



# 强化学习原理及应用 Reinforcement Learning (RL): Theories & Applications

*DCS6289 Spring 2022*

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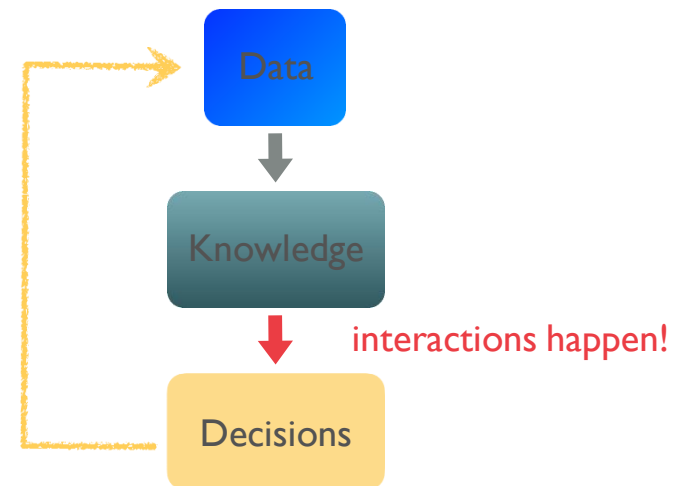
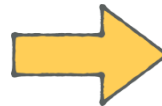
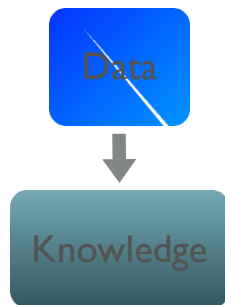
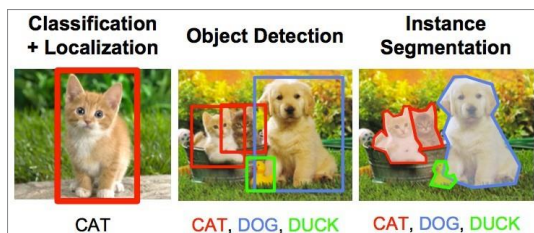
# Lecture 9: Multi-Agent RL

10<sup>th</sup> May. 2022

- ❑ What is the multi-agent reinforcement learning(MARL)?
- ❑ What is the difference between single-agent reinforcement learning and multi-agent reinforcement learning?
- ❑ What is the difficulty in the multi-agent reinforcement learning?
- ❑ What are the categories of multi-agent reinforcement learning?

# Multi-Agent RL

- ❑ What is the multi-agent reinforcement learning(MARL)?
  - ❑ Reinforcement learning turns data/knowledge into closed-loop decision making.
  - ❑ Multi-agent learning deal with interactions among the learning agents.



## □ What is the multi-agent reinforcement learning(MARL)?

- MARL addresses the sequential decision-making problem of multiple autonomous agents that operate in a common environment, each of which aims to optimize its own long-term return by interacting with the environment and other agents
- A group of agents work together to optimize team performance
- Multiagent systems include a set of autonomous entities(agents) that share a common environment and where each agent can independently perceive environment, act acting to its individual objectives and as a consequence, modify the environment
- In an multiagent system, agents must compete or cooperate to obtain the best overall results.

## □ What is the multi-agent reinforcement learning(MARL)?



Robots



Drone Delivery



Games



Autonomous  
Vehicles



Smart Grids



MALib

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❑ What is the difference between single-agent reinforcement learning and multi-agent reinforcement learning?

❑ Single-agent RL:

- Only one agent
- State、local action、single reward

❑ Multi-agent RL:

- At least two agents
- Local observation、joint action、team reward
- Agents communicate with each other and interact with environment at the same time.

