

#### **Motivations**

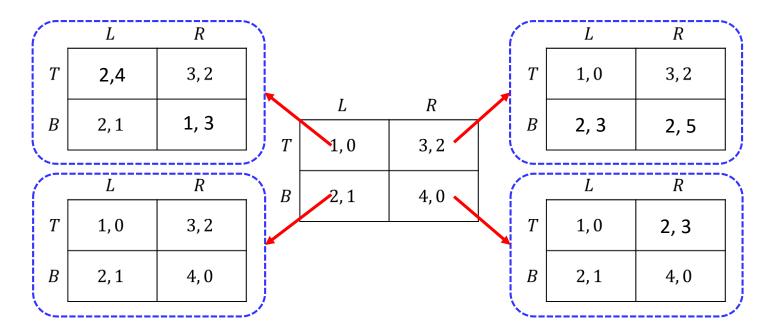
## What if we didn't always repeat back to the same stage game?

- A stochastic game is a generalization of repeated games
  - agents repeatedly play games from a set of normal-form games
  - the game played at any iteration depends on the previous game played and on the actions taken by all agents in that game

### What if there are multiple decision makers in Markov Decision Process?

- A stochastic game is a generalized Markov decision process
  - there are multiple players one reward function for each agent
  - the state transition function and reward functions depend on the action choices of both players

#### **Motivations**



- Stochastic game is a moral general setting where learning is taking place
  - The game transits to another game depending on the joint actions by agents
  - Same players and same actions sets are used through games
- Most of the techniques discussed in the context of repeated games are applicable more generally to stochastic games
  - ✓ specific results obtained for repeated games do not always generalize.

#### **Formal Definition**

### **Definition (Stochastic game)**

A stochastic game is a tuple (N, S, A, R, T), where

- *N* is a finite set of *n* players
- S is a finite set of states (stage games),
- $A = A_1 \times \cdots \times A_n$ , where  $A_i$  is a finite set of actions available to player i,
- $T: S \times A \times S \mapsto [0,1]$  is the transition probability function; T(s, a, s') is the probability of transitioning from state s to state s' after joint action a,
- $R = r_1 \dots, r_n$ , where  $r_i : S \times A \mapsto \mathbb{R}$  is a real-valued payoff function for player i

#### **Transition model**

- All agents (1, ..., n) share the joint state s
- The transition equation is similar to the Markov Decision Process decision transition:

$$MDP: \sum_{s'} T(s, \boldsymbol{a}, s') = \sum_{s'} p(s'|\boldsymbol{a}, s) = 1, \forall s \in S, \forall a \in A$$

SG: 
$$\sum_{s'} T(s, a_1, ..., a_i, ..., a_n, s') = \sum_{s'} p(s'|a_1, ..., a_i, ..., a_n, s) = 1$$

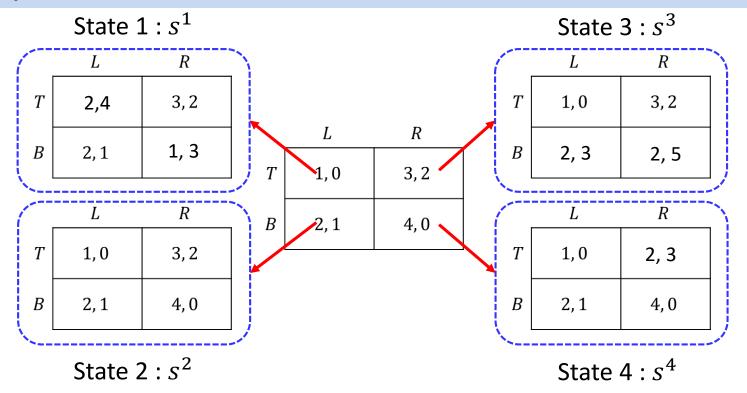
$$\forall s \in S, \forall a_i \in A_i, i = (1, ..., n)$$

#### **Reward function**

• Reward function  $r_i$  for agent i depends on the current joint state s, the joint action  $a = (a_1, ..., a_n)$ , and the next joint future state s'

SG: 
$$r_i(s, a_1, ..., a_i, ..., a_n, s')$$

### **Policy**



• Policy  $\pi_1$  will give the action that will be taken by player 1 at a given state (stage game):

$$a_1 = \pi_1(s), \ a_1 \in \{T, B\}$$

#### **Value function**

- As we did in MDP, we can define value function
- Let  $\pi_i$  be the policy of player  $i \in N$ . For a given initial state s, the value of state s for player i is defined as

$$V_{i}(s, \pi_{1}, ..., \pi_{i}, ..., \pi_{n}) = \sum_{t=0}^{\infty} \gamma^{t} E[r_{i,t} | \pi_{1}, ..., \pi_{i}, ..., \pi_{n}, s_{0} = s]$$

- > The accumulated rewards depends on the policies of other agents
- $\blacktriangleright$  The immediate reward is expressed as expected value, because some policy  $\pi_i$  can be stochastic
- In a *discounted stochastic game*, the objective of each player is to maximize the discounted sum of rewards, with discount factor  $\gamma \in [0,1)$ .

