

Robust H_∞ Pitch Control Applied to a Fixed-Wing Unmanned Aerial Vehicle

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Abstract

This project studies the design and evaluation of a robust H_∞ controller for the pitch-attitude dynamics of a fixed-wing unmanned aerial vehicle (UAV). The work is based on the paper “Robust H_∞ Control Applied on a Fixed Wing Unmanned Aerial Vehicle,” where robust longitudinal and lateral controllers are developed for a quarter-scale Piper J3-Cub UAV. In this project, the focus is restricted to the longitudinal elevator-to-pitch channel to reproduce the main robust control ideas in a clear and organized way. A linear longitudinal model is built from the state-space matrices reported in the paper, and parametric uncertainty is introduced for selected aerodynamic derivatives and control-effectiveness terms. A mixed-sensitivity H_∞ controller is then synthesized in MATLAB using weighting functions motivated by the paper. Simulation results show that the nominal closed-loop system is stable and achieves fast pitch tracking with no overshoot. Uncertain closed-loop simulations also remain stable across all sampled plants. Robust-stability analysis confirms that the controller preserves stability for the modeled uncertainty, although robust-performance analysis shows that performance is not guaranteed for the entire uncertainty set. Overall, the controller achieves strong robust stability and good nominal performance, but only limited robust performance under the selected uncertainty description.

1 Introduction

Fixed-wing unmanned aerial vehicles are widely used in surveillance, inspection, environmental monitoring, and defense-related applications. Because these systems operate under varying flight conditions and are affected by modeling errors, disturbances, and sensor noise, classical fixed-gain control methods may experience degraded performance when the actual plant departs from its nominal model. For this reason, robust control methods are attractive for UAV applications.

One important robust control framework is H_∞ control, which allows the designer to shape closed-loop performance while explicitly accounting for uncertainty and disturbances. In the selected reference paper, robust H_∞ controllers are developed for pitch, roll, and yaw autopilots of a fixed-wing UAV using a μ -synthesis-based design framework. The results in the paper show that robust control improves stability and tracking performance under perturbations, disturbances, and noise.

The purpose of this project is to reproduce and study part of that work in MATLAB. Instead of implementing the full longitudinal and lateral design, this project focuses only on the pitch-attitude control problem. This narrower scope allows the main modeling assumptions, uncertainty description, controller design steps, and simulation results to be presented clearly while still meeting the objective of the course project.

2 Problem Formulation

The goal of this project is to design a robust controller for the longitudinal pitch dynamics of a fixed-wing UAV. More specifically, the control input is the elevator deflection and the controlled output is the pitch angle. The design objectives are:

- stabilize the nominal longitudinal plant,
- achieve fast and well-damped pitch-angle tracking,
- maintain acceptable closed-loop behavior under plant uncertainty,
- reduce sensitivity to measurement noise and modeling errors.

The selected paper develops robust controllers for the full UAV attitude dynamics, but in this project only the pitch channel is considered. This corresponds to the elevator-to-pitch subsystem extracted from the longitudinal model. The problem is therefore treated as a single-input single-output robust control problem based on the reported linearized longitudinal dynamics.

3 Modeling and Uncertainty Description

The paper provides the linear longitudinal state-space model of the UAV around a trim condition. The state vector is taken as

$$x = [\Delta u \quad \Delta w \quad \Delta q \quad \Delta \theta]^T,$$

where Δu and Δw are perturbations in the body-axis velocity components, Δq is the pitch-rate perturbation, and $\Delta \theta$ is the pitch-angle perturbation.

The nominal longitudinal model used in this project is

$$\dot{x} = Ax + Bu, \quad y = Cx + Du \tag{1}$$

with

$$A = \begin{bmatrix} -0.06729 & 1.052 & 0 & -9.806 \\ -1.346 & -8.613 & 15 & 0 \\ 0.4848 & -8.606 & -20.95 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix},$$

$$B = \begin{bmatrix} 0 \\ -11.43 \\ -107.5 \\ 0 \end{bmatrix}, \quad C = [0 \ 0 \ 0 \ 1], \quad D = 0.$$

The system input u is the elevator deflection δ_e , and the output y is the pitch angle θ .