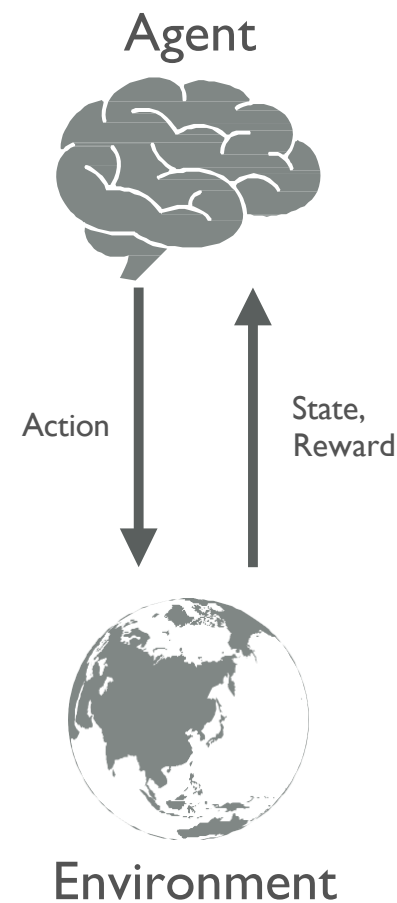


## □ What is the difference between single-agent reinforcement learning and multi-agent reinforcement learning?

### □ Problem Formulation: single-agent RL:

Markov Decision Process(MDP)  $(S, A, R, T, P_0, \gamma)$

- $S$  denotes the state space
- $A$  is the action space
- $R = R(s, a)$  is the reward function
- $T: S \times A \times S \rightarrow [0,1]$  is the state transition function
- $P_0$  is the distribution of the initial state
- $\gamma$  is a discount factor
- Goal: find the optimal policy that maximizes expected reward



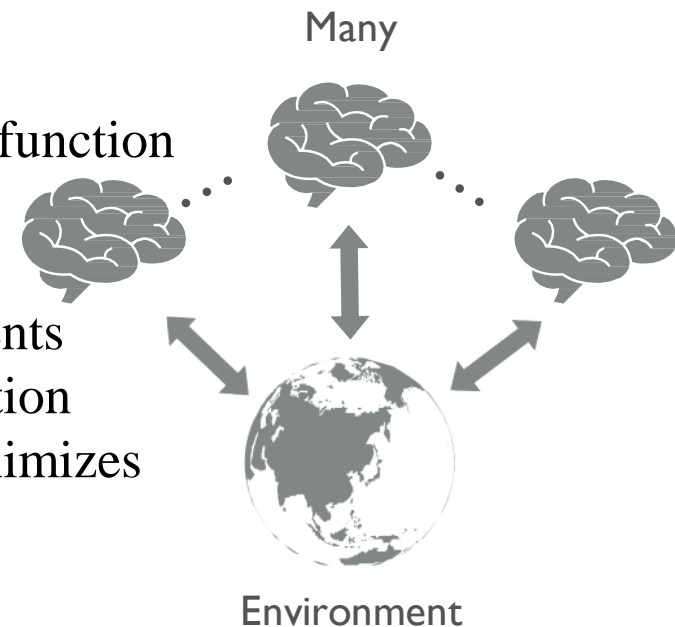
# Multi-Agent RL

## ❑ What is the difference between single-agent reinforcement learning and multi-agent reinforcement learning?

### ❑ Problem Formulation: multi-agent RL:

Partially Observable Markov Decision Process(POMDP)( $S, A, R, T, P_0, Z, O, n, \gamma$ )

- $n$  agents in the environment
- $S$  denotes the state space
- $A$  is the joint action space  $A^1 \times \dots \times A^n$
- $R = R(S, A)$  is the share reward function
- $T: S \times A \times S \rightarrow [0,1]$  is the state transition function
- $P_0$  is the distribution of the initial state
- $\gamma$  is a discount factor
- $Z$  is the individual observation for each agents
- $O(s, a): S \times A \rightarrow Z$  is the observation function
- Goal: find the optimal joint policy that maximizes expected team reward



- ❑ What is the multi-agent reinforcement learning(MARL)?
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- ❑ What are the categories of multi-agent reinforcement learning?

## □ What is the difficulty in the multi-agent reinforcement learning?

- Non-stationarity:
  - an agent observes not only the outcomes of its own action but also the behavior of other agents
  - Learning among the agents is complex because all agents potentially interact with each other and learn concurrently
- Partial observability:
  - The agents only capture partial information about the environment before making decision
- Dimension catastrophic:
  - Joint action space and Joint state space
  - Large-scale multi-agent decision-making、
- Credit assignment:
  - Lazy agent
- Sample efficiency、 Exploration and Exploitation、 complex mixed environment, etc.

