

Unsupervised Curricula for Visual Meta-Reinforcement Learning

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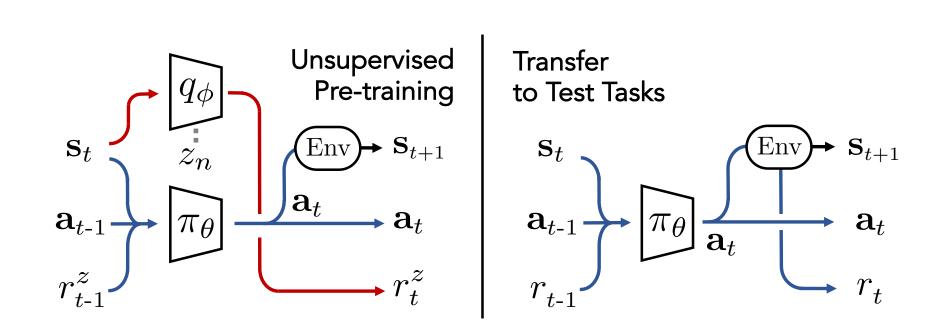
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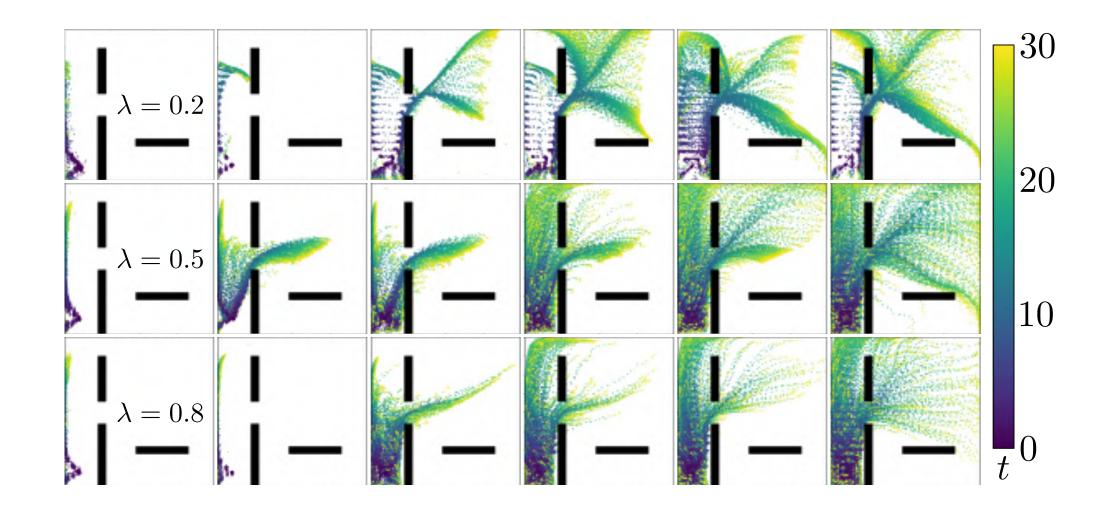