To account for model uncertainty, selected longitudinal derivatives were modeled as uncertain real parameters. Specifically, uncertainty was introduced in aerodynamic and control-effectiveness terms associated with the entries of the A and B matrices. A 20% uncertainty level was applied to the following nominal parameters:

$$X_u, X_w, Z_u, Z_w, M_u, M_w, M_q, Z_{\delta_e}, M_{\delta_e}.$$

This uncertainty model does not reproduce the exact uncertainty structure of the original paper, but it provides a reasonable parametric uncertainty description suitable for evaluating robust stability and robust performance in MATLAB. The uncertain plant therefore represents a family of longitudinal pitch-channel models around the nominal trim condition.

4 Robust Control Method

To design the controller, a mixed-sensitivity H_∞ formulation was used. The objective is to find a controller $K(s)$ that minimizes

$$\left\| \begin{bmatrix} W_1 S \\ W_2 K S \\ W_3 T \end{bmatrix} \right\|_\infty,$$

where

$$S = \frac{1}{1 + GK}, \quad T = \frac{GK}{1 + GK}.$$

Here, S is the sensitivity function and T is the complementary sensitivity function. The weighting functions were chosen based on the pitch-control case reported in the paper:

$$W_1(s) = \frac{0.1s + 5.973}{s + 0.176},$$

$$W_2(s) = \frac{0.8s + 42.91}{s + 119.2},$$

$$W_3(s) = 0.01.$$

The interpretation of these weights is as follows:

- W_1 penalizes tracking error and shapes the low-frequency performance,
- W_2 penalizes excessive control effort,
- W_3 limits high-frequency amplification and helps reduce noise sensitivity.

The controller was synthesized in MATLAB using the `mixsyn` command. This yields an H_∞ controller that balances tracking performance, control effort, and robustness.

The frequency-domain behavior of the nominal closed-loop transfer function $T(s)$ is shown in Fig. 1. The plot indicates that the closed-loop response has strong attenuation at high frequencies, which is desirable for limiting the effect of high-frequency uncertainty and measurement noise.

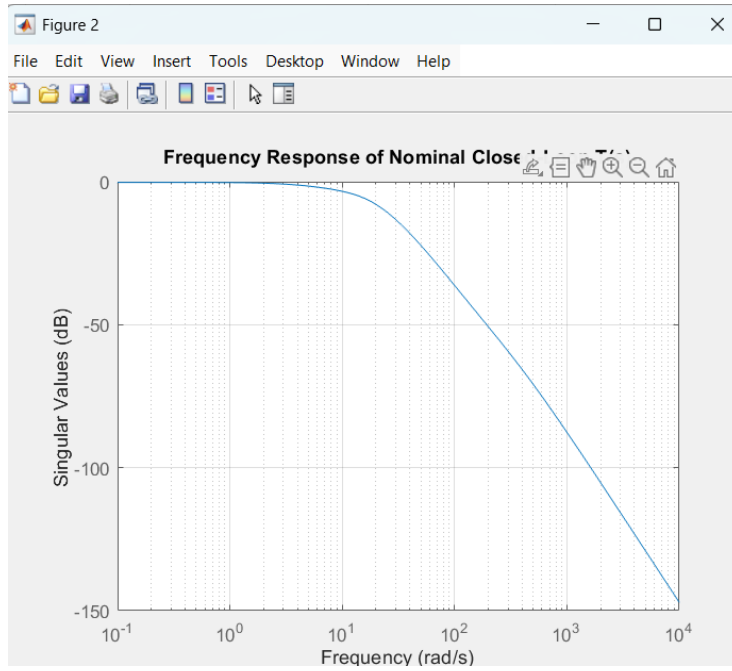


Figure 1: Frequency response of the nominal closed-loop transfer function $T(s)$.

