
Introduction to Aerial Robotics

A Robust Control Approach

Jiayao Wang and Vinzent Rudolf

Institute for Systems Theory and Automatic Control
University of Stuttgart



- **Introduction and Theoretical Background**
- *System Dynamics*
 - Approximated System
 - Actuator Uncertainties
 - Nonlinear Terms
 - Noisy Measurements
 - External Inputs
 - Position Controller for x and y
- *Results*
 - Controller Design
 - Robust Stability and Robust Performance
 - Simulation Study
- Conclusion



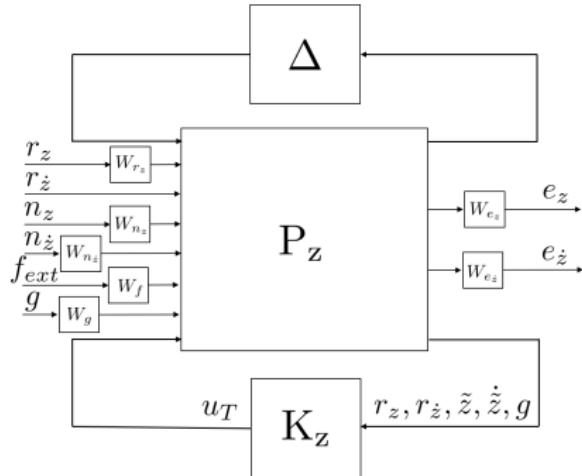
Project Target

apply *Robust Control Theory* to a *Multirotor Aerial Vehicle* (MAV), so that the control loop is robust stable and still achieves good performance regarding

- model uncertainties
- measurement noise
- external disturbances



Robust Control Theory is well known for linear time invariant systems





$$P \star K = N = \begin{pmatrix} M & N_{12} \\ N_{21} & N_{22} \end{pmatrix}$$

$$\Delta \star N = N_{22} + N_{21}\Delta(I - M\Delta)^{-1}N_{12}$$

starting from this representation, following theorem can be proved

Theorem

Suppose M is a proper and stable transfer matrix. If

$$\det(I - M(i\omega)\Delta) \neq 0 \quad \forall \Delta \in \Delta, \omega \in \mathbb{R} \cup \{\infty\}$$

then $I - M\Delta$ has a proper and stable inverse for all $\Delta \in \Delta$ [2].

major advantage of this theorem:

- suffices to check imaginary axes for stability and not the full right-half plane!
- can be solved very efficiently
- able to take the structure of Δ into account



- famous *Structured Singular Value* (SSV) is a simple test for theorem 1, which evaluates M and Δ at specific frequencies
- fortunately, one can calculate a *lower* and *upper* bound for the SSV pretty efficiently [2]
- lower bound indicates robust stability for the closed-loop for all $\|\Delta\|_\infty < \frac{1}{\gamma_l}$
- upper bound implies that there exists a destabilizing uncertainty with $\|\Delta\|_\infty < \frac{1}{\gamma_u}$ [3]
- uncertainty weights W_Δ are used to normalize the input and output of Δ
- an additional SSV test can be used to check robust performance, this is known as the *Main Loop Theorem* [2]



nominal controller for P is minimizing the H_∞ -norm of $P \star K$

$$\begin{aligned} & \min_K \|P \star K\|_\infty \\ & \text{s.t. } K \text{ stabilizes } P \end{aligned} \tag{1}$$

- performance weights W_P are used to weight frequencies differently
- H_∞ problem does not take uncertainties into account, hence so called *DK-Synthesis* is used
- approach uses so called *D-Scalings* in the controller design, which are used to compute better upper bounds of the SSV test



- Introduction and Theoretical Background
- **System Dynamics**
 - Approximated System
 - Actuator Uncertainties
 - Nonlinear Terms
 - Noisy Measurements
 - External Inputs
 - Position Controller for x and y
- *Results*
 - Controller Design
 - Robust Stability and Robust Performance
 - Simulation Study
- Conclusion



