Multi-task Reinforcement Learning with Soft Modularization

Ruihan Yang, Huazhe Xu, Yi Wu, Xiaolong Wang

project page: <u>https://rchalyang.github.io/SoftModule</u>





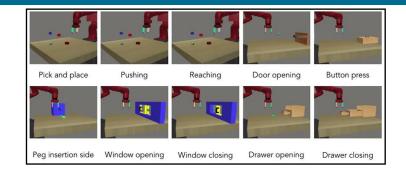


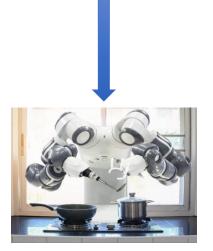
Motivation

Most RL: specialized

How? Multi-task RL

Goal: Apply to real-world

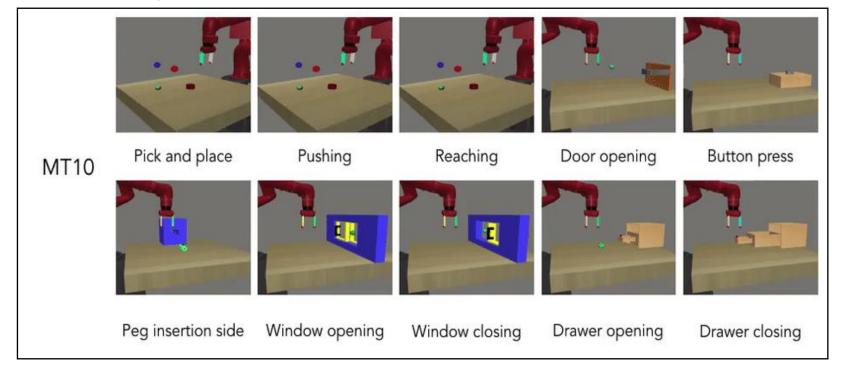




Picture source: https://robotage.guru/robot-chef-robotic-kitchen/

Current Multi-task RL Benchmark: Meta-World

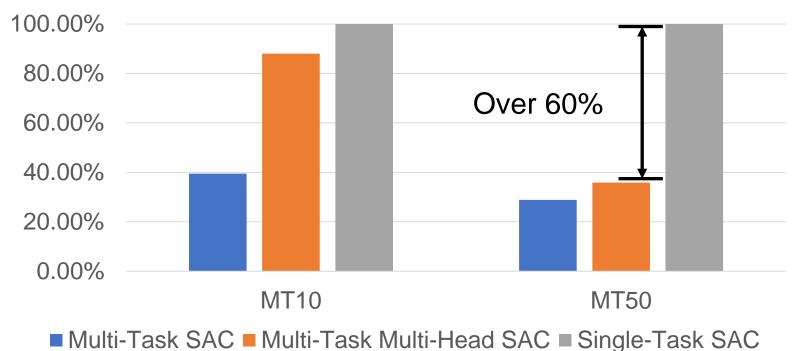
Containing dozens of robotics manipulation tasks.



Yu, T et al. Meta-World: A Benchmark and Evaluation for Multi-Task and Meta Reinforcement Learning, 2019

Meta-World

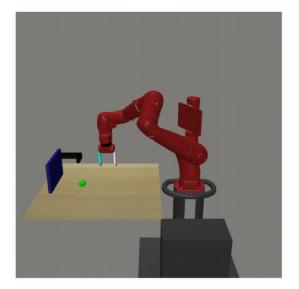
Success Rate

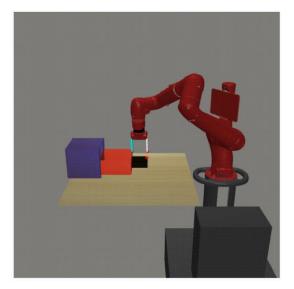


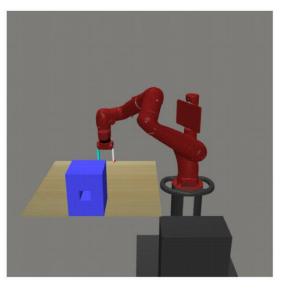
Haarnoja, T et al. Soft Actor-Critic: Off-Policy Maximum Entropy Deep Reinforcement Learning with a Stochastic Actor, 2018



