



POLITECNICO
MILANO 1863

Robust design of attitude and position control system for a multirotor UAV

Aerospace Control Systems
Final exam project
AA 2022-2023

Pietro Dal Lago
Lorenzo Cucchi
Alberto Armanni

Summary



- Formulation of the problem
- Attitude control of simplified model
- Attitude control with motor and battery dynamics
- Robust analysis
- Monte Carlo analysis

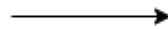
Formulation of the problem

Given the lateral-directional dynamics of a quadrotor drone, in terms of a statespace LTI system with stability and control derivatives, the purpose of the project is to design a control system for the roll angle ϕ .

Lateral dynamics:

Dynamic equation

$$\begin{cases} \dot{x} = Ax + Bu \\ y = Cx + Du \end{cases}$$



$$\begin{bmatrix} \dot{v} \\ \dot{p} \\ \dot{\phi} \end{bmatrix} = \begin{bmatrix} Y_v & Y_p & g \\ L_v & L_p & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} v \\ p \\ \phi \end{bmatrix} + \begin{bmatrix} Y_d \\ Y_d \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} p \\ \varphi \\ a_y \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ Y_v & Y_p & 0 \end{bmatrix} \begin{bmatrix} v \\ p \\ \phi \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ Y_\delta \end{bmatrix}$$

Formulation of the problem

Followed procedure

Simplified model

Task 1

- Nominal performance
- Control effort

Task 2

- Robustness analysis

Complete model

Task 2

- Nominal performance
- Control effort
- Robustness analysis

Task 3

- Observer design

Task 4

- Monte Carlo Validation

Formulation of the problem

Data

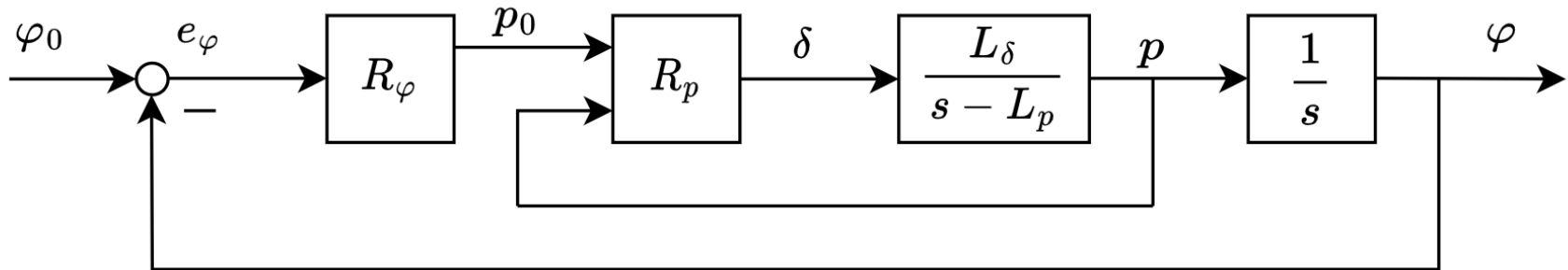
Derivatives	Nominal Value	Units	Standard deviation [%]
Y_v	-0.1068	s^{-1}	4.26
Y_p	0.1192	$m s^{-1} rad$	2.03
L_v	-5.9755	$rad s m^{-1}$	1.83
L_p	-2.6478	s^{-1}	2.01
Y_d	-10.1647	$m s^{-2}$	1.37
L_d	450.785	$rad s^{-2}$	0.81

The nominal values for stability and control derivatives are listed in the table on the left

For each one the standard deviations is expressed as percentage of the corresponding nominal values (assuming Gaussian distribution for each derivative)

Attitude control

(with assumptions)



Control-oriented assumptions:

- $Y_\delta = 0$
- $Y_p = 0$
- $L_v = 0$

Resulting equations:

- $\dot{\phi} = p$
- $\dot{p} = L_p p + L_\delta \delta$
- $\dot{v} = Y_v + g\phi$

Transfer functions:

- $\varphi = \frac{1}{s} p$
- $p = \frac{L_\delta}{s - L_p} \delta$
- $v = \frac{g}{s - Y_v} \varphi$

Requirements:

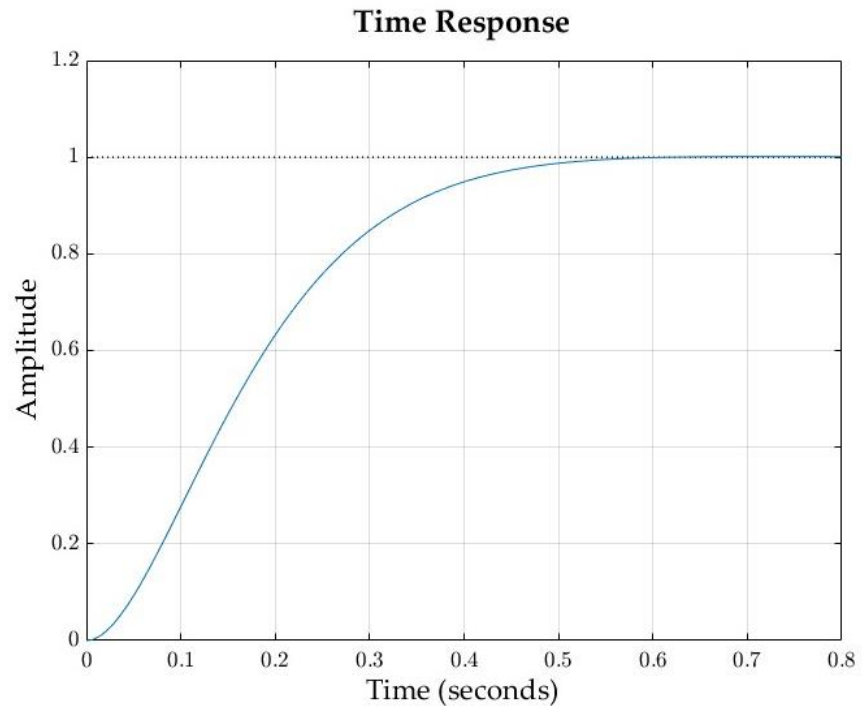
- Nominal performance: response of ϕ to its setpoint equivalent to a second order system with $\omega_n \geq 10$ rad/s and $\xi \geq 0.9$
- Control effort limitation: 5% for a doublet setpoint (+10° between 1÷3 sec and -10° between 3÷5 sec)
- Robust stability
- Robust performance

Task 1 (with assumptions)

weight selection for nominal response

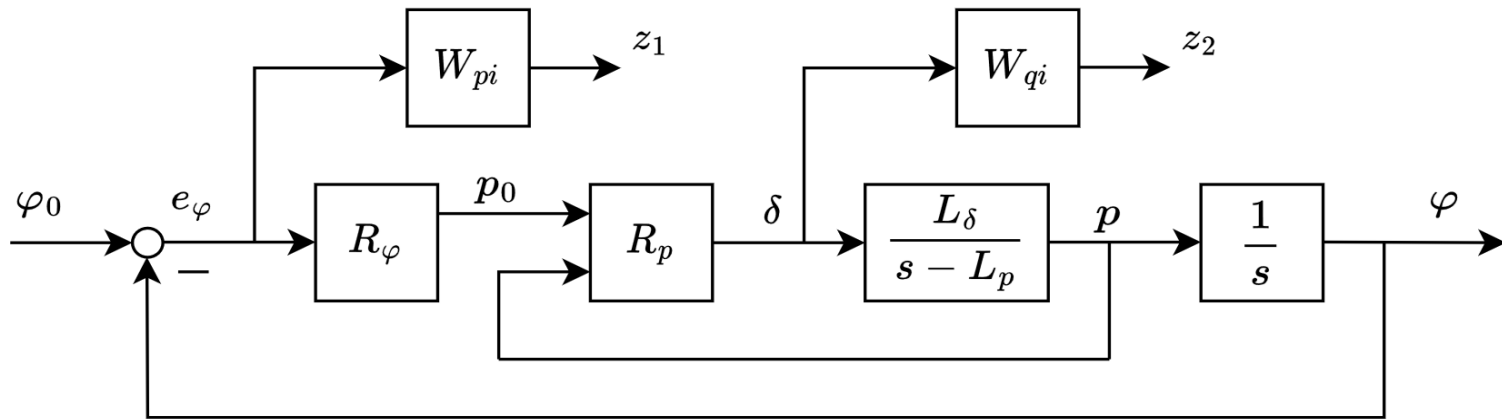
$$F_{\varphi^0 \rightarrow \varphi}^{target} = \frac{\omega_n^2}{s^2 + 2\xi\omega_n s + \omega_n^2}$$

Step Info	
Rise Time	0.2883
Transient Time	0.4700
Settling Time	0.4700
Settling Min	0.9024
Settling Max	1.0015
Overshoot	0.1524
Undershoot	0
Peak	1.0015
Peak Time	0.7215



Task 1 (with assumptions)

Augmented Plant, tuning



- The control-oriented assumptions allow a cascade architecture
- R_φ is a proportional controller
- R_p is a 2-degree of freedom PID
- Two weighting functions are added to the augmented plant for the structured H^∞ synthesis of the controllers

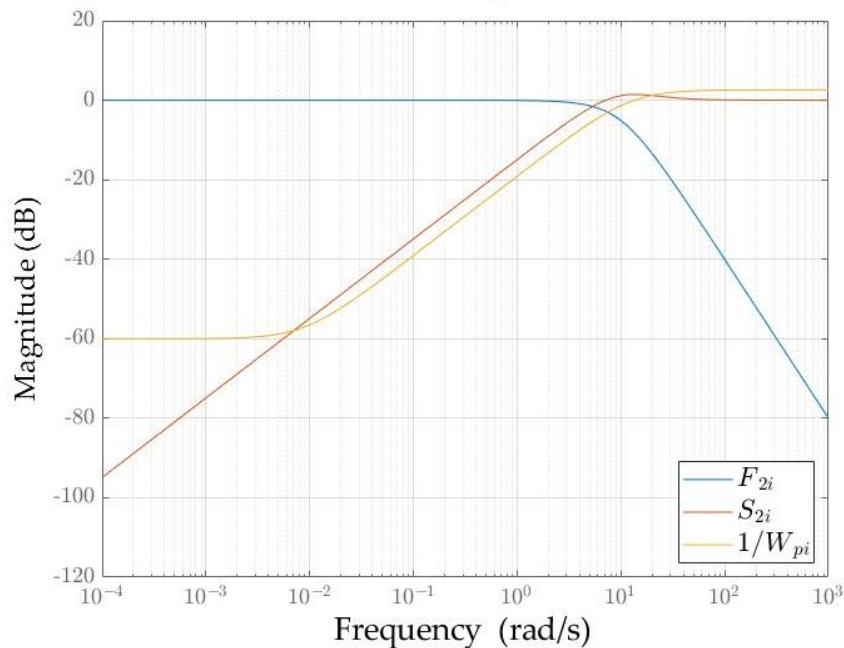
Task 1 (with assumptions)

