

UNIVERSITY OF
WATERLOO



UNIVERSITY OF WATERLOO

FACULTY OF ENGINEERING

ECE 488 - Project 3 & 4

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17 March 2021

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1 Problem P3: Controller Design by Youla Parameterization for the SISO aiming system

1.1 (a) Controller Design $C_1(s)$ with Youla Parameterization

In order to achieve perfect steady-state tracking of steps, the final $C_1(s)$ must have an integrator in it. To achieve this, we may augment the plant with an integrator term (Equation (1)), and design an augmented controller $C_1^{aug}(s)$ for the augmented plant $P_1^{aug}(s)$ through Youla Parameterization.

$$P_1^{aug}(s) = \frac{1}{s} P_1(s) = \frac{8}{s(s-4.427)(s+4.427)} \quad (1)$$

To perform Youla Parameterization, coprime factorization is performed with a scale factor of 10 to avoid numerical issues. The resultant coprimes can be found below:

$$M(s) = \frac{s(s-4.427)(s+4.427)}{(s+30)(s+20)(s+10)} \quad (2)$$

$$N(s) = \frac{8}{(s+30)(s+20)(s+10)} \quad (3)$$

$$X(s) = \frac{255510(s^2 + 7.861s + 17.61)}{(s+30)(s+20)(s+10)} \quad (4)$$

$$Y(s) = \frac{(s+65.16)(s^2 + 54.84s + 2246)}{(s+30)(s+20)(s+10)} \quad (5)$$

For simplicity, we assume the $Q(s) = 1$, which may result a complex augmented controller:

$$C_1^{aug}(s) = \frac{(s+255500)(s^2 + 7.861s + 17.61)}{(s+65.15)(s^2 + 54.85s + 2246)} \quad (6)$$

We may now obtain the final controller by applying the integrator term to the augmented controller obtained just now:

$$C_1(s) = \frac{1}{s} C_1^{aug}(s) = \frac{(s+255500)(s^2 + 7.861s + 17.61)}{s(s+65.15)(s^2 + 54.85s + 2246)} \quad (7)$$

The detailed implementation can be found in Code 1.

1.2 (b) Closed-loop step response for controller designed in Section 1.1

We may now run the simulation with a step input of 0.5 [rad] same as the Lab 1 and Lab 2. We may find the closed-loop system is indeed stable and the steady-state tracking error is zero from Figure 1-1 below.

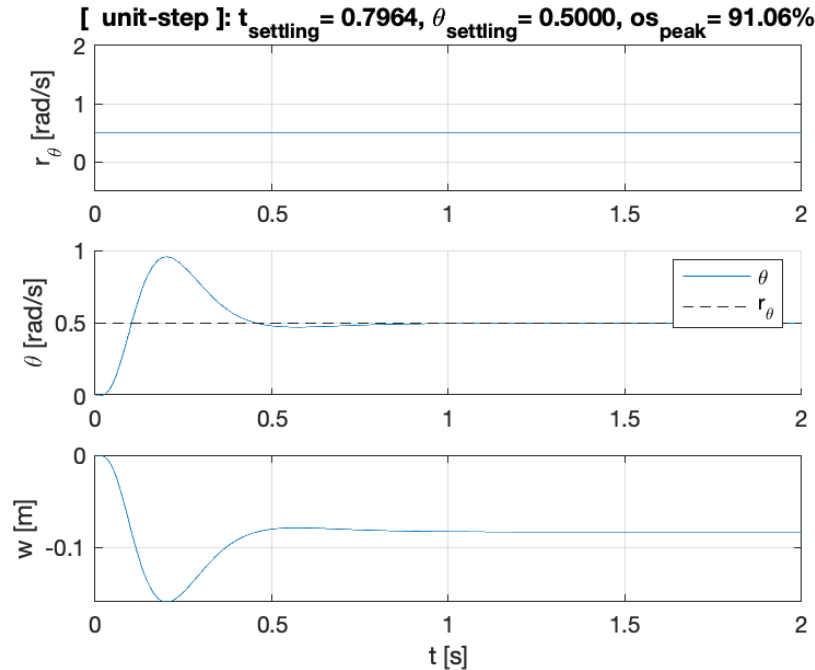


Figure 1-1. Closed-loop step response for controller $C_1(s)$ designed in Section 1.1

As expected, a simple $Q(s) = 1$ results a stable system, but there is no guarantee on the transient performance. From the Figure 1-1, we may observe a 91% overshoot and 0.7964 [s] settling time. The overshoot is quite significant.

To further analyze the system, a bode plot is generated for the closed-loop system, as shown in Figure 1-2 below.

