

MAS Practice Exam Dec 2023

① Game Theory

(Q1)

		A	B
		q	1-q
		P	
A		2, 2	-1, -1
B		-1, -1	1, 1
1-p			

PNE:
 $(A, A), (B, B)$

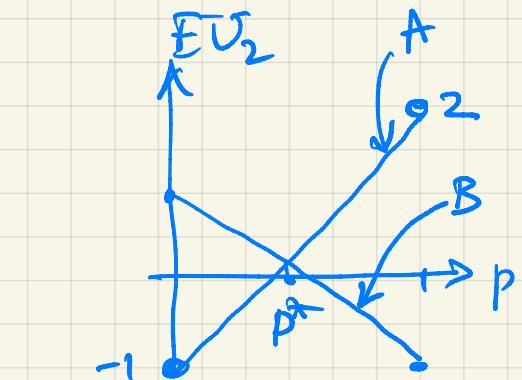
MNE:

$$EU_2(p, A) = EU_2(p, B)$$

$$2p - (1-p) = -p + (1-p)$$

$$3p - 1 = -2p + 1$$

$$5p = 2 \Rightarrow p^* = \frac{2}{5}$$



$$q^* = \frac{2}{5}$$

(symmetric game)

Corresponding utility:

$$EU_2(p^*, A) = EU_2(p^*, B) = 2 \cdot \frac{2}{5} - \frac{3}{5} = \frac{1}{5}$$