### **Equilibrium strategy**

### **Definition (Nash equilibrium policy in Stochastic game)**

In a stochastic game  $\Gamma = (N, S, A, R, T)$ , a Nash equilibrium policy is a tuple of n policies  $\pi^* = (\pi_1^*, \dots, \pi_n^*)$  such that for all  $s \in S$  and  $i = 1, \dots n$ ,

$$V_i(s, \pi_1^*, ..., \pi_i^*, ..., \pi_n^*) \ge V_i(s, \pi_1^*, ..., \pi_i^*, ..., \pi_n^*)$$
 for all  $\pi_i \in \Pi_i$ 

- A Nash equilibrium is a joint policy where each agent's policy is a best response to the others
- For a stochastic game, each agent's policy is defined over the entire time horizon of the game
- A Nash equilibrium state value  $V_i(s, \pi_1^*, ..., \pi_n^*)$  is defined as the sum of discounted rewards when all agents following the Nash equilibrium policies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ 
  - Notations:  $V_i^*(s) = V_i^{\pi^*}(s) = V_i(s, \pi_1^*, ..., \pi_n^*)$

### **Equilibrium policy**

### Theorem (Fink 1964)

Every n —player discounted stochastic game processes at least one Nash equilibrium policy in stationary policies

- Action selection rule for non-stationary policy is different depending on time
  - $\pi_t(s) \neq \pi_{t+1}(s)$
- There are generally a great multiplicity of non-stationary equilibria, whose fact is partially demonstrated by Folk Theorems

# Single agent

#### **Q-values**

 $Q^{\pi}(s,a)$ : The expected utility of taking action a from state s, and then following policy  $\pi$ 

$$Q^{\pi}(s,a) = \mathbb{E}_{\pi} \left( \sum_{k=0}^{\infty} \gamma^k r_{t+k} \middle| S_t = s, A_t = a \right)$$

#### **Optimal Q-values**

$$\begin{split} Q^*(s, a) &= \max_{\pi} Q^{\pi}(s, a) \\ &= \max_{\pi} \mathbb{E}[r(s, a, s') + \gamma V^{\pi}(s') | s_t = s, a_t = a] \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\pi} V^{\pi}(s') | s_t = s, a_t = a\right] \\ &= \mathbb{E}[r(s, a, s') + \gamma V^*(s') | s_t = s, a_t = a] \\ &= \mathbb{E}\left[r(s, a, s') + \gamma V^*(s') | s_t = s, a_t = a\right] \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} V^{\pi}(s') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t = a\right] \\ &: V^*(s') \equiv \max_{\alpha'} Q^*(s', a') \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \max_{\alpha'} Q^*(s', a') | s_t = s, a_t$$

Optimization over policy becomes greedy optimization over action!

 Optimal Q-value for a single-agent is the sum of the current reward and future discounted rewards when playing the optimal strategy from the next period onward

#### Optimal Q-values → Nash Q-Values

## Multi agents

#### **Q**-values for agent i

 $Q_i^{\pi}(s, a_1, ..., a_n)$ : The expected utility of taking joint action  $(a_1, ..., a_n)$  from state s, and then following policy  $\pi$   $Q_i^{\pi}(s, a_1, ..., a_n) = \mathbb{E}_{\pi} \left( \sum_{k=0}^{\infty} \gamma^k r_{i,t+k} \mid S_t = s, A = (a_1, ..., a_n) \right)$ 

#### **Optimal Q-values for agent** *i*

$$\begin{split} Q_i^*(s, a_1, \dots, a_n) &= \max_{\pi_1, \dots, \pi_n} Q_i^\pi(s, a_1, \dots, a_n) \\ &= \max_{\pi_1, \dots, \pi_n} \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \max_{\pi_1, \dots, \pi_n} V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1^*, \dots, \pi_n^*) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \max_{a_1, \dots, a_n} Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \end{split}$$

- Optimal Q-value for agent i occurs when all agents are jointly coordinating to maximize agent i's accumulated reward
  - Rarely occurs! : Optimal Q-values for all agents are not achieved simultaneously

#### Optimal Q-values → Nash Q-Values

## Multi agents

#### Q-values for agent i

 $Q_i^{\pi}(s, a_1, ..., a_n)$ : The expected utility of taking joint action  $(a_1, ..., a_n)$  from state s, and then following policy  $\pi$   $Q_i^{\pi}(s, a_1, ..., a_n) = \mathbb{E}_{\pi}(\sum_{k=0}^{\infty} \gamma^k r_{i,t+k} \mid S_t = s, A = (a_1, ..., a_n))$ 

### Optimal Q-values for agent i

$$\begin{split} Q_i^*(s, a_1, \dots, a_n) &= \max_{\pi_1, \dots, \pi_n} Q_i^{\pi}(s, a_1, \dots, a_n) \\ &= \max_{\pi_1, \dots, \pi_n} \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \max_{\pi_1, \dots, \pi_n} V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1^*, \dots, \pi_n^*) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \max_{a_1, \dots, a_n} Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \end{split}$$

- Optimal Q-value for agent i occurs when all agents are jointly coordinating to maximize agent i's accumulated reward
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#### Optimal Q-values → Nash Q-Values