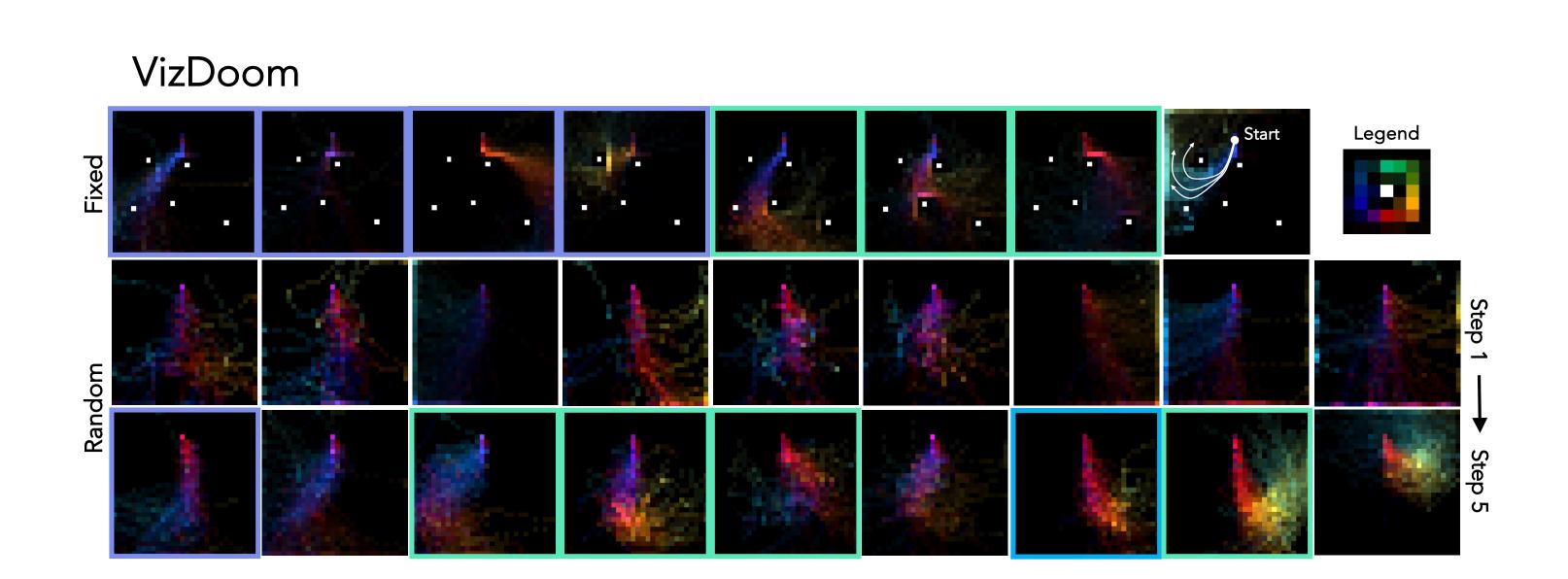
Motivation: Can useful meta-RL tasks be discovered in an unsupervised manner?



1. Organize
Update behavior model $q_{\phi}(\mathbf{s}) = \sum_{\mathbf{z}} q_{\phi}(\mathbf{s}|\mathbf{z})p(\mathbf{z})$ Tasks $\mathbf{z}. \text{ Meta-Train}$ Acquire skills and explore $r_{\mathbf{z}}(\mathbf{s}) = \lambda \log q_{\phi}(\mathbf{s}|\mathbf{z}) - \log q_{\phi}(\mathbf{s})$

Idea: Search for useful tasks via information maximization between a meta-learner and a self-generated task distribution.

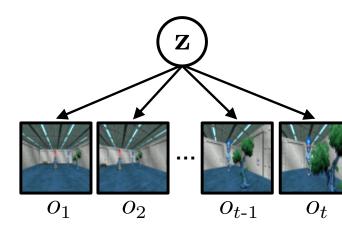




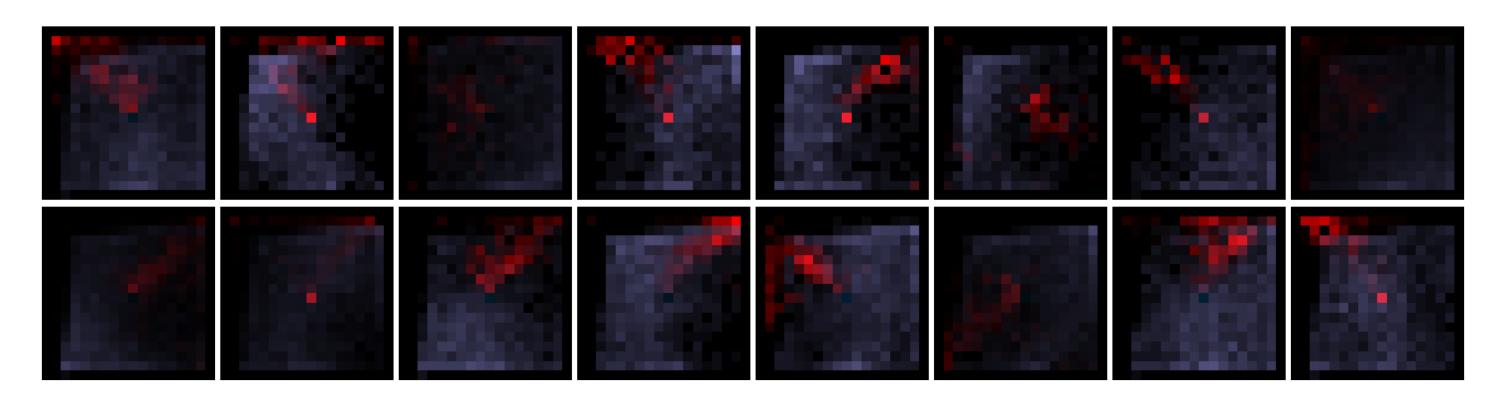
Key Challenges:

- 1. Mode-collapse of task distribution in high-dimensional observation spaces.
- 2. Joint task acquisition and meta-learning.
- 3. Structure and Diversity in task distribution.