Weight selection for nominal response

$$W_p = \frac{\frac{s}{M} + \omega_b}{s + A\omega_b}$$

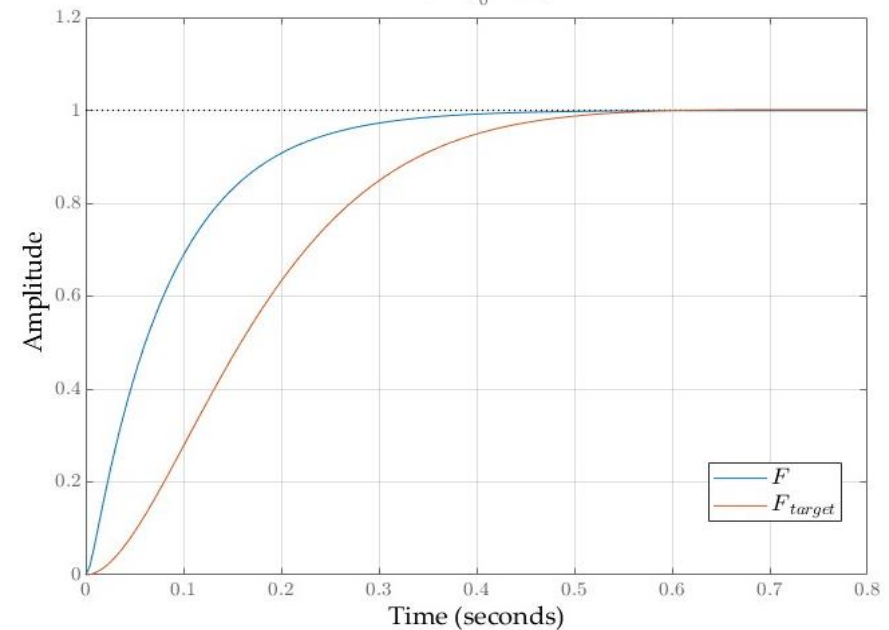
$$\begin{aligned} M &= 1.35 \\ A &= 0.001 \\ \omega_b &= 0.9\omega_n \end{aligned}$$

Bode Diagram



Step Response

From: ϕ_0 To: ϕ



Task 1 (with assumptions)

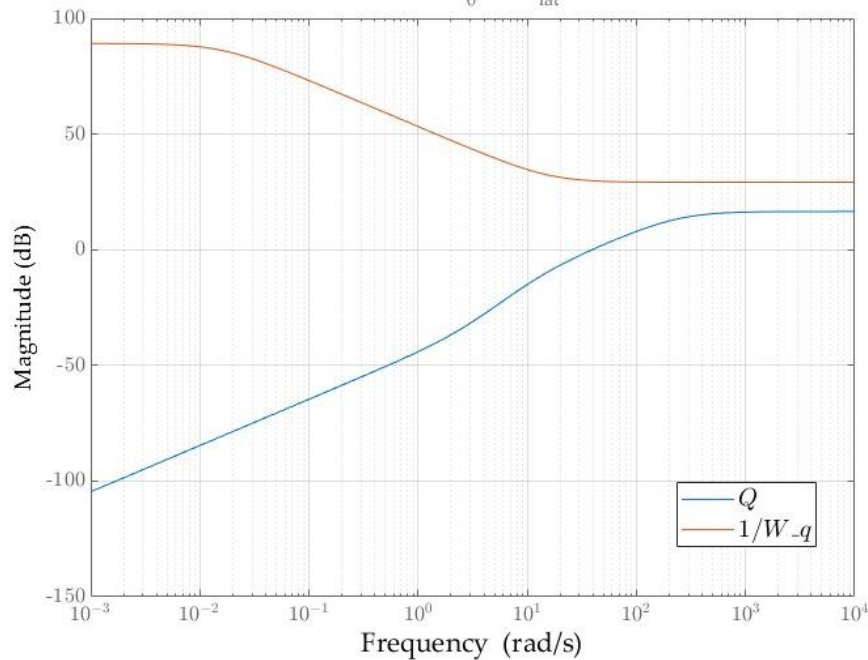
Weight selection for control effort limitation

$$W_q = \alpha \frac{s + \omega_{act} 10^{-3}}{s + \omega_{act}}$$

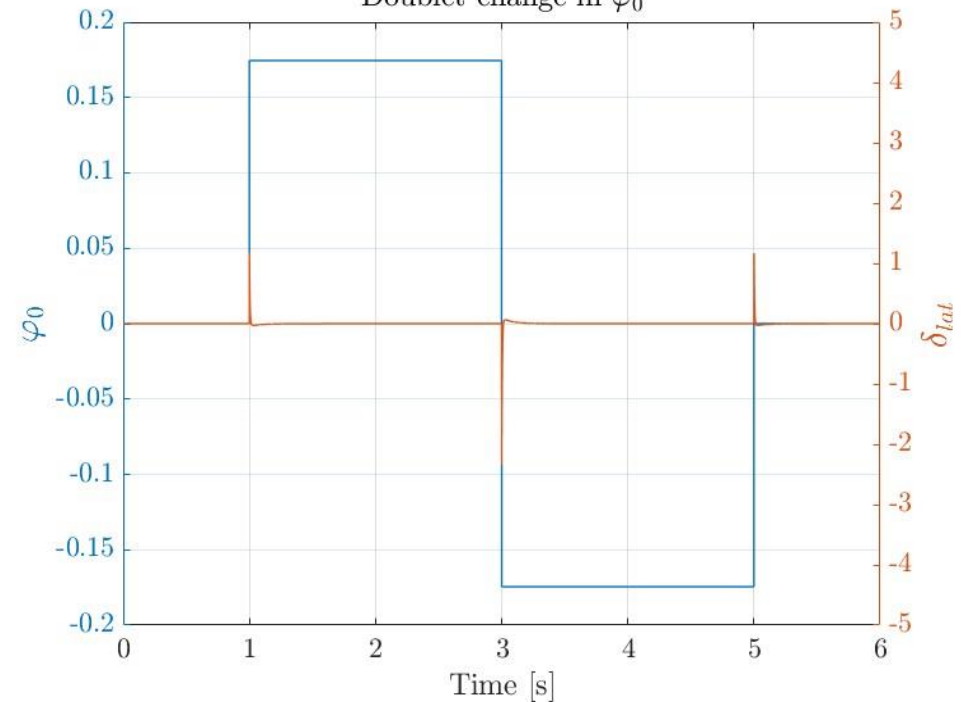
$$\alpha = \frac{10 \frac{180}{\pi}}{5}$$
$$\omega_{act} = 16$$

Bode Diagram

From: ϕ_0 To: δ_{lat}

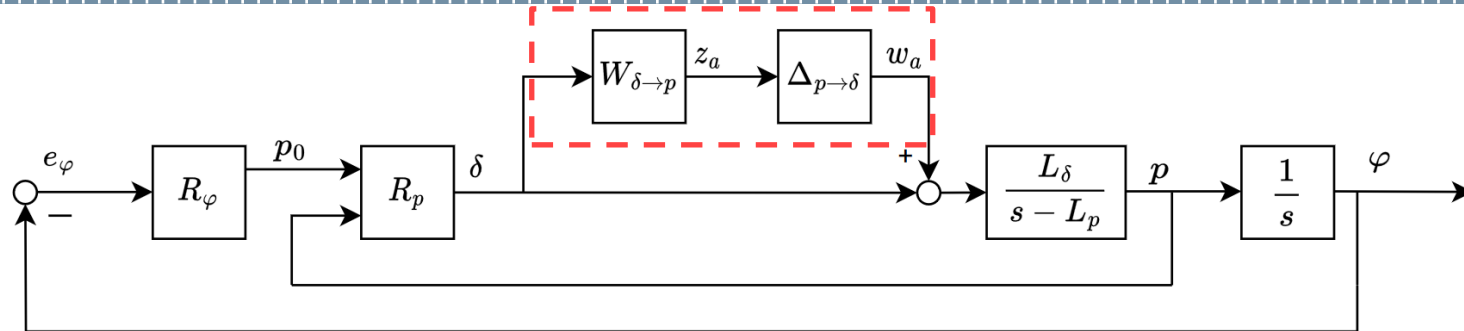


Doublet change in φ_0



Task 1 (with assumptions)

Robustness Verification



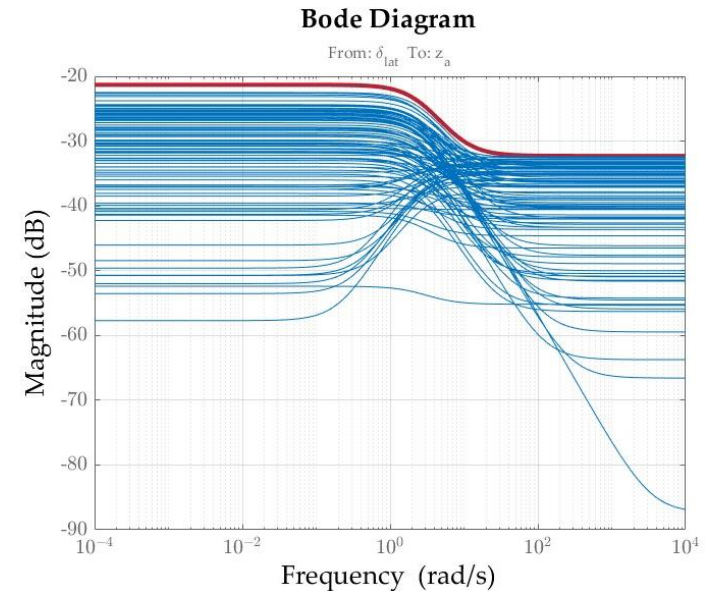
The weight for robust stability and robust performance is constructed with 'usample' and 'ucover' as an upper limit for the uncertain model

```
delta_p_array = usample(sys_delta_p,100);
[~,Info] = ucover(delta_p_array, sys_delta_p_nom, 3);
W_delta_p = Info.W1;
```

```
SumInner2 = sumblk('uili = \delta_{lat} + w_a')
```

```
CL = connect(R_phi, R_p, sys_p_phi_nom, sys_delta_p_nom, SumInner1,...
SumInner2, W_delta_p, 'w_a', 'z_a',{'\delta_{lat}','\phi'});
```

```
M = getIOTransfer(CL, 'w_a', 'z_a');
```