## **Multi agents**

#### **Q**-values for agent i

 $Q_i^{\pi}(s, a_1, ..., a_n)$ : The expected utility of taking joint action  $(a_1, ..., a_n)$  from state s, and then following policy  $\pi$   $Q_i^{\pi}(s, a_1, ..., a_n) = \mathbb{E}_{\pi} \left( \sum_{k=0}^{\infty} \gamma^k r_{i,t+k} \mid S_t = s, A = (a_1, ..., a_n) \right)$ 

#### Nash Q-values for agent i

$$\begin{split} Q_i^*(s, a_1, \dots, a_n) &= \underset{\pi_1, \dots, \pi_n}{\operatorname{Nash}} \, Q_i^\pi(s, a_1, \dots, a_n) \\ &= \underset{\pi_1, \dots, \pi_n}{\operatorname{Nash}} \, \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{\pi_1, \dots, \pi_n}{\operatorname{Nash}} \, V_i(s', \pi_1, \dots, \pi_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1^*, \dots, \pi_n^*) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \,$$

• A Nash Q value  $Q_i^*(s, a_1, ..., a_n)$  is the expected sum of discounted rewards when all agents take the joint action  $a = (a_1, ..., a_n)$  at given state s and follow a Nash equilibrium strategy  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ 

#### **Nash Bellman equation**

#### For single agent:

$$V^{*}(s') = \max_{a} Q^{*}(s', a)$$

$$Q^{*}(s, a) = \mathbb{E}[r(s, a, s') + \gamma V^{*}(s') | s_{t} = s, a_{t} = a]$$

$$= \mathbb{E}\left[r(s, a, s') + \gamma \max_{a'} Q^{*}(s', a') | s_{t} = s, a_{t} = a\right]$$

#### For multiple agents:

$$\begin{split} V_i(s', \pi_1^*, ..., \pi_n^*) &= \underset{a_1, ..., a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, ..., a_n) \\ Q_i^*(s, a_1, ..., a_n) &= \mathbb{E}[r(s, a, s') + \gamma V_i(s', \pi_1^*, ..., \pi_n^*) | s_t = s, a_t = a] \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \underset{a_1, ..., a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, ..., a_n) \, | s_t = s, a_t = (a_1, ..., a_n)\right] \end{split}$$

### How to compute A Nash (equilibrium) state value $V_i(s, \pi_1^*, \dots, \pi_n^*)$

#### For multiple agents:

$$\begin{split} V_i(s', \pi_1^*, \dots, \pi_n^*) &= \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) \\ Q_i^*(s, a_1, \dots, a_n) &= \mathbb{E}[r(s, a, s') + \gamma V_i(s', \pi_1^*, \dots, \pi_n^*) | s_t = s, a_t = a] \\ &= \mathbb{E}\left[r(s, a, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} \, Q_i^*(s', a_1, \dots, a_n) \, | s_t = s, a_t = (a_1, \dots, a_n)\right] \end{split}$$

- Nash **equilibrium** Q value  $\underset{a_1,...,a_n}{\operatorname{Nash}} Q_i^*(s',a_1,...,a_n)$  can be computed by computing player ith
  - Nash equilibrium value for the stage game  $[Q_i^*(s', a_1, ..., a_n), ..., Q_n^*(s', a_1, ..., a_n)]$ 
    - $\triangleright$  for example when i = 1,2

$$a_{1}^{1} \qquad a_{2}^{2}$$

$$a_{1}^{1} \qquad Q_{1}^{*}(s', a_{1}^{1}, a_{2}^{1}), Q_{2}(s', a_{1}^{1}, a_{2}^{1}) \qquad Q_{1}^{*}(s', a_{1}^{1}, a_{2}^{2}), Q_{2}(s', a_{1}^{1}, a_{2}^{2})$$

$$a_{1}^{2} \qquad Q_{1}^{*}(s', a_{1}^{2}, a_{2}^{1}), Q_{2}(s', a_{1}^{2}, a_{2}^{1}) \qquad Q_{1}^{*}(s', a_{1}^{2}, a_{2}^{2}), Q_{2}(s', a_{1}^{2}, a_{2}^{2})$$
Nash equilibrium

### **Simplifying Notation**

#### For multiple agents:

$$r_i(s, a_1, ..., a_n, s') \to r_i(s, \vec{a}, s')$$
 $V_i(s, \pi_1^*, ..., \pi_n^*) \to V_i^*(s)$ 
 $Q_i^*(s, a_1, ..., a_n) \to Q_i^*(s', \vec{a})$ 

$$\begin{split} Q_i^*(s, a_1, \dots, a_n) &= \mathbb{E}[r_i(s, a_1, \dots, a_n, s') + \gamma V_i(s', \pi_1^*, \dots, \pi_n^*) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= \mathbb{E}\left[r_i(s, a_1, \dots, a_n, s') + \gamma \underset{a_1, \dots, a_n}{\operatorname{Nash}} Q_i^*(s', a_1, \dots, a_n) | s_t = s, a_t = (a_1, \dots, a_n)\right] \end{split}$$

$$Q_{i}^{*}(s', \vec{a}) = \mathbb{E}[r_{i}(s, \vec{a}, s') + \gamma V_{i}^{*}(s') | s_{t} = s, a_{t} = \vec{a}]$$

$$= \mathbb{E}[r_{i}(s, \vec{a}, s') + \gamma \operatorname{Nash} Q_{i}^{*}(s') | s_{t} = s, a_{t} = \vec{a}]$$

$$\underset{a_1,...,a_n}{\operatorname{Nash}} Q_i^*(s', a_1, ..., a_n) = Q_i^*(s', \vec{a}_{NE}) = \operatorname{Nash} Q_i^*(s')$$