System Dynamics

dynamics of a MAV are given by [5]

$$u = \begin{pmatrix} u_T & \tau_x & \tau_y & \tau_z \end{pmatrix}^T \in \mathbb{R}^4$$

$$\dot{x} = \begin{pmatrix} \dot{p} \\ \ddot{p} \\ \dot{\eta} \\ \dot{\omega} \end{pmatrix} = \begin{pmatrix} \dot{p} \\ ge_z \\ T(\eta)\omega \\ -J_B^{-1}[\omega] \wedge J_B\omega \end{pmatrix} + \begin{pmatrix} 0 \\ -\frac{1}{m}R(\eta)e_z u_T \\ 0 \\ J_B^{-1}\tau \end{pmatrix} \in \mathbb{R}^{12} \quad (2)$$

⇒ dynamics are nonlinear. Therefore, the problem to design a robust controller for (2) is splitted up into different sub tasks

- ① decouple dynamics of (2) into altitude and attitude dynamics
- ② design robust controller for altitude and attitude independently from each other
- ③ design cascaded position controller for x and y

System Dynamics - Approximated System



$$\begin{aligned}\dot{x} &= \begin{pmatrix} \dot{p}_z \\ \ddot{p}_z \\ \dot{\eta} \\ \dot{\omega} \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \Delta_{T(\eta)} \\ 0 & 0 & 0 & -J_B^{-1} \Delta_{[\omega] \wedge J_B} \end{pmatrix} \begin{pmatrix} p_z \\ \dot{p}_z \\ \eta \\ \omega \end{pmatrix} \\ &+ \begin{pmatrix} 0 & 0 & 0 & 0 \\ -\frac{1}{m}(1 + W_m \Delta_m)(1 + W_{u_T} \Delta_{u_T}) & -\frac{1}{m} & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & J_B^{-1}(1 + W_\tau \Delta_\tau) & J_B^{-1} \end{pmatrix} \begin{pmatrix} u_T \\ f_{Ext} \\ g \\ \tau \\ \tau_{Ext} \end{pmatrix} \\ \tilde{x} &= \begin{pmatrix} \tilde{p}_z \\ \dot{\tilde{p}}_z \\ \tilde{\eta} \\ \tilde{\omega} \end{pmatrix} = \begin{pmatrix} p_z \\ \dot{p}_z \\ \eta \\ \omega \end{pmatrix} + \begin{pmatrix} n_z \\ n_{\dot{z}} \\ n_\eta \\ n_\omega \end{pmatrix}\end{aligned}$$

System Dynamics - Approximated System



approximated system contains following simplifications and extensions:

- ① difference between commanded torques / thrusts and applied torques / thrusts is covered by dynamic and parametric uncertainties
- ② uncertainties for nonlinear terms are introduced
- ③ all measurements are noisy
- ④ external torques and external forces can also occur as inputs
- ⑤ gravity vector g is used as additional input to the system

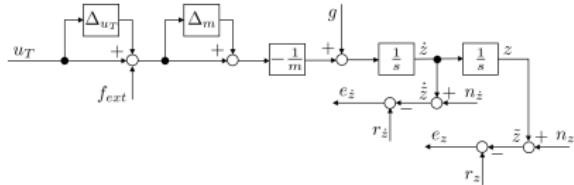


Fig.: MAV altitude dynamic model with uncertainties.

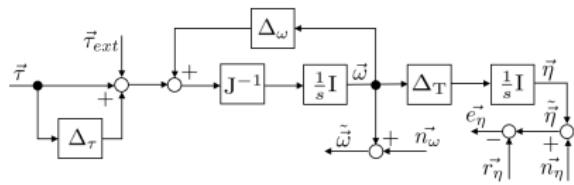


Fig.: MAV rotational dynamic model with uncertainties.



