## Transfer Learning by Modeling A Distribution over Policies



McGill

Disha Shrivastava\*<sup>,1,2</sup>, Eeshan Gunesh Dhekane<sup>\*,1,2</sup>, Riashat Islam<sup>1,3</sup>

Mila<sup>1</sup>, Université de Montréal<sup>2</sup>, McGill University<sup>3</sup>

### Introduction

- VFunc: Deep Bayesian Learning framework [1]:
  - $\circ~$  Model joint distribution p(f,z) over function space using latent variable z
  - $\circ~$  Efficient function sampling by marginalizing ~z
  - Training entropy-regularized objective with variational lower bound
- Transfer Learning: Modeling a distribution over policies can lead to improved exploration in the target environment

#### Methodology

#### **Results: GridWorld**

• Transfer Results on GridWorld [3] [REINFORCE]



Transfer on Grid 2



 $H(f) \ge H(z) + \mathop{\mathbb{E}}_{(f,z) \sim p(f,z)} [\log q(z|f)] + H(f|z)$ 

\* Taken from [1]

• Schematics of Transfer Learning with VFunc







#### **Results: MiniGrid**

• Multi-Room Envs [2]:

**Static** 

Dynamic

Name	# Rooms	Room Size	Туре
N2S4	2	4	Source
N2S6	2	6	Target
N3S4	3	4	Target

• Transfer Results on MiniGrid [A2C]:



#### Transfer on N2S6 (Dynamic)



Sample Heatmaps for Transfer with VFunc on Grid 3

#### Conclusions

- Learned Target policy by pre-training with VFunc on source leads to faster convergence as compared to:
  - Pre-training with other PG algorithms on source
  - Training from scratch on target
- Difference more apparent with increasing difficulty level and environment dynamics
- Faster exploration in target task with diverse set of policies

# [1] Bachman, P., Islam, R., Sordoni, A., and Ahmed, Z. Vfunc: a deep generative model for functions. arXiv preprint arXiv:1807.04106, 2018. [2] Easy MDP: <a href="https://github.com/zafarali/emdp">https://github.com/zafarali/emdp</a> [3] Chevalier-Boisvert, M., Willems, L., and Pal, S. Minimalistic gridworld environment for openai gym, 2018. <a href="https://github.com/maximecb/gym-minigrid">https://github.com/maximecb/gym-minigrid</a>