



Exploration-Exploitation in Multi-Agent Competition

Convergence with Bounded Rationality

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Introduction: Multi-Agent Competition

Motivation

Many recent ML and AI advances involve competitive interactions between 2-agents

- generative adversarial networks (GANs)
- actor-critic systems
- competitive game-playing: chess, Go

Modelled as strictly-competitive, 2-agent, zero-sum games or variants thereof

- multiple equilibria but unique value
- equilibrium strategies are exchangeable
- optimization-driven algorithms perform well

What happens beyond these settings?







Motivation: Open Questions

In multi-agent competition, many properties of the 2-agent settings collapse

- multiple but payoff-diverse equilibria
- exploration-exploitation for equilibrium selection

Multi-agent vs 2-agent competition

- not only significantly harder
- but also qualitatively different

Research goals: in networks of strictly competitive games

- convergence of exploration-exploitation dynamics
- equilibrium selection with payoff-diverse equilibria







Game-Theoretic Model

Weighted Zero-sum Polymatrix Games

A weighted zero-sum polymatrix game (WZPG), $\Gamma = ((V, E), (S_k, w_k)_{k \in V}, (\mathbf{A}_{kl})_{[k,l] \in E})$

$$egin{aligned} u_k(\mathbf{x}) &:= \mathbf{x}_k^ op \sum_{[k,l]\in E} \mathbf{A}_{kl} \mathbf{x}_l = \mathbf{x}_k^ op r_k\left(\mathbf{x}_{-k}
ight) \ &\sum_{k\in V} w_k u_k(\mathbf{x}) = 0, \ \ ext{for all } \mathbf{x}\in\Delta. \end{aligned}$$



Nash Equilibrium (NE): a strategy profile, $\mathbf{p} = (p_k)_{k \in V} \in \Delta$, with one strategy for each agent $k \in V$, $\mathbf{p}_k = (p_{ki})_{i \in S_k} \in \Delta_k$ such that

$$u_k(\mathbf{p}) \ge u_k(x_k, \mathbf{p}_{-k}), \text{ for all } x_k \in \Delta_k, k \in V.$$

Properties: WZPGs capture complexities of multi-agent competition

• multiple NE with non-unique payoff values and non-exchangeable NE strategies

Joint Learning Model

Joint Learning Model

Q-Learning Dynamics (QLD)

Q-value updates and Boltzmann selection probabilities for all agents $k \in V$

$$\frac{\dot{\mathbf{x}}_{ki}}{\mathbf{x}_{ki}} = \underbrace{\mathbf{r}_{ki}(\mathbf{x}_{-k}) - \mathbf{x}_{k}^{\top}\mathbf{r}_{k}\left(\mathbf{x}_{-k}\right)}_{\text{exploitation}} - T_{k}\underbrace{\left[\ln\left(\mathbf{x}_{ki}\right) - \mathbf{x}_{k}^{\top}\ln\left(\mathbf{x}_{k}\right)\right]}_{\text{exploration}},\tag{1}$$

Exploration rates *T_k*:

- $T_k = 0$: select action with highest *Q*-value (exploitation)
- $T_k \rightarrow \infty$: uniformly randomize over actions (exploration)

Interpretation of *T_k*'s:

- physics: temperature of the system
- behavioral: agents bounded rationality or discounting of past payoffs
- algorithmic: regularization to avoid boundary or local optima

Solution Concept: QRE

Quantal Response Equilibria

Quantal Response Equilibria (QRE), $\mathbf{p} = (p_k)_{k \in V}$, of Γ

- standard solution concept in games with bounded rationality
- logit (softmax) form that depends on exploration rates

$$p_{ki} = \frac{\exp\left(r_{ki}/T_k\right)}{\sum_{j \in S_k} \exp\left(r_{kj}/T_k\right)}, \quad \text{for all } i \in S_k, k \in V.$$
(2)

• may be very different from NE, but not when T_k are close to 0.

Theorem (Interior Fixed Points of QLD)

The interior fixed points, $\mathbf{p} = (p_k)_{k \in V}$, of the *Q*-learning dynamics in an arbitrary game Γ with positive exploration rates, $T_k > 0$, always exist and coincide with the QRE of Γ .

A strategy profile $\mathbf{p} = (p_k)_{k \in V}$ is an interior fixed point of QLD if the RHS in (1) is 0.

Main Result

Main Result

Convergence of Q-Learning to QRE in Multi-Agent Competition

Main Theorem (Informal)

Let Γ be a WZPG, with positive exploration rates, $T_k > 0$, for all $k \in V$. There exists a unique QRE, **p**, such that any trajectory, $\mathbf{x}(t)$, of the Q-learning dynamics starting from an arbitrary interior point, converges to **p** exponentially fast.

Takeaways

- despite the diversity of NE, we have uniqueness of QRE
- as $T_k \rightarrow 0$, QRE approaches a NE of Γ : way out of tight spot of equilibrium selection

Remarks

- tight assumptions: if $T_k = 0$ for some $k \in V$, then QLD may converge to the boundary even for interior starting points.
- prior work: QLD provably converges in multi-agent coordination.

Visualization of the QRE Manifold

Asymmetric Matching Pennies (AMPs): 2-agent, weighted zero-sum game with

$$\mathbf{A} = \begin{pmatrix} 2 & -2 \\ 0 & 2 \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} 4 & 0 \\ -4 & -4 \end{pmatrix},$$

so that $\mathbf{A} + 0.5 \cdot \mathbf{B}^{\top} = 0$ and a unique interior NE at $(\mathbf{p}, \mathbf{q}) = ((1/3, 2/3), (2/3, 1/3))$.



- ETE: explore-then-exploit
- CLR-1: cyclical learning rate (1-cycle)



• QRE manifold and exploration path

Excursion: QLD in Multi-Agent Coordination

Multi-agent learning in coordination settings (prior work)¹

- QLD provably converges in multi-agent weighted potential games
- multiple QRE, but bifurcation phenomena explain equilibrium selection
- equilibrium selection after exploration depends on a game's geometry



¹S. Leonardos, G. Piliouras, Exploration-Exploitation in Multi-Agent Learning: Catastrophe Theory Meets Game Theory, AAAI-21, Best paper award.

Convergence to QRE

Match-Mismatch Game (MMG): line-network WZPG with



•
$$\mathbf{A}_{+} = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$$
, $\mathbf{A}_{-} = -\mathbf{A}_{+}$ and $\mathbf{A}_{1} = \mathbf{A}_{2} = (1, -1)$

- first and last are dummy agents with fixed actions
- goal: mismatch the previous and match the next agent
- infinite many NE: (T, H/T, T, H/T, ...)

Convergence to QRE

Match-Mismatch Game (MMG): line-network WZPG with





Convergence result is tight

Conclusions

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Takeaways: Q-Learning and Quantal Response Equilibria

Multi-agent competition

- despite the diversity of NE, we have uniqueness of QRE
- QLD converges to QRE and solves the equilibrium selection problem

Multi-agent coordination (prior work)

- QLD converges to QRE in multi-agent weighted potential games
- even with multiple QRE, bifurcation phenomena explain equilibrium selection

Next steps

- Can we go beyond that: mixed games with both cooperation and competition?
- Effects of exploration on individual/social welfare after equilibrium selection?

Thank you