### **Computing Nash Q-values analytically**

• If we know Nash equilibrium policy  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ , we can compute the Nash equilibrium state values  $V_i(s, \pi_1^*, ..., \pi_n^*)$  (i.e., policy evaluation)

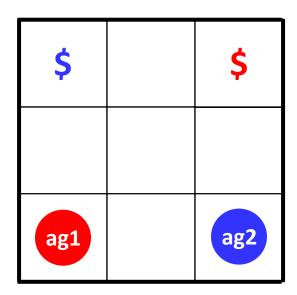
$$V_{i}(s, \pi_{1}^{*}, ..., \pi_{n}^{*}) = \sum_{t=0}^{\infty} \gamma^{t} E[r_{i,t} | \pi_{1}^{*}, ..., \pi_{n}^{*}, s_{0} = s]$$

• If we know Nash equilibrium state value  $V_i(s, \pi_1^*, ..., \pi_n^*)$  and transition models  $p(s'|s, a_1, ..., a_n)$ , we can compute Nash Q-values (i.e., Nash Q-function) using backward induction (analytical approach)

$$Q_i^*(s, a_1, ..., a_n) = \mathbb{E}[r_i(s, a_1, ..., a_n, s') + \gamma V_i(s', \pi_1^*, ..., \pi_n^*) | s_t = s, a_t = (a_1, ..., a_n)]$$

$$= r_i(s, a_1, ..., a_n, s') + \sum_{s'} p(s'|s, a_1, ..., a_n) V_i(s', \pi_1^*, ..., \pi_n^*)$$

#### **Grid Game 1**



- Grid game has deterministic moves
- Two agents start from respective lower corners, trying to reach their goal cells in the top row
- Agent can move only one cell a time, and in four possible directions: Left, Right, Up, Down
- If two agents attempt to move into the same cell (excluding a goal cell), they are bounced back to their previous cells
- The game ends as soon as an agent reaches its goal
  - The objective of an agent in this game is therefore to reach its goal with a minimum No. of steps
- Agents do not know
  - the locations of their goals at the beginning of the learning period
  - their own and the other agents' payoff functions
- Agent choose their action simultaneously and observe
  - the previous actions of both agents and the current joint state
  - the immediate rewards after both agents choose their actions

<b>\$</b>	7	8 \$
3	4	5
0 ag1	1	2 ag2

- The action space of agent i, i = 1,2, is  $A_i = \{Left, Right, Down, Up\}$
- The sate space is  $S = \{(0,1), (0,2), ..., (8,7)\}$ 
  - $s = (l_1, l_2)$  represents the agents' joint location
  - $l_i \in \{0, 2, ..., 8\}$  is the indexed location
- The reward function is, for i = 1, 2,

$$r_i = \begin{cases} 100 \text{ if } L(l_i, a_i) = Goal_i \\ -1 \text{ if } L(l_1, a_1) = L(l_2, a_2) \text{ and } L(l_i, a_i) \neq Goal_i \text{ for } i = 1,2 \\ 0 \text{ otherwise} \end{cases}$$

 $l_i^\prime = L(l_i, a_i)$  is the next location when executing  $a_i$  at  $l_i$ 

6 <b>\$</b>	7	<b>\$</b>
3	4	5
0 ag1	1	2 ag2

• 
$$s = (l_1, l_2) = (0.2)$$

• 
$$a = (a_1, a_2) = (Up, Left)$$

<b>\$</b>	7	8 \$
3 ag1	4	5
0	1 ag2	2

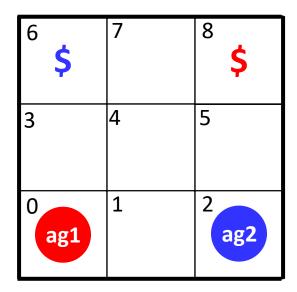
- $s = (l_1, l_2) = (0.2)$
- $a = (a_1, a_2) = (Up, Left)$



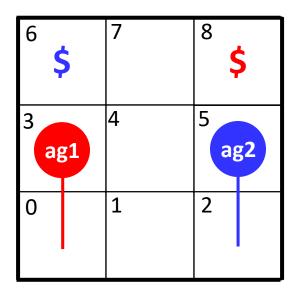
• 
$$s' = (L(l_1, a_1), L(l_2, a_2)) = (3, 1)$$
  