The tradeoff between disturbance rejection and bandwidth can be examined more directly through the sensitivity and complementary sensitivity functions. Fig. 2 shows that $S(s)$ is small at low frequencies, which supports good tracking and disturbance rejection, while $T(s)$ rolls off at high frequencies, which helps attenuate noise and unmodeled dynamics.

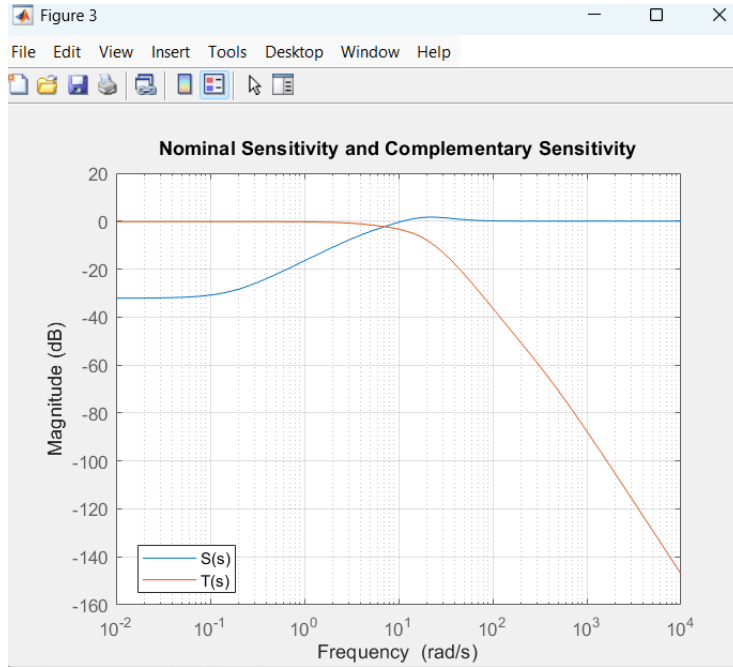


Figure 2: Nominal sensitivity and complementary sensitivity functions.

5 Results

The synthesized controller achieved a mixed-sensitivity value of

$$\gamma = 0.8784.$$

The nominal closed-loop poles were located in the open left-half plane, which confirms nominal closed-loop stability. The nominal step response is shown in Fig. 3. The response is fast, well-damped, and has no overshoot. The main time-domain characteristics were:

- rise time: 0.2398 s,
- settling time: 0.5776 s,
- overshoot: 0%,
- peak value: 0.9749.

These values indicate that the nominal controller produces fast tracking with no overshoot and good damping.

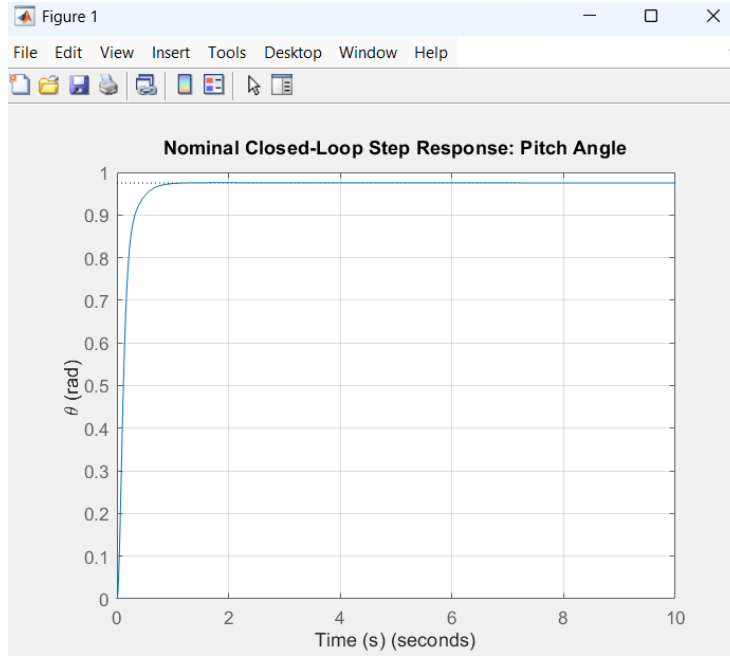


Figure 3: Nominal closed-loop step response for the pitch angle.

