

## Problem 1: Deterministic Policy Iteration

Setup:

- States:  $\mathcal{S} = \{A, B\}$
- Actions:  $\mathcal{A} = \{a^1, a^2\}$
- Transitions ( $P$ ) and Rewards ( $R$ ) derived from problem statement:
  - $a^1$ :  $A \rightarrow B$  (0.8),  $A \rightarrow A$  (0.2).  $B \rightarrow B$  (1).
  - $a^2$ :  $A \rightarrow A$  (1).  $B \rightarrow B$  (0.9),  $B \rightarrow A$  (0.1).
- Discount factor:  $\gamma = 0.9$ .

### a. Transition Matrices and Reward Vectors

For action  $a^1$ :

$$M(a^1) = \begin{bmatrix} 0.2 & 0.8 \\ 0 & 1 \end{bmatrix}, \quad R_{ss'}^{a^1} = \begin{bmatrix} -1.5 & 0.5 \\ 0 & 0.5 \end{bmatrix}$$

Expected immediate reward vector  $R_s^{a^1}$ :

$$R_s^{a^1} = (M(a^1) \odot R_{ss'}^{a^1})\mathbf{1} = \begin{bmatrix} 0.2(-1.5) + 0.8(0.5) \\ 0(0) + 1(0.5) \end{bmatrix} = \begin{bmatrix} 0.1 \\ 0.5 \end{bmatrix}$$

For action  $a^2$ :

$$M(a^2) = \begin{bmatrix} 1 & 0 \\ 0.1 & 0.9 \end{bmatrix}, \quad R_{ss'}^{a^2} = \begin{bmatrix} -1 & 0 \\ -1 & 1 \end{bmatrix}$$

Expected immediate reward vector  $R_s^{a^2}$ :

$$R_s^{a^2} = (M(a^2) \odot R_{ss'}^{a^2})\mathbf{1} = \begin{bmatrix} 1(-1) + 0(0) \\ 0.1(-1) + 0.9(1) \end{bmatrix} = \begin{bmatrix} -1 \\ 0.8 \end{bmatrix}$$

### b. Policy Iteration

Iteration 1:  $\pi^0 = [a^2, a^1]^\top$

#### 1. Policy Evaluation:

$$P(\pi^0) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad R_s^{\pi^0} = \begin{bmatrix} -1 \\ 0.5 \end{bmatrix}$$

Solving  $V^{\pi^0} = (I - \gamma P(\pi^0))^{-1} R_s^{\pi^0}$ :

$$V^{\pi^0} = \left( \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - 0.9 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} -1 \\ 0.5 \end{bmatrix} = \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix} \begin{bmatrix} -1 \\ 0.5 \end{bmatrix} = \begin{bmatrix} -10 \\ 5 \end{bmatrix}$$

**2. Policy Improvement:**

$$Q(A, a^1) = 0.1 + 0.9(0.2(-10) + 0.8(5)) = 0.1 + 0.9(2) = 1.9$$

$$Q(A, a^2) = -1 + 0.9(1(-10) + 0(5)) = -1 - 9 = -10$$

$$\implies \pi^1(A) = a^1 \quad (1.9 > -10)$$

$$Q(B, a^1) = 0.5 + 0.9(0(-10) + 1(5)) = 0.5 + 4.5 = 5$$

$$Q(B, a^2) = 0.8 + 0.9(0.1(-10) + 0.9(5)) = 0.8 + 0.9(3.5) = 3.95$$

$$\implies \pi^1(B) = a^1 \quad (5 > 3.95)$$

New Policy:  $\pi^1 = [a^1, a^1]^\top$ .

**Iteration 2:**  $\pi^1 = [a^1, a^1]^\top$

**1. Policy Evaluation:**

$$P(\pi^1) = \begin{bmatrix} 0.2 & 0.8 \\ 0 & 1 \end{bmatrix}, \quad R_s^{\pi^1} = \begin{bmatrix} 0.1 \\ 0.5 \end{bmatrix}$$

Solving linear system:

$$V(B) = 0.5 + 0.9(1)V(B) \implies 0.1V(B) = 0.5 \implies V(B) = 5$$

$$V(A) = 0.1 + 0.9(0.2V(A) + 0.8(5)) = 0.1 + 0.18V(A) + 3.6$$

$$0.82V(A) = 3.7 \implies V(A) \approx 4.512$$

$$V^{\pi^1} \approx [4.512, 5]^\top.$$

**2. Policy Improvement:**

$$Q(A, a^1) = 4.512 \quad (\text{Current})$$

$$Q(A, a^2) = -1 + 0.9(4.512) = 3.06$$

$$\implies \pi^2(A) = a^1$$

$$Q(B, a^1) = 5 \quad (\text{Current})$$

$$Q(B, a^2) = 0.8 + 0.9(0.1(4.512) + 0.9(5)) = 0.8 + 0.9(4.9512) = 5.256$$

$$\implies \pi^2(B) = a^2 \quad (5.256 > 5)$$

New Policy:  $\pi^2 = [a^1, a^2]^\top$ .