Baseline failed to generalize to different tasks





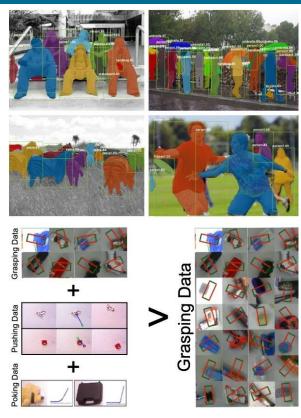




Close Drawer

Insert Peg

General Multi-Task Learning



Computer Vision: Detection + Segmentation

Robot Learning: Pushing + Grasping + Poking > Grasping

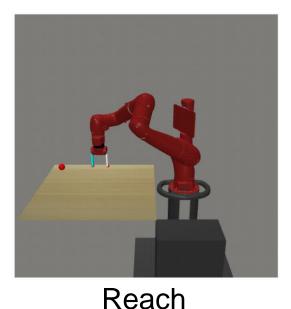
He, K et al. Mask R-CNN, 2017

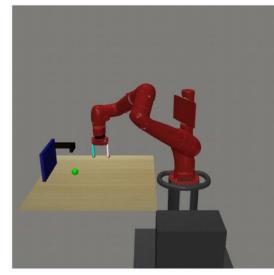
Pinto, L et al. Learning to Push by Grasping: Using multiple tasks for effective Learning, 2017

Two Challenges in Multi-Task Reinforcement Learning

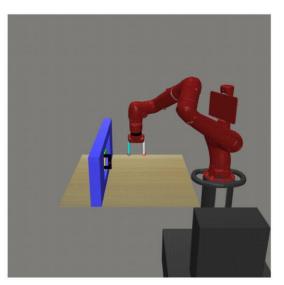
Avoid negatively interference between irrelevant tasks

Reuse shared components across similar tasks



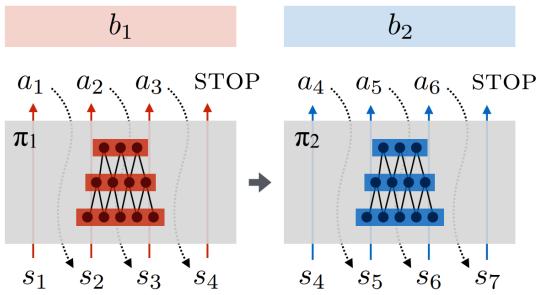






Open Window

Modularization



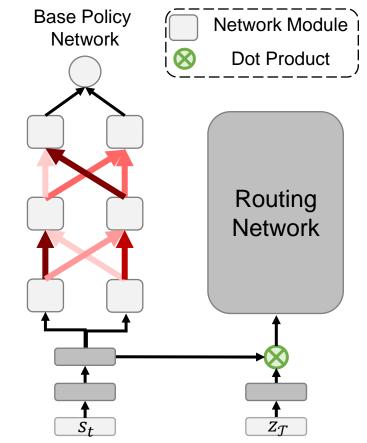
Previous Modular network for multi-task RL

