Stabilizing Contrastive RL: Techniques for Robotic Goal Reaching from Offline Data

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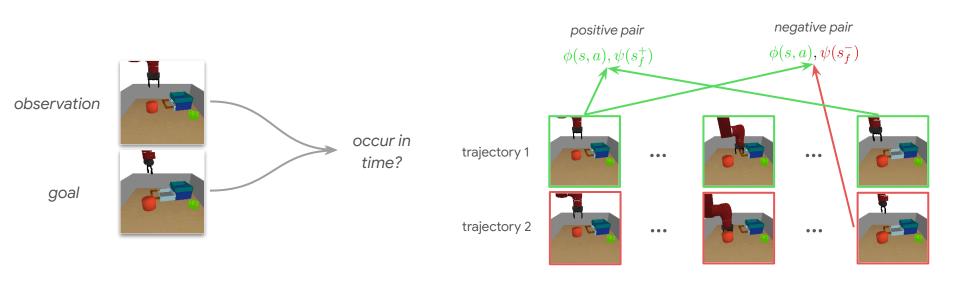




Reward labeling is challenging for long-horizon offline RL tasks.



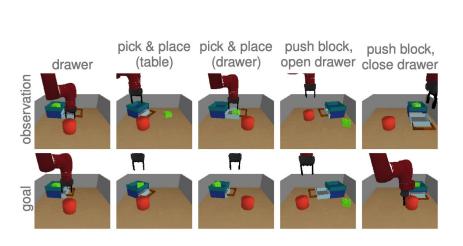
Prior contrastive method avoids reward engineering.

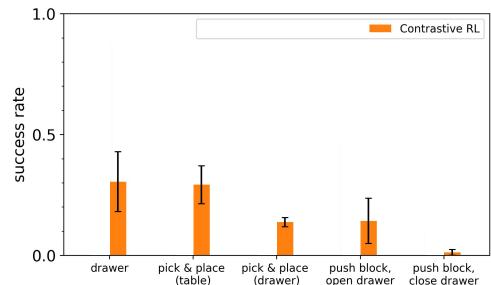


^[1] Eysenbach et al. Contrastive Learning as Goal-Conditioned Reinforcement Learning. 2022.

^[2] Eysenbach et al. C-Learning: Learning to Reach Goal via Recursive Classification. 2020.

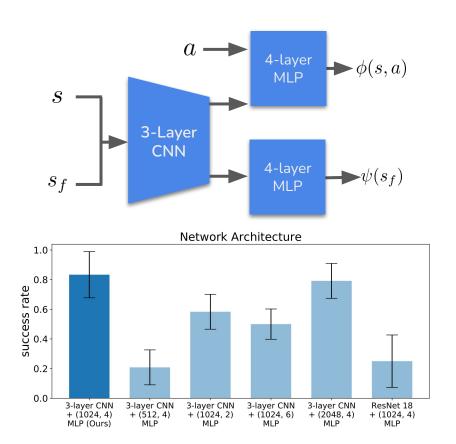
Prior contrastive method fails to scale.





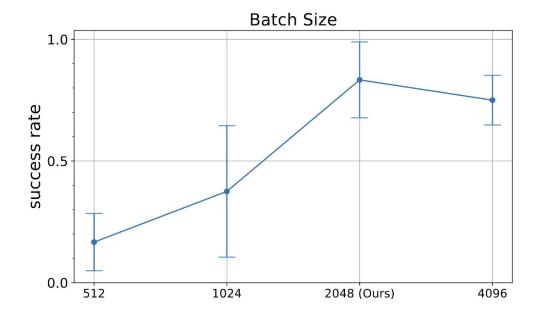
Design decisions

- 1. Network architecture
- 2. Batch size
- 3. Layer normalization
- 4. Data augmentation
- Cold initialization



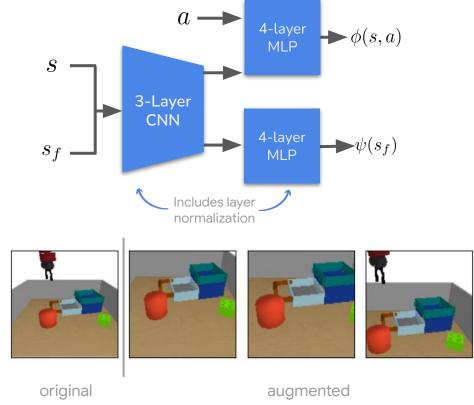
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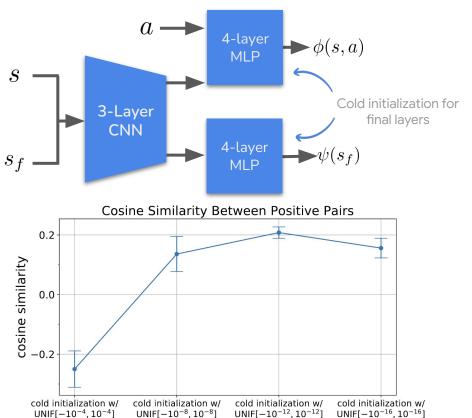
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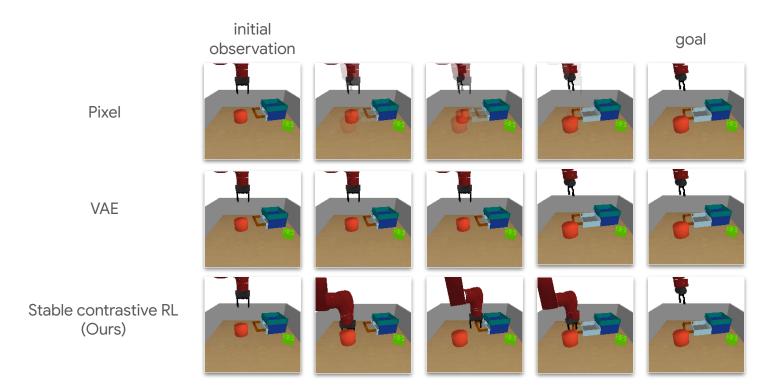
(ours)

Solving robotic tasks

pick and place spoon



Our learned representations capture both contents and causal relationships.



Future directions

Checkout the poster to learn more

- How about even larger size of dataset?
- How to mitigate overfitting for large scale vision backbones
- Is there any structure of the learned representation in theory?
- What about tasks beyond goal-reaching?

Video, code, and paper!



https://chongyi-zheng.github.io/stable_contrastive_rl