Learning to Communicate in Multi Agent Settings

Joel Oskarsson, joeos014@student.liu.se

Reinforcement Learning

- Partial observability
- Collaborative



Communication



Deep RL Algorithms

• Goal: Learn a policy $\pi(a_i|\omega_i)$

Policy Gradient Methods (e.g. REINFORCE)

• Policy is Neural Network: $\pi(a_i|\omega_i) = NN_ heta(\omega_i)$

$$\omega_i \longrightarrow NN_{\theta} \longrightarrow \pi(a_i | \omega_i)$$

[Sutton et al.]

Deep Q-learning (DQN)

• Q-function is Neural Network: $Q(\omega_i,a_i)=[NN_ heta(\omega_i)]_{a_i}$

$$\omega_i \longrightarrow NN_{ heta} \longrightarrow Q(\omega_i, a_i)$$

[Mnih et al.]

Discrete Messages

• Binary messages $m_i \in \{0,1\}^L$

Differentiable Inter-Agent Learning (DIAL)

[Foerster *et al.*]

- DQN for actions: $Q(\omega_i, a_i)$
- Messaging trained implicitly





Continuous Messages

• Real-valued messages $m_i \in \mathbb{R}^L$

Communication Neural Net (CommNet)

[Sukhbaatar et al.]

- REINFORCE
- Dynamic amount of agents



Extensions

IC3Net [Singh et al.]

• Learning when to communicate

TarMAC [Das et al.]

• Targeted communication



Thanks for listening! Questions are welcome!

References

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