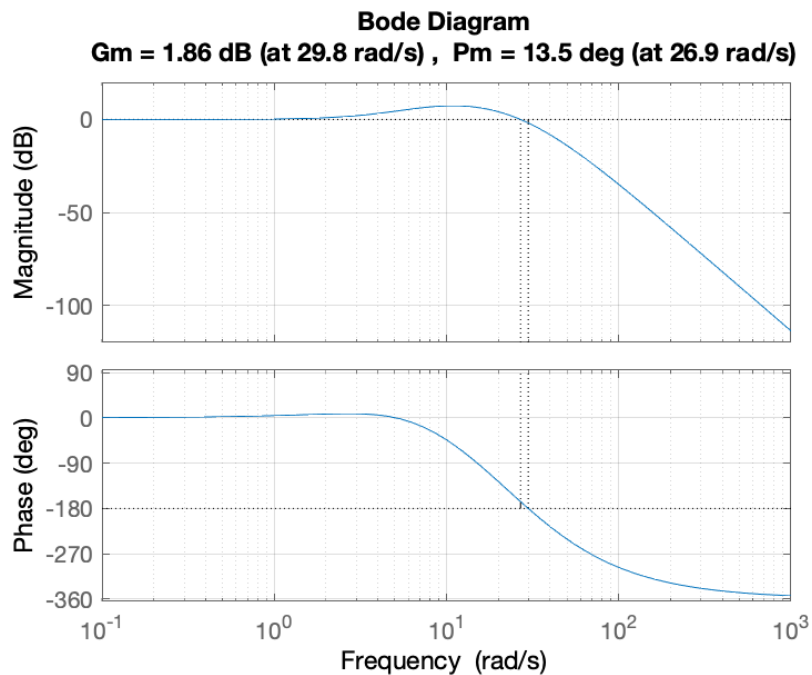


Figure 1-2. Bode plot for the closed-loop system designed in Section 1.1

From Figure 1-2, both gain margin (1.86 [dB]) and phase margin (13.5 [deg]) are extremely small, hence the sys-

tem is quite sensitive to noise and disturbances. To improve the transients and phase margin, we can adjust the parameter $Q(s)$ such that the desired criterion is met. Based on the research, many system utilizes Youla Parameterization with stochastic gradient descent, graph theory, and genetic algorithm to improve the transient performance by manipulating the $Q(s)$. Since, any proper transfer function of $Q(s)$ would result a stable controller from the Youla formulation, hence, all we need to do is to adaptively find the optimal solution via various adaptive and un-supervised learning approach.

1.3 (c) [Optional] Simulation Visualization

The performance of the controller is visualized with the provided simulation, as shown in Figure 1-3 below.

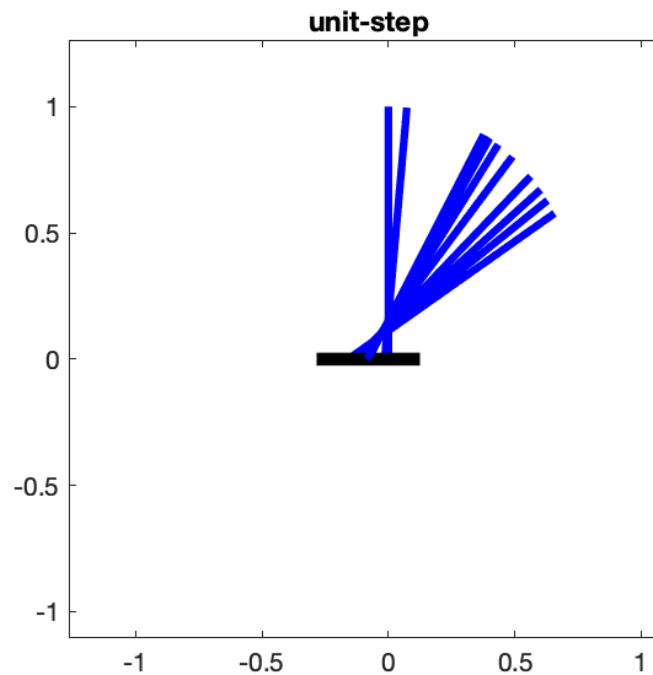


Figure 1-3. Final simulation result of a unit-step response for the closed-loop system designed in Section 1.1

2 Problem 4: Performance limitations associated with the SISO aiming system

2.1 (a) Performance limitations of one-rod system

2.1.1 (i) Overshoot y_{os} bound

For a unit-step of 0.5 [rad/s] , the overshoot bound can be formulated in Equation (8), and its corresponding bounding curve can be seen in Figure 2-1 below.

$$y_{os} \geq (1 - 0.9y_{\infty})(e^{pt_r} - 1) > 0 \quad (8)$$

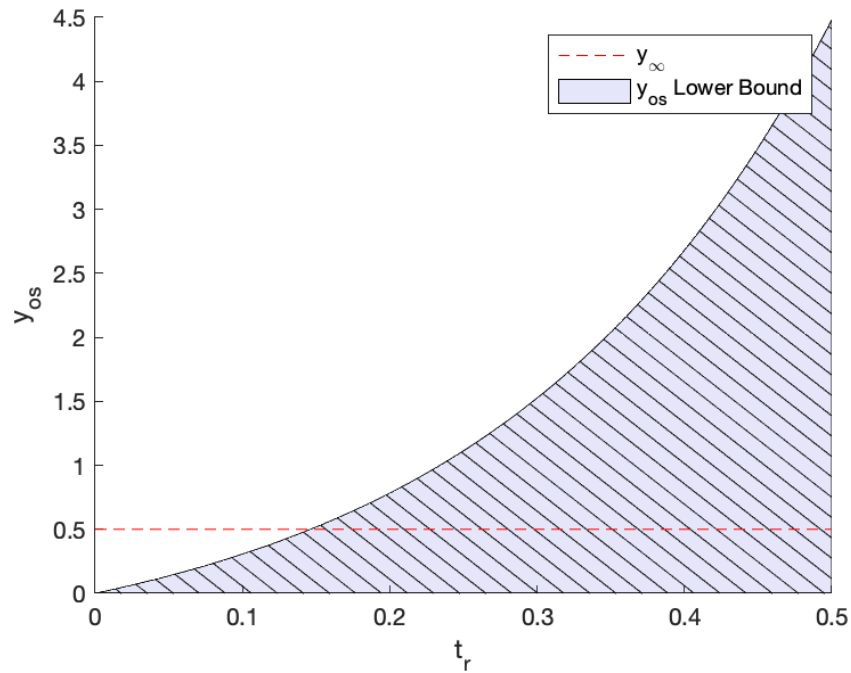


Figure 2-1. Overshoot y_{os} bound

2.1.2 (ii) Ω bound

Three additional design specifications are imposed:

- I For tracking performance, $|S(j\omega)| \leq -20 [\text{dB}]$ for $\omega \in [0, 0.1] [\text{rad/s}]$
- II For robust stability, $\max_{\omega} |S(j\omega)| \leq 5 [\text{dB}]$
- III "Shut off" $\forall \omega > \Omega [\text{rad/s}]$

Since the controller stabilizes the closed loop system, and the loop gain has a relative degree of at least 2, the BSI (Bode Sensitivity Integral) holds. We may now derive the *Omega* boundary (Equation (15)) from the BSI inequality for the sensitivity function, as derived below:

$$\int_0^{\infty} \ln |S(j\omega)| d\omega = \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) \quad (9)$$

$$\text{Spec III \& Spec I} \Rightarrow \int_0^{0.1} \ln |S(j\omega)| d\omega + \int_{0.1}^{\Omega} \ln |S(j\omega)| d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) \quad (10)$$

$$\text{Spec I} \Rightarrow \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \int_0^{0.1} 1 d\omega + \int_{0.1}^{\Omega} \ln |S(j\omega)| d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) \quad (11)$$

$$\int_{0.1}^{\Omega} \ln |S(j\omega)| d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) - 0.1 \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \quad (12)$$

$$\text{Spec II} \Rightarrow \ln \left\{ \max_{\omega \in [0.1, \Omega]} |S(j\omega)| \right\} \int_{0.1}^{\Omega} d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) - 0.1 \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \quad (13)$$

$$\ln \left\{ \max_{\omega \in [0.1, \Omega]} |S(j\omega)| \right\} (\Omega - 0.1) \geq \pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) - 0.1 \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \quad (14)$$

$$\Omega \geq \frac{\pi \cdot \sum_{i=1}^{N_p} \text{Re}(p_i) - 0.1 \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\}}{\ln \left\{ \max_{\omega \in [0.1, \Omega]} |S(j\omega)| \right\}} + 0.1 \quad (15)$$

Now, we may solve the derived boundary equation as stated in Equation (15) numerically:

$$\Omega \geq \frac{\pi * 4.427 - 0.1 * \ln(0.1)}{\ln(1.7783)} + 0.1 = 24.66 [\text{rad/s}] \quad (16)$$

$$\therefore \Omega \geq 24.66 [\text{rad/s}] \quad (17)$$

2.1.3 (iii) Ω bound if there were no unstable plant pole

Now, let's pretend there is no unstable plant pole, as a result the Ω bound is much smaller than the one in part (ii). This is because there is a need for extra increase in sensitivity as a cost of having to stabilize the unstable poles in the plant, resulting the positive area in high frequency regions for sensitivity much larger than the negative area. Consequently, the Ω is much larger for unstable pole to spread the large area over a wider range of frequencies, as the peak of the sensitivity is capped as Spec II.

Conversely, if there is no unstable plant pole, it only needs a smaller range of high frequencies to compensate the loss in low frequencies.

$$\Omega \geq \frac{0 - 0.1 * \ln(0.1)}{\ln(1.7783)} + 0.1 = 0.5 \text{ [rad/s]} \quad (18)$$

$$\therefore \Omega \geq 0.5 \text{ [rad/s]} \quad (19)$$

2.2 (b) Performance limitations of two-rod system

2.2.1 (i) Overshoot y_{os} bound

For the given two-rod plant below:

$$P_2(s) = \frac{110.4(s + 5.539)(s - 5.539)}{(s - 4.331)(s + 4.331)(s - 10.52)(s + 10.52)} \quad (20)$$

There are two unstable poles and one unstable zeros in the plant. We may believe the unstable pole further into the ORHP is more sever than the pole closer to the origin. The unstable pole would result overshoot, and the unstable zero would result undershoot. In addition, it is believe that the current system is extremely undesirable as the severe unstable pole is on the right side of the unstable zero (as seen in Figure 2-2 below), resulting the system hard to stabilize nicely.

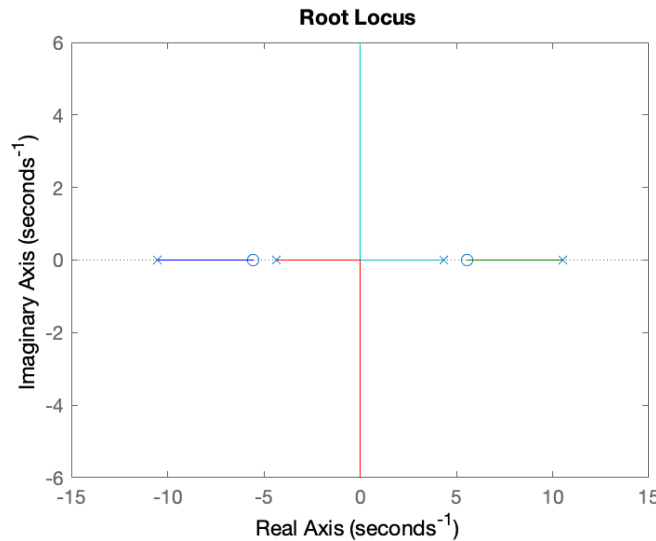


Figure 2-2. Root locus of the two-rod plant

Similar to Section 2.1.1, we would compute the overshoot y_{os} bound curve based on the most sever pole at the far right side ($s = 10.52$). After laying over against the one-rod system (in blue), as shown in Figure 2-3, we may see the overshoot bound for two-rod plant is much sever than the one-rod. As a result the two-rod system needs a much faster system to control.