Task 1 (with assumptions)

Robustness analysis

$$R_\varphi = k_p(\varphi_0 - \varphi)$$

$$R_p = k_p(bp_0 - p) + \frac{k_i}{s}(p - p_0) + k_d \frac{s}{Ts + 1}$$

$$T = 0.001s, \quad b = 1, \quad c = 1$$

Proportional controller

$$k_p = 11.5857$$

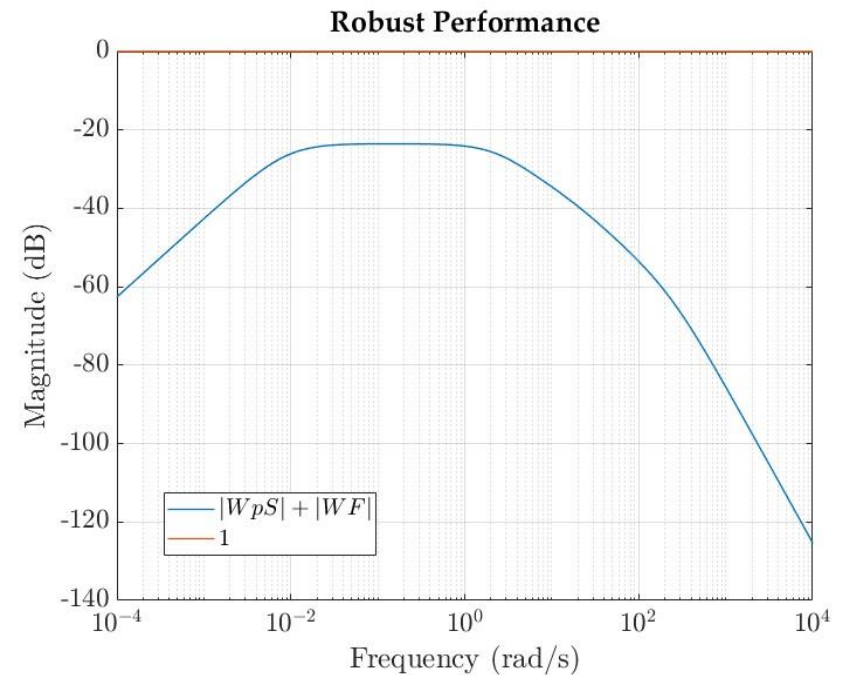
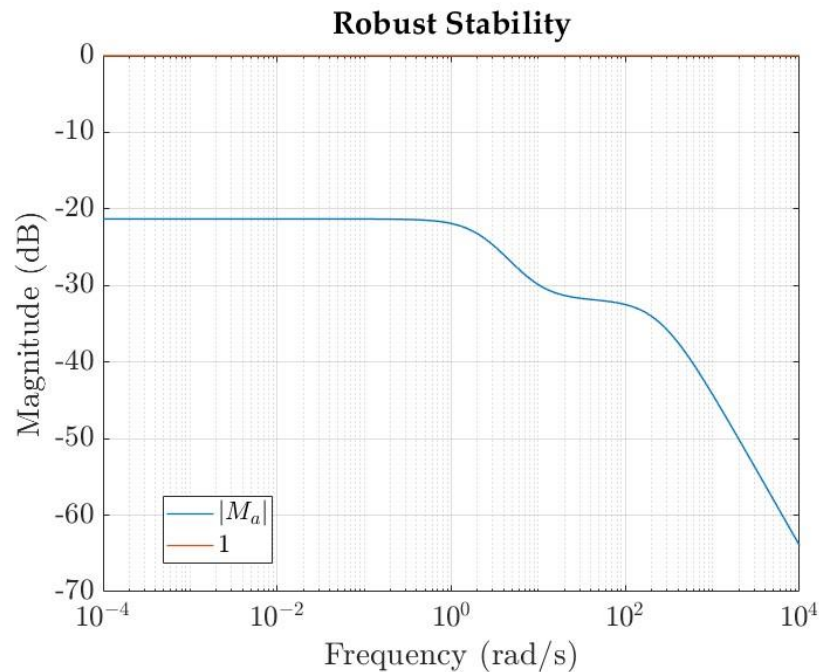
PID

$$k_p = 0.5782$$

$$k_d = 1.5613$$

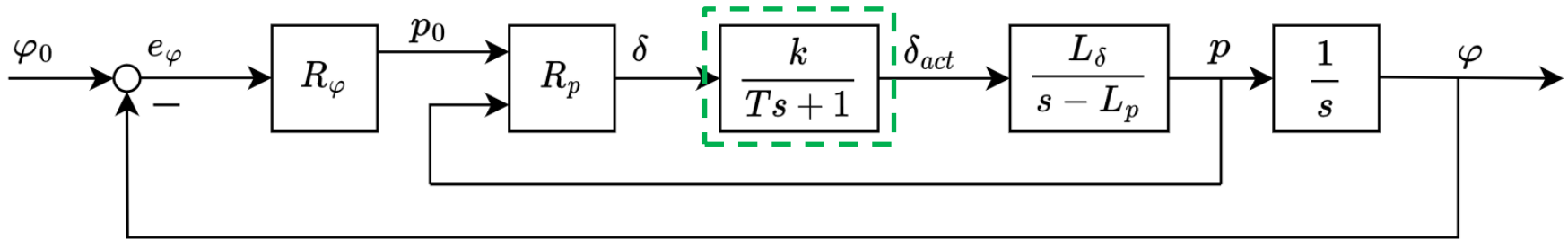
$$k_i = 2.745e-7$$

$$\text{Settling Time} = 0.3261 \text{ [s]}$$



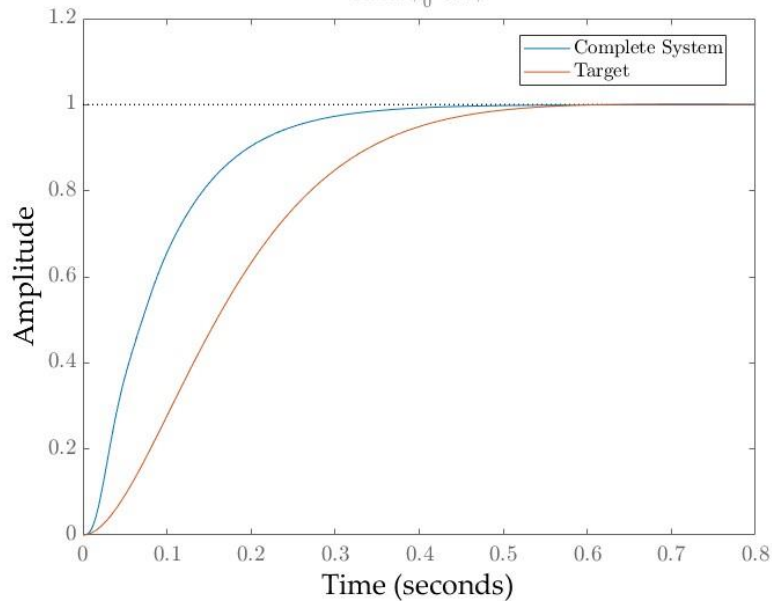
Task 2 (with assumptions)

Propulsive unit and battery addition, tuning



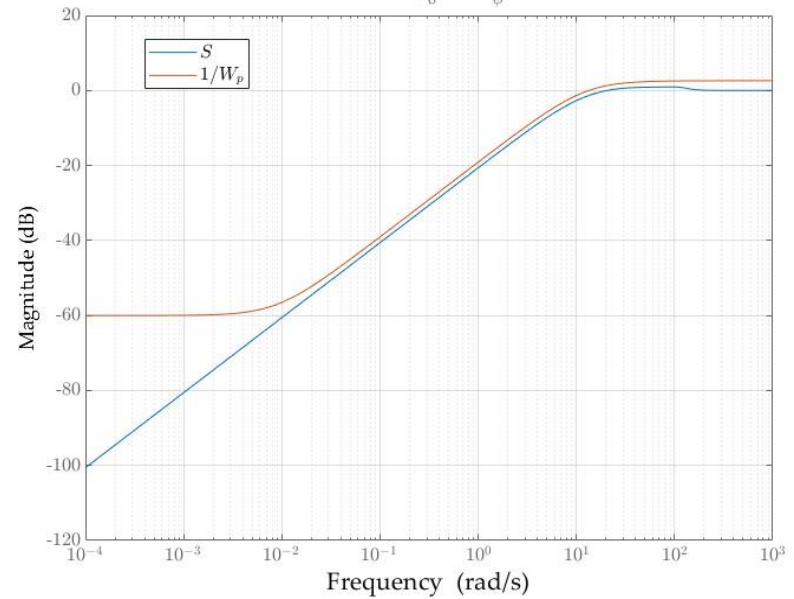
Step Response

From: ϕ_0 To: ϕ



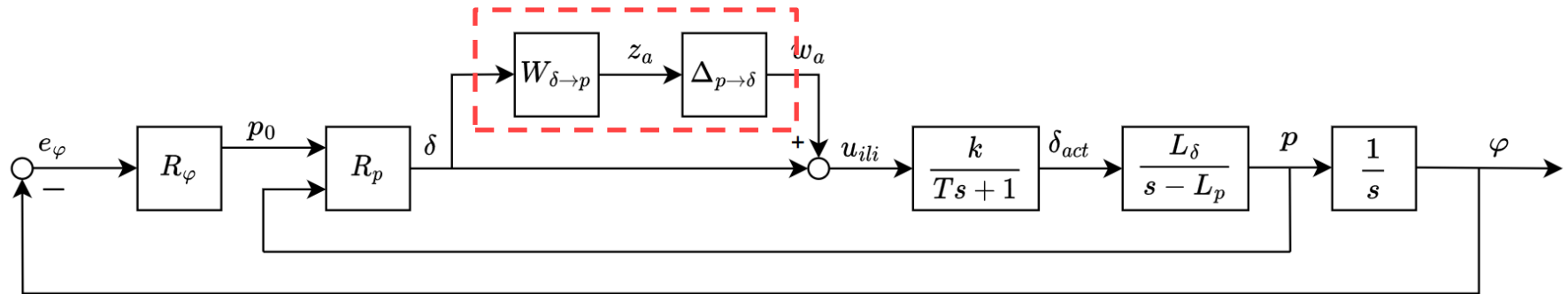
Bode Diagram

From: ϕ_0 To: e_ϕ



Task 2 (with assumptions)