## □ What is the difficulty in the multi-agent reinforcement learning?

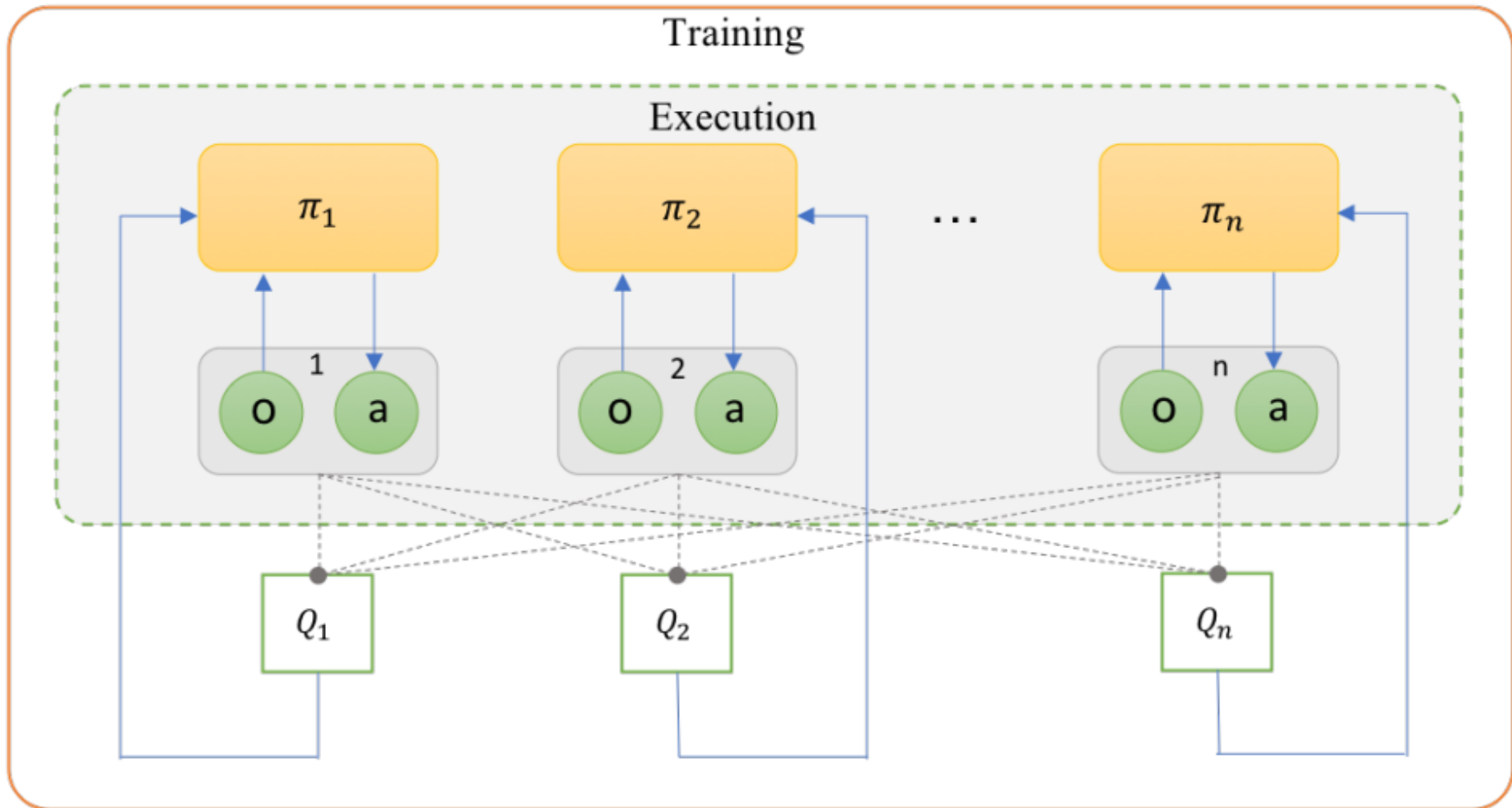
### Solution

- Non-stationarity:
  - Centralized Train and Decentralized execution(CTDE), e.g. MADDPG
  - Communication, e.g. CommNet
- Partial observability:
  - RNN、GRU、LSTM, e.g. DRQN
- Dimension catastrophic:
  - CTDE, e.g. VDN、QMIX
  - Mean-Field, e.g. MFAC
- Credit assignment:
  - Counterfactual mechanisms, e.g. COMA
- Exploration and Exploitation:
  - Reward shaping(intrinsic reward、novelty)
  - UCB
  - Influence(mutual information)

# Multi-Agent RL

❑ What is the difficulty in the multi-agent reinforcement learning?