Mass Uncertainty

- altitude dynamics consist of a real parametric uncertainty Δ_m , which describes a non-dynamic relative mass deviation $\frac{\hat{m} - m}{\hat{m}}$

$$\frac{1}{\hat{m}} = \frac{1}{m}(1 + W_m \Delta_m) \Leftrightarrow W_m \Delta_m = \frac{\hat{m} - m}{\hat{m}}$$

- it is assumed that the nominal mass m does not differ more than 10% from the real (actual / uncertain) mass $\hat{m} \Rightarrow W_m = 0.1$



Motor Uncertainty

- direct feedthrough of the thrust u_T and torques τ are assumed
- in practice, the desired force can not be applied exactly and motors have internal dynamics

$$G = \frac{1+k}{Ts+1} = 1 + W_{u_T} \Delta_{u_T} \Leftrightarrow W_{u_T} \Delta_{u_T} = \frac{k-Ts}{Ts+1}$$

- internal dynamics are designed as PT1 filter with a maximum time constant T
- gain k describes a fixed relation between desired and actual force, which can be compared to a tilting of the characteristic curve of the motor
- thrust: $T = 10\text{ms}$, $|k| < 0.5$
- torques: $T = 40\text{ms}$, because multiple motors have to create thrust, so that a torque around the body axes is applied; $|k| < 0.5$



System Dynamics - Nonlinear Terms

- Robust Control Theory can only be applied to linear systems, the system dynamics (2) have to be brought into a linear form
- nonlinear terms are $R(\eta)$, $T(\eta)$ and $[\omega] \wedge J_B \omega$
- in order to not simply ignore these terms, they are introduced as uncertain systems into the uncertain system dynamics

$R(\eta)$

- no additional uncertainty object is introduced
- uncertainty of this term is already implicitly captured with Δ_m and Δ_{u_T}



System Dynamics - Nonlinear Terms

$T(\eta)$

- $T(\eta)$ is approximated by $\Delta_{T(\eta)}$.
- therefore, $T(\eta)$ is linearized around $(\phi \quad \theta) = (0 \quad 0)$
- new linearized variables $\Delta\phi$ and $\Delta\theta$ are used as dynamic uncertainties
- maximum roll and pitch angle of $|\phi|, |\theta| < 10^\circ$ is assumed, which is used as output weight W_η .

$$\begin{aligned}\Delta_{T(\eta)} &= T(\eta) \Big|_{\phi,\theta=0} + \frac{\partial T(\eta)}{\partial(\phi, \theta)} \Big|_{\phi,\theta=0} \begin{pmatrix} W_\eta \Delta_\phi \\ W_\eta \Delta_\theta \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & W_\eta \Delta_\theta \\ 0 & 1 & -W_\eta \Delta_\phi \\ 0 & W_\eta \Delta_\phi & 1 \end{pmatrix} \quad (3)\end{aligned}$$



System Dynamics - Nonlinear Terms

$[\omega] \wedge J_B \omega$

- inertia matrix J_B is assumed to have only entries on the main diagonal, whereby $J_{xx} \equiv J_{yy}$

$$\begin{aligned} [\omega] \wedge J_B \omega &= \begin{pmatrix} (J_{zz} - J_{yy})\omega_z \omega_y \\ (J_{xx} - J_{zz})\omega_z \omega_x \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 0 & (J_{zz} - J_{yy})W_{\omega_z}\Delta_{\omega_z} & 0 \\ (J_{xx} - J_{zz})W_{\omega_z}\Delta_{\omega_z} & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \omega \quad (4) \\ &= \Delta_{[\omega] \wedge J_B \omega} \end{aligned}$$

- nonlinear term can be rearranged, so that the inner matrix only depends on ω_z , which is exchanged with a dynamic uncertainty Δ_{ω_z}
- output weight $W_{\omega_z} = \frac{2}{s+1}$: 2 is the assumed max. body rate $|\omega_z|$, PT1 filter describes correlation between ω_z and $\omega_{x,y}$: two body rates do not change simultaneously very fast