•  $r_1 = 0$ 

• 
$$r_1 = 0$$

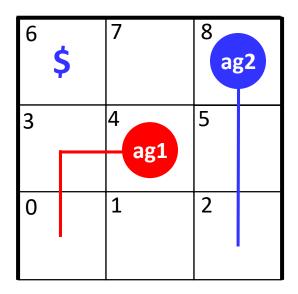
• 
$$r_2 = 0$$



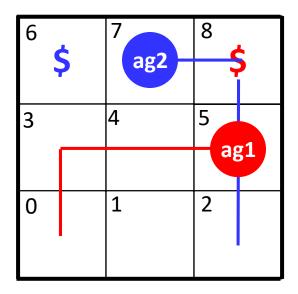
**Nash Equilibrium strategies** 



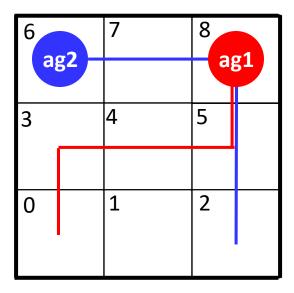
**Nash Equilibrium strategies** 



**Nash Equilibrium policies** 



**Nash Equilibrium policies** 

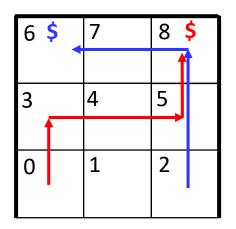


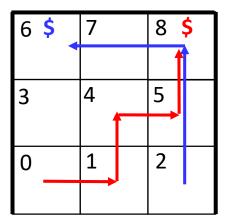
**Nash Equilibrium policies** 

State s	$\pi_1(s)$
(0, any)	U
(3, any)	Right
(4, any)	Right
(5, any)	Up

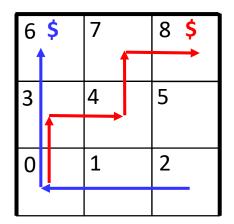
Nash strategy for agent 1

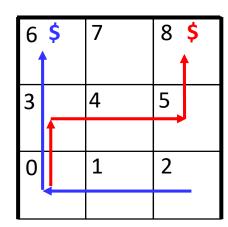
## **All Nash Equilibrium policies**





6 \$	7	8	\$
		1	
3	4	5	
0	1	2	
			-





### Nash Q values for the initial state $s_0 = (0.2)$

• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium polices  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s_0, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

In Grid game 1 and initial state  $s_0 = (0,2)$ , this becomes, given  $\gamma = 0.99$ ,

$$V_1(s_0, \pi_1^*, \pi_2^*) = 0 + 0.99 \times 0 + 0.99^2 \times 0 + 0.99^3 \times 100$$
  
= 97.0

$$s_0 = (0,2)$$

<sup>6</sup> \$	7	8 \$
3	4	5
0 ag1	1	2 ag2

#### Nash Q values for the initial state $s_0 = (0.2)$

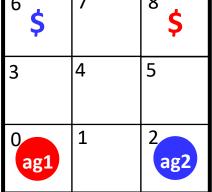
$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Right, Left) = -1 + 0.99 \times V_1(s' = (0,2), \pi_1^*, \pi_2^*)$$
  
= -1 + 0.99 × 97 = 95.1

 $s_0 = (0,2)$ 





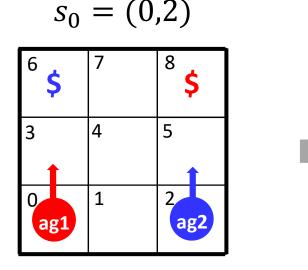
 $s' = s_0 = (0,2)$ 

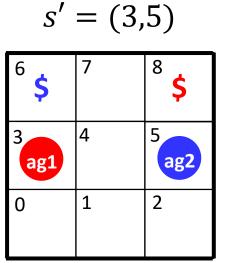
#### Nash Q values for the initial state $s_0 = (0.2)$

$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Up, Up) = 0 + 0.99 \times V_1(s' = (3,5), \pi_1^*, \pi_2^*)$$
$$= 0 + 0.99 \times \{0 + 0.99 \times 0 + 0.99^2 \times 100\} = 97.0$$



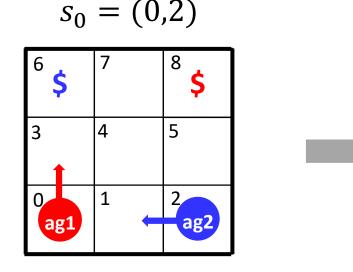


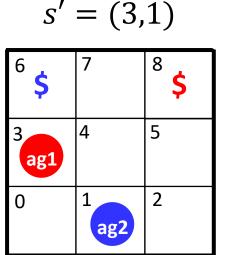
#### Nash Q values for the initial state $s_0 = (0.2)$

$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Up, Left) = 0 + 0.99 \times V_1(s' = (3,1), \pi_1^*, \pi_2^*)$$
  
=  $0 + 0.99 \times \{0 + 0.99 \times 0 + 0.99^2 \times 100\} = 97.0$ 



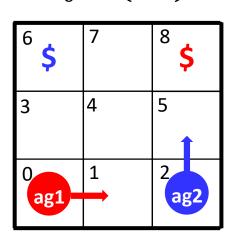


#### Nash Q values for the initial state $s_0 = (0.2)$

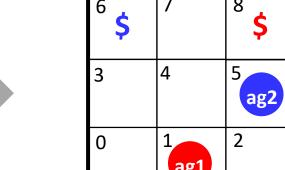
$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Right, Up) = 0 + 0.99 \times V_1(s' = (1,5), \pi_1^*, \pi_2^*)$$
$$= 0 + 0.99 \times \{0 + 0.99 \times 0 + 0.99^2 \times 100\} = 97.0$$