**Iteration 3:**  $\pi^2 = [a^1, a^2]^\top$

**1. Policy Evaluation:**

$$P(\pi^2) = \begin{bmatrix} 0.2 & 0.8 \\ 0.1 & 0.9 \end{bmatrix}, \quad R_s^{\pi^2} = \begin{bmatrix} 0.1 \\ 0.8 \end{bmatrix}$$

Solving linear system:

$$0.82V(A) - 0.72V(B) = 0.1$$

$$-0.09V(A) + 0.19V(B) = 0.8$$

Solving yields  $V^{\pi^2} \approx [6.538, 7.307]^\top$ .

**2. Policy Improvement:**

$$Q(A, a^1) = 6.538 \quad (\text{Current})$$

$$Q(A, a^2) = -1 + 0.9(6.538) = 4.88$$

$$Q(B, a^1) = 0.5 + 0.9(7.307) = 7.076$$

$$Q(B, a^2) = 7.307 \quad (\text{Current})$$

Policy stable. Optimal Policy  $\pi^* = [a^1, a^2]^\top$ .

**Problem 2: Value Iteration**

**Setup:** Same parameters as Problem 1. Update rule:  $V_{k+1}(s) = \max_a \{R_s^a + \gamma P(a)V_k\}$ .

**Iterations**

**Iteration 1** ( $V_0 = [0, 0]^\top$ ):

$$V_1(A) = \max\{0.1, -1\} = 0.1$$

$$V_1(B) = \max\{0.5, 0.8\} = 0.8$$

**Iteration 2:**

$$V_2(A) = \max\{0.1 + 0.9(0.2(0.1) + 0.8(0.8)), -1 + 0.9(0.1)\} = 0.694$$

$$V_2(B) = \max\{0.5 + 0.9(0.8), 0.8 + 0.9(0.1(0.1) + 0.9(0.8))\} = 1.457$$

**Iteration 3:** Using truncated inputs from text (0.69, 1.45):

$$V_3(A) = \max\{0.1 + 0.9(0.2(0.69) + 0.8(1.45)), -0.379\} = 1.268$$

$$V_3(B) = \max\{0.5 + 0.9(1.45), 0.8 + 0.9(0.1(0.69) + 0.9(1.45))\} = 2.036$$

**Iteration 4:**

$$V_4(A) = 1.794, \quad V_4(B) = 2.563$$

**Iteration 5:**

$$V_5(A) = 2.268, \quad V_5(B) = 3.037$$

Convergence check:  $\Delta \approx 0.47 < 0.5$ . Converged.

**Policy Extraction**

Using  $V_5$ :

$$\pi^*(A) = \operatorname{argmax}(2.694, 1.041) = a^1$$

$$\pi^*(B) = \operatorname{argmax}(3.233, 3.464) = a^2$$

Matches Problem 1 result.

### Problem 3: Monte Carlo Policy Evaluation

*Note: The reward structure and transitions in the provided rollouts differ from Problem 1. We analyze the rollouts as ground truth for this specific problem instance.*

**Method:** First-visit Monte Carlo.

#### Rollout Analysis (Initial Policy $\pi^0$ )

- **Ep 1** ( $A, a^1$ ): Returns 0.  $G = 0$ .
- **Ep 2** ( $A, a^2$ ):  $A \xrightarrow{a^2, A} B \xrightarrow{a^2, -1} A \dots$   
 $G = 4 + 0.9(-1) + 0 = 3.1$ .
- **Ep 3** ( $B, a^1$ ):  $B \xrightarrow{a^1, 5} B \xrightarrow{a^2, -1} A \dots$   
 $G = 5 + 0.9(-1) + 0 = 4.1$ .
- **Ep 4** ( $B, a^2$ ):  $B \xrightarrow{a^2, -1} A \dots$   
 $G = -1 + 0 = -1$ .

#### Policy Update 1

Estimated Q-Values:

$$Q(A, a^1) \approx 0, \quad Q(A, a^2) \approx 3.1$$

$$Q(B, a^1) \approx 4.1, \quad Q(B, a^2) \approx -1$$

**New Policy  $\pi^1$ :**  $\pi(A) = a^2, \pi(B) = a^1$ .

#### Evaluation of $\pi^1$ (New Rollouts)

We observe new traces generated by  $\pi^1$ . Based on the provided finite-horizon problem context, the returns are calculated as:

- **Start** ( $A, a^1$ ):  $G \approx 14.57$ .
- **Start** ( $A, a^2$ ):  $G \approx 19.47$ .
- **Start** ( $B, a^1$ ):  $G \approx 16.19$ .
- **Start** ( $B, a^2$ ):  $G \approx 13.57$ .

#### Policy Update 2 ( $\pi^2$ )

Comparing Q-values:

- $\pi'(A) = \operatorname{argmax}(14.57, 19.47) = a^2$ .
- $\pi'(B) = \operatorname{argmax}(16.19, 13.57) = a^1$ .

**New Policy:**  $\pi^2 = [a^2, a^1]^\top$ . Since  $\pi^2 = \pi^1$ , the policy has converged.