Centralized Train and Decentralized execution(CTDE)



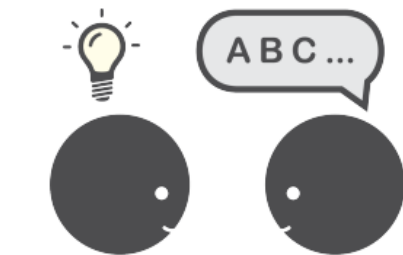
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## □ What are the categories of multi-agent reinforcement learning?

- Analysis of emergent learning:
  - Simply using Single-agent RL algorithm in multi-agent scenarios
- Learning communication:
  - Learning communication protocols among agents
- Learning cooperation:
  - Learning to cooperate using only actions and local observation
- Agents modeling agents:
  - Reasoning about others



(a) Analysis of emergent behaviors



(b) Learning communication



(c) Learning cooperation

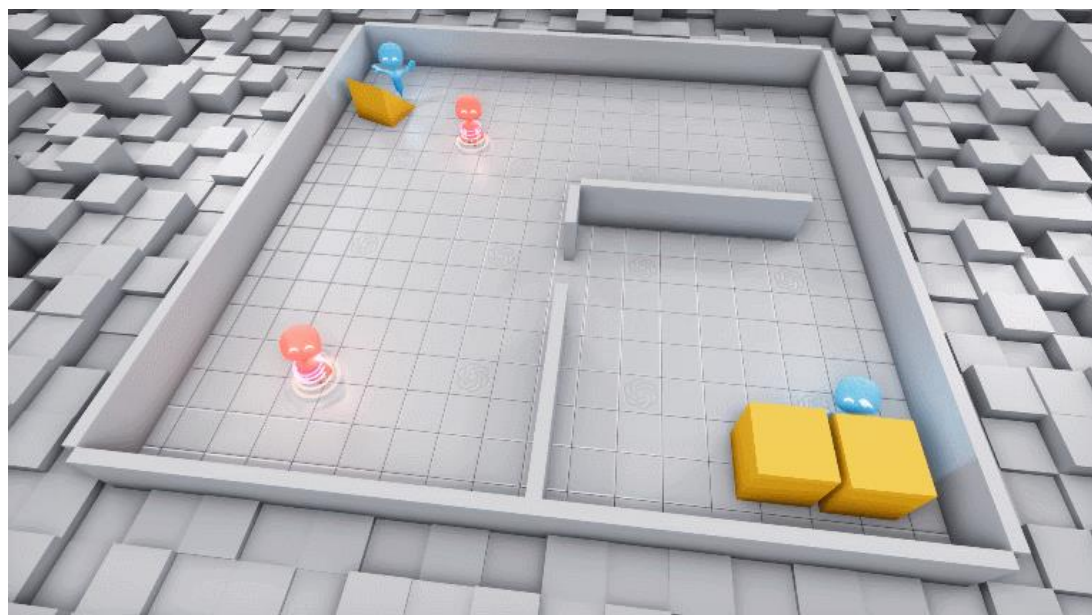
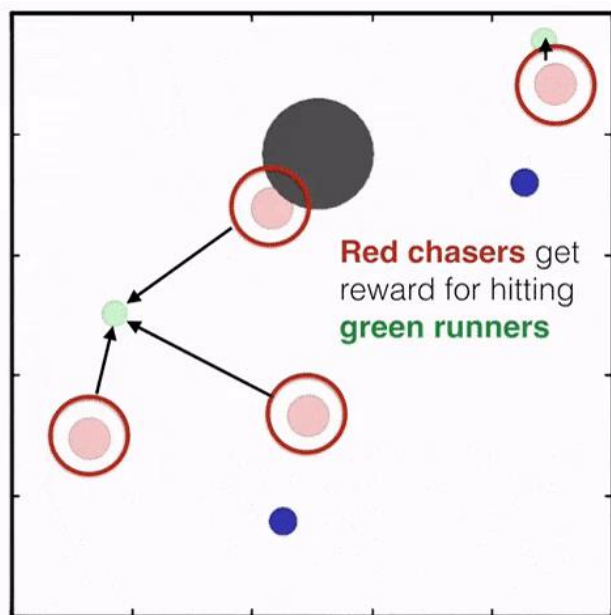


