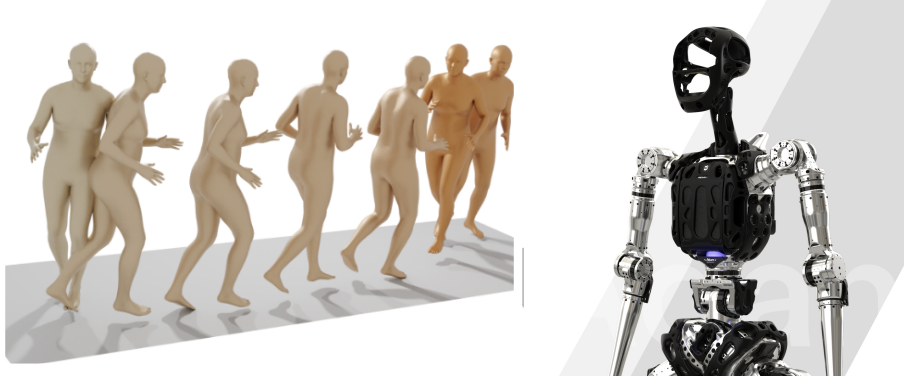


**Master Thesis****Motion Imitation and Control for Humanoid Robot using Diffusion****Introduction**

To achieve effective and natural interaction, humanoid robots may need to closely imitate human motion, which encompasses walking, object manipulation, and environmental interactions. Motion capture data [1], which captures human motion with high precision, serves as an excellent resource for training robotic systems to replicate human movements. Diffusion models [2] are a class of generative models designed to handle multi-modal distributions, making them highly suitable for complex motion generation tasks. Recent state-of-the-art methods use diffusion to produce human like motions for character animation [3] or to imitate human expert data for controlling robotic arms [4].

The goal of this project is to explore diffusion approaches for imitating motion data from humans to obtain control policies for humanoid robots.

**Task Description**

The student will:

- Conduct a thorough literature review. Particularly, study prior work on RL, motion imitation, Diffusion [5, 3, 4].
- Leverage physics-based simulators such as Isaac Gym[6] for training.
- Use motion capture data as reference for imitation.
- Possibly deploy trained policies on humanoid hardware platforms.
- Write a report and give an oral presentation at the end of the project.

**Skills**

- Very good programming skills in Python, and familiarity with Pytorch.
- Experience in Machine Learning and RL.
- Experience with physics-engines or motion capture data is plus.
- Willingness to work on cutting edge methods and algorithms.

**Remarks**

This thesis is overseen by Prof. Dr. Stelian Coros.

## Contact

For further information or application for the thesis project, please contact Fatemeh Zargarbashi ([fatemeh.zargarbashi@inf.ethz.ch](mailto:fatemeh.zargarbashi@inf.ethz.ch)) and submit a copy of your CV and your transcripts.

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