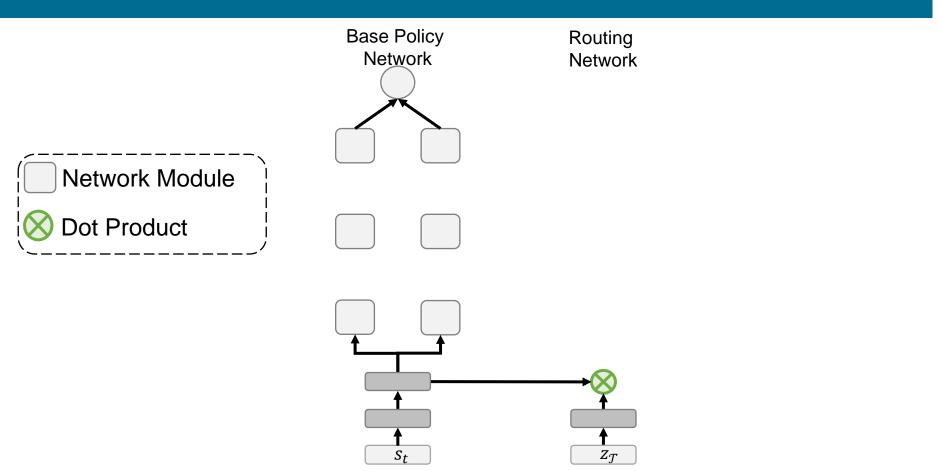
In hierarchical manner

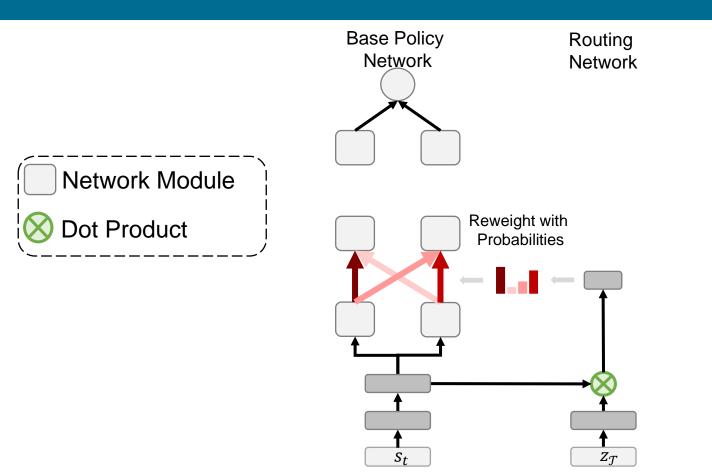
Andreas, J., Klein, D., and Levine, S. Modular multi-task reinforcement learning with policy sketches, 2017