 $s_0 = (0,2)$ 



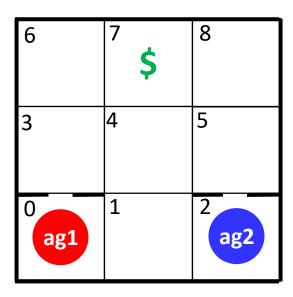
s' = (1,5)

# Nash Q values for the initial state $s_0 = (0.2)$

	$a_2 = Left$	$a_2 = Up$
$a_1 = Right$	$Q_1^*(s_0, R, L), Q_2^*(s_0, R, L)$	$Q_1^*(s_0, R, U), Q_2^*(s_0, R, U)$
$a_2 = Up$	$Q_1^*(s_0, U, L), Q_2^*(s_0, U, L)$	$Q_1^*(s_0, U, U), Q_2^*(s_0, U, U)$
	$a_2 = Left$	$a_2 = Up$

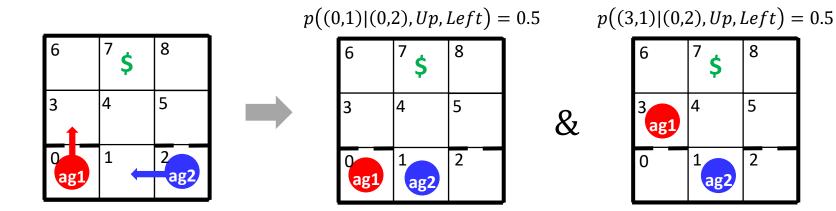
	$a_2 = Lej t$	$a_2 = op$
$a_1 = Right$	95.1, 95.1	97.0,97.0
$a_2 = Up$	97.0, 97.0	97.0,97.0

#### **Grid Game 2**

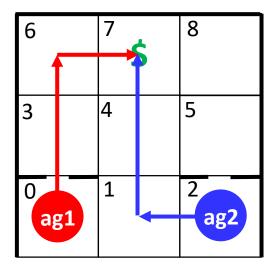


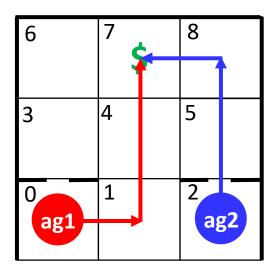
- First to reach goal gets \$100
- If both reaches the money at the same time, both win
- Semi wall (50% go through)
- Cannot occupy the same grid

- Grid game has both stochastic and deterministic moves
- If agent choses Up from position 0 or 2, it moves up with probability 0.5 and remains in its previous position with probability 0.5



#### **Grid Game 2**





- There are two Nash equilibrium paths
- The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

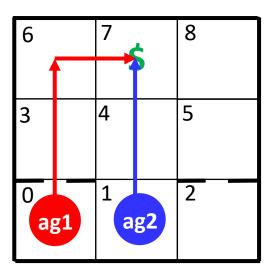
$$V_1(s, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

- $V_1((0,1), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 0 + 0.99^2 \times 0 = 0$
- $V_1((0,x),\pi_1^*,\pi_2^*)=0$  for x=3,...,8
- $V_1((1,2), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 100 = 99$
- $V_1((1,3), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 0 + 0.99^2 \times 0 = 0$
- $V_1((1,x),\pi_1^*,\pi_2^*) = 0 + 0.99 \times 0 + 0.99^2 \times 0 = 0$

• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

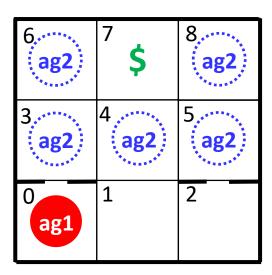
•  $V_1((0,1), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 0 + 0.99^2 \times 0 = 0$ 



• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

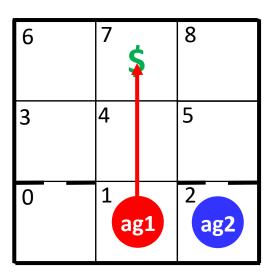
•  $V_1((0,x),\pi_1^*,\pi_2^*)=0$  for x=3,...,8



• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

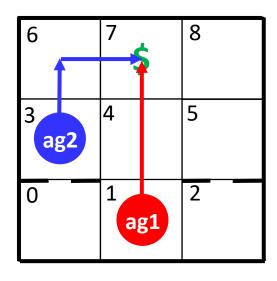
•  $V_1((1,2), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 100 = 99$ 

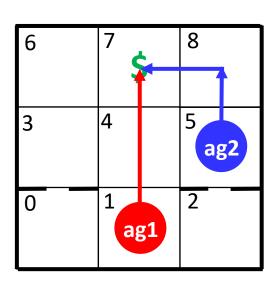


• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s, \pi_1^*, \pi_2^*) = \sum_{t=0}^{\infty} \gamma^t E[r_{i,t} | \pi_1^*, \dots, \pi_n^*, s_0 = s]$$