Robustness Verification



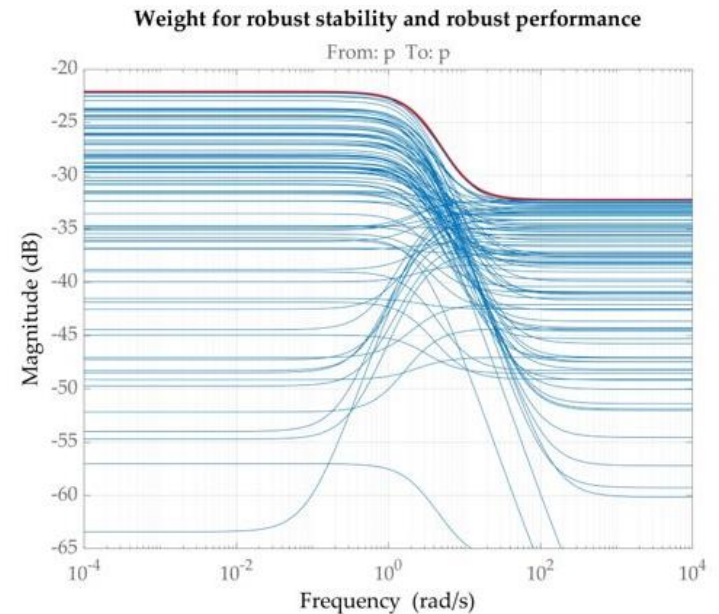
```

delta_p_array = usameple(sys_delta_p,100);
[~,Info] = ucover(delta_p_array, sys_delta_p_nom, 3);
W_delta_p = Info.W1;

SumInner2 = sumblk('uili = \delta_{lat} + w_a')

CL = connect(R_phi, R_p, sys_d_db_nom, sys_db_da, sys_delta_p_nom,...
SumInner1, SumInner2, W_delta_p, 'w_a', 'z_a',{'\delta_{lat}','\phi'});

M = getIOTransfer(CL, 'w_a', 'z_a');
    
```



Task 2 (with assumptions)

Propulsive unit and battery addiction, tuning

$$R_\varphi = k_p(\varphi_0 - \varphi)$$

$$R_p = k_p(bp_0 - p) + \frac{k_i}{s}(p - p_0) + k_d \frac{s}{Ts + 1}$$

$$T = 0.001s, b = 1, c = 1$$

Proportional controller

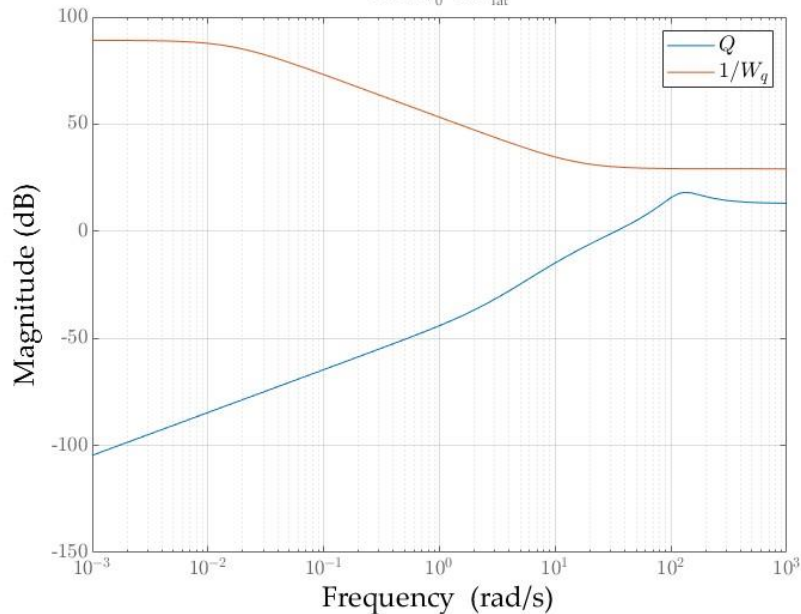
k_p	10.8
-------	------

PID

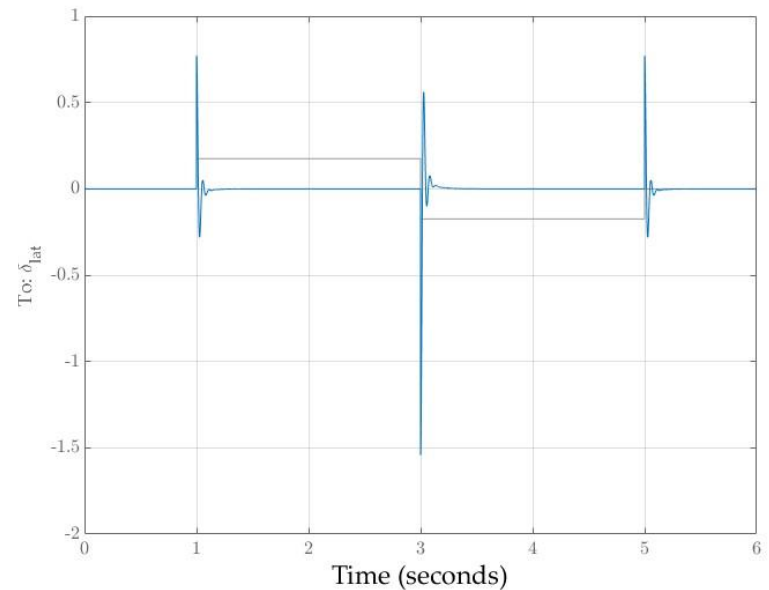
k_p	0.47
k_i	4.3e-8
k_d	0.00303

Bode Diagram

From: ϕ_0 To: δ_{lat}

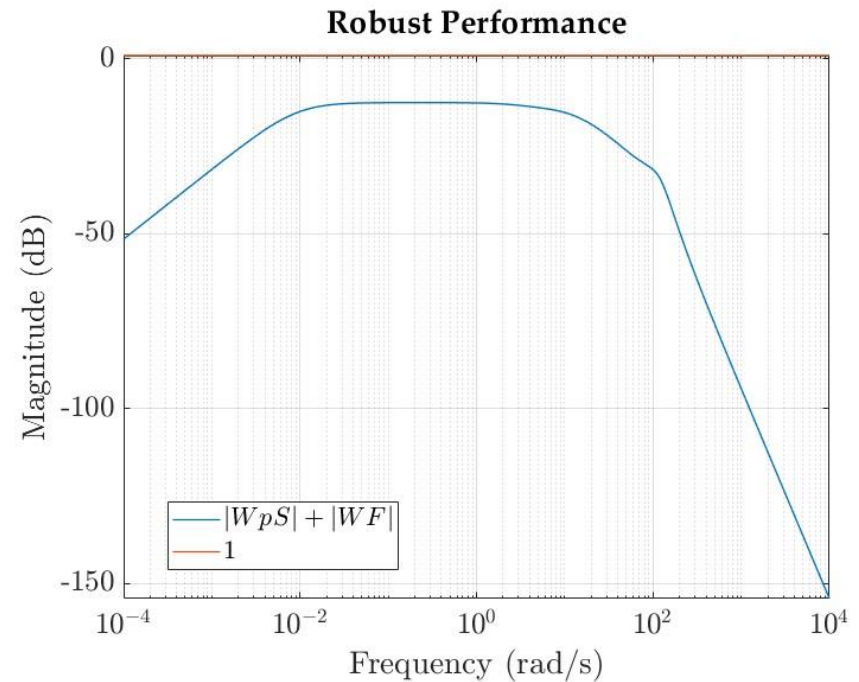
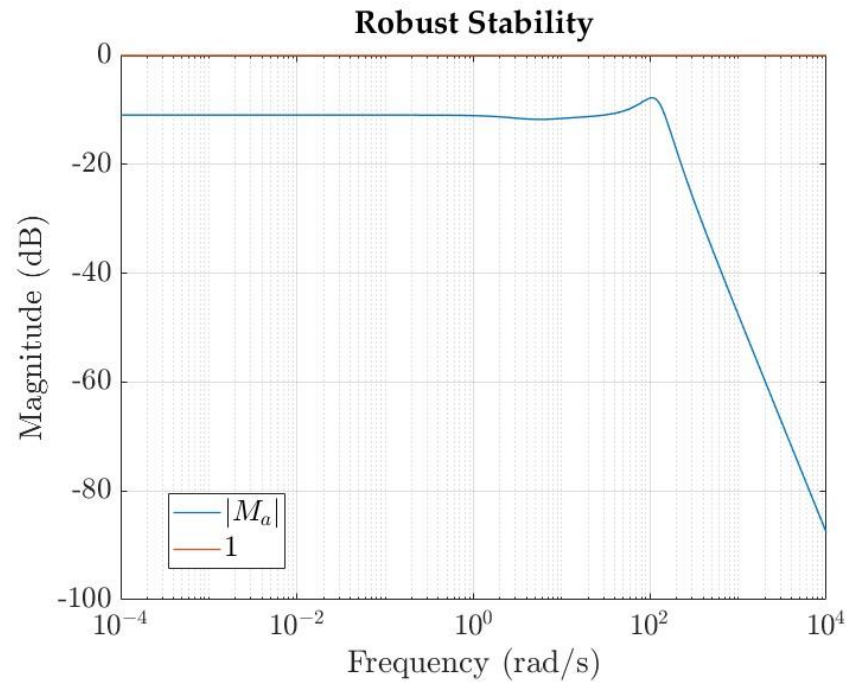


Linear Simulation Results



Task 2 (with assumptions)