(d) Agents modeling agents



## □ What are the types of multi-agent reinforcement learning?

- Analysis of emergent learning:
  - Three major settings: **cooperative**, **competitive**, **mixed scenarios**
    - **Cooperative**: working together and coordinating their actions、maximizing a shared team reward
    - **Competitive**: self-interested(maximizing an individual reward)、opposite rewards、zero-sum games
    - **Mixed scenarios**: general-sum games



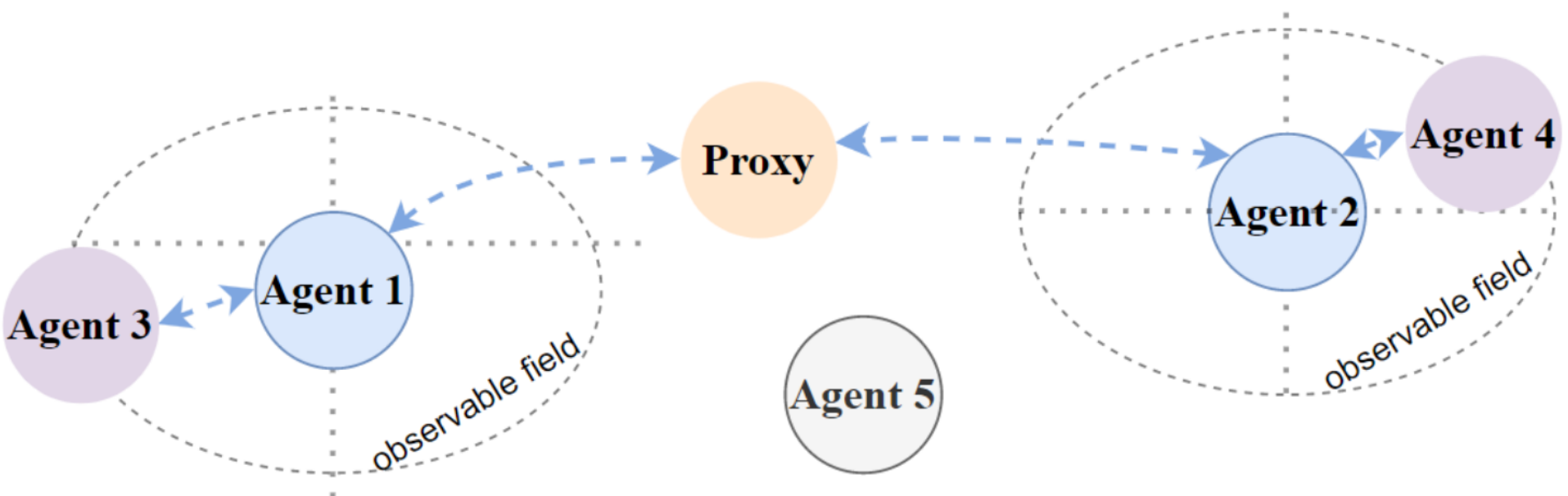
## □ What are the types of multi-agent reinforcement learning?

- Analysis of emergent learning:
  - These works, in general, do not propose learning algorithms—their main focus is to analyze and evaluate DRL algorithms, e.g., DQN, PPO and others, in a multiagent environment

| Summary  | Learning | Setting |
|--|----------|---------|
| Train DQN agents to play Pong.   | VB       | CO&CMP  |
| Train DQN agents to play sequential social dilemmas.                                       | VB       | Mixed   |
| Propose DRL agents able to cooperate in social dilemmas.                                   | VB       | Mixed   |
| Propose Malthusian reinforcement learning which extends self-play to population dynamics.  | VB       | Mixed   |
| Train PPO agents in competitive MuJoCo scenarios.  | PG       | CMP     |
| Train PPO, A3C, and DQN agents in attacker-defender games.                                 | VB, PG   | CMP     |
| Train agents represented with NN to learn a communication language.                        | PG       | CO      |
| Learn communication with an end-to-end differentiable model to train with backpropagation. | PG       | CO      |