•  $V_1((1,3), \pi_1^*, \pi_2^*) = 0 + 0.99 \times 100 = 99 = V_1((1,5), \pi_1^*, \pi_2^*)$ 

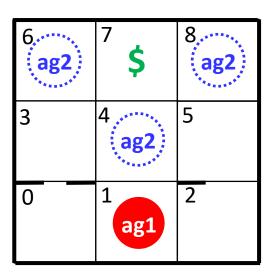




• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s,\pi_1^*,\pi_2^*) = \sum\nolimits_{t=0}^{\infty} \gamma^t E\big[r_{i,t}|\pi_1^*,\dots,\pi_n^*,s_0=s\big]$$

•  $V_1((1,x),\pi_1^*,\pi_2^*)=0$  for x=4,6,8



• The value of the game for agent 1 is defined as its accumulated reward when both agents follow their Nash equilibrium strategies  $\pi^* = (\pi_1^*, ..., \pi_n^*)$ :

$$V_1(s,\pi_1^*,\pi_2^*) = \sum\nolimits_{t=0}^{\infty} \gamma^t E\big[r_{i,t}|\pi_1^*,\dots,\pi_n^*,s_0=s\big]$$

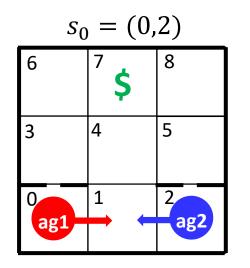
- $V_1((0,2), \pi_1^*, \pi_2^*) = V_1(s_0, \pi_1^*, \pi_2^*)$  can be computed only in expectation
- We solve  $V_1(s_0, \pi_1^*, \pi_2^*)$  from the state game  $(Q_1^*(s_0, a_1, a_2), Q_2^*(s_0, a_1, a_2))$

6	<sup>7</sup> \$	8
3	4	5
0 ag1	1	2 ag2

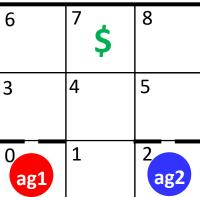
## Nash Q values for the initial state $s_0 = (0.2)$

$$\begin{aligned} Q_1^*(s_0, a_1, a_2) &= \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)] \\ &= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*) \end{aligned}$$

$$Q_1^*(s_0 = (0,2), Right, Left) = -1 + 0.99 \times V_1(s_0, \pi_1^*, \pi_2^*)$$





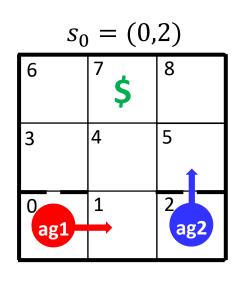


 $s' = s_0 = (0.2)$ 

$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Right, Up) = 0 + 0.99 \times \left\{ \frac{1}{2} V_1((1,2), \pi_1^*, \pi_2^*) + \frac{1}{2} V_1((1,5), \pi_1^*, \pi_2^*) \right\}$$
$$= 0 + 0.99 \times (0.5 \times 99 + 0.5 \times 99) = 98$$





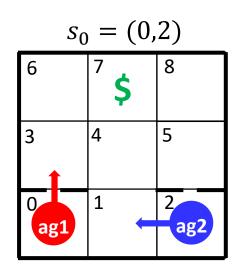
s' = (1,2)			
6	<sup>7</sup> \$	8	
3	4	5	
0	1ag1	2 ag2	

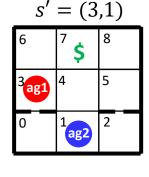
s' = (1,5)			
6	<sup>7</sup> \$	8	
3	4	5 <b>ag2</b>	
0	1 ag1	2	

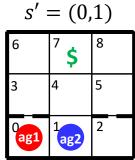
$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Up, Left) = 0 + 0.99 \times \left\{ \frac{1}{2} V_1((3,1), \pi_1^*, \pi_2^*) + \frac{1}{2} V_1((0,1), \pi_1^*, \pi_2^*) \right\}$$
$$= 0 + 0.99 \times (0.5 \times 99 + 0.5 \times 0) = 49$$



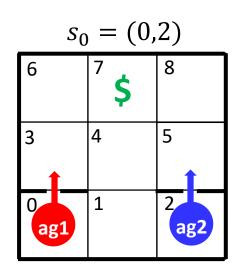




$$Q_1^*(s_0, a_1, a_2) = \mathbb{E}[r_1(s_0, a_1, a_2) + \gamma V_1(s', \pi_1^*, \pi_2^*) | s_t = s, a_t = (a_1, \dots, a_n)]$$