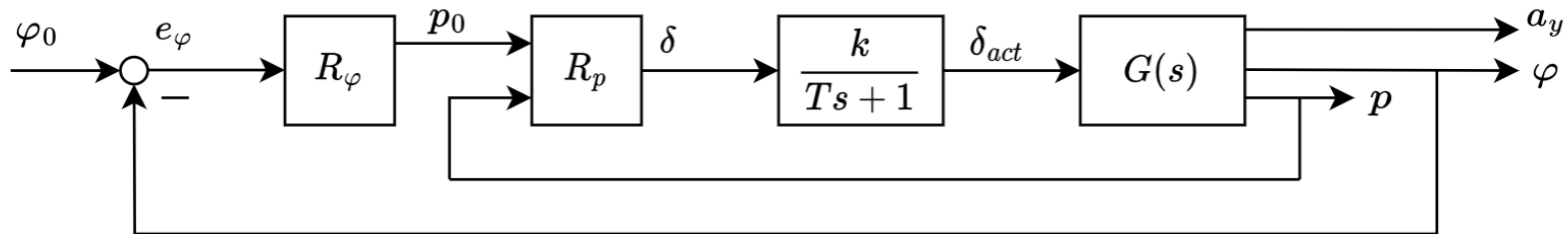
Results and comparison with the requirements



Relaxing assumptions

Task 2 (complete)

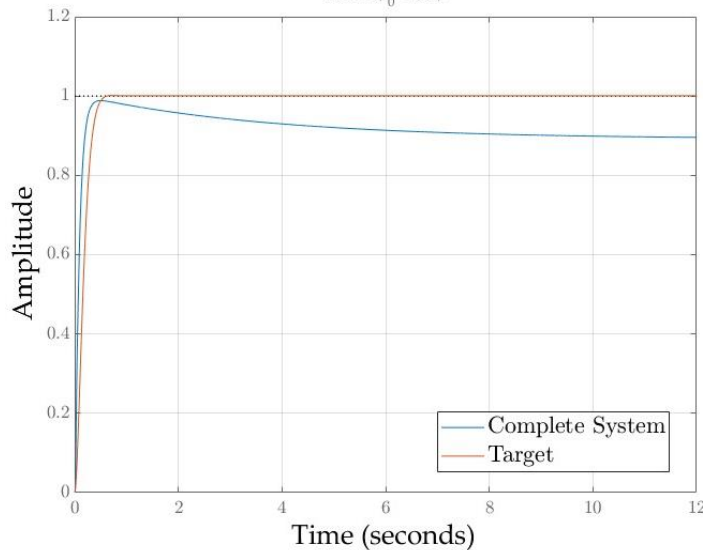
Pid from simplified system



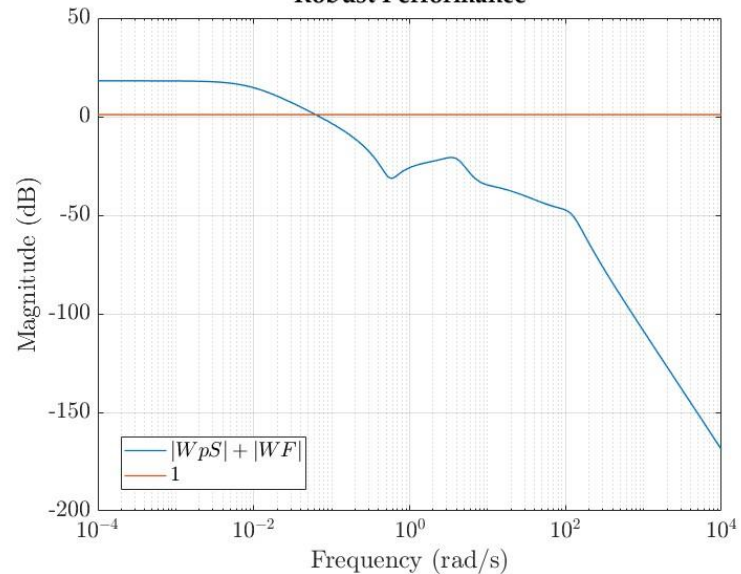
The Integral gain of the R_p pid obtained from the simplified system is 0, with the complete system we need an integral gain to track the error.

Step Response

From: ϕ_0 To: ϕ



Robust Performance



Task 2 (complete)

Results and comparison with requirements

$$R_\varphi = k_p(\varphi_0 - \varphi)$$

$$R_p = k_p(bp_0 - p) + \frac{k_i}{s}(p - p_0) + k_d \frac{s}{Ts + 1}$$

$$T = 0.001s, b = 1, c = 1$$

Proportional controller

$$k_p = 11.1$$

PID

$$k_p = 0.438$$

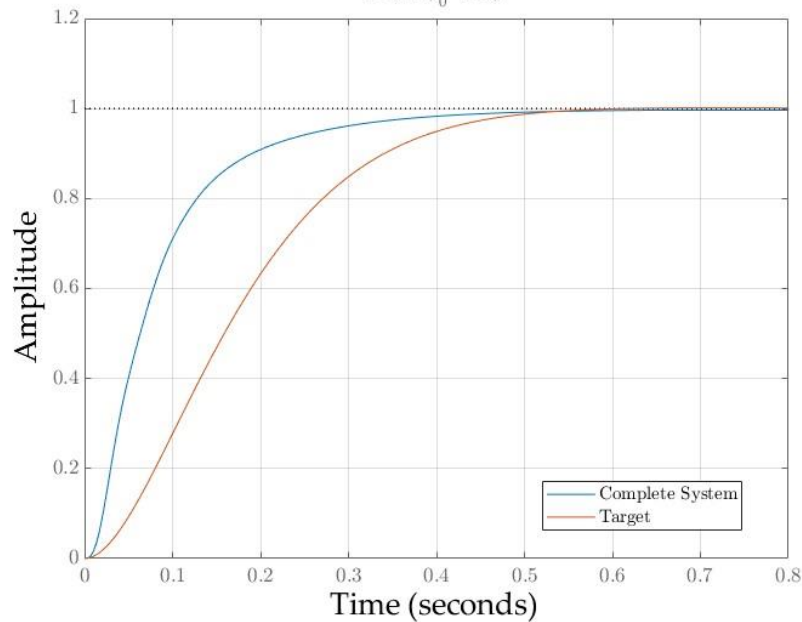
$$k_i = 5.67$$

$$k_d = 0.00415$$

$$\text{Settling Time} = 0.3805 \text{ [s]}$$

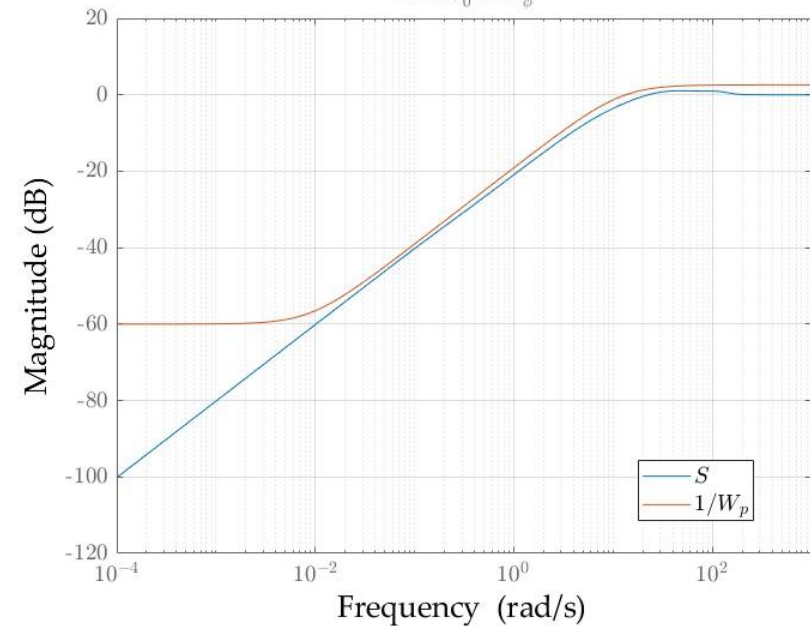
Step Response

From: ϕ_0 To: ϕ



Bode Diagram

From: ϕ_0 To: e_ϕ

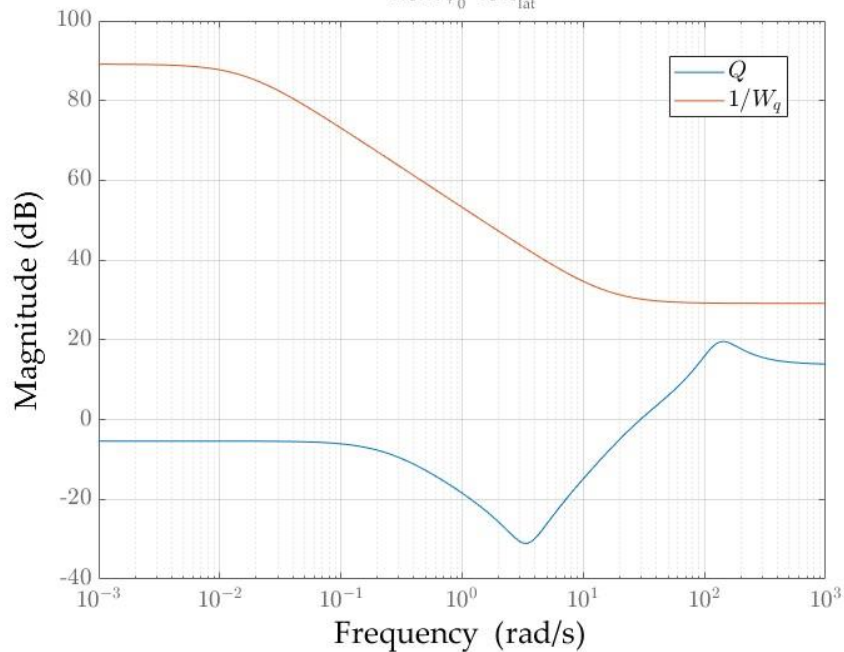


Task 2 (complete)

