}(s) = \sum_a \underline{\pi(a | s)} \underline{Q_\pi(s, a)}$$

$$\pi \rightarrow \begin{matrix} \underline{a_1} \rightarrow \pi(a_1 | s) \\ a_2 \rightarrow \pi(a_2 | s) \end{matrix}$$

$$V_\pi(s) = \pi(a_1 | s) Q_\pi(s, a_1) + \pi(a_2 | s) Q_\pi(s, a_2)$$

$$Q_{\pi}(s, a) \quad V_{\pi}(s) \quad G_t = R_{t+1} + \gamma R_{t+2} + \gamma^2 R_{t+3} + \dots$$

$$G_{t+1} = R_{t+2} + \gamma R_{t+3} + \gamma^2 R_{t+4} + \dots$$

$$Q_{\pi}(s, a) = E_{\pi} \left[ G_t \mid s_t = s, A_t = a \right]$$

$$= E_{\pi} \left[ R_{t+1} + \gamma G_{t+1} \mid s_t = s, A_t = a \right]$$

$$= \underbrace{E_{\pi} [R_{t+1} \mid s_t = s, A_t = a]}_{\downarrow} + \gamma \underbrace{E_{\pi} [G_{t+1} \mid s_t = s, A_t = a]}_{\downarrow \downarrow}$$

$$= R_s^a + \gamma$$

$$E[X+Y] = E[X] + E[Y]$$



# Relating $Q_\pi$ and $V_\pi$

$\bullet Q_\pi(s, a) = \mathbb{E}_\pi[G_t \mid S_t = s, A_t = a]$   
 $= \mathbb{E}_\pi[R_{t+1} + \gamma G_{t+1} \mid S_t = s, A_t = a]$   
 $= \mathbb{E}_\pi[R_{t+1} \mid S_t = s, A_t = a] + \mathbb{E}_\pi[G_{t+1} \mid S_t = s, A_t = a]$   
 $= R_s^a + \gamma \sum_{s'} P_{ss'}^a \mathbb{E}_\pi[G_{t+1} \mid S_{t+1} = s', S_t \neq s, A_t \neq a]$   
 $\underline{Q_\pi(s, a)} = \underline{R_s^a + \gamma \sum_{s'} P_{ss'}^a V_\pi(s')}$

Handwritten notes:  $V_\pi \leftrightarrow Q_\pi$ ,  $V_\pi = \begin{pmatrix} V_\pi(A) \\ V_\pi(B) \\ V_\pi(C) \end{pmatrix}$ , Markov property.

- $\bullet$  Substitute this in  $\underline{V_\pi(s)} = \sum_a \pi(a \mid s) \underline{Q_\pi(s, a)}$  to get  $V_\pi$  in terms of  $V_\pi$
- Handwritten notes: 1, 2.

# Bellman Expectation (BE) equation

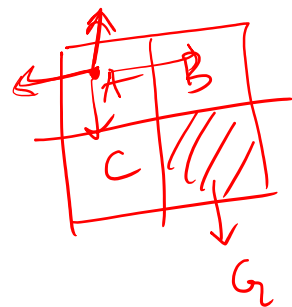
- $V_\pi$  in terms of  $V_\pi$  : (Useful to compute  $V_\pi$  from  $P_{ss'}^a$  and  $R_s^a$ )

$$V_\pi(s) = R_s^\pi + \sum_{s'} P_{ss'}^\pi V_\pi(s')$$

Immediate reward

Remaining Return

- ▶  $R_s^\pi := \sum_a R_s^a \pi(a | s)$
- ▶  $P_{ss'}^\pi := \sum_a P_{ss'}^a \pi(a | s)$

Example $\pi \rightarrow$  uniform Random PolicyBE

$$V_{\pi}(s)$$

$$V_{\pi}(s) = R_s^{\pi} + \gamma \sum_{s'} P_{ss'}^{\pi} \underline{V_{\pi}(s')}$$

① Immediate Reward      ② Remaining Return

3 unknowns  
3 linear equations

$$\underline{V_{\pi}(A)}$$

$$① R_A^{\pi} = -1$$

$$② P_{AB}^{\pi} = \frac{1}{4}$$

$$P_{AC}^{\pi} = \frac{1}{4}, \quad P_{AA}^{\pi} = \frac{1}{2}$$

$$\boxed{Ax = b}$$

$$\boxed{x = A^{-1}b}$$

$$① \begin{cases} \checkmark V_{\pi}(A) = -1 + \gamma \left( \overset{\downarrow}{P_{AB}^{\pi}} V_{\pi}(B) + \overset{\downarrow}{P_{AC}^{\pi}} V_{\pi}(C) + \overset{\downarrow}{P_{AA}^{\pi}} V_{\pi}(A) \right) \\ \checkmark V_{\pi}(B) = -1 + \dots \\ \checkmark V_{\pi}(C) = -1 + \dots \end{cases}$$

$$\underline{10^6 \times 10^6}$$