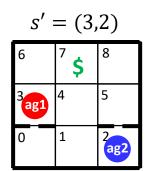
$$= r_1(s_0, a_1, a_2) + \gamma \sum_{s'} p(s' | s_0, a_1, a_2) V_1(s', \pi_1^*, \pi_2^*)$$

$$Q_1^*(s_0 = (0,2), Up, Up) = 0 + 0.99 \times \left\{ \frac{1}{4} V_1^*((0,2)) + \frac{1}{4} V_1^*((0,5)) + \frac{1}{4} V_1^*((3,2)) + \frac{1}{4} V_1^*((3,5)) \right\}$$
$$= 0 + 0.99 \times \left\{ \frac{1}{4} V_1^*(s_0) + \frac{1}{4} \times 0 + \frac{1}{4} \times 99 + \frac{1}{4} \times 99 \right\} = 0.99 \times \frac{1}{4} V_1^*(s_0) + 49$$



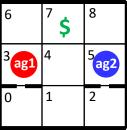


s' = (0,2)			
6	<sup>7</sup> \$	8	
3	4	5	
0ag1	1	2 ag	



s' = (0,5)			
6	<sup>7</sup> <b>\$</b>	8	
3	4	5 <b>ag2</b>	
0ag1	1	2	

s'=(3	3,5)
-------	------



	$a_2 = Left$	$a_2 = Up$
$a_1 = Right$	$Q_1^*(s_0, R, L), Q_2^*(s_0, R, L)$	$Q_1^*(s_0, R, U), Q_2^*(s_0, R, U)$
$a_2 = Up$	$Q_1^*(s_0, U, L), Q_2^*(s_0, U, L)$	$Q_1^*(s_0, U, U), Q_2^*(s_0, U, U)$

$$a_{2} = Left a_{2} = Up$$

$$a_{1} = Right -1 + 0.99V_{1}^{*}(s_{0}), -1 + 0.99V_{2}^{*}(s_{0}) 98,49$$

$$a_{2} = Up 49,98 49 + \frac{0.99}{4}V_{1}^{*}(s_{0}), 49 + \frac{0.99}{4}V_{2}^{*}(s_{0})$$

# Nash Q values for the initial state $s_0 = (0.2)$

$$a_{2} = Left a_{2} = Up$$

$$a_{1} = Right -1 + 0.99V_{1}^{*}(s_{0}), -1 + 0.99V_{2}^{*}(s_{0}) 98,49$$

$$a_{2} = Up 49,98 49 + \frac{0.99}{4}V_{1}^{*}(s_{0}), 49 + \frac{0.99}{4}V_{2}^{*}(s_{0})$$

$$V_1^*(s_0) = \text{Nash}\{Q_1^*(s_0, a_1, a_2), Q_2^*(s_0, a_1, a_2)\}$$

**Case 1**:  $V_1^*(s_0) = 49$ 

	Left	Up	
Right	47,96	98, 49	
Up	49,98	61,73	

$$a_{2} = Left a_{2} = Up$$

$$a_{1} = Right -1 + 0.99V_{1}^{*}(s_{0}), -1 + 0.99V_{2}^{*}(s_{0}) 98,49$$

$$a_{2} = Up 49,98 49 + \frac{0.99}{4}V_{1}^{*}(s_{0}), 49 + \frac{0.99}{4}V_{2}^{*}(s_{0})$$

$$V_1^*(s_0) = \text{Nash}\{Q_1^*(s_0, a_1, a_2), Q_2^*(s_0, a_1, a_2)\}$$

Case 2: 
$$V_1^*(s_0) = 98$$

	Left	Up	
Right	96,47	98,49	
Up	49,98	73,61	

$$V_1^*(s_0) = \text{Nash}\{Q_1^*(s_0, a_1, a_2), Q_2^*(s_0, a_1, a_2)\}$$

Case 3: 
$$\{\pi_1(s_0), \pi_2(s_0)\} = (\{p(R) = 0.97, p(U) = 0.03\}, \{p(L) = 0.97, p(U) = 0.03\})$$

Left		Up	
Right	47.48, 47.48	98, 49	
Up	49,98	61.2, 61.2	

# **Optimal Q-function v.s. Nash Q-function**

# **Definition (Optimal Q-function)**

Optimal Q function is defined as

$$Q^*(s,a) = r(s,a,s') + \gamma \sum_{s' \in S} p(s'|s,a) V^*(s')$$

- $\triangleright V^*(s') = \max_{a} Q^*(s', a)$
- With **optimum** policy  $\pi^*(s) = \underset{a}{\operatorname{argmax}} Q^*(s, a)$

## **Definition (Nash Q-function)**

Nash-Q function is defined as

$$Q_{i}^{*}(s, \vec{a}) = r_{i}(s, \vec{a}, s') + \gamma \sum_{s' \in S} p(s'|s, \vec{a}) \frac{V_{i}^{*}(s')}{Nash Q_{i}(s')}$$

 $V_i^*(s') = \text{Nash } Q_i^*(s')$  is Nash equilibrium value that can be computed by solving the following state game

$$(Q_1^*(s',\vec{a}),...,Q_n^*(s',\vec{a}))$$

# **Definition (Nash equilibrium policy in Stochastic game)**

Compute the Nash equilibrium policies  $\pi^* = (\pi_1^*, \pi_2^*)$  such that for all  $s \in S$  and i = 1, ... 2,

$$V_i(s, \pi_i^*, \pi_{-i}^*) \ge V_i(s, \pi_i^*, \pi_{-i}^*)$$
 for all  $\pi_i \in \Pi_i$