## □ What are the types of multi-agent reinforcement learning?

- Learning communication:
  - These works explore a sub-area in which agents can share information with communication protocols, for example through direct message or via a shared memory
  - This setting usually considers a set of cooperative agents in partially observable environment



## □ What are the types of multi-agent reinforcement learning?

- Learning communication—**Dimensions and Targeted Problems**

| Dimensions                | Targeted Problems   |
|---------------------------|---|
| Communication Type        | Which type of agents to communication with?                       |
| Communication Policy      | When and how to build communication links among agent?            |
| Communicated Messages     | Which piece of information to share?                              |
| Message Combination       | How to combine received messages?                                 |
| Inner Integration         | How to integrate combined message into learning models?           |
| Communication Constraints | How to fulfill realistic requirements?                            |
| Communication learning    | How to train and improve communication?                           |
| Training Scheme           | How to utilize collected experience from agents?                  |
| Controlled Goals          | What kind of behaviours are desired to emerge with communication? |

## □ What are the types of multi-agent reinforcement learning?

- Learning communication—Communicatee Type

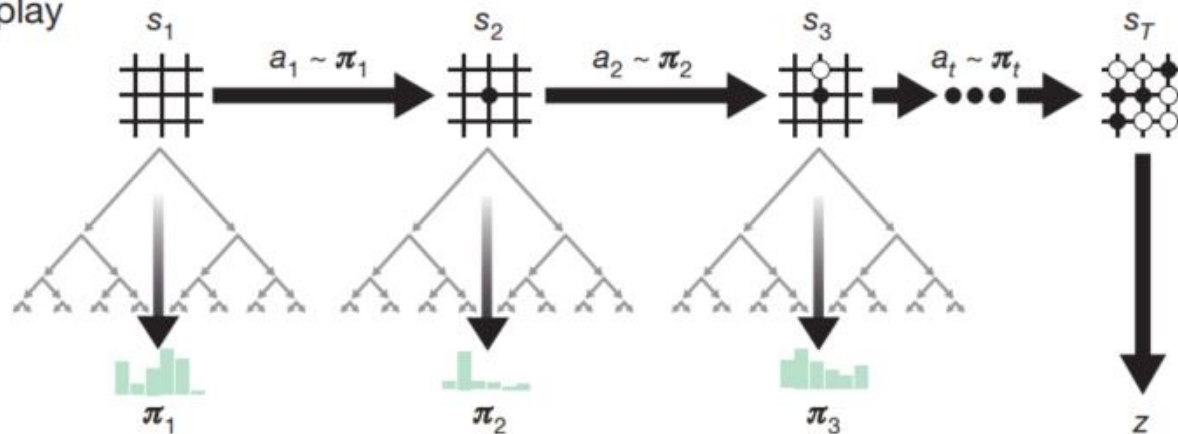
| Types  | Subtypes   | Methods   |
|--|--|---|
| <p>Agents in the MAS:</p> <ul style="list-style-type: none"><li>• communicatees is composed of agents</li><li>• agent directly communicate with each other</li></ul> | <p>Nearby Agents:<br/>Communication is only allowed between neighbors</p>  | DGN; MAGNet-SA-GS; Agent-Entity Graph; LSC; NeurComm; FlowComm; GAXNet          |
|  | <p>Other Agents:<br/>Nearby agents are not identified, the set of communicatees is simply composed of other agents</p> | DIAL; RIAL; CommNet; BiCNet; TarMAC; MADDPG-M; IC3Net; DCC-MD; VBC; ETCNet; TMC |
| <p>Proxy:<br/>A proxy is a visual agent who plays an essential role in communication, but does not have direct effect on the environment</p>                         |  | MS-MARL-GCM; ATOC; MD-MADDPG; IMAC; GA-Comm; Gated-ACML; HAMMER; MAGIC          |