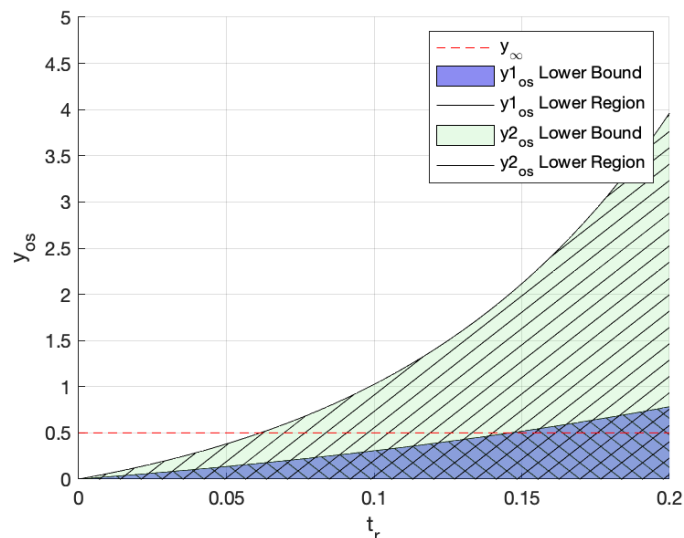


Figure 2-3. Overshoot bound curve ($y1_{os}$: one-rod, $y2_{os}$: two-rod) vs. rise time

However, it is also not ideal to have a too fast system, since, the unstable zero needs a slower system to minimize the undershoot, as shown in Figure 2-4.

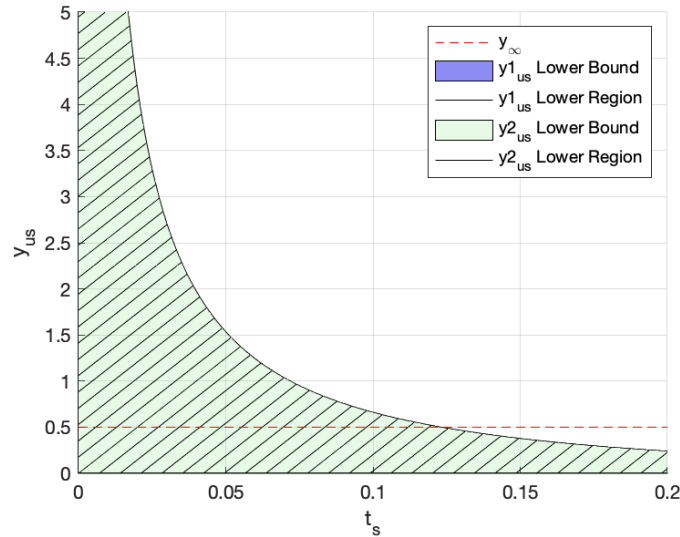


Figure 2-4. Undershoot bound curve ($y1_{us} = 0$: one-rod, $y2_{us}$: two-rod) vs. settling time

As a result, the two-rod system is more difficult to control than the one-rod system, since the two-rod system would have to suffer either both severe undershoot and overshoot, whereas, the one-rod system can achieve simply by making the system faster.

Now, let's compute a meaningful bound on overshoot for the two-rod system using (1) on page 161. We may realize the severe pole results a negative bound which is meaningless, and the less severe pole results a positive bound of 3.5853.

$$y2_{os}^{severe-pole} = \frac{p}{z-p} = -2.112 \quad (21)$$

$$y2_{os}^{moderate-pole} = \frac{p}{z-p} = 3.5853 \quad (22)$$

Since the pg161 equation is only insightful for $p > z$, and in our case, we only know the fact that the transient performance is considerably pretty bad for the meaningful pole, since the pole and zero is close to each other. In addition, there exists a pole that is on right side of zero, resulting a negative overshoot, indicating, the situation is hopeless, and we would expect a pretty bad system. This bound tells us a combined effects of unstable zeros and poles, whereas the curve bounds tell us more insight about how the overshoot bound changes depending on rise time. Both bounds are telling us how bad this system is, but from two different perspectives.

2.2.2 (ii) Ω bound with BSI

Similar to Section 2.1.2, we may derive the lower bound on Ω from BSI, using the formulation of Equation (15). As a result, we get a lower bound of 81.55 [rad/s]. As expected, the effects of double unstable more severe poles result the system needs a much larger positive region to compensate the reduced sensitivity in the low frequency region.

$$\Omega \geq \frac{\pi * (4.3310 + 10.5200) - 0.1 * \ln(0.1)}{\ln(1.7783)} + 0.1 = 81.55 \text{ [rad/s]} \quad (23)$$

$$\therefore \Omega \geq 81.55 \text{ [rad/s]} \quad (24)$$

As a result, we may conclude, the two-rod system is definitely more difficult to control, since it requires increased width of high frequency regions with increasing sensitivity to stabilize the system nicely.