$$EU_1(A, q^*) = EU_1(B, q^*) = \frac{1}{5}$$

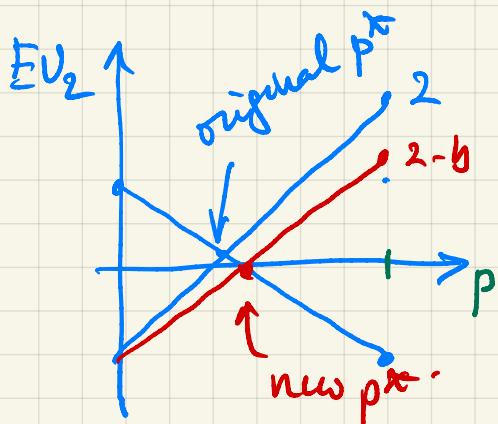
(Q2): MNE is Pareto dominated by PNE

$$\text{MNE} < (B, B) < (A, A)$$

Q3, $u_2(A, A) = 2-b$ $(0 < b < 1)$

		A q	B b
		2, 2-b	-1, -1
		-1, -1	1, 1
A	p	2, 2-b	-1, -1
B	1-p	-1, -1	1, 1

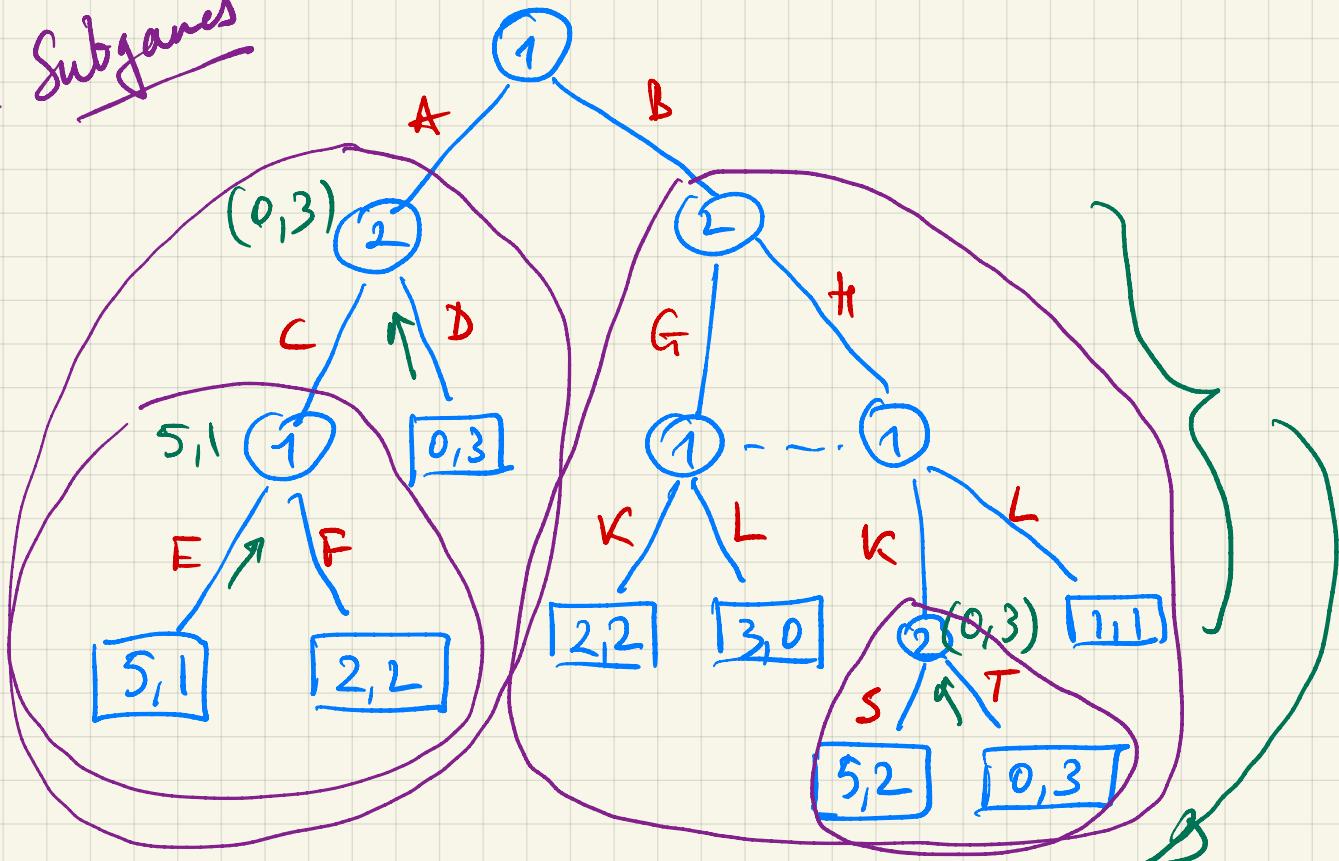
PNE unchanged.



As $b \uparrow \Rightarrow p^* \uparrow$

NB: q^* is unaffected!

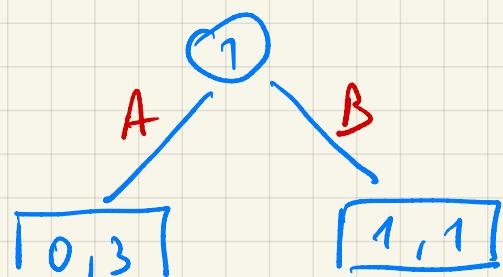
4 Subgames



G	H
K	2, 2 0, 3
L	3, 0 1, 1

$\rightarrow \{L, H\}$ is NE

We can reduce the game tree to!

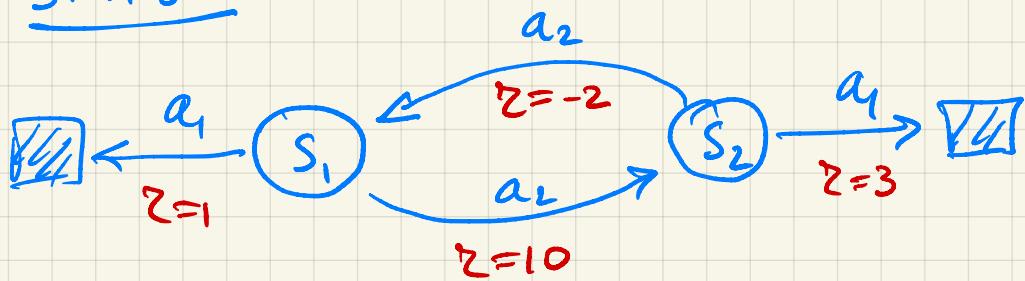


$\Rightarrow 1$ will choose B
 2 — H
 1 — L

Payoff = (1, 1).

