

Emergent Communication - First steps



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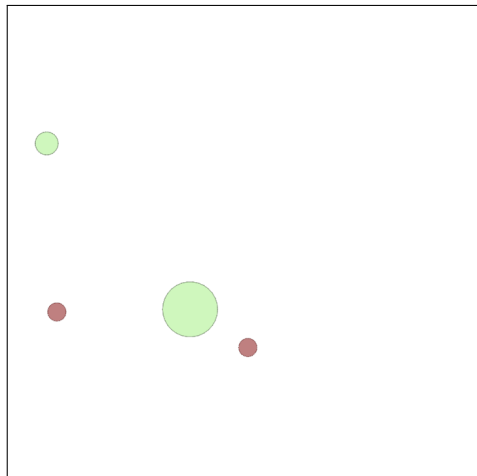
December 2nd, 2021

Fixing the environment:

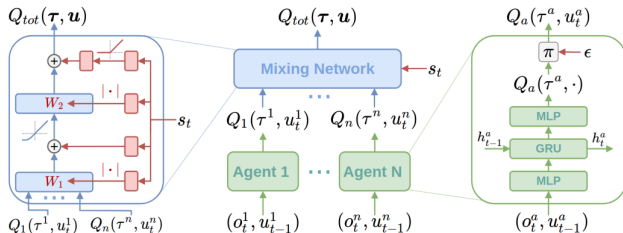
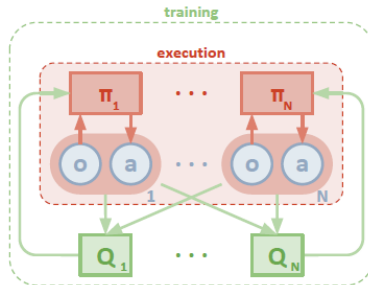
- Bigger agents and object
- More steps (100 steps maximum)
- Reward:
 - *obj*: $-1 \times \text{squared distance}$ between object and landmark
 - *dist*: $-1 \times \text{average distance}$ between agents and object
 - + penalty for collision between agents

Closing the environment:

- Walls on each side of the frame
- Agents can't go beyond $[-1, 1]^2$
⇒ Should be easier to solve



- MADDPG
- CMA-ES
- QMIX

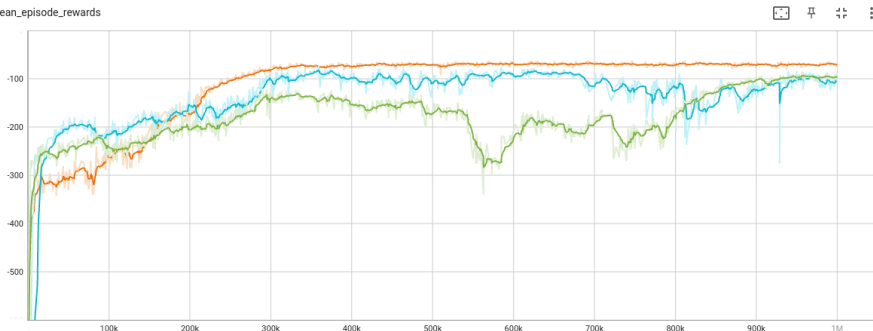


Opened environment:

Fully observable

- MADDPG

mean_episode_rewards



Best run: 2 agents, $lr=0.005$, discrete actions

Near optimal mean episode reward $\cong -60$

Opened environment:

Fully observable, 2 agents

- CMA-ES (160 parameters, population = 19)

mean_episode_rewards

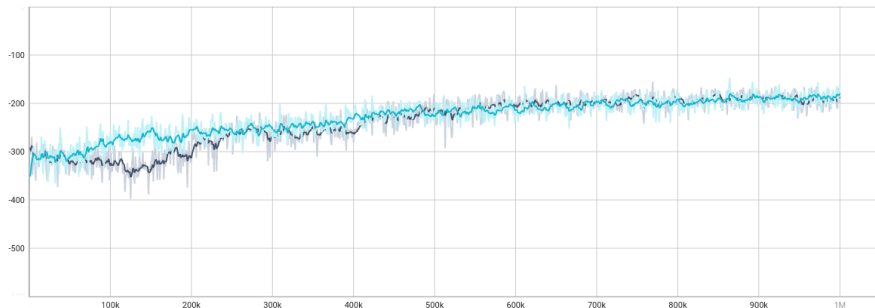


Opened environment:

Fully observable, 2 agents

- QMIX

average_episode_rewards

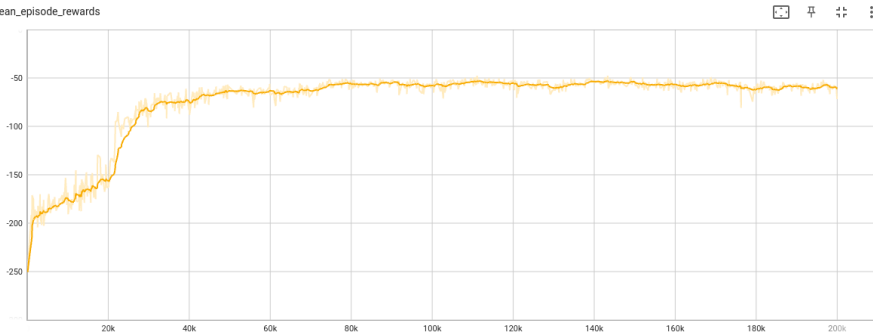


Closed environment:

Fully observable, 2 agents, no collision penalty

- MADDPG

mean_episode_rewards



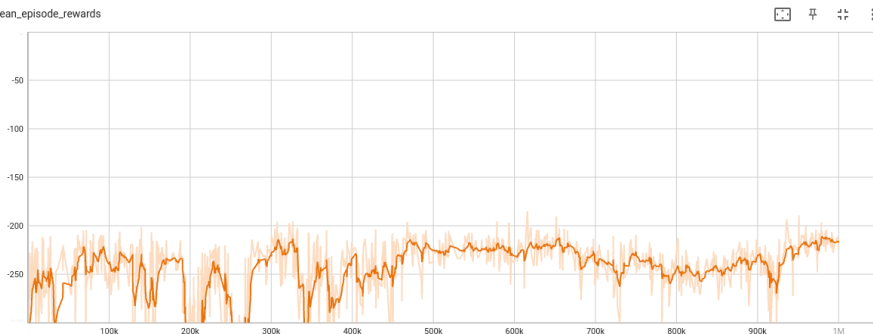
Best run so far: $lr=0.005$, continuous actions, train every 2500 steps

Closed environment:

Fully observable, 2 agents

- CMA-ES

mean_episode_rewards

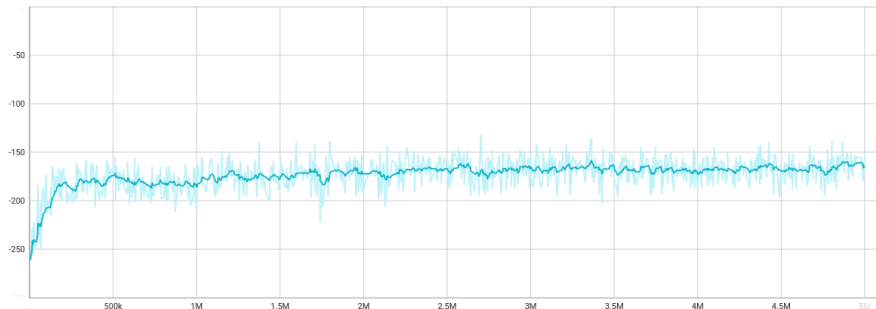


Closed environment:

Fully observable, 2 agents

- QMIX

average_episode_rewards



Fixing issues:

- Fix CMA-ES
- Fix QMIX
- Find better reward function

Testing harder versions of the environment:

- Partial observability
- Heavier object
- More agents

- How to learn emergent communication well ?
 - Sukhbaatar et al., Learning Multiagent Communication with Backpropagation, 2016
 - Das et al., TarMAC: Targeted Multi-Agent Communication, 2019
 - Learning to Ground Multi-Agent Communication with Autoencoders, 2021
- Use transformers ?
 - Parisotto et al., Stabilizing Transformers for Reinforcement Learning, 2020
 - Karch et al., Grounding Spatio-Temporal Language with Transformers, 2021
- Path to discrete communication
 - Language modeling
 - Language-Augmented Reinforcement Learning

Thank you!