NONLINEAR MODELING WITH LEARNED PARAMETER REFINEMENTS FOR NMPC ON A REAL-WORLD AERODYNAMIC SYSTEM Jonas Schlagenhauf, Tobias Schöls, Moritz Diehl Systems Control and Optimization Laboratory - IMTEK - University of Freiburg

Summary

- ► Comparison of classical and machine-learning modeling methods
- ► Design of a nonlinear model predictive controller
- ► Hard real-time implementation in C++



► Evaluation on a real-world AWE prototype

Modeling

► Classic Approach: Quadratic Model & LLS Fitting

 $\dot{x}_{i} = x^{\top} A_{i}^{\text{quad}} x + A_{i}^{\text{lin}} x + B_{i} u + C_{i}$ **linear**: $A_{i}^{\text{quad}} = 0, C_{i} = 0$ **lasso**, **elastic** and **huber** use respective **affine**: $A_{i}^{\text{quad}} = 0$ regularizer with learned weighing parameters

► Novel approach:

- -approximate system dynamics via neural network (NN)
- -nonlinear state augmentation (quadratic and trigonometric)



Real-world setup

- ► Reduced-DOF plane for rotational start experiments
- ► Custom hardware designed in-house for sensor/actuator IO
- ► Fully equipped for lift-mode energy production

Conclusion

Control

- ► Nonlinear Model Predictive Control (NMPC)
- ► Direct Multiple Shooting using explicit RK4
- ► Gauss-Newton Hessian approximation + condensing
- ► Constrained control surface position and speed
- \blacktriangleright C++ Code Generation from MATLAB via ACADO / QPOASES
- ► 4 states, 1 control, 100 steps, @20Hz, 5 sec Horizon

- \blacktriangleright Robust results with linear regression fit
- ► Quadratic elements can reduce error, but are easy to overfit
- ► ML refinements achieve further improvement reduce overfitting
- Neural networks achieve good short time prediction and are suitable for integration in NMPC
- Evaluation on real-world setup shows adequate performance of machine learning approaches in a real-time control setup
- Future research aims to apply the presented approaches to tethered flight operation

Experimental Results