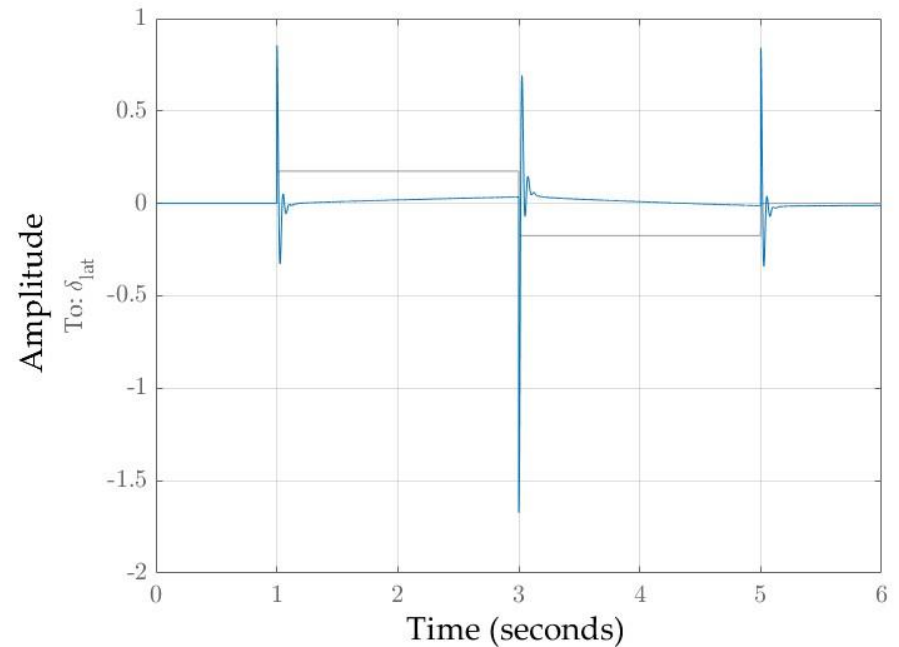
Re-tuning and compliance with requirements

Bode Diagram

From: ϕ_0 To: δ_{lat}

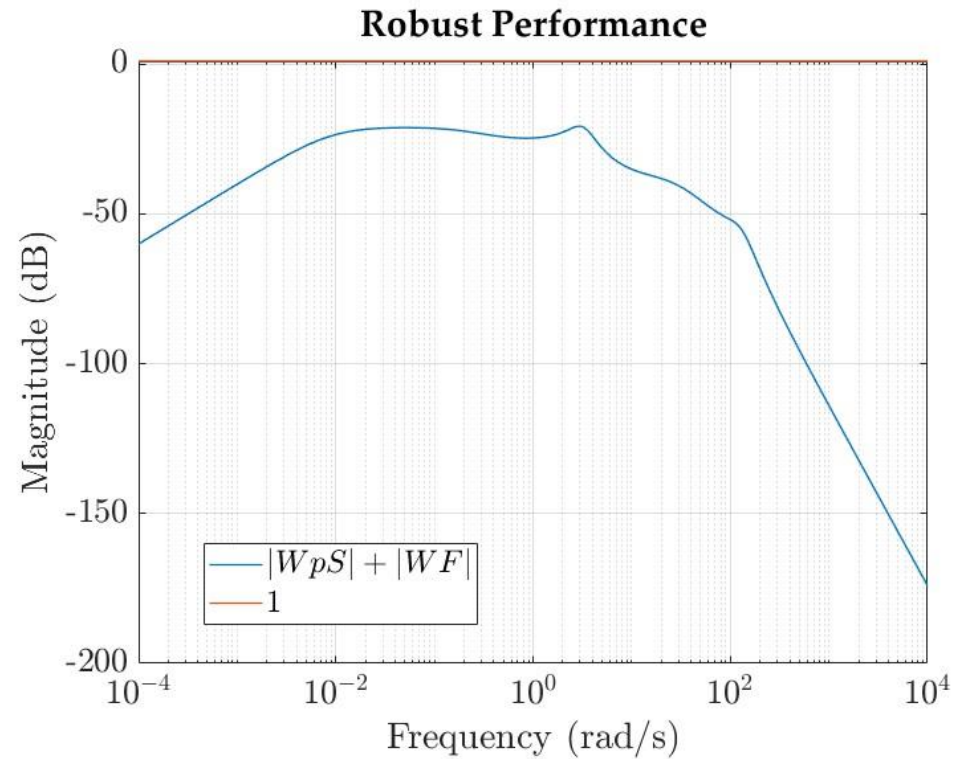
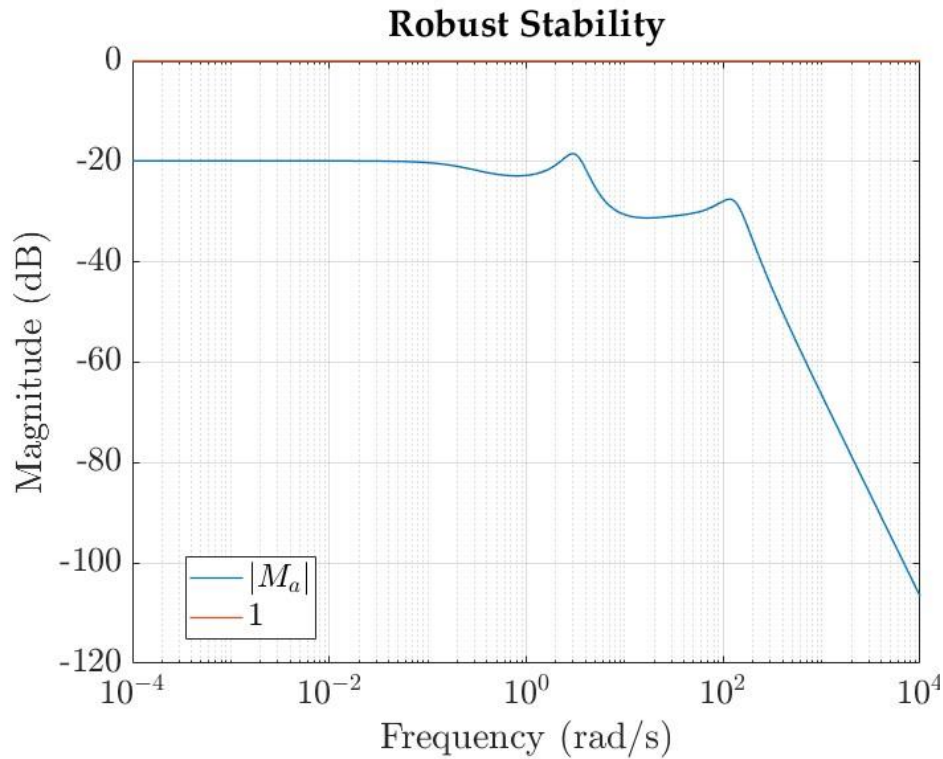


Linear Simulation Results



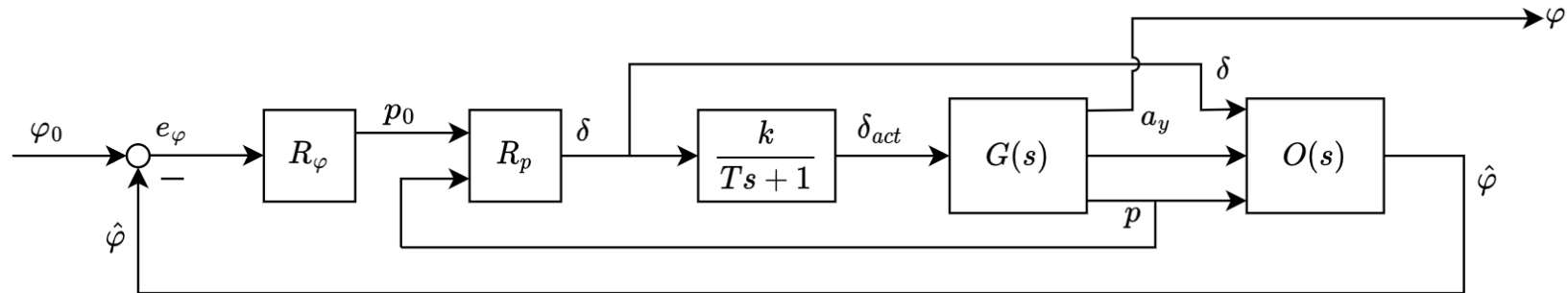
Task 2 (complete)

Robustness analysis



Task 3 (complete)

Observer design



A Luenberger state observer θ estimates ϕ through the measurements of δ , a_y and p .

The dynamics of the observer is governed by the state matrix $(A - L^*C)$.

The problem of defining the L matrix is an eigenvalue assignment, to solve it Matlab's command *place* is used