Time Variance of Uncertainties

- dynamic uncertainties of the nonlinear terms are very unspecific, so that nearly any possible (stable) transfer function could be obtained by them
- approach has the advantage that all possible situations are covered under the assumption that the maximum values fit
- however, these uncertainties are in general time variant, because the exact transfer function depends on the current state vector x , which is formally not covered in Robust Control Theory
- nevertheless, **simulation shows satisfactory results**



System Dynamics - Noisy Measurements

- to create a system simulation close to real world applications noise is added to the feedback channels of p_z , \dot{p}_z , Euler-angles η and body rates ω
- taken into account in the controller design by introducing additional performance channels from input noise to controller error
- accuracy values for p_z , \dot{p}_z and the heading are taken from the datasheet of a GNSS receiver [4]
- due to missing data, the accuracy of ϕ and θ is estimated
- variance of the normal distributed noise n_ω is taken from the IMU datasheet [1]



System Dynamics - External Inputs

- two additional input channels are added into the system structure to take external forces and external torques like those created by wind, into account
- for simulation purposes following external inputs are assumed

$$f_{Ext} = 1 \cdot \sin(1t), \quad \tau_{Ext} = 1 \cdot \sin(0.1t)$$



- it is assumed, that the MAV is in a state in which almost all Euler-angles are close to zero, hence altitude and the rotation can be controlled separately
- position controller for x and y is implemented in the same way as for the near hovering controller
- unfortunately, it is not possible to apply the robust control theory to this structure
- nevertheless, possible extension is proposed in our paper to be able to apply robust control theory also to position controller



Content

- Introduction and Theoretical Background
- *System Dynamics*
 - Approximated System
 - Actuator Uncertainties
 - Nonlinear Terms
 - Noisy Measurements
 - External Inputs
 - Position Controller for x and y
- **Results**
 - Controller Design
 - Robust Stability and Robust Performance
 - Simulation Study
- Conclusion

Controller Design

final robust controllers are each a LTI system with following properties:

- ① The altitude controller consists of 16 states, 5 inputs ($g, p_{des_z}, \dot{p}_{des_z}, \tilde{p}_z, \dot{\tilde{p}}_z$) and 1 output (thrust u_T).
- ② The attitude controller has got 35 states, 9 inputs ($r_\eta, \tilde{\eta}, \tilde{\omega}$) and 3 outputs (torques τ)

