



# Learning Diverse Behaviors: Diffusion Meets RL

**Master Thesis** 



## Description

This thesis will tackle the challenge of multi-modality in robot learning by integrating diffusion models into reinforcement learning (RL). We will transfer selected tasks from the D3IL [1] benchmark to the Isaac Lab [2] simulation environment and apply state-of-the-art RL algorithms enhanced with diffusion [3] models. The goal is to better capture and generalize diverse behaviors in complex, multi-modal tasks. The research will involve adapting D3IL tasks to Isaac Lab, implementing diffusion-augmented RL models, and developing metrics to evaluate their performance. The outcome of this thesis would advance the capability of RL to handle multi-modal behavior in sophisticated simulation environments.

#### References:

Jia, Xiaogang, et al., "Towards diverse behaviors: A benchmark for imitation learning with human demonstrations." ICLR 2024.
Mittal, Mayank et al., "Orbit: A Unified Simulation Framework for Interactive Robot Learning Environments", IEEE RA-L 2023
Ho, Jonathan et al., "Denoising Diffusion Probabilistic Models", NeurIPS 2020

### Work packages

- Literature research on diffusion policies
- Designing simulation environments in Isaac Lab and generating data
- Training and improving diffusion policy architecture for multi-modality in interaction data
- Performing ablations and comparisons to alternate methods

### **Requirements**

- Highly motivated and autonomous student
- Experience with machine learning and controls
- Excellent knowledge of Python and PyTorch
- Experience with simulators and robot hardware is a bonus