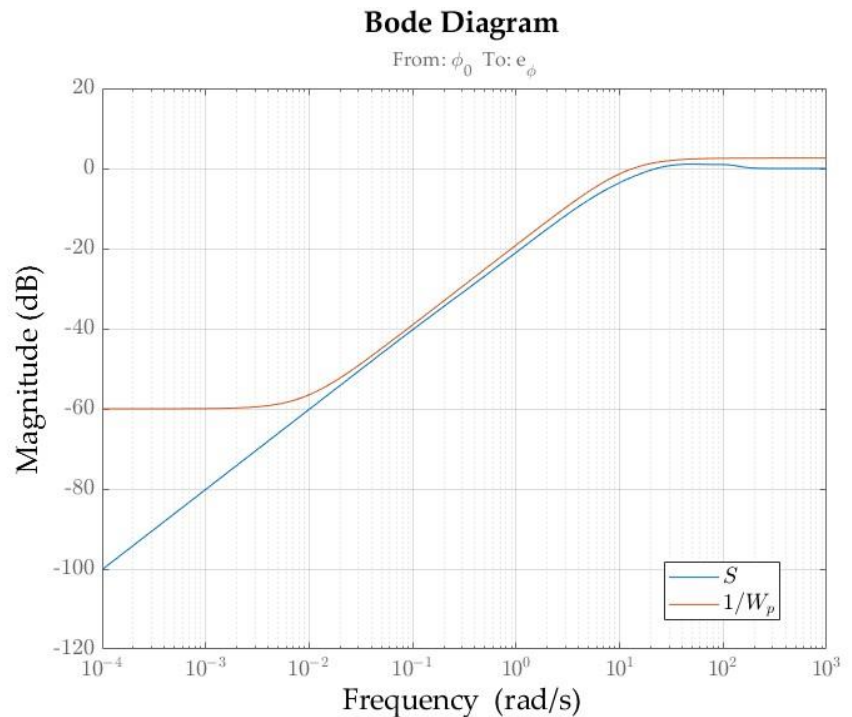
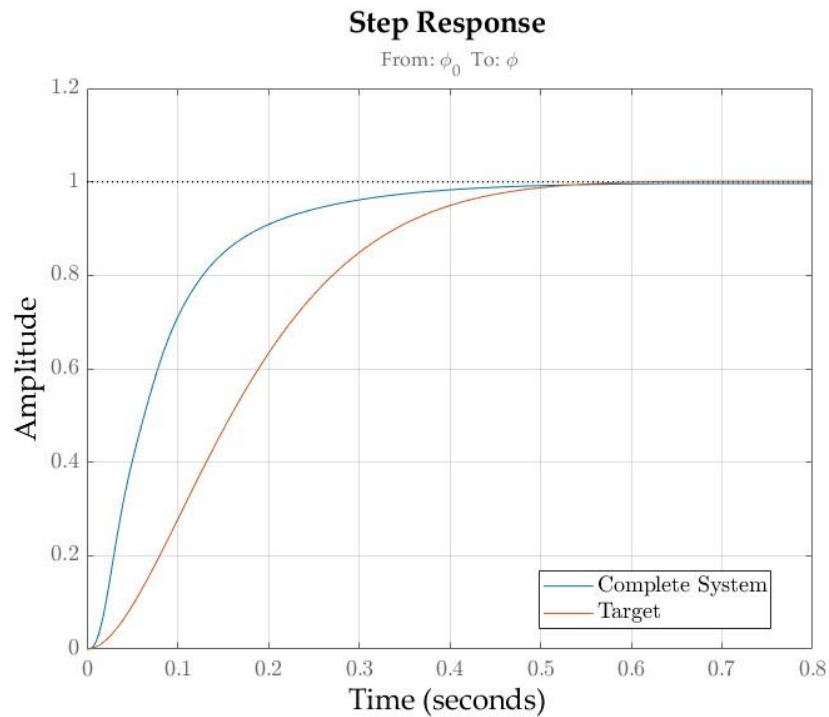
$$\begin{cases} \dot{\hat{x}} = (A - LC)\hat{x} + (B - LD)u + Ly \\ \hat{y} = C\hat{x} + Du \end{cases}$$

$$L = \begin{bmatrix} 0.5692 & -3.7753 \\ 0.6829 & 55.9504 \\ 1.000 & -0.0048 \end{bmatrix}$$

$$P = \begin{bmatrix} -0.5 \\ -0.01 \\ -10 \end{bmatrix}$$

Task 3 (complete)

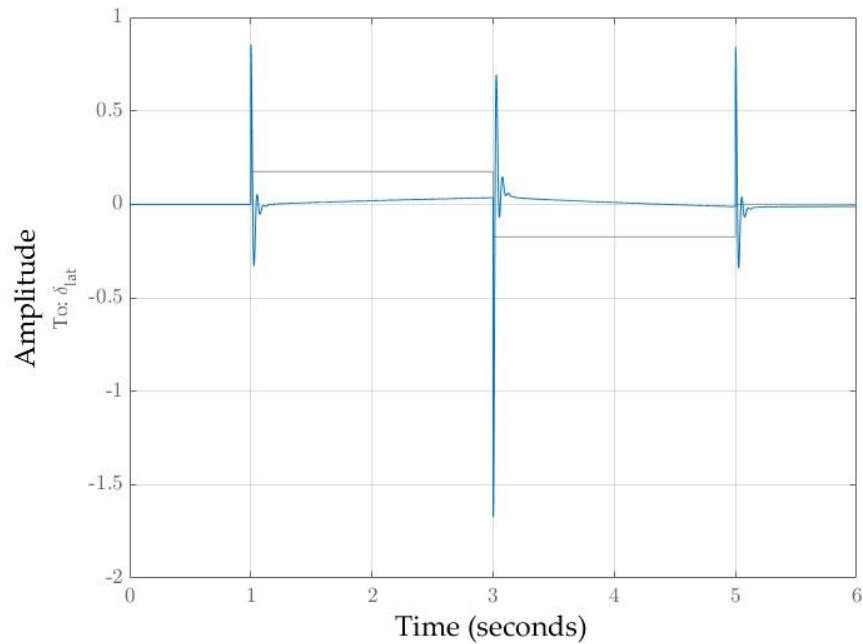
Robust stability analysis



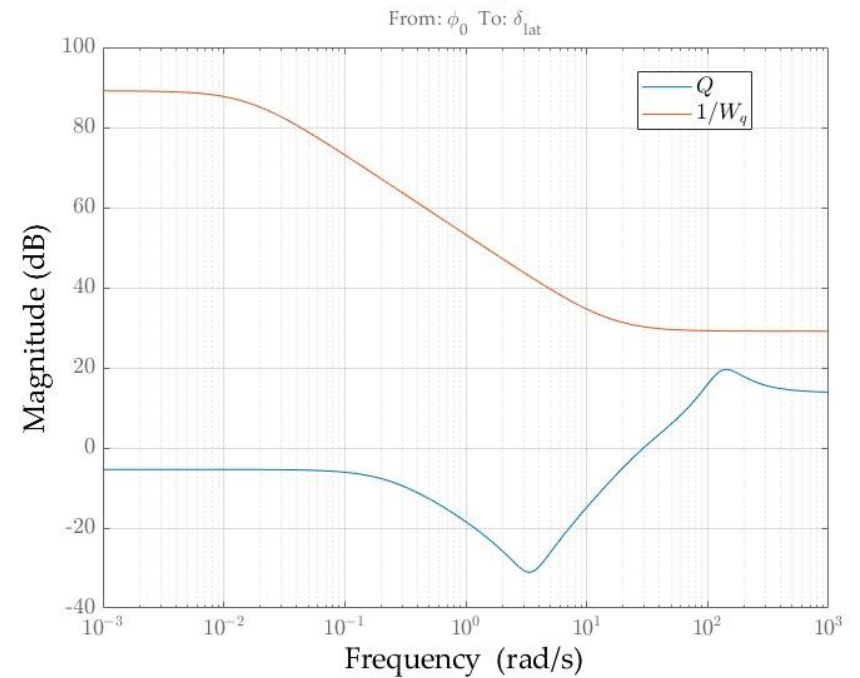
Task 3 (complete)

Compliance with the requirements

Linear Simulation Results



Bode Diagram



A Montecarlo method with 10000 iterations is used to validate the observer design and the system tuning:

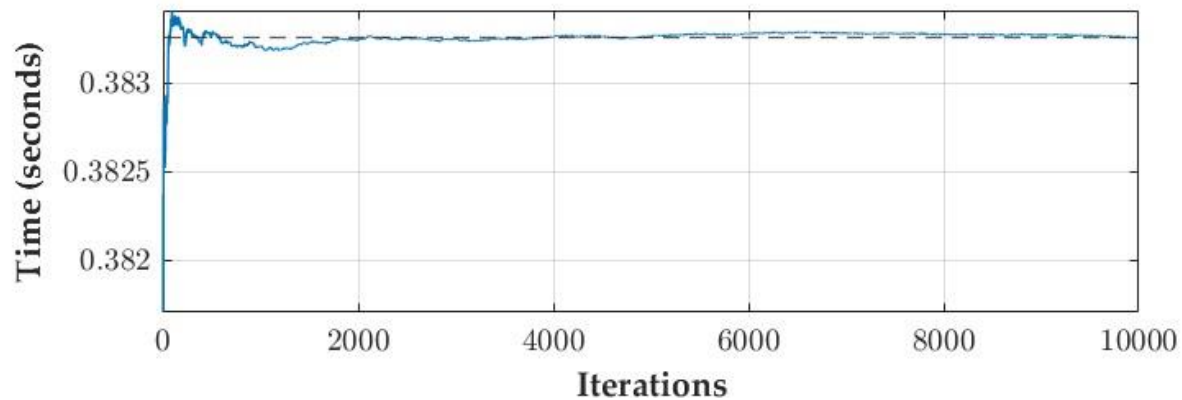
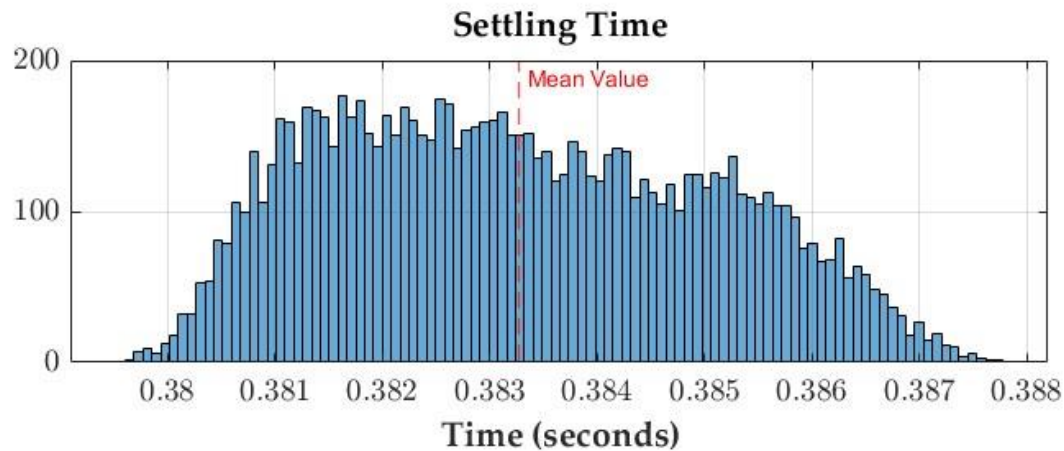
- Step response of ϕ in terms of percentage overshoot and settling time
- Control effort

The plots present a graph of the convergence of the mean value, this is used to show after how many iteration the results are stable.

