

Action Trading for Self-Interested Multi-Agent Reinforcement Learning in a Smart Factory Setting

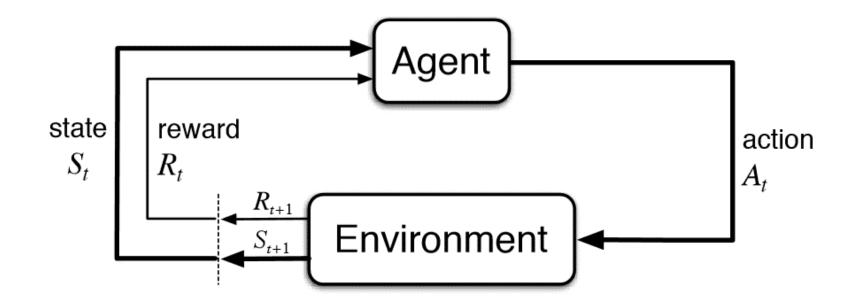
Arnold Unterauer



Reinforcement Learning







Source: Richard S. Sutton and Andrew G. Barto Reinforcement Learning: An Introduction





Motivation



Multi Agent Systems:

agents act selfish

no cooperation between agents

unused potential

Solution:

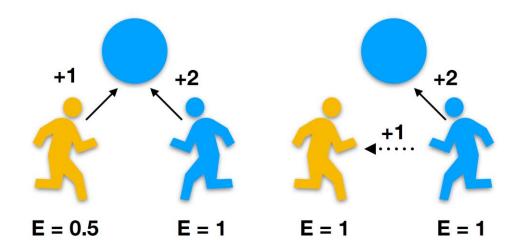
Cooperative Game Theory

motivate agents to cooperate with each other



Related Work – Action Trading





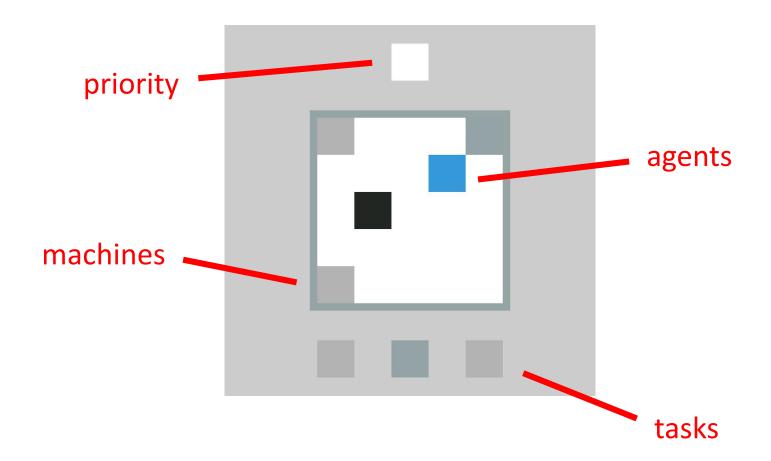
agents trade reward for the following action
agents are able to trade with extended action space
cooperating agents outperformed selfish ones

Source: Kyrill Schmid, Lenz Belzner, Thomas Gabor, and Thomy Phan Action Markets in Deep Multi-Agent Reinforcement Learning



Environment – Smart Factory



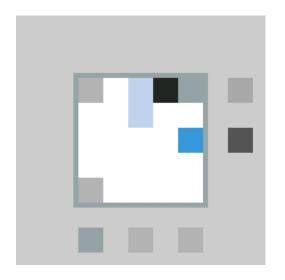


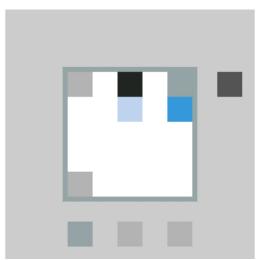


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Action Trading







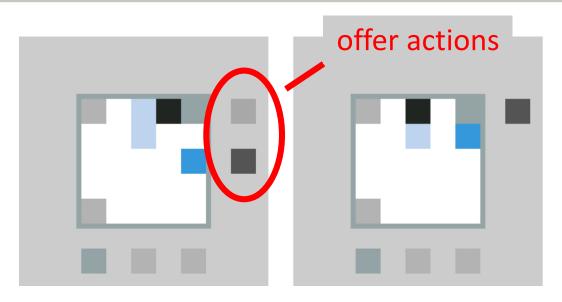


agent makes an offer consisting of n offer actions the other agent can decide to perform the offer actions by executing all offer actions the agent gets paid