2.2.3 (iii) Poisson Integral

Since there exists a real ORHP zero at $s = 5.539$, hence the PoI holds, as a result we may derive a sensitivity boundary function.

Let's assume the max peak sensitivity in the positive region of the Poisson Integral is $M = \max_{0.1 < \omega < \Omega} |S(j\omega)|$. For simplicity, we let $\varepsilon = \max_{\omega \in [0, 0.1]} |S(j\omega)|$.

We may now derive an relationship of the M relative to the choice of attenuation frequency Ω :

$$\int_0^\infty \ln |S(j\omega)| W(\omega) d\omega = \pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| \quad (25)$$

$$\text{Spec III \& Spec I} \Rightarrow \int_0^{0.1} \ln |S(j\omega)| W(\omega) d\omega + \int_{0.1}^\Omega \ln |S(j\omega)| W(\omega) d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| \quad (26)$$

$$\text{Spec I} \Rightarrow \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \int_0^{0.1} W(\omega) d\omega + \ln \left\{ \max_{\omega \in [0, 0.1]} |S(j\omega)| \right\} \int_{0.1}^\Omega W(\omega) d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| \quad (27)$$

$$\ln(\varepsilon) \int_0^{0.1} W(\omega) d\omega + \ln(M) \int_{0.1}^\Omega W(\omega) d\omega \geq \pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| \quad (28)$$

$$\ln(\varepsilon) \left[2 \tan^{-1} \left(\frac{\omega}{z} \right) \right]_0^{0.1} + \ln(M) \left[2 \tan^{-1} \left(\frac{\omega}{z} \right) \right]_{0.1}^\Omega \geq \pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| \quad (29)$$

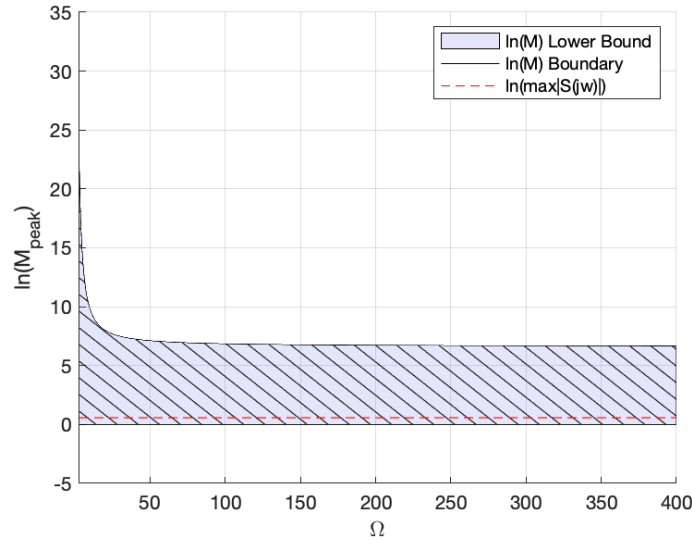
After rearrange:

$$\left[2 \tan^{-1} \left(\frac{\omega}{z} \right) \right]_{0.1}^\Omega \geq 0 \Rightarrow \ln(M) \geq \frac{\pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| - \ln(\varepsilon) \left[2 \tan^{-1} \left(\frac{\omega}{z} \right) \right]_0^{0.1}}{\left[2 \tan^{-1} \left(\frac{\omega}{z} \right) \right]_{0.1}^\Omega} \quad (30)$$

$$\ln(M) \geq \ln(M) \frac{\pi \cdot \sum_{i=1}^{N_p} \ln \left| \frac{p_i + z}{p_i^* - z} \right| - \ln(\varepsilon) \left[\tan^{-1} \left(\frac{0.1}{z} \right) \right]}{\left[\tan^{-1} \left(\frac{\Omega}{z} \right) \right] - \left[\tan^{-1} \left(\frac{0.1}{z} \right) \right]} \quad (31)$$

$$\ln(M) \geq \frac{10.266}{\tan^{-1}(0.18054\Omega) - 0.018052} \quad (32)$$

As a result, we can now plot the boundary equation Equation (32), and overlay the max peak requirement from Spec II:

Figure 2-5. M peak bound vs. Ω

From the plot in Figure 2-5, we may find that no matter how we increase the attenuating frequency Ω , the peak of the high frequency region would exceed the max peak sensitivity requirement (Spec II). As Equation (33) suggested, we may find the fact that $\ln(M(\Omega)) \gg \ln(\epsilon) \forall \Omega$.

$$\ln(M)_{\Omega \rightarrow \infty} \geq \frac{10.266}{\tan^{-1}(0.18054(\Omega \rightarrow \infty)) - 0.018052} = \frac{10.266}{\pi/2 - 0.018052} = 6.6115 \gg \ln(\epsilon) = 0.5756 \quad (33)$$

In short, we can conclude it is impossible to satisfy the three design specifications given in Spec I, Spec II, and Spec III.

Glossary

BSI Bode Sensitivity Integral.

LHP Left Hand Plane.

ORHP Open Right Hand Plane.

PoI Poisson Integral.

RHP Right Hand Plane.