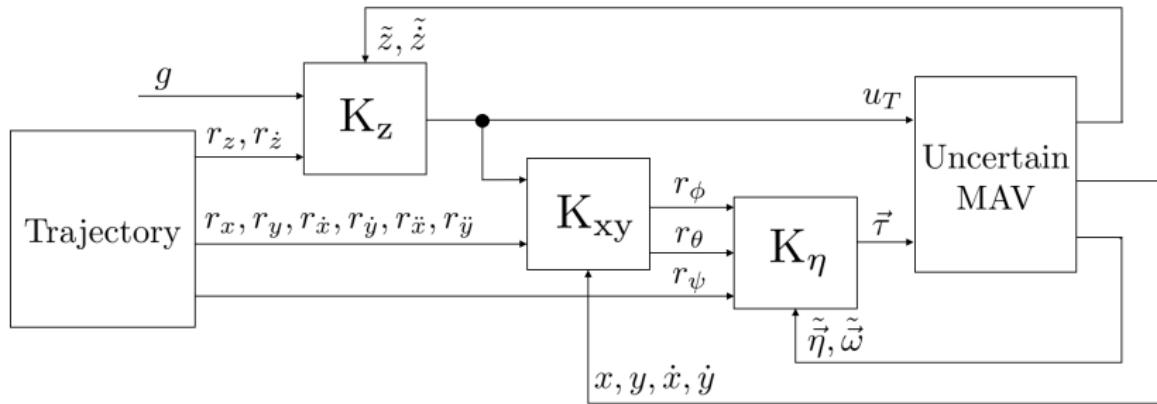


Fig.: Sketch of the robust control loop

Controller Design - General Framework

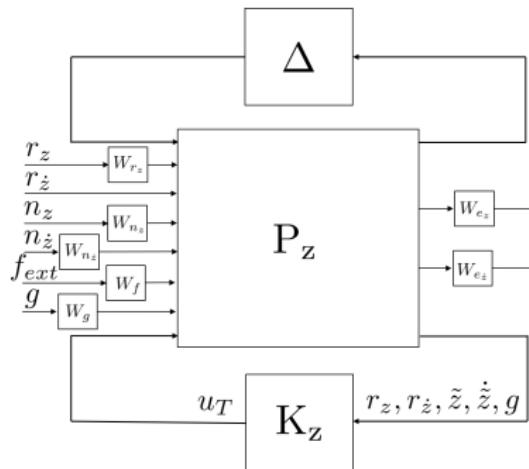


Fig.: Δ - P - K structure for altitude control

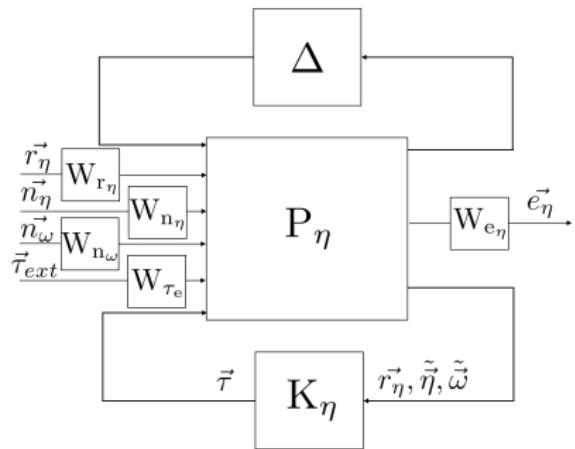


Fig.: Δ - P - K structure for attitude control



Results - Robust Performance

Robust Performance

- during weight design it turns out, that the *dksyn* command is able to compute better controllers, if different performance weights than the desired dynamic behavior are used
- performance has only been analyzed during simulation studies and using the bode plot



Results - Robust Performance

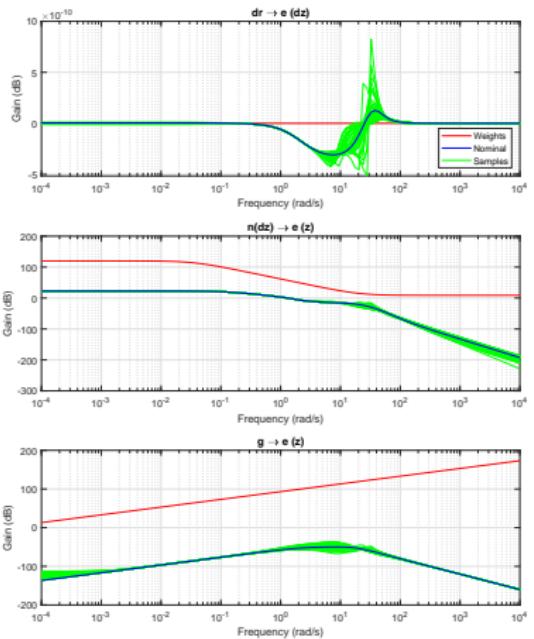
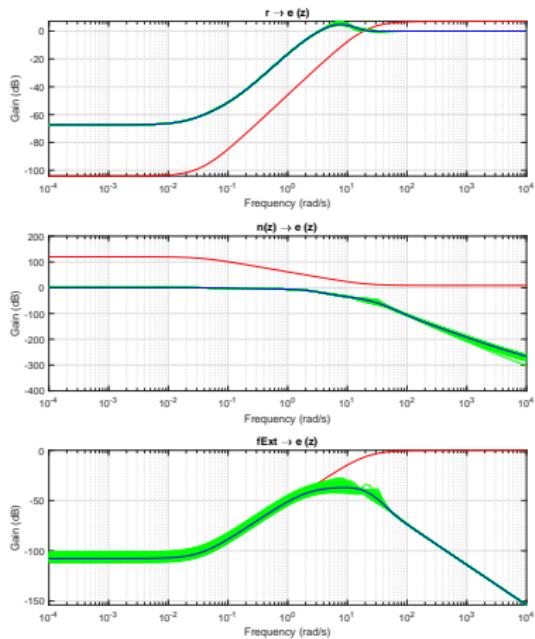