Action Trading





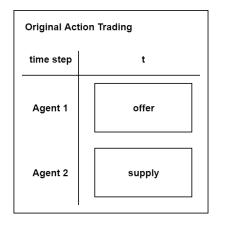


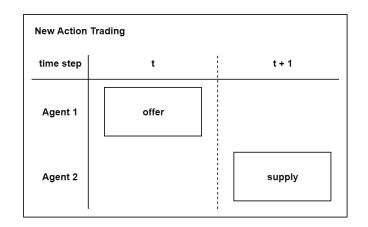
agent makes an offer consisting of n offer actions
the other agent can decide to perform the offer actions
by executing all offer actions the agent gets paid



Modification on Action Trading







Original Action Trading:

offer and supply match at same time step fixed compensation not scalable offer actions

New Action Trading:

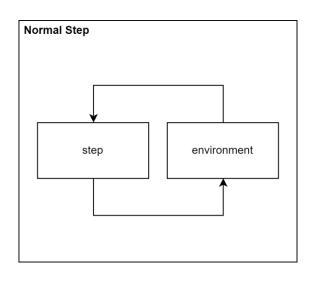
offer and supply in separate time steps compensation based on state scalable offer actions

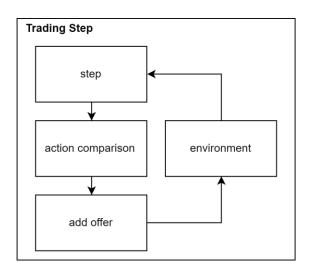


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Environment with Action Trading







Extending normal environment step:

action comparison: checks if last action matched offer action

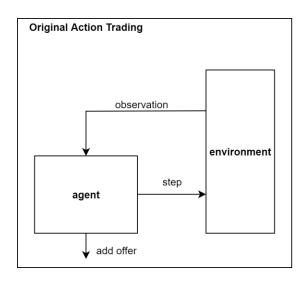
add offer: agent makes a new offer

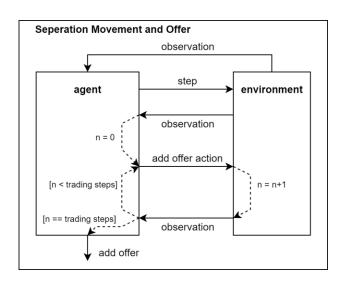


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Trading Modes







Original Action Trading:

one single action determines movement and offer

 \rightarrow exponentially growing action space \rightarrow constant action space

Separation Movement and Offer:

seperates the selection of movement and offer actions

offer actions are added one at a time



Compensation





Offer:

OFFER ACTION 0

LEFT

Expected Rewards:

Q0 UP	Q1 DOWN	Q2 LEFT	Q3 RIGHT
0.05	0.01	- 0.07	- 0.04

Difference D:

$$Q_{max} - Q_{offer}$$

$$0.05 - (-0.07) = 0.12$$

Compensation in Future:

$$D/\gamma$$

$$0.12/0.95 = 0.126$$

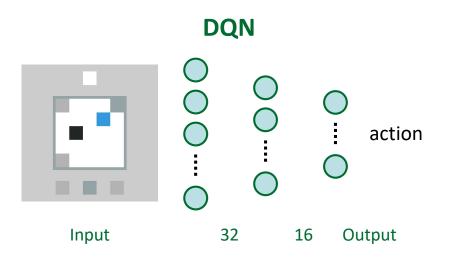
$$\longrightarrow Compensation = Mc * \sum_{l=0}^{n-1} \frac{Q_{max}(t+l) - Q_{offer}(t+l)}{\gamma}$$