Appendix A Code for Main

```

1 close all;
2 clear all;
3 clc;
4
5 %% init.
6 helper.createFolder("output", false);
7 helper.createFolder("output/p3", false);
8 helper.createFolder("output/p4", false);
9 disp("=== Output Folder Initd for P3 and P4 ===")
10
11 %% User Param:
12 ENV_P3 = true;
13 ENV_P4 = true;
14 EN_SIM = true;
15
16 % P3:
17 COPRIME_SCALE_FACTOR = 10;
18 P3_Q = 1;
19
20 % P4:
21
22
23 % MISC: Conditional Params [For Simulation]
24 FIG_SIZE = [400, 300]
25 Y_peak = 0.5;
26 T = 0.001;
27 T_END = 2; %[s]
28
29 %% TF : Common for P3 and P4
30 fprintf("=== Init MATLAB [Simulation:%d] ... \n", EN_SIM);
31 % tf
32 s = tf('s');
33 P1 = 8/(s-4.427)/(s+4.427);
34 P1w = -8/6/(s-4.427)/(s+4.427);
35 P2 = 110.4*(s+5.539)*(s-5.539)/(s-4.331)/(s+4.331)/(s-10.52)/(s+10.52);
36
37 % Generate Test Data
38 t = 0:T:T_END;
39 r_theta = Y_peak * ones(1, size(t,2)); % step @ 0.5 [rad/s]
40 % r_theta_sin_lf = Y_peak * sin((t ./ 2) * 2 * pi); % sin wave @ 0.5 [Hz]
41 % r_theta_sin_hf = Y_peak * sin((t .* 2) * 2 * pi); % sin wave @ 2 [Hz]
42 % r_theta_sin_hf2 = Y_peak * sin((t .* 10) * 2 * pi); % sin wave @ 10 [Hz]
43 % r_theta_sin_hf100 = Y_peak * sin((t .* 100) * 2 * pi); % sin wave @ 100 [Hz]
44
45
46 %% Problem 3 Solution %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
47 fprintf("=== Perform Computation [P3:%d] ... \n", ENV_P3);
48 if ENV_P3
49     % augment the plant with an integrator for perfect steady-state
50     % tracking:
51     P1_aug = P1 / s;
52     % fetch coprime:
53     [M, N, X, Y] = coprime(P1_aug, COPRIME_SCALE_FACTOR);
54     % report:
55     zpk(M)
56     zpk(N)

```

```

57 zpk(X)
58 zpk(Y)
59 % define Q(s)
60 Q = P3-Q;
61 % construct C_1-aug(s) for P_1-aug
62 C_1_aug = (X + M * Q)/(Y - N * Q);
63 C_1_aug = minreal(C_1_aug);
64 fprintf("\n\n>>> [C1-augmented]:");
65 zpk(C_1_aug)
66 % de-aug C_1-aug(s) to get C1(s) for P1(s)
67 C_1 = C_1_aug / s;
68 C_1 = minreal(C_1);
69 fprintf("\n\n>>> [C1]:");
70 zpk(C_1)
71
72 % compute TF:
73 L_1 = minreal(P1 * C_1);
74 TF_r2theta = minreal( 1 - 1 / (1 + L_1) ); % sensitivity
75 TF_r2w = minreal( Plw * (C_1) / (1 + L_1) );
76 % bode plot:
77 helper.bode_plot_margin(TF_r2theta, "bode_plot_r2theta", "p3")
78 helper.bode_plot_margin(TF_r2w, "bode_plot_r2w", "p3")
79 % Simulation
80 helper.simulation_and_plot(TF_r2theta, TF_r2w, r_theta, t, "unit-step", EN_SIM, "p3", true)
81 end
82
83 %% Problem 4 Solution %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
84 fprintf("=== Perform Computation [P4:%d] ... \n", ENV_P4);
85 if ENV_P4
86     %% a)
87     % i) Minimum bound on y-os vs. t-r :
88     y_inf = Y_peak;
89     p_ORHP = 4.427;
90     T_end4 = 0.5;
91     t_r = 0:T:T_end4;
92     y_os = (1 - 0.9 * y_inf) * (exp(p_ORHP * t_r) - 1);
93     % plot:
94     figure()
95     hold on;
96     plot([0, T_end4], [y_inf, y_inf], 'r--');
97     patch1 = patch([t_r T_end4 0], [y_os 0 0], [0.5,0.5,0.9]);
98     hatchfill(patch1, 'single', 'HatchAngle', -45, 'SpeckleWidth', 10);
99     alpha(patch1, .2)
100     plot([0, T_end4], [y_inf, y_inf], 'r--');
101     legend(["y-{\infty}", "y-{\os} Lower Bound"]);
102     xlabel("t-r");
103     ylabel("y-{\os}");
104     helper.saveFigure(FIG_SIZE, "p4", "y-os-bound");
105
106     % ii)
107     Omega2 = (pi * 4.427 - 0.1 * log(0.1))/log(1.7783) + 0.1
108
109     % iii)
110     Omega3 = (0 - 0.1 * log(0.1))/log(1.7783) + 0.1
111     %% b) Two-rods System
112     % i) Minimum bound on y-os and y-us vs. t-r :
113     p2_ORHP = [4.331, 10.52];
114     z2_ORHP = 5.539;
115
116     t_s = t_r;
117     y2_os = (1 - 0.9 * y_inf) * (exp(max(p2_ORHP) * t_r) - 1);
118     y2_us = (0.98 * y_inf) ./ (exp(z2_ORHP * t_s) - 1);
119
120     % plot os:
121     figure()
122     hold on;
123     grid on;
124     plot([0, T_end4], [y_inf, y_inf], 'r--');
125     patch1 = patch([t_r T_end4 0], [y_os 0 0], [0.1,0.1,0.9]);
126     patch2 = patch([t_r T_end4 0], [y2_os 0 0], [0.5,0.9,0.5]);
127     h1 = hatchfill(patch1, 'single', 'HatchAngle', -45, 'SpeckleWidth', 10);
128     h2 = hatchfill(patch2, 'single', 'HatchAngle', 45, 'SpeckleWidth', 10);