Fig.: Bode Plot of Altitude Controller

Results - Robust Performance

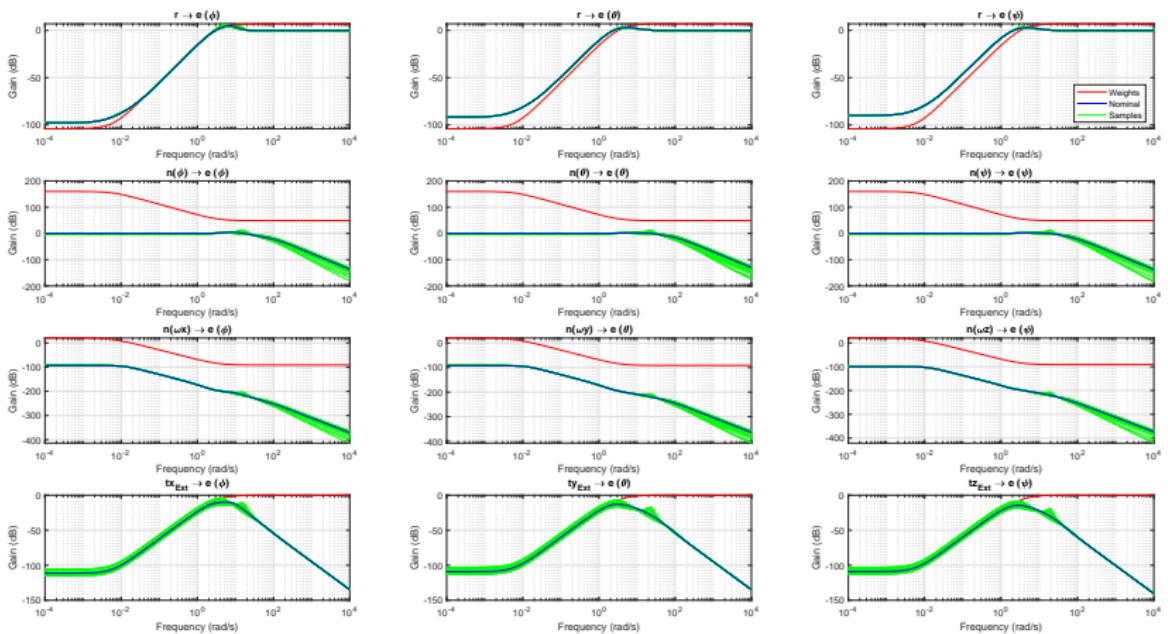


Fig.: Bode Plot of Attitude Controller



Results - Robust Stability

Robust Stability

- robust stability does not depend on the weights of the performance channels
- stability of linear systems does not depend on the inputs (in contrast to nonlinear systems)
- Therefore, it is possible to give stability margins $r = \frac{1}{\gamma}$

	lower bound	upper bound
Altitude Controller	0.97966	0.98165
Attitude Controller	1.00184	1.00387