Experiments







Decaying ϵ -greedy: $\epsilon = 1 \rightarrow \epsilon_{decay} = 8e^{-6} \rightarrow \epsilon_{min} = 0.01$

Discount rate: $\gamma = 0.95$

20 runs: 1500 training episodes 2000 evaluation episodes



Experiments



Action Trading vs No Action Trading

Trading Mode: Original Action Trading

Trading Mode: Separation of Movement and Offer

Comparison between Trading Modes

Mark up

Budget

Payment Timing

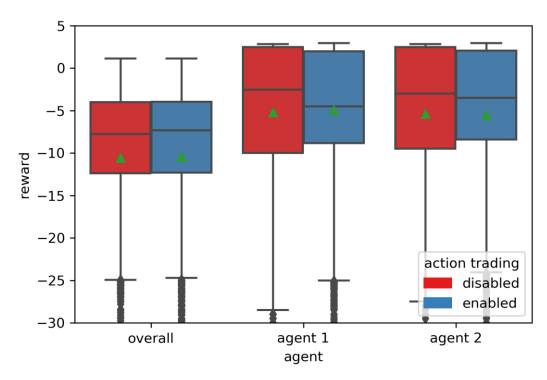
Partial Payment



Experiments – Trading vs No Trading



Trading Mode: Original Action Trading

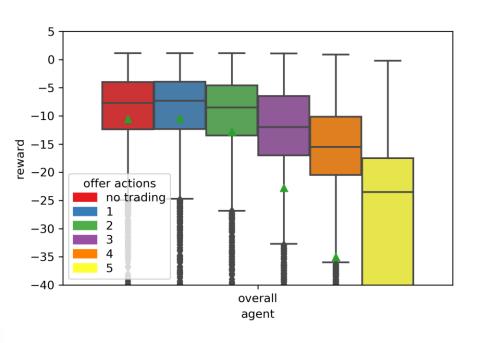


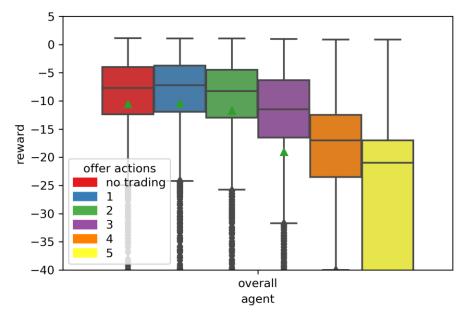


Experiments – Original Action Trading



Trading Mode: Original Action Trading





mark up: 1.00

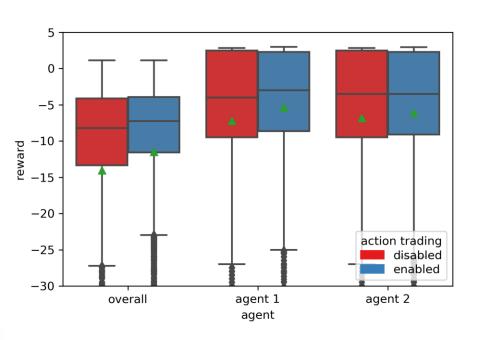
mark up: 1.05

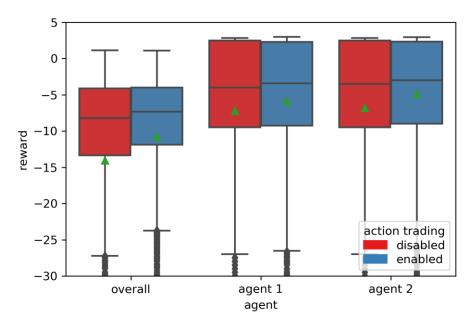


Experiments – Separation Movement and Offer



Trading Mode: Separation Movement and Offer





mark up: 1.00

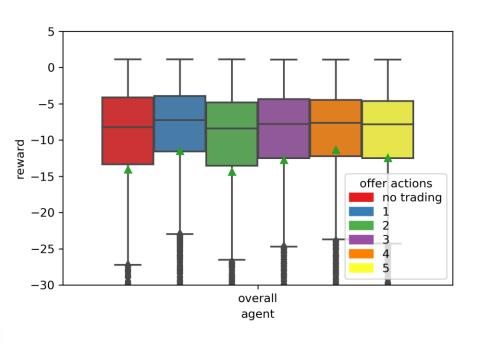
mark up: 1.05

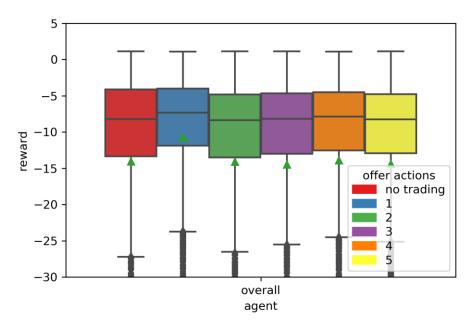