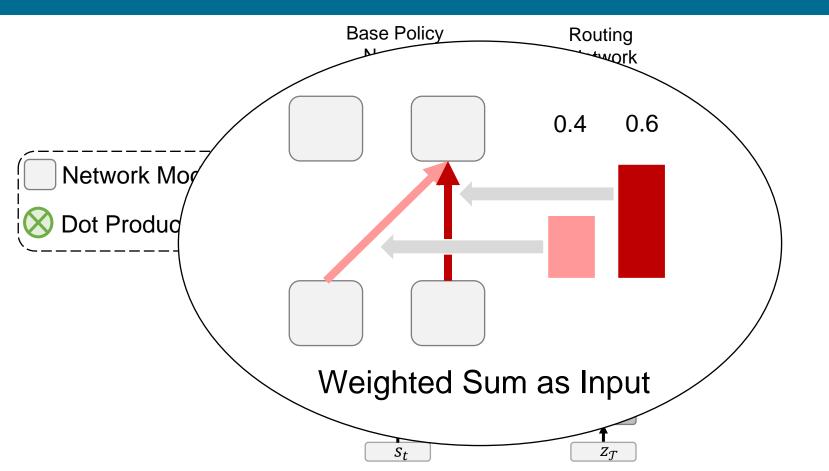
Base Policy Network + Routing Network

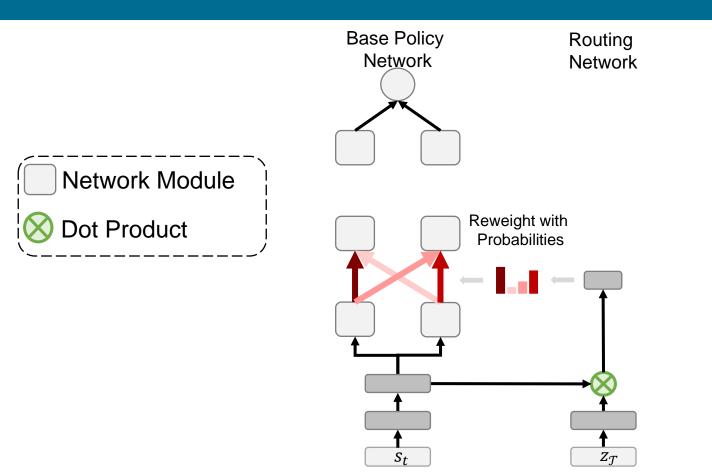
Differentiable

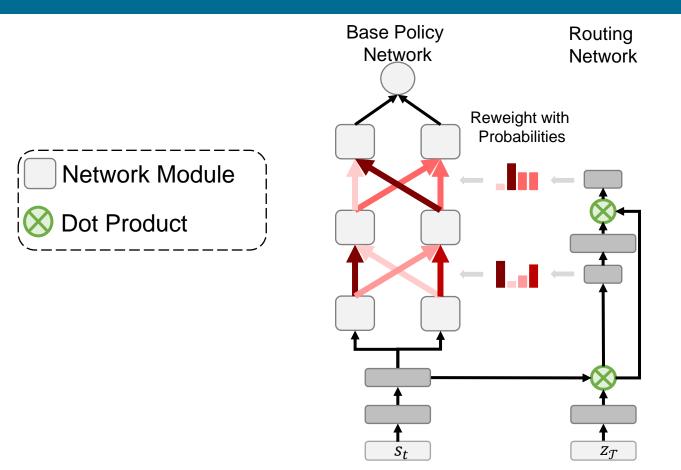


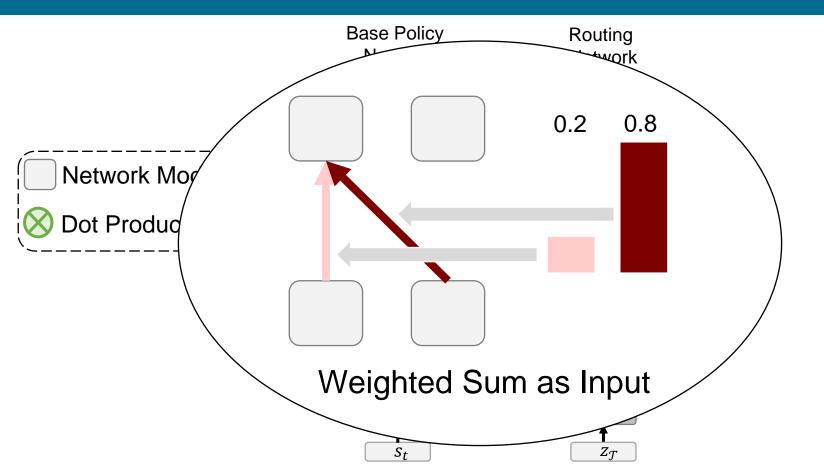


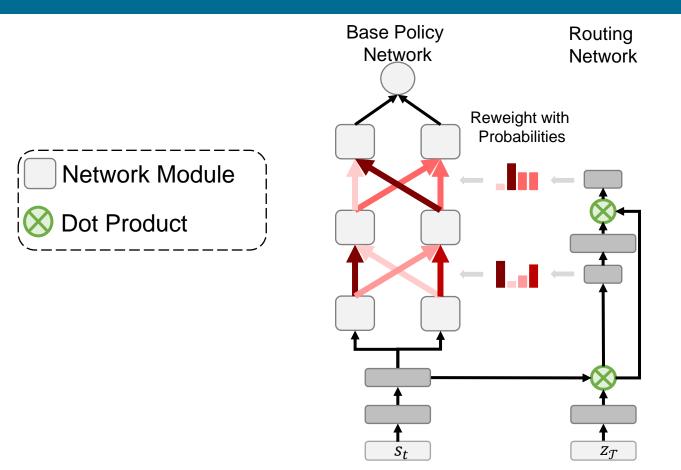












Temperature weight for multi-task RL:

$$J_{\pi}(\phi) = \mathbb{E}_{\mathcal{T} \sim p(\mathcal{T})}[w_{\mathcal{T}} \cdot J_{\pi,\mathcal{T}}(\phi)]$$
(1)

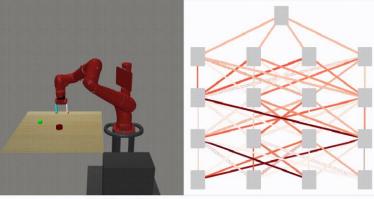
Temperature adjustment in SAC:

$$J(\alpha) = E_{a_t \sim \pi_{\phi}}[-\alpha \log \pi_{\phi}(a_t|s_t) - \alpha \overline{\mathcal{H}}]$$
(2)