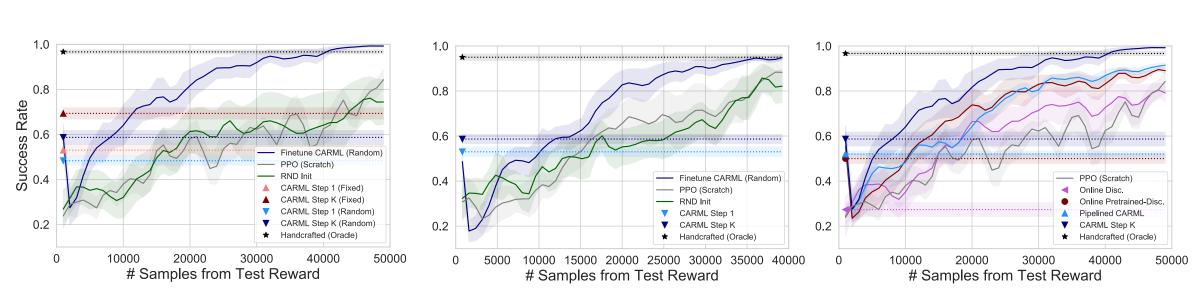
Simplifying Assumption. Conditional independence of observations along our trajectories



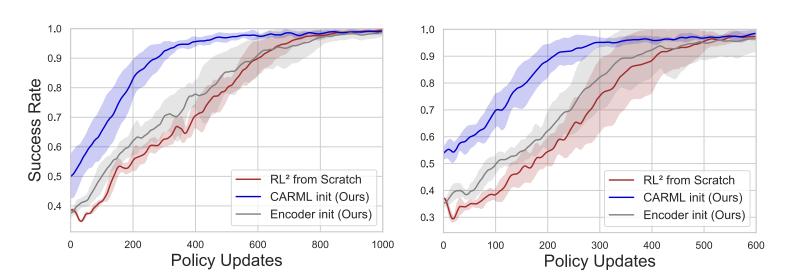




Experiments The acquired meta-RL strategies transfer to downstream test task distributions

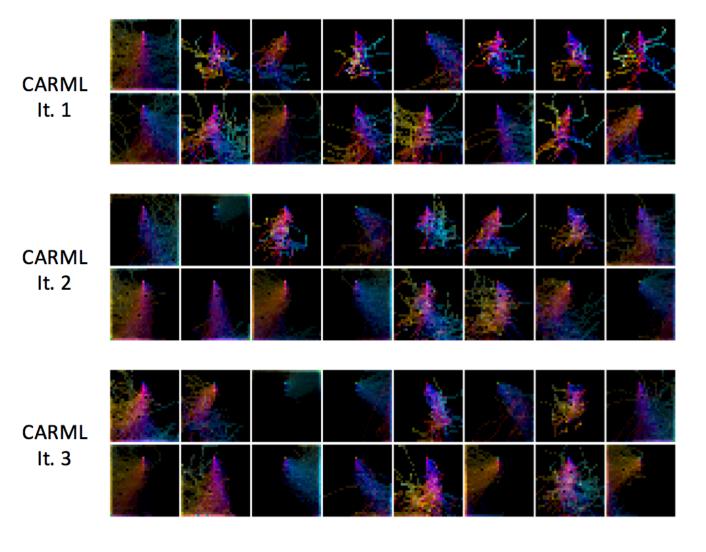


Direct Transfer and Task-Specific Finetuning



Meta-Pretraining: Accelerate Supervised Meta-RL

Evolution of the Task Distribution



Looking Forward

- 1. No free lunch -- Semi-supervised!
- 2. Non-visual task semantics
- 3. Convergence and Early Stopping
- 4. Forgetting