Experiments – Separation Movement and Offer



Trading Mode: Separation Movement and Offer





mark up: 1.00

mark up: 1.05



Experiments – Comparison Trading Modes





Original Action Trading

	_		_		
n	mark up	action space	trades/episode	mean reward	median reward
-	-	4	-	-10.57	-7.74
1	1.0	20	2.80	-10.45	-7.32
2	1.0	68	0.60	-12.82	-8.5
3	1.0	260	0.13	-22.77	-12.00
4	1.0	1028	0.03	-35.02	-15.50
5	1.0	4100	0.00	-86.71	-23.50
1	1.05	20	2.78	-10.40	-7.20
2	1.05	68	0.62	-11.74	-8.24
3	1.05	260	0.13	-19.03	-11.50
4	1.05	1028	0.03	-49.30	-17.00
5	1.05	4100	0.00	-70.66	-21.00

Separation Movement and Offer

n	morle un	action appea	tradas/anisada	moon roward	median reward
n	mark up	action space	trades/episode	mean reward	median reward
-	-	8	-	-14.04	-8.22
1	1.0	8	1.88	-11.49	-7.24
2	1.0	8	0.40	-14.34	-8.4
3	1.0	8	0.16	-12.73	-7.8
4	1.0	8	0.11	-11.33	-7.64
5	1.0	8	0.04	-12.45	-7.84
1	1.05	8	1.70	-10.75	-7.34
2	1.05	8	0.51	-14.07	-8.36
3	1.05	8	0.17	-14.48	-8.16
4	1.05	8	0.09	-13.91	-7.88
5	1.05	8	0.05	-14.41	-8.26