The sensitivity norms were also computed as

$$\|S\|_{\infty} = 1.2153, \quad \|T\|_{\infty} = 0.9752.$$

These results indicate acceptable closed-loop robustness and limited amplification of the complementary sensitivity response.

To evaluate uncertain performance, twelve samples of the uncertain plant were generated and closed with the same controller. The corresponding responses are shown in Fig. 4. All twelve uncertain closed-loop systems remained stable, and their trajectories remained close to the nominal response. This indicates that the controller provides good robustness with respect to the modeled parameter variations.

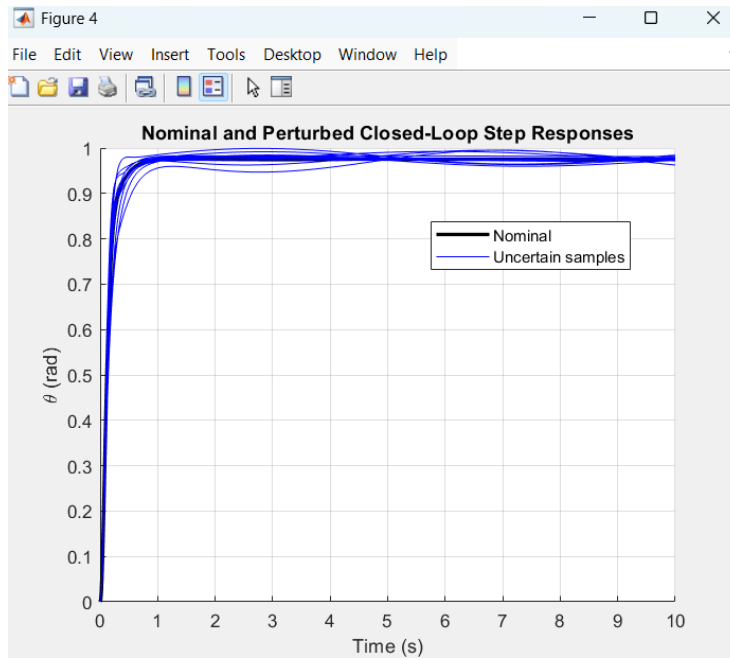


Figure 4: Nominal and perturbed closed-loop step responses for the pitch angle.

The corresponding control effort for a unit step pitch reference is shown in Fig. 5. The elevator command exhibits a strong initial transient and then decays smoothly toward its steady value. This confirms that the controller achieves good tracking without requiring unbounded actuator action.

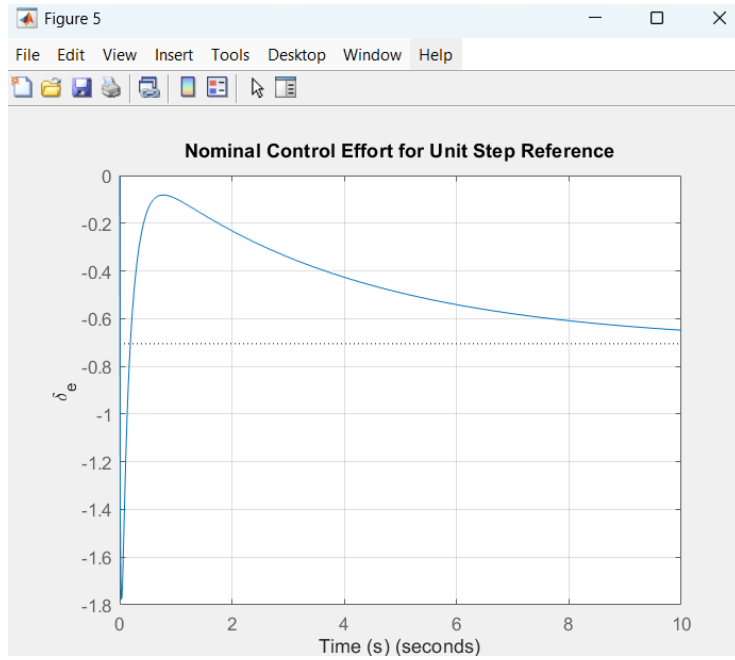


Figure 5: Nominal control effort for a unit step pitch-angle reference.