```

```

129 alpha(patch1,.5)
130 alpha(patch2,.2)
131 plot([0, T_end4], [y_inf, y_inf], 'r--');
132 legend(["y-\infty", "y1-{os} Lower Bound", "y1-{os} Lower Region", ...
133        "y2-{os} Lower Bound", "y2-{os} Lower Region"]);
134 xlabel("t_r");
135 ylabel("y-{os}");
136 xlim([0,0.2]);
137 ylim([0,5]);
138 helper.saveFigure(FIG_SIZE, "p4", "y2-os-bound");
139
140 % plot us:
141 figure()
142 hold on;
143 grid on;
144 plot([0, T_end4], [y_inf, y_inf], 'r--');
145 patch1 = patch([0 T_end4 T_end4 0], [0 0 0 0], [0.1,0.1,0.9]);
146 patch2 = patch([0 t_s(10:length(t_s)) T_end4 0], [5 y2_us(10:length(t_s)) 0 0], [0.5,0.9,0.5]);
147 h1 = hatchfill(patch1, 'single', 'HatchAngle', -45, 'SpeckleWidth', 10);
148 h2 = hatchfill(patch2, 'single', 'HatchAngle', 45, 'SpeckleWidth', 10);
149 alpha(patch1,.5)
150 alpha(patch2,.2)
151 plot([0, T_end4], [y_inf, y_inf], 'r--');
152 legend(["y-\infty", "y1-{us} Lower Bound", "y1-{us} Lower Region", ...
153        "y2-{us} Lower Bound", "y2-{us} Lower Region"]);
154 xlabel("t_s");
155 ylabel("y-{us}");
156 xlim([0,0.2]);
157 ylim([0,5]);
158 helper.saveFigure(FIG_SIZE, "p4", "y2-us-bound");
159
160 % compute bound with 161:
161 y2_us_min = max(z2_ORHP ./ (p2_ORHP - z2_ORHP))
162 y2_os_min = max(p2_ORHP ./ (z2_ORHP - p2_ORHP))
163 % root locus ? (not req.)
164 figure()
165 rlocus(P2)
166 helper.saveFigure(FIG_SIZE, "p4", "y2-root-locus");
167
168 disp("Yes, its more difficult, faster response for overshoot, and there exists undershoot")
169 %% ii) BSI => lower bound on \Omega
170 MAX_SENSITIVITY = db2mag(5);
171 omega_lower_bound_BSI = (pi * sum(p2_ORHP) - 0.1 * log(0.1)) / (log(MAX_SENSITIVITY)) + 0.1
172 %% iii) PI => lower bound on \Omega
173 Omega_max = 400;
174 Omega = 0.1:0.1:Omega_max;
175 z = z2_ORHP;
176 Const = pi * sum(log(abs((p2_ORHP + z2_ORHP) ./ (p2_ORHP - z2_ORHP))))
177 M_lower_bound_log = (((Const - log(MAX_SENSITIVITY) * atan(0.1/z))) ./ (atan(Omega / z) - atan(0.1/z)))
178 figure()
179 hold on;
180 grid on;
181 % plot([0.1, 0.1], [0, 1.8], 'r--');
182 patch1 = patch([Omega(20:length(Omega)) Omega_max 0], ...
183               [M_lower_bound_log(20:length(Omega)) 0 0], [0.5,0.5,0.9]);
184 hatchfill(patch1, 'single', 'HatchAngle', -45, 'SpeckleWidth', 10);
185 plot([0, Omega_max], [log(MAX_SENSITIVITY), log(MAX_SENSITIVITY)], 'r--');
186 alpha(patch1,.2)
187 legend(["ln(M) Lower Bound", "ln(M) Boundary", "ln(max|S(jw)|)"]);
188 xlabel("\Omega")
189 ylabel("ln(M-{peak})")
190 xlim([3, Omega_max])
191 helper.saveFigure(FIG_SIZE, "p4", "M-bound");
192 end