Fig.: Robust Stability Margins



Results - Robust Stability

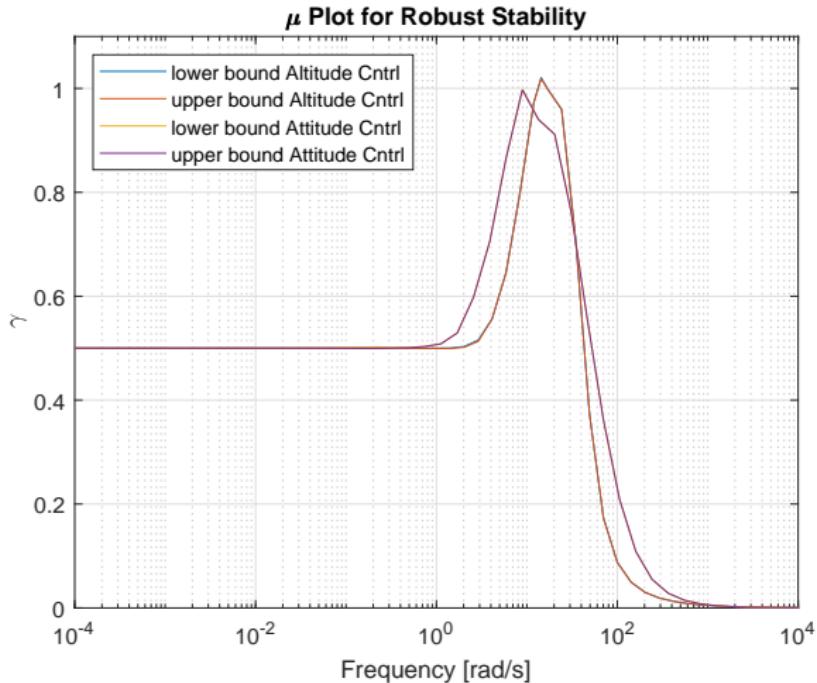


Fig.: μ Plot for Robust Stability



Results - Simulation Study

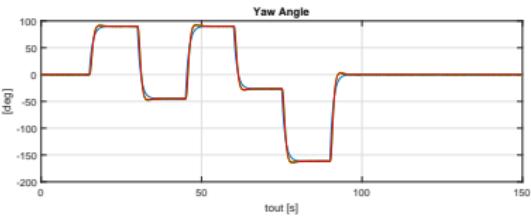
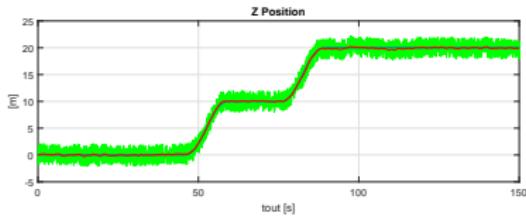
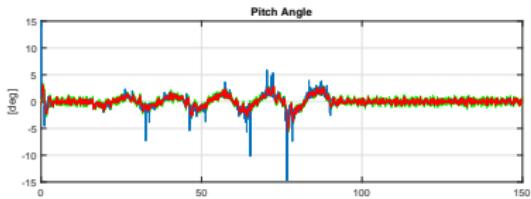
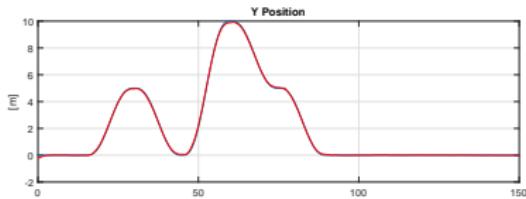
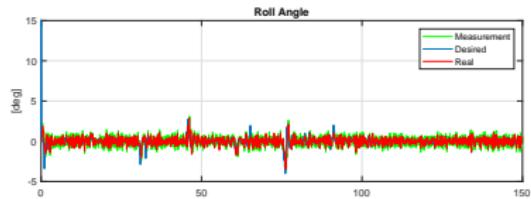
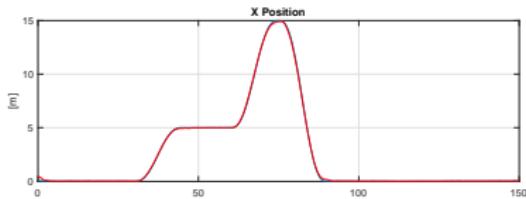