The relationship between reference tracking and measurement-noise transmission is shown in Fig. 6. At low frequencies, the closed loop preserves good reference tracking, while at high frequencies the transmission rolls off strongly, indicating strong attenuation of measurement noise.

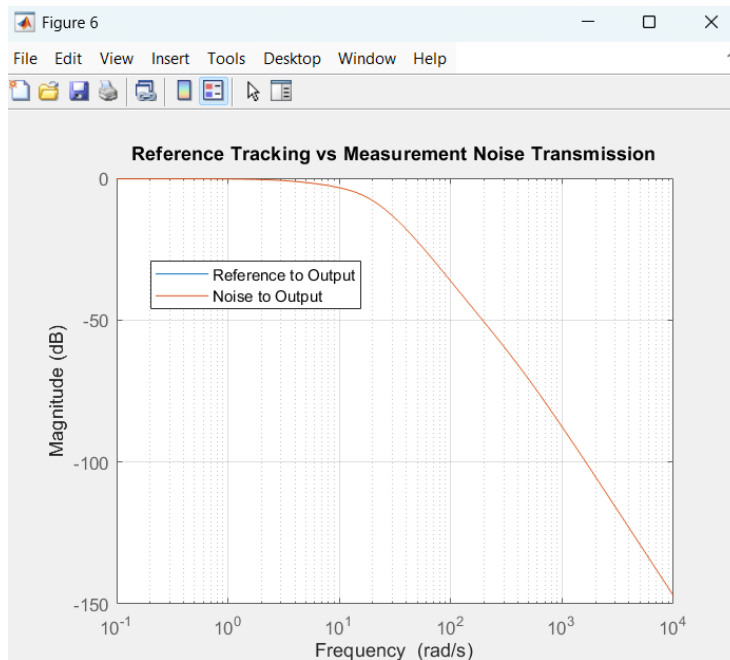


Figure 6: Reference tracking versus measurement-noise transmission.

A more formal robust-stability analysis was performed using `robuststab`. The result showed that the system is robustly stable for the modeled uncertainty and can tolerate approximately 196% of the uncertainty set before losing stability. The destabilizing perturbation was predicted near 0.682 rad/s. This is a strong indication that the controller provides a significant stability margin.

However, robust-performance analysis using `robustperf` showed that the uncertain system does not satisfy the desired performance objectives over the full uncertainty set. The result indicated that performance robustness is lost at about 79% of the modeled uncertainty. In other words, the system remains stable, but the closed-loop gain and tracking performance are not guaranteed to satisfy the chosen bounds for all uncertain plants.

6 Discussion

The results show a clear difference between robust stability and robust performance. From a stability point of view, the controller performs very well. The nominal plant is stable, all sampled uncertain plants are stable, and the robust-stability margin is greater than one. This means the closed-loop system preserves stability even when the uncertainty is significantly larger than the modeled level.

From a performance point of view, the result is more limited. While the

nominal and sampled step responses shown in Figs. 3 and 4 appear satisfactory, the formal robust-performance test indicates that the selected weighting functions and controller do not guarantee the desired performance over the entire uncertainty set. Therefore, there is a tradeoff between conservatism, uncertainty size, and achievable closed-loop performance.

The frequency-domain plots in Figs. 1, 2, and 6 are consistent with this interpretation. The controller provides good low-frequency tracking and strong high-frequency attenuation, but these advantages do not fully guarantee robust performance for every uncertain plant in the modeled family.

