# **Offset-free learning of predictive controllers**



### **Master's Thesis**

Model Predictive Control (MPC) is an advanced control method that can naturally handle nonlinear systems subject to constraints. However, for MPC, we need to repeatedly solve a complex nonlinear programming problem (NLP) online. Consequently, MPC is often out of reach for real-time application. One approach to counter the computational demand is to use supervised learning methods to approximate the MPC law [2]. However, approximation errors can often cause a steady state offset when employing the learned controller [1, 3].

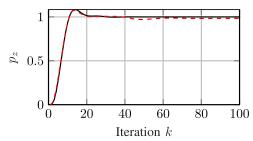
The goal of this thesis is to develop methods to learn an offset-free predictive controller, while maintaining low computational complexity.

#### **Requirements:**

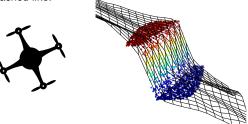
- Critical and creative thinking
- Programming skills in Matlab or Python
- Model predictive control and machine learning

## Tasks:

- Literature review on offset-free MPC.
- Propose methods for offset-free approximation of MPC.
- Illustrate the proposed methods through case studies.
- Evaluate, compare and document the results.



**Figure**: The quadrocopter controlled by the predictive controller flies from the ground position to a height of 1 meter as outlined with the black line. The learned controller never steers the drone to a height of 1 meter but stops with a small offset as indicated with the red dashed line.



#### **Resources:**

 Kimberly J. Chan, Deep Learning-based Approximate Nonlinear Model Predictive Control with Offset-free Tracking for Embedded Applications.
A. Rose. Learning a Gaussian Process Approximation of a Model Predictive Controller with Guarantees.
G. Pannocchia, Offset-free MPC explained: novelties, subtelties, and applications.

#### M. Sc. Alexander Rose

E-Mail: alexander.rose@iat.tu-darmstadt.de Web: https://www.ccps.tu-darmstadt.de

