

# MECH 6327 - Homework 5

Jonas Wagner

2021, May 3

## Contents

<b>1</b>	<b>Control Design via SDP</b>	<b>2</b>
1.1	$\mathcal{H}_2$ -norm . . . . .	2
1.2	$\mathcal{H}_\infty$ -norm . . . . .	2
<b>2</b>	<b>Model Predictive Control (MPC)</b>	<b>3</b>
2.1	LQR Implementation . . . . .	3
2.2	MPC Implementation . . . . .	5
2.2.1	$T_h = 10$ . . . . .	5
2.2.2	$T_h = 5$ . . . . .	5
2.2.3	$T - h = 3$ . . . . .	5
2.2.4	$T - h = 20$ . . . . .	5
<b>A</b>	<b>MATLAB Code:</b>	<b>10</b>
A.1	Problem 1 Code . . . . .	11
A.2	Problem 2 Code . . . . .	13

# 1 Control Design via SDP

Consider the linearized model of a Cessna Citation Aircraft (at 5000m, at speed 128.2 m/s)

$$\dot{x} = \begin{bmatrix} -1.2822 & 0 & 0.98 & 0 \\ 0 & 0 & 1 & 0 \\ -5.4293 & 0 & -1.8366 & 0 \\ -128.2 & 128.2 & 0 & 0 \end{bmatrix} x + \begin{bmatrix} -0.3 \\ 0 \\ -17 \\ 0 \end{bmatrix} u + \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} wy = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u \quad (1)$$

with control input  $u$  = elevator angle, states  $x_1$  = angle of attack,  $x_2$  = pitch angle,  $x_3$  = pitch rate, and  $x_4$  = altitude (all relative to nominal), performance output  $y$ , and pitch rate disturbance input  $w$ .

**Problem:** Use the SDP formulations discussed in class to compute the optimal state feedback controller to minimize both the  $\mathcal{H}_2$ - and  $\mathcal{H}_\infty$ -norm from disturbance input  $w$  to output  $y$ . Also provide the optimal value for the respective closed-loop system norms.

## 1.1 $\mathcal{H}_2$ -norm

The procedure outlined in class was implemented in CVX as seen in Appendix A.1, which resulted with the following results:

K\_H2 =

-44.6546    56.8903    1.7838    1.0000

Norm\_H2 =

0.5000

## 1.2 $\mathcal{H}_\infty$ -norm

The procedure outlined in class was implemented in CVX as seen in Appendix A.1, which resulted with the following results:

K\_Hinfy =

1.0e+05 \*

-0.9457    1.1643    0.0299    0.0281

Norm\_Hinfy =

0.1330

## 2 Model Predictive Control (MPC)

Suppose the elevator angle input is limited to  $\pm 0.262$  rad ( $\pm 15$  degrees), the elevator angle rate is limited to  $\pm 0.524$  rad/s ( $\pm 30$  degrees/s), and we would like to limit the pitch angle to  $\pm 0.349$  rad ( $\pm 20$  degrees). Consider the time discretized system with sampling period  $dt = 0.25$ s, and discrete-time dynamics matrices  $A = e^{A_c dt}$ ,  $B = dt B_c + \frac{1}{2} dt A_c B_c + \frac{1}{6} dt^2 A_c^2 B_c$ , where  $A_c$  and  $B_c$  are the above continuous time state space matrices.

### 2.1 LQR Implementation

**Problem:** Compute the optimal infinite-horizon LQR controller by solving the discrete-time algebraic Riccati equation with  $Q = I$  and  $R = 10$ . Simulate the closed-loop system from initial state  $x_0 = \begin{bmatrix} 0 & 0 & 0 & 10 \end{bmatrix}^T$  but saturate the LQR controller to the input (magnitude and rate) constraints.

**Solution:** The calculation of an LQR controller gain was computed in MATLAB (Appendix A.2) and resulted in the following controller gain:

K\_LQR =

2.6795    -3.6639    -0.1890    -0.0447

This was then simulated with the required saturation constraints as can be seen in Figure 1.

It is clear that the LQR controller had difficulty stabilizing the system given the saturation requirements on the elevator position and rate. It is not only clear that the control signal was saturated and limited by the maximum rate requirements, but that state  $x_2$  is well outside of the desired bounds. (It is also just obvious that the system is unstable and will likely continue with increasing magnitudes periodic movements)

