

Joint meeting

Andrea Pierré March 4th, 2024

Brown University

Outline

1. Online Deep RL training

2. Generalization experiment

3. Discussion

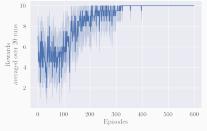
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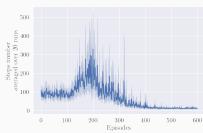
1. Online Deep RL training

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Rewards & steps





- · Lights cues in the state?
- Start training once replay buffer is full (5000 transitions) instead of when there are enough transitions for a batch (32 transitions)
- Soft update of the networks weights (instead of sharp transition)
- Huber loss instead of mean squared error → should be less sensible to outliers
- · Remove ReLU on output layer.

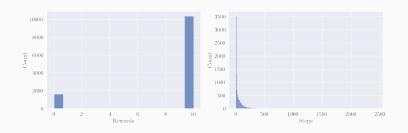
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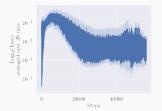
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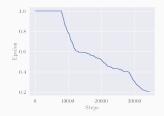
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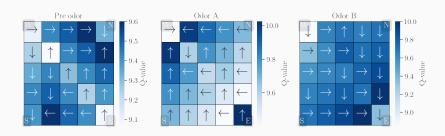
Loss, rewards & steps distributions, exploration/exploitation rate







Policy learned



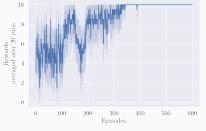
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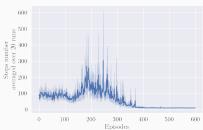
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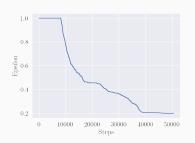
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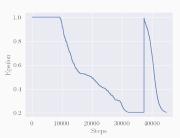
Training only in the lower triangle

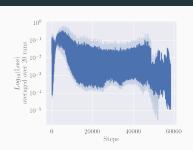


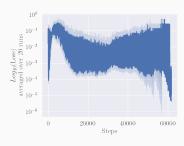


Training only in the lower triangle then switch to the upper triangle

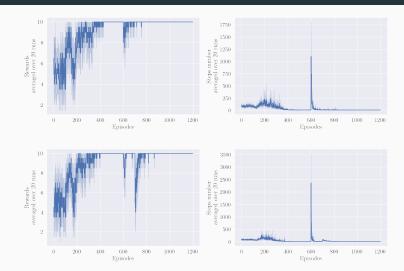








Training only in the lower triangle then switch to the upper triangle



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1. Online Deep RL training

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- · Debrief from the meeting with Thomas
- Topics of discussion for future meetings?
 - How to compare neural data with simulation data?
 Journal club (e.g. MINDS paper, etc.)
 Any other topics to add?

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