One limitation of this project is that the uncertainty model was simplified to parametric perturbations in selected state-space entries. Also, only the pitch channel was studied instead of the full longitudinal and lateral UAV dynamics considered in the original paper. Nevertheless, the project successfully reproduces the main robust control ideas of the reference work and demonstrates how MATLAB robust control tools can be used to evaluate both stability and performance.

7 Conclusion

This project investigated robust H_∞ pitch control for a fixed-wing UAV based on the paper “Robust H_∞ Control Applied on a Fixed Wing Unmanned Aerial Vehicle.” A longitudinal elevator-to-pitch model was implemented, parametric uncertainty was introduced, and a mixed-sensitivity H_∞ controller was synthesized in MATLAB.

The results showed that the controller achieves nominal closed-loop stability, fast pitch tracking, and no overshoot. Uncertain simulations further showed that the controller maintains stable behavior across all sampled plants. Robust stability analysis confirmed that the system remains stable for the modeled uncertainty with a substantial stability margin. However, robust performance analysis showed that performance objectives are not guaranteed for the entire uncertainty set.

Overall, the controller can be considered successful in achieving robust stability and good nominal behavior, but only partial robust performance. This highlights an important lesson in robust control: a system may remain stable under uncertainty even when its performance specifications are no longer fully satisfied.

Appendix: MATLAB Code

```
1 %% ECE 854 Final Project
2 % Option 2
3 %
4 % Julio Quiroga Galan
5 %
```

```

6 % Paper:
7 % "Robust H-infinity Control Applied on a Fixed Wing
  Unmanned Aerial Vehicle."
8 %
9 % Toolboxes required:
10 % - Control System Toolbox
11 % - Robust Control Toolbox
12
13 clear; clc; close all;
14
15 %% -----
16 % 1) Nominal longitudinal model from the paper
17 % States: x = [Delta u; Delta w; Delta q; Delta theta]
18 % Input used: elevator deflection delta_e
19 % Output used: pitch angle theta
20 %
21 % The model and weights are stated in the selected paper.
22
23 A_nom = [ -0.06729  1.052  0  -9.806;
24           -1.346  -8.613  15  0;
25           0.4848  -8.606  -20.95  0;
26           0  0  1  0 ];
27
28 B_nom_full = [ 0  3.6;
29               -11.43  0;
30               -107.5  0;
31               0  0 ];
32
33 B_elev = B_nom_full(:,1);
34 C_theta = [0 0 0 1];
35 D_theta = 0;
36
37 G_nom = ss(A_nom, B_elev, C_theta, D_theta);
38
39 disp('Nominal elevator-to-pitch plant:');
40 G_nom
41
42 %% -----
43 % 2) Uncertainty description
44 % The assignment requires an uncertain model or uncertainty
  description.
45 % Here we use uncertainty in selected entries of A and B.
46 %
47 % This is a reasonable course-project uncertainty model
  around the nominal
48 % plant and is consistent with the paper's robust control
  motivation under
49 % uncertainty, disturbances, and sensor noise.
50
51 Xu = ureal('Xu', -0.06729, 'Percentage', 20);

