Deep Reinforcement Learning for Robotic Grasping from Octrees Learning Manipulation from Compact 3D Observations

Andrej Orsula Robotics

Master's Thesis





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STUDENT REPORT

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Deep Reinforcement Learning for Robotic Grasping from Octrees

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Resumé

TODO: Resumé in Danish

Preface

TODO: Preface

Test citation (TODO: remove test citation): (Name, 2000)

Glossary

- **2D** Two-dimensional
- 3D Three-dimensional

1 Introduction

2 Related Work

- 2.1 Robotic Grasping
- 2.1.1 Empirical Approaches
- 2.1.2 Learning-Based Approaches
- 2.2 Learning from 3D
- 2.2.1 3D Data Representations

Mesh

Point Cloud

Voxel Grid

Octree

3 Background

- 3.1.1 Markov Chain
- 3.1.2 Markov Decision Process
- 3.1.3 Q-Learning
- 3.1.4 Value-Based Reinforcement Learning
- 3.1.5 Policy-Based Reinforcement Learning
- 3.2 Actor Critic
- 3.2.1 Deep Deterministic Policy Gradient (DDPG)
- 3.2.2 Twin Delayed Deep Deterministic (TD3)
- 3.2.3 Soft Actor Critic (SAC)
- 3.2.4 Truncated Quantile Critics (TQC)
- 3.3 Function Approximation
- 3.3.1 Neural Networks

4 Problem Formulation

- 4.1 Observation Space
- 4.1.1 Observation Stacking
- 4.2 Action Space
- 4.3 Reward Function

5 Methods

- 5.1 Curriculum Learning
- 5.2 Demonstration Bootstrapping
- 5.3 Domain Randomization

6 Implementation

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6.1	Simi	IIATIAN	$ \mathbf{n}$	onment
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6.1.1 Selection

Mu.JoCo

PyBullet

Gazebo Classic

Ignition Gazebo

6.1.2 Simulating with Ignition Gazebo

Controller

Middleware - ROS 2

Motion Planning - MoveIt 2

6.2 OpenAl Gym Environment

- 6.2.1 Gym-Ignition
- 6.3 Stable Baselines3
- 6.4 Network Architecture
- 6.4.1 PyTorch
- 6.4.2 Feature Extractor
- 6.4.3 Actor Critic Networks
- 6.5 Hyperparameter Optimisation with Optuna

7 Experimental Evaluation

- 7.1 Experimental Setup
- 7.2 Results
- 7.3 Ablation Studies

8 Discussion

9 Conclusion

10 Future Work

Bibliography

Name, R. (2000). Fascinating title. Journal of Testing 420(69), 69-420.

Appendices

- **A** Low-Level Controller
- **B** Dataset
- **C** Hyperparameters
- **D** Full Results