# Multi-Agent RL

## □ What are the types of multi-agent reinforcement learning?

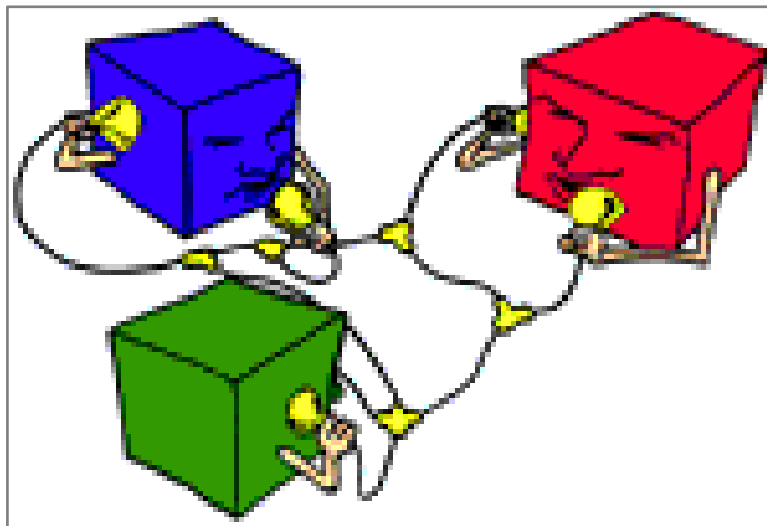
- Agents modeling agents :
  - An importance ability for agents to have is to reason about the behaviors of other agents by constructing models that make predictions about the modeled agents
  - Opponent model: utilizing network to learn a representation of the opponent's policy or q-value, e.g., DRON; SOM
  - Self-play: agents learn to improve their performance by playing “against themselves”, e.g., NFSP; PSRO
  - Minimax: minimizing the worst case scenario (maximum loss), e.g., M3DDPG

Self-play



## □ What are the types of multi-agent reinforcement learning?

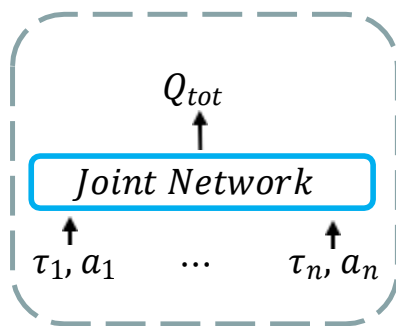
- Learning cooperation :
  - Open, dynamic, persistent systems
  - Decentralized control
  - Large scale
  - Partial observability
  - No real-time global reward signal
  - Communication delay



# Multi-Agent RL

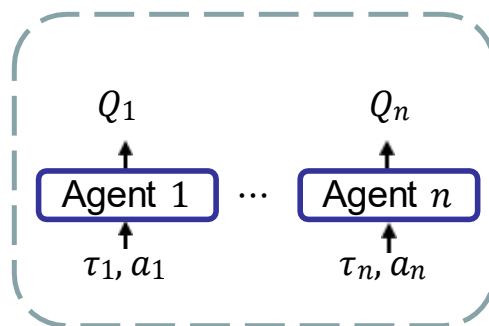
## □ What are the types of multi-agent reinforcement learning?

- Learning cooperation :
  - Value-Based Methods
    - Paradigm: CTDE
    - Methods: VDN, QMIX, QPLEX
  - Policy Gradient Methods
    - Paradigm: Centralized Critic and Decentralized Actors
    - Methods: MADDPG, COMA, DOP




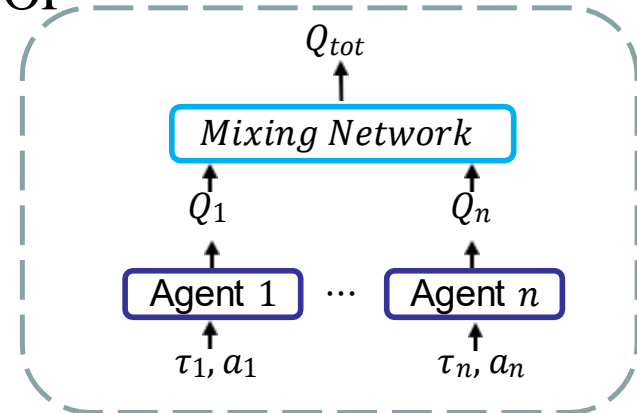
Centralized Value Functions

Scalability 




Decentralized Value Functions

Non-stationarity  
Credit assignment 



Factorized Value Functions

Centralized training  
Decentralized execution 



# Multi-Agent RL

## □ Summary