```

Code 1: Main Lab Contents

Appendix B Code for Helper Class

```
1 %% Helper Functions %%
```

```

2 classdef helper
3     methods(Static)
4         function createFolder(path, clear_folder)
5             if ~exist(path)
6                 mkdir(path)
7                 fprintf("[HELPER] Folder created!\n");
8             else
9                 if ~isempty(path) & clear_folder
10                     rmdir(path, 's');
11                     mkdir(path);
12                     fprintf("[HELPER] Folder is emptied, %s\n", path);
13                 else
14                     fprintf("[HELPER] Folder already existed!\n");
15                 end
16             end
17         end
18         function RH_criterion(coeffs) % [ n , .... , 0 ]
19             num = size(coeffs,2);
20             n = ceil(num / 2);
21
22             if mod(num,2) == 1 % odd number
23                 A = [coeffs, 0];
24             else
25                 A = coeffs;
26             end
27
28             RH_mat = reshape(A, 2, n);
29
30             for j = 1:n
31                 b = sym(zeros(1, n));
32                 for i = 1:n-1
33                     b(i) = RH_mat(j, 1) * RH_mat(j+1, i+1);
34                     b(i) = RH_mat(j+1, 1) * RH_mat(j, i+1) - b(i);
35                     b(i) = b(i)/RH_mat(j+1, 1);
36                     b(i) = simplifyFraction(b(i))
37                 end
38                 RH_mat = [RH_mat; b];
39             end
40             disp(RH_mat)
41         end
42         function saveFigure(DIMENSION, FOLDER, FILE_NAME)
43             set(gcf, 'units', 'points', 'position', [0, 0, DIMENSION(1), DIMENSION(2)]);
44             exportgraphics(gcf, sprintf('output/%s/%s.png', ...
45                 FOLDER, FILE_NAME), 'BackgroundColor', 'white');
46         end
47         function sisoPlot(L_TF, DIMENSION, FOLDER, TAG)
48             % Plot
49             figure()
50
51             subplot(2, 2, [1,3])
52             margin(L_TF)
53             grid on
54
55             subplot(2, 2, 2)
56             rlocus(L_TF)
57
58             subplot(2, 2, 4)
59             G_TF = minreal(L_TF/(L_TF + 1));
60             step(G_TF)
61             grid on
62
63             helper.saveFigure(DIMENSION, FOLDER, sprintf("siso_plot_%s", TAG))
64         end
65         %% Simulation
66         function simulation_and_plot(TF_r2theta, TF_r2w, r_theta, t, tag, ifsim, FOLDER, verbose)
67             fprintf("=== SIMULATION [%s:%s] ===\n", FOLDER, tag)
68             % sim
69             y_theta = lsim(TF_r2theta, r_theta, t);
70             y_w = lsim(TF_r2w, r_theta, t);
71
72             % analysis
73             rinfo = stepinfo(r_theta, t);

```



```

74     yinfo = stepinfo(y_theta,t);
75     t_delay = (yinfo.PeakTime - rinfo.PeakTime);
76     e_peak = (yinfo.Peak - rinfo.Peak);
77     os = e_peak/ rinfo.Peak * 100;
78     info_str = sprintf("[ %10s ]: t_{settling}= %.4f, \\theta_{settling}= %.4f, os_{peak}= %.2f
%%", ...
79         tag, yinfo.SettlingTime, y_theta(length(y_theta)), os)
80     if verbose
81         disp("r_theta info")
82         disp(rinfo)
83         disp("y_theta info")
84         disp(yinfo)
85     end
86
87     % Plot
88     figure()
89
90     subplot(3, 1, 1)
91     plot(t, r_theta)
92     grid on;
93     ylabel("r_{\\theta} [rad/s]")
94     title(info_str)
95
96     subplot(3, 1, 2)
97     hold on;
98     plot(t, y_theta)
99     plot(t, r_theta, '--', 'color', '#222222')
100    grid on;
101    ylabel("\\theta [rad/s]")
102    legend(["\\theta", "r_{\\theta}"])
103
104    subplot(3, 1, 3)
105    plot(t, y_w)
106    grid on;
107    ylabel("w [m]")
108    xlabel("t [s]")
109
110    helper.saveFigure([400, 300], FOLDER, sprintf("step_response_%s", tag))
111
112    % Simulate
113    if ifsim
114        figure()
115        single.pend_fancy_sim(t, [y_w, y_theta], [zeros(length(t),1) r_theta'], 1, 50, tag);
116        helper.saveFigure([300, 300], FOLDER, sprintf("sim_%s", tag))
117    end
118 end
119 %% Custom Bode
120 function bode_plot_custom(TF, tag, FOLDER, verbose)
121     fprintf("=== BODE PLOT [%s:%s] ===\\n", FOLDER, tag);
122     % custom bode plot
123     [mag, phase, wout] = bode(TF);
124     figure()
125     subplot(2,1,1)
126     semilogx(wout, mag2db(abs(mag(:))))
127     grid on
128     ylabel("Magnitude [dB]")
129     hold on
130     yline(0, 'r--')
131     legend(["W(s)", "0 dB"])
132
133     subplot(2,1,2)
134     semilogx(wout, phase(:))
135     grid on
136     ylabel("Phase [deg]")
137     xlabel("\\omega [rad/s]")
138
139     if verbose
140         m = allmargin(TF);
141         disp(m)
142         title(sprintf("BODE PLOT [%s:%s]", FOLDER, tag));
143     end
144

```

```
145     helper.saveFigure([400, 300], FOLDER, tag)
146 end
147 function bode_plot_margin(TF, tag, FOLDER)
148     fprintf('=== BODE PLOT [%s:%s] ===\n', FOLDER, tag);
149     % custom bode plot
150     figure()
151     margin(TF)
152     grid on
153     helper.saveFigure([400, 300], FOLDER, tag)
154 end
155 %% misc
156 function tflatex(TF)
157     [num, den] = tfdata(TF);
158     syms s
159     t_sym = poly2sym(cell2mat(num), s) / poly2sym(cell2mat(den), s);
160     latex(vpa(t_sym, 5))
161 end
162 end
163 end
```

Code 2: Helper and commonly used functions by main