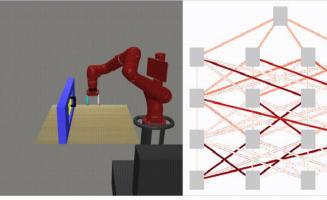
Temperature weight for multi-task RL:

$$w_i = \frac{\exp(-\alpha_i)}{\sum_{j=1}^{M} \exp(-\alpha_i)}$$
(3)

Push

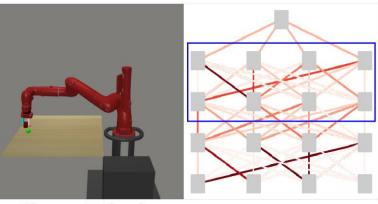


Close Window

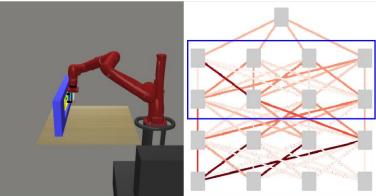


Module Sharing : Knowledge Sharing.

Push

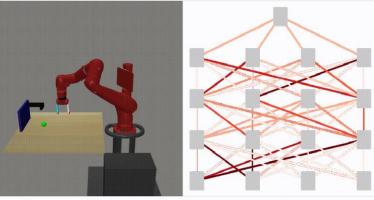


Close Window

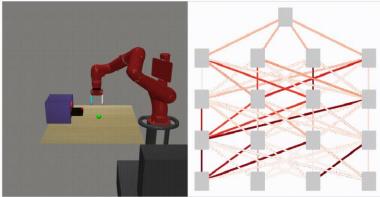


Module Sharing : Knowledge Sharing.

Open Door

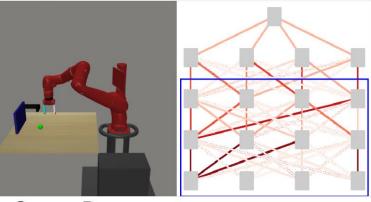


Open Drawer

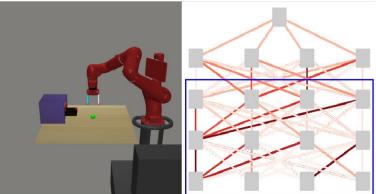


Module Sharing : Knowledge Sharing.

Open Door



Open Drawer

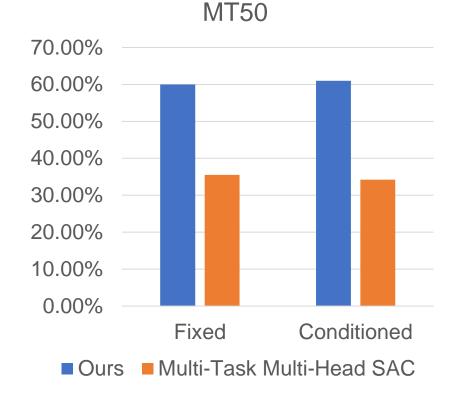


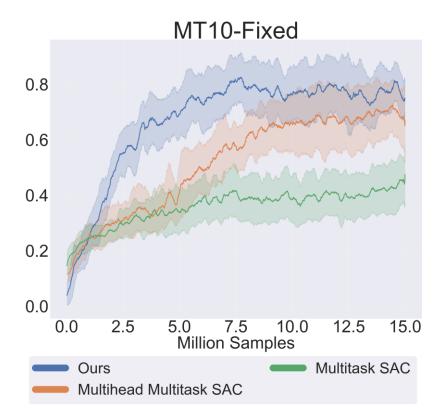
Original Meta-World: Fixed goal. MT10-Fixed / MT50-Fixed

More realistic and more challenging: Goal conditioned MT10-Conditioned / MT50-Conditioned