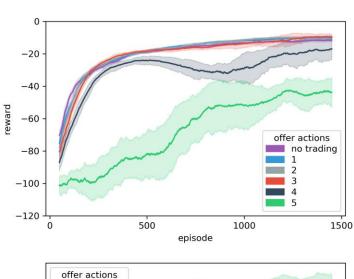


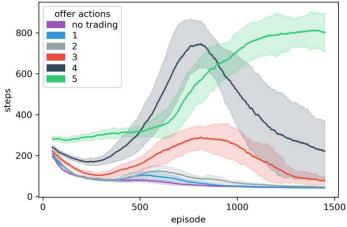
Experiments – Comparison Trading Modes



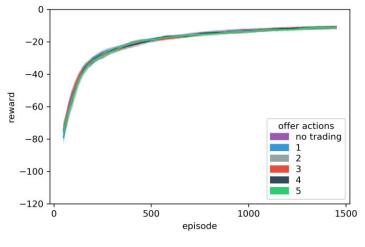


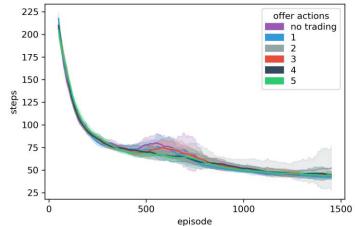
Original Action Trading





Separation Movement and Offer





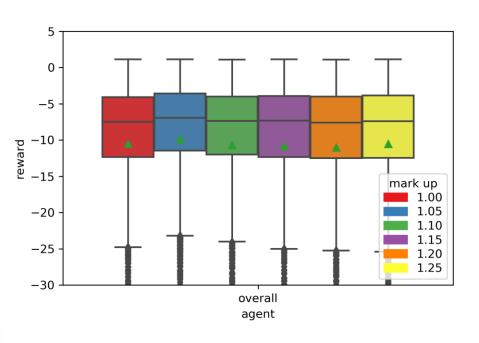


Experiments - Mark up

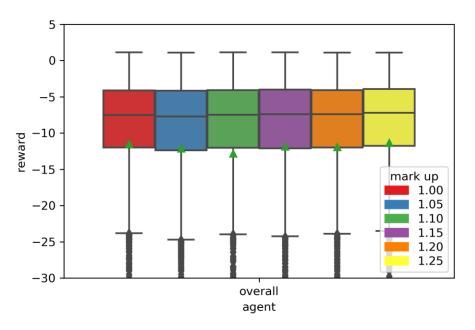




Original Action Trading



Separation Movement and Offer

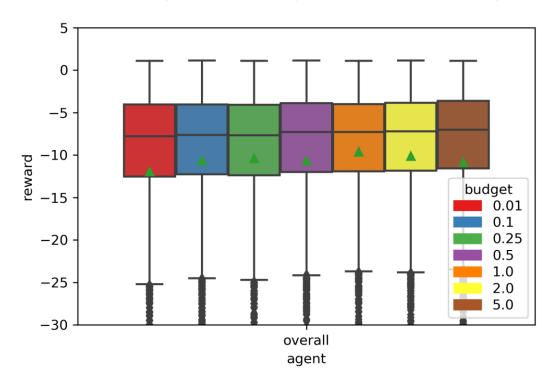




Experiments – Budget



Trading Mode: Original Action Trading

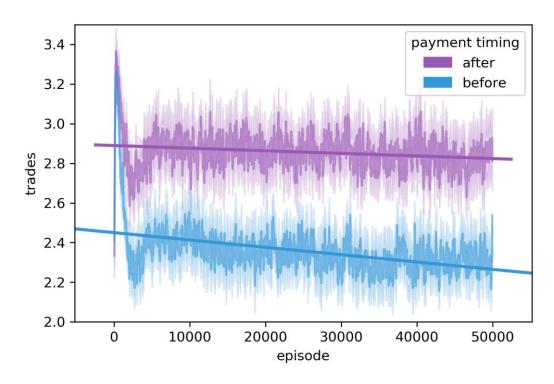




Experiments – Payment Timing



Trading Mode: Original Action Trading



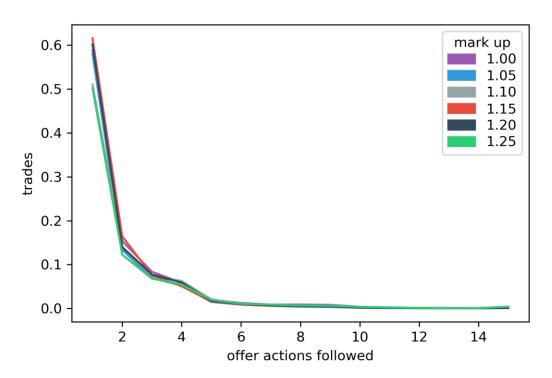


Experiments – Partial Payment





Trading Mode: Separation Movement and Offer





Conclusion





Action Trading outperforms No Action Trading

Trading Modes have different advantages:

Original Action Trading: more cooperation

Separation Movement and Offer: stable across n

Existing optimal mark up

Budget has only negative impact

Less cooperation between agents who pay beforehand

Agents rather cooperate short- than long-term







Thank you for your attention