Fig.: Simulation Results: Robust Controller



Results - Simulation Study

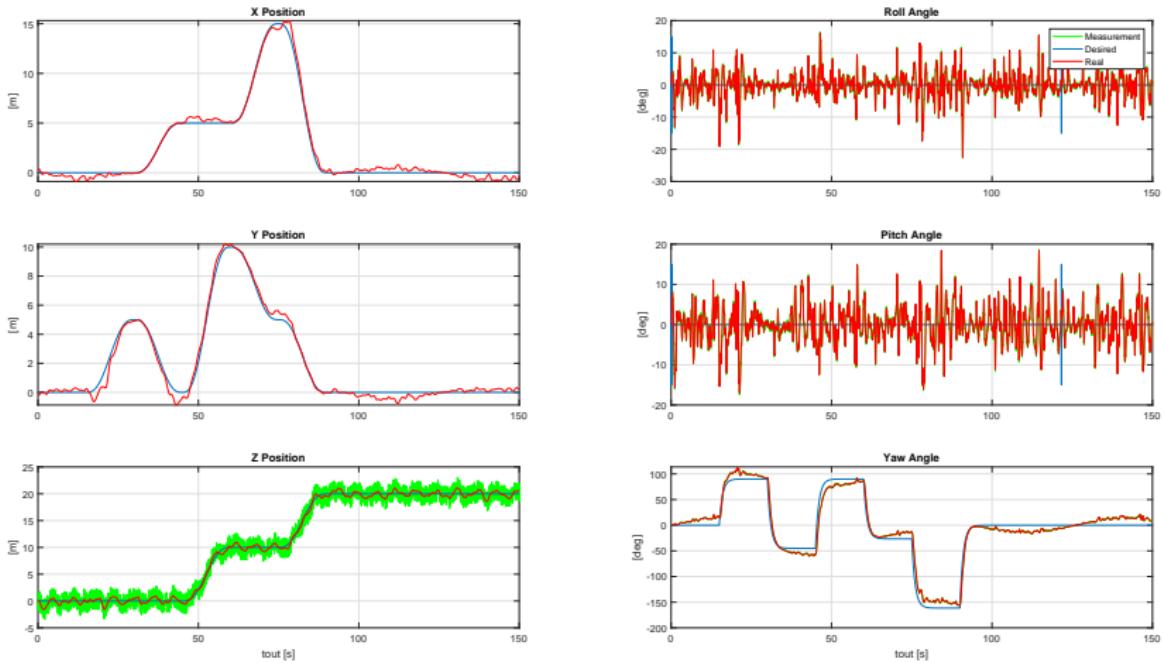


Fig.: Simulation Results: NHC



Results - Simulation Study



Conclusion and Outlook

Conclusion

- Approach is promising but extendable
- Robust Controller is derived
- Introduced uncertainties are justified
- RC can compete with NHC
- Stability guarantees can be given

Outlook

- Implementation of further performance channels, e.g. reference → controller output
- Implementation of x and y controller as robust controller
 - Handling of noise and external inputs on these values possible
 - Weights have to be found
- **Adding more weights will get more difficult!**



References

Bosch Sensortec GmbH.

Small, low power inertial measurement unit, 10 2018.
Rev. 0.9.

C. W. Scherer.

Theory of robust control.
Lecture Notes, 2018.

Inc. The MathWorks.

robstab - robust stability of uncertain system, 2019.
<https://de.mathworks.com/help/robust/ref/robstab.html#bveh4hp-1-wcu>, visited 2019-03-01.

u-blox AG.

u-blox 8 GNSS modules, 7 2016.
Rev. 03.

Dr.-Ing. Burak Yüksel.

Introduction to aerial robotics.
Lecture Notes, 2018.