Indicatively two thousands iteration are enough to have an accurate convergence of the mean and of the standard deviation

Task 4 (complete)

Monte Carlo validation

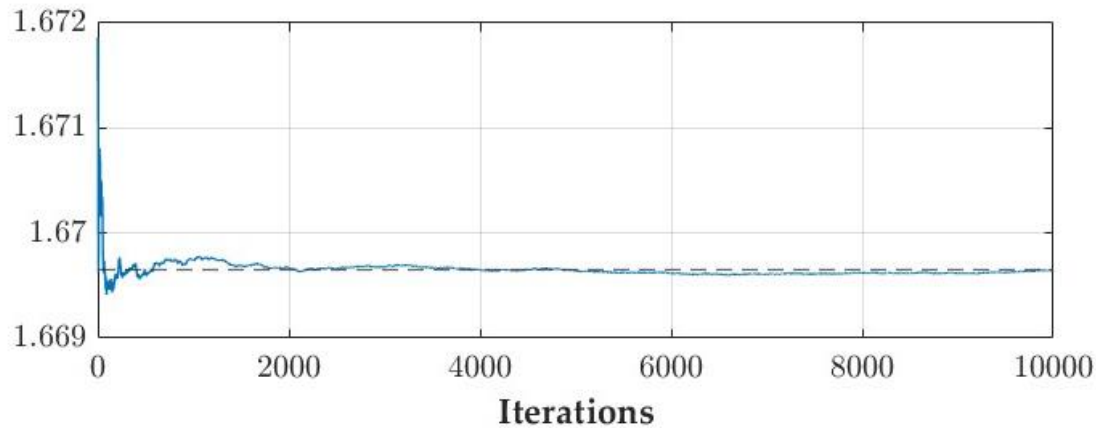
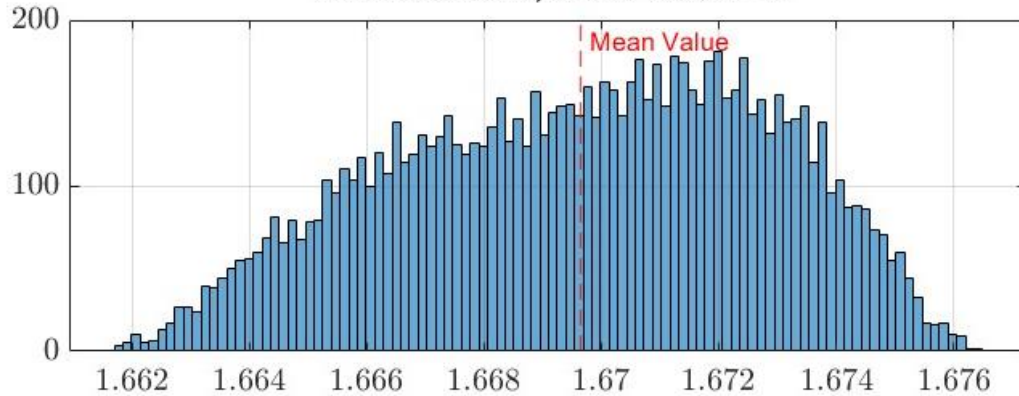


Settling Time	
Mean Value:	0.383262 [s]
Standard Deviation:	0.001755 [s]

Task 4 (complete)

Monte Carlo validation

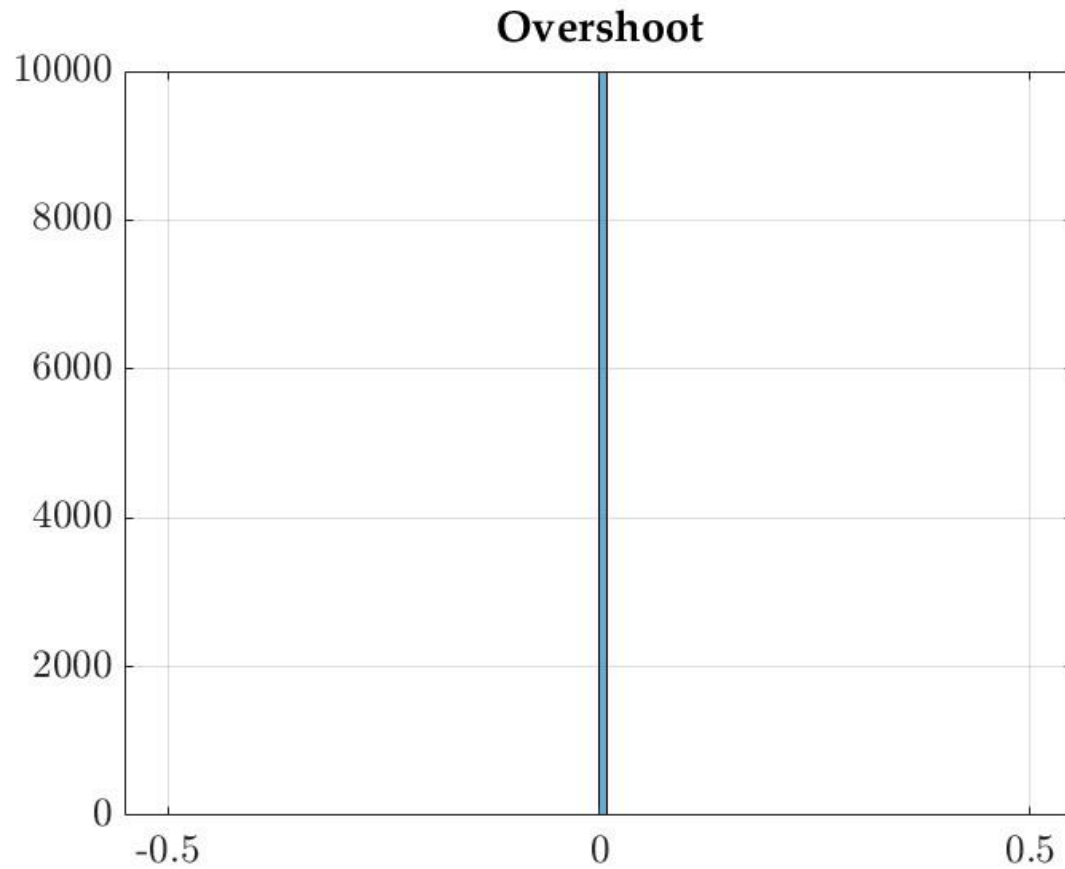
Control effort, limit value = 5



Control Effort	
Mean Value:	1.669646 [deg]
Standard Deviation:	0.003148 [deg]

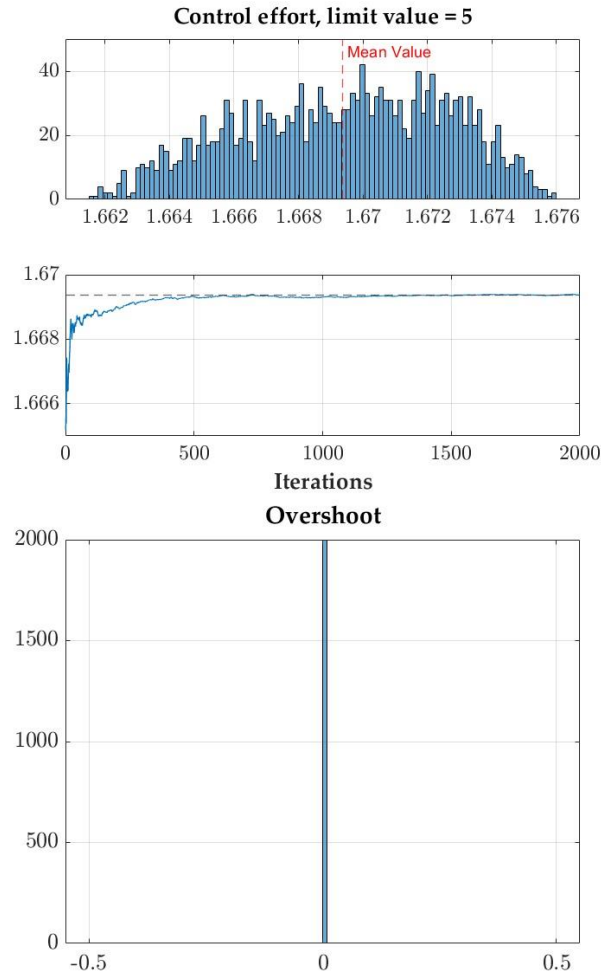
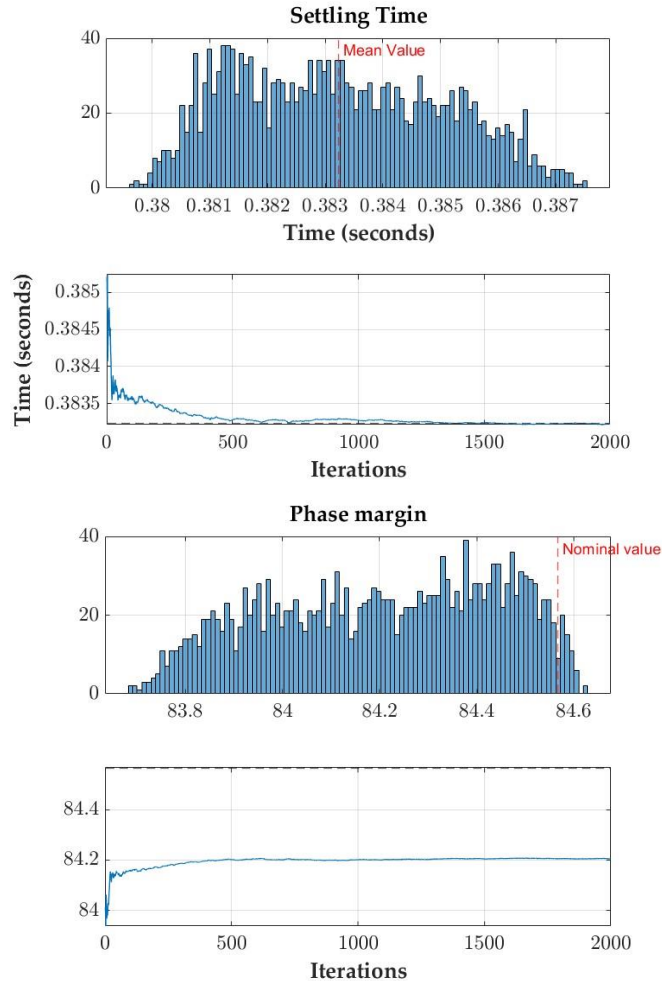
Task 4 (complete)

Monte Carlo validation



Task 4 (complete)

Monte Carlo validation



Samples	2000
---------	------

Step Info

Mean Value:	0.383232 [s]
Standard Deviation:	0.001794 [s]

Control Effort

Mean Value:	1.669373
Standard Deviation:	0.003209

Phase Margin

Mean Value:	84.20500 [deg]
Standard Deviation:	0.235579 [deg]



Fine