$$\begin{aligned}
 \text{Q3: } 1 &\rightarrow \{A, B\} \times \{C, D\} \times \{G, H\} = \{AEK, AE, \\
 &= \{AEK, AEL, AFL, AFK, BEK, BEL, BFK, BFL\} \\
 2 &\rightarrow \{C, D\} \times \{G, H\} \times \{S, T\} = \dots
 \end{aligned}$$

3. MDP



Q1: Infinite horizon (could keep on cycling)

Q2: $\gamma = 0.9 \rightarrow$ optimal policy?

4 possible (deterministic) policies.

$\pi_1: S_1 \rightarrow a_1, S_2 \rightarrow a_1$

$$v(S_1) = 1, v(S_2) = 3$$

$\pi_2: S_1 \rightarrow a_2, S_2 \rightarrow a_1$

$$v(S_1) = 10 + 3\gamma = 10 + 2.7 = 12.7$$

$$v(S_2) = 3$$

$\pi_3: S_1 \rightarrow a_1, S_2 \rightarrow a_2$

$$v(S_1) = 1$$

$$v(S_2) = -2 + \gamma = -2 + 0.9 = -1.1$$

$\pi_4: S_1 \rightarrow a_2, S_2 \rightarrow a_2$

$$v(S_1) = 10 + (-2)\gamma + 10\gamma^2 + (-2)\gamma^3 + \dots$$

$$= 10 \underbrace{(1 + \gamma^2 + \gamma^4 + \dots)}_{1 - \gamma^2} - 2\gamma(1 + \gamma^2 + \dots) \approx \underline{\underline{43.2}}$$

(10 - 1.8) . 5.3

$$\frac{1}{1 - \gamma^2} = \frac{1}{1 - 0.9^2} = \frac{1}{1 - 0.81} = \frac{1}{0.19} \approx 5.3$$

$$\begin{aligned}
 v(S_2) &= -2 + \gamma \cdot 10 + (-2)\gamma^2 + 10\gamma^3 + \dots \\
 &= -2(1 + \gamma^2 + \dots) + 10\gamma(1 + \gamma^2 + \dots) \\
 &= \underbrace{(10\gamma - 2)}_{7} \underbrace{\left(\frac{1 + \gamma^2 + \gamma^4 + \dots}{1 - \gamma^2}\right)}_{\simeq 5.3} \simeq \underline{\underline{37.1}}
 \end{aligned}$$

Conclusion, for $\gamma = 0.9$ optimal policy: $\bar{T}^x = \bar{T}_4$.

4. Reinforcement Learning

① Recall, $v(s) = \sum_a q(s, a) \pi(a | s)$

$$v(2) = 5 = \underbrace{q(2, R)}_x \underbrace{\pi(R | 2)}_{1/4} + \underbrace{q(2, L)}_4 \underbrace{\pi(L | 2)}_{3/4}$$

$$\frac{1}{4}x + \frac{3}{4} \cdot 4 = 5 \Rightarrow x = 20 - 12 = 8$$

$$q(2, R) = 8$$

Similarly, $q(3, L) = 9$

② Q-Learning : $2 \xrightarrow{R} 3$

$$q(2, R) \leftarrow q(2, R) + \alpha \left[r(2, R) + \gamma \max_{a'} q(3, a') - q(2, R) \right]$$

$$\leftarrow 8 + 0.9 \left[\underbrace{-1 + \frac{2}{3} \cdot 9 - 8}_{-3} \right] = 8 - 2 \cdot 7 = \underline{\underline{5.3}}$$