```

```

52 Xw = ureal('Xw', 1.052, 'Percentage', 20);
53
54 Zu = ureal('Zu', -1.346, 'Percentage', 20);
55 Zw = ureal('Zw', -8.613, 'Percentage', 20);
56
57 Mu = ureal('Mu', 0.4848, 'Percentage', 20);
58 Mw = ureal('Mw', -8.606, 'Percentage', 20);
59 Mq = ureal('Mq', -20.95, 'Percentage', 20);
60
61 Zde = ureal('Zde', -11.43, 'Percentage', 20);
62 Mde = ureal('Mde', -107.5, 'Percentage', 20);
63
64 A_unc = [ Xu    Xw    0    -9.806;
65           Zu    Zw    15    0;
66           Mu    Mw    Mq    0;
67           0     0     1     0 ];
68
69 B_unc = [ 0;
70           Zde;
71           Mde;
72           0 ];
73
74 G_unc = ss(A_unc, B_unc, C_theta, D_theta);
75
76 disp('Uncertain plant created.');
```

```

78 %% -----
79 % 3) Weighting functions
80 % From Table 4 in the paper for the pitch case:
81 % Perf_theta = (0.1 s + 5.973)/(s + 0.176)
82 % W_elev     = (0.8 s + 42.91)/(s + 119.2)
83 % N_theta    = 0.01
84 %
85 % These are used here as mixed-sensitivity weights.
86
87 s = tf('s');
```

```

88
89 W1 = (0.1*s + 5.973) / (s + 0.176);    % performance weight
90 W2 = (0.8*s + 42.91) / (s + 119.2);    % control effort
91 W3 = 0.01;                             % noise /
92                                         complementary sensitivity weight
93
94 disp('Weights defined.');
```

```

95 %% -----
96 % 4) H-infinity mixed-sensitivity design
97 % The assignment requires a robust control or robustness
98 % analysis method.
```

```

99 try
100     [K, CL, gamma_opt] = mixsyn(G_nom, W1, W2, W3);
101 catch ME
102     error(['mixsyn failed. Check that Robust Control Toolbox
103           is installed. Error: ', ME.message]);
104 end
105 disp('Controller K obtained. ');
106 K
107 fprintf('Achieved gamma = %.4f\n', gamma_opt);
108
109 %% -----
110 % 5) Closed-loop systems
111 L_nom = minreal(G_nom*K);
112 S_nom = minreal(feedback(1, L_nom));      % S = 1/(1+L)
113 T_nom = minreal(feedback(L_nom, 1));      % T = L/(1+L)
114 Tr_nom = T_nom;                          % reference to
115     output
116 disp('Nominal closed-loop poles: ');
117 disp(pole(Tr_nom));
118
119 if isstable(Tr_nom)
120     disp('Nominal closed-loop is stable. ');
121 else
122     disp('Warning: nominal closed-loop is NOT stable. ');
123 end
124
125 %% -----
126 % 6) Basic norms
127 try
128     gS = norm(S_nom, inf);
129     gT = norm(T_nom, inf);
130     fprintf('||S||_inf = %.4f\n', gS);
131     fprintf('||T||_inf = %.4f\n', gT);
132 catch
133     disp('Could not compute H-infinity norms exactly. ');
134 end
135
136 %% -----
137 % 7) Time vector
138 t = 0:0.01:10;
139
140 %% -----
141 % 8) Nominal step response
142 figure;
143 step(Tr_nom, t);
144 grid on;
145 title('Nominal Closed-Loop Step Response: Pitch Angle');
146 xlabel('Time (s)');