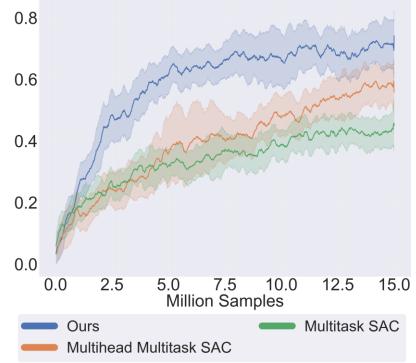
Experiments : Over All

MT10 100.00% 80.00% 60.00% 40.00% 20.00% 0.00% Fixed Conditioned Ours Multi-Task Multi-Head SAC

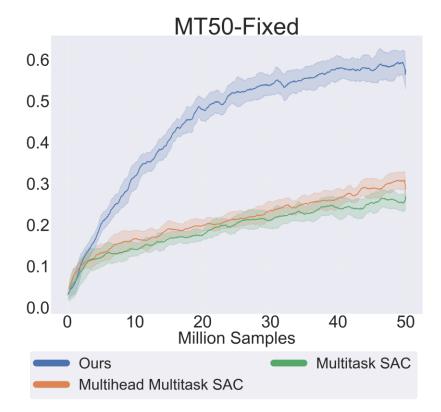




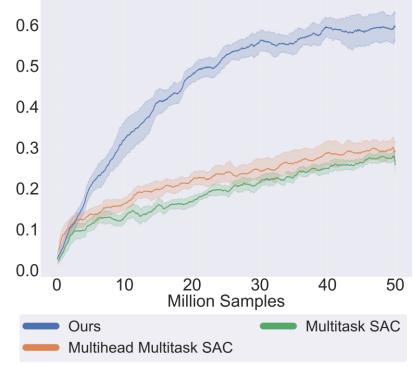
MT10-Conditioned



Experiments : MT50



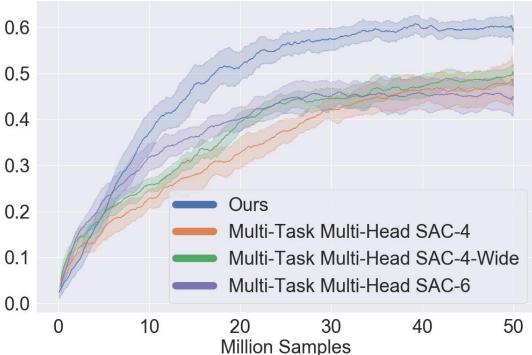
MT50-Conditioned



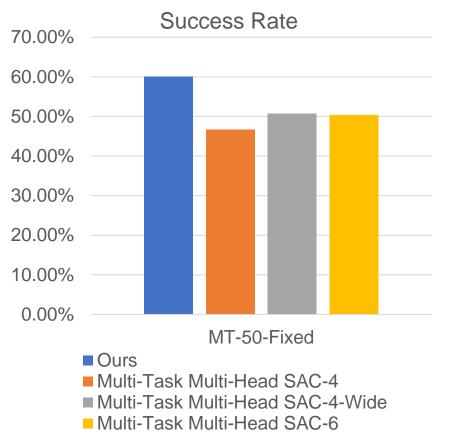
Effects on Network Capacity

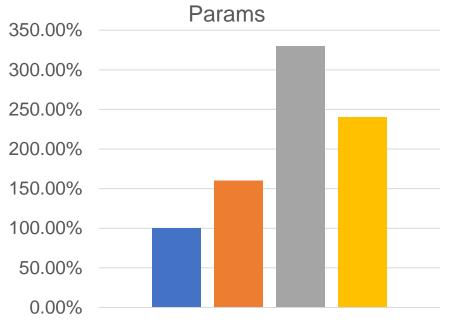
Performed on MT50-Fixed ^C

Larger network can not solve multi-task RL.



Effects on Network Capacity

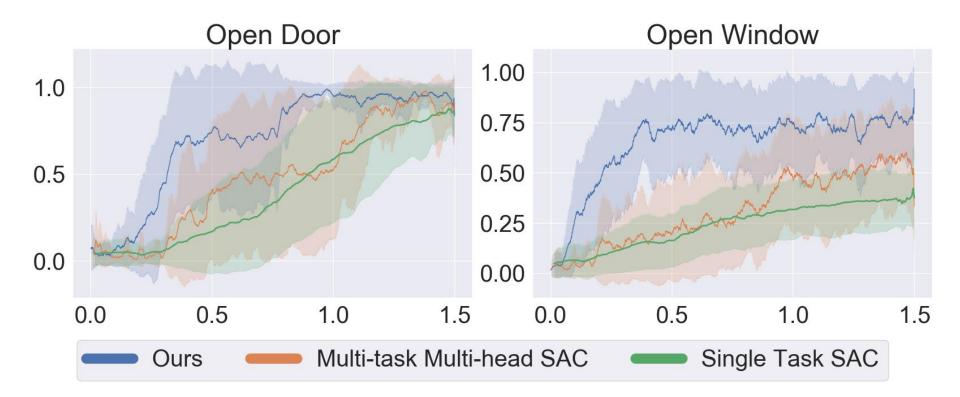




Ours

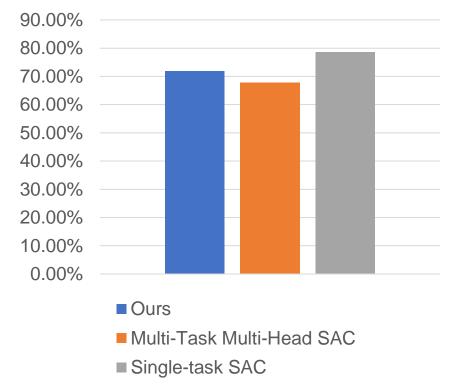
- Multi-Task Multi-Head SAC-4
- Multi-Task Multi-Head SAC-4-Wide
- Multi-Task Multi-Head SAC-6

Comparison with Single Task Policy



Comparison with Single Task Policy

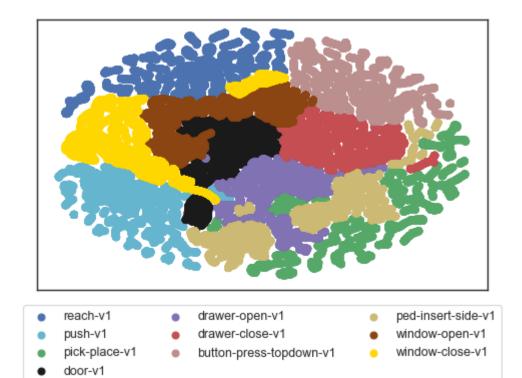
MT10-Conditioned



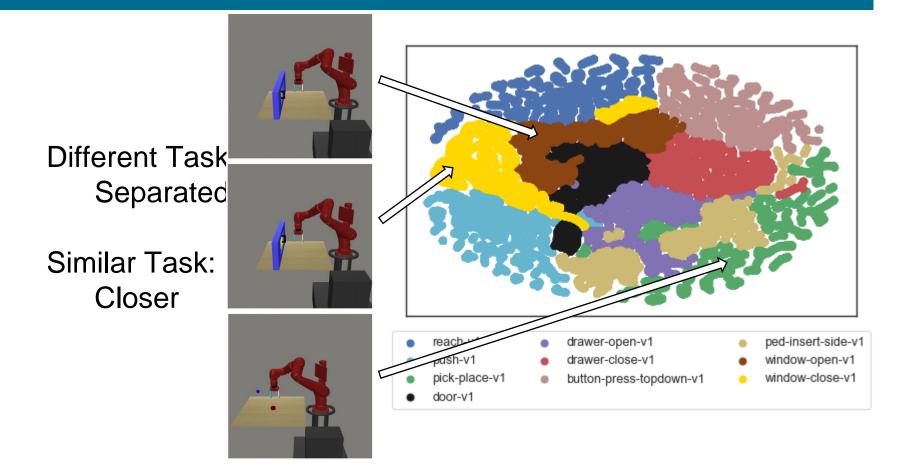
Routing Visualization

Different Task: Separated

Similar Task: Closer



Routing Visualization



Thanks!

Our project page: https://rchalyang.github.io/SoftModule