```

```

147 ylabel('\theta (rad)');
148
149 try
150     info_nom = stepinfo(Tr_nom);
151     disp('Nominal step response info:');
152     disp(info_nom);
153 catch
154     disp('stepinfo could not be computed.');
```

```

155 end
156
157 %% -----
158 % 9) Singular value / magnitude plot
159 % sigma may not always be ideal for SISO if you want a
160 % simple plot, but it runs.
161 figure;
162 sigma(T_nom);
163 grid on;
164 title('Frequency Response of Nominal Closed-Loop T(s)');
```

```

165 %% -----
166 % 10) Bode magnitude of sensitivity functions
167 figure;
168 bodemag(S_nom, T_nom);
169 grid on;
170 legend('S(s)', 'T(s)', 'Location', 'best');
171 title('Nominal Sensitivity and Complementary Sensitivity');
```

```

172 %% -----
173 % 11) Perturbed / uncertain closed-loop step responses
174 N = 12;
175 stable_count = 0;
176 infos_unc = cell(N,1);
177
178 try
179     G_samples = usample(G_unc, N);
180 catch ME
181     error(['usample failed. Error: ', ME.message]);
182 end
183
184 figure;
185 hold on;
186 grid on;
187
188 % Plot nominal first
189 [y_nom, t_nom] = step(Tr_nom, t);
190 plot(t_nom, squeeze(y_nom), 'k', 'LineWidth', 2);
191
192 for k = 1:N
193     Gk = G_samples(:, :, k);
194     Tk = minreal(feedback(Gk*K, 1));
```

```

196
197     if isstable(Tk)
198         stable_count = stable_count + 1;
199
200         try
201             [y_k, t_k] = step(Tk, t);
202             plot(t_k, squeeze(y_k), 'b');
203             infos_unc{k} = stepinfo(Tk);
204         catch
205             disp(['Could not plot uncertain sample ',
206                 num2str(k), '.']);
207         end
208     else
209         disp(['Sample ', num2str(k), ' closed loop is
210             unstable.']);
211     end
212 end
213
214 title('Nominal and Perturbed Closed-Loop Step Responses');
215 xlabel('Time (s)');
216 ylabel('\theta (rad)');
217 legend('Nominal', 'Uncertain samples', 'Location', 'best');
218
219 fprintf('Stable uncertain closed-loop samples: %d out of %d\
220         n', stable_count, N);
221
222 %% -----
223 % 12) Control effort response
224 % u/r = K*S
225 U_over_R = minreal(K*S_nom);
226
227 figure;
228 step(U_over_R, t);
229 grid on;
230 title('Nominal Control Effort for Unit Step Reference');
231 xlabel('Time (s)');
232 ylabel('\delta_e');
233
234 %% -----
235 % 13) Reference tracking vs measurement-noise transmission
236 % For measurement noise entering at the sensor, y/n = -T
237 figure;
238 bodemag(Tr_nom, -T_nom);
239 grid on;
240 legend('Reference to Output', 'Noise to Output', 'Location', '
241         best');
242 title('Reference Tracking vs Measurement Noise Transmission'
243       );
244 %% -----

```

```

241 % 14) Robust stability check
242 % To avoid failures I used try catch
243 CL_unc = minreal(feedback(G_unc*K, 1));
244
245 try
246     [stabmarg, destabunc, report_stab] = robuststab(CL_unc);
247     disp('Robust stability report:');
248     disp(report_stab);
249 catch ME
250     disp('robuststab could not be completed on this system/
251         version. ');
252     disp(ME.message);
253 end
254 %% -----
255 % 15) Robust performance check
256 try
257     [perfmarg, worstunc, report_perf] = robustperf(CL_unc);
258     disp('Robust performance report:');
259     disp(report_perf);
260 catch ME
261     disp('robustperf could not be completed on this system/
262         version. ');
263     disp(ME.message);
264 end
265 %% -----
266 % 16) Summary printed in command window
267 fprintf('\n
268     -----\n');
269 fprintf('PROJECT SUMMARY\n');
270 fprintf('-----\n');
271 fprintf('Selected paper: Robust H-infinity Control Applied
272     on a Fixed Wing UAV\n');
273 fprintf('Channel used      : Elevator to pitch angle\n');
274 fprintf('Method              : H-infinity mixed-sensitivity (
275     mixsyn)\n');
276 fprintf('Nominal stable : %d\n', isstable(Tr_nom));
277 fprintf('Gamma achieved : %.4f\n', gamma_opt);
278 fprintf('Stable uncertain samples: %d / %d\n', stable_count,
279     N);
280 if exist('stabmarg','var')
281     try
282         fprintf('Robust stability lower bound: %.4f\n',
283             stabmarg.LowerBound);
284         fprintf('Robust stability upper bound: %.4f\n',
285             stabmarg.UpperBound);
286     catch

```

```
282     end
283 end
284
285 fprintf('-----\n\n');
```

Listing 1: MATLAB code for robust H_∞ pitch control of the fixed-wing UAV

References

- [1] C. Uyulan and M. T. Yavuz, "Robust H_∞ control applied on a fixed wing unmanned aerial vehicle," *Advances in Aircraft and Spacecraft Science*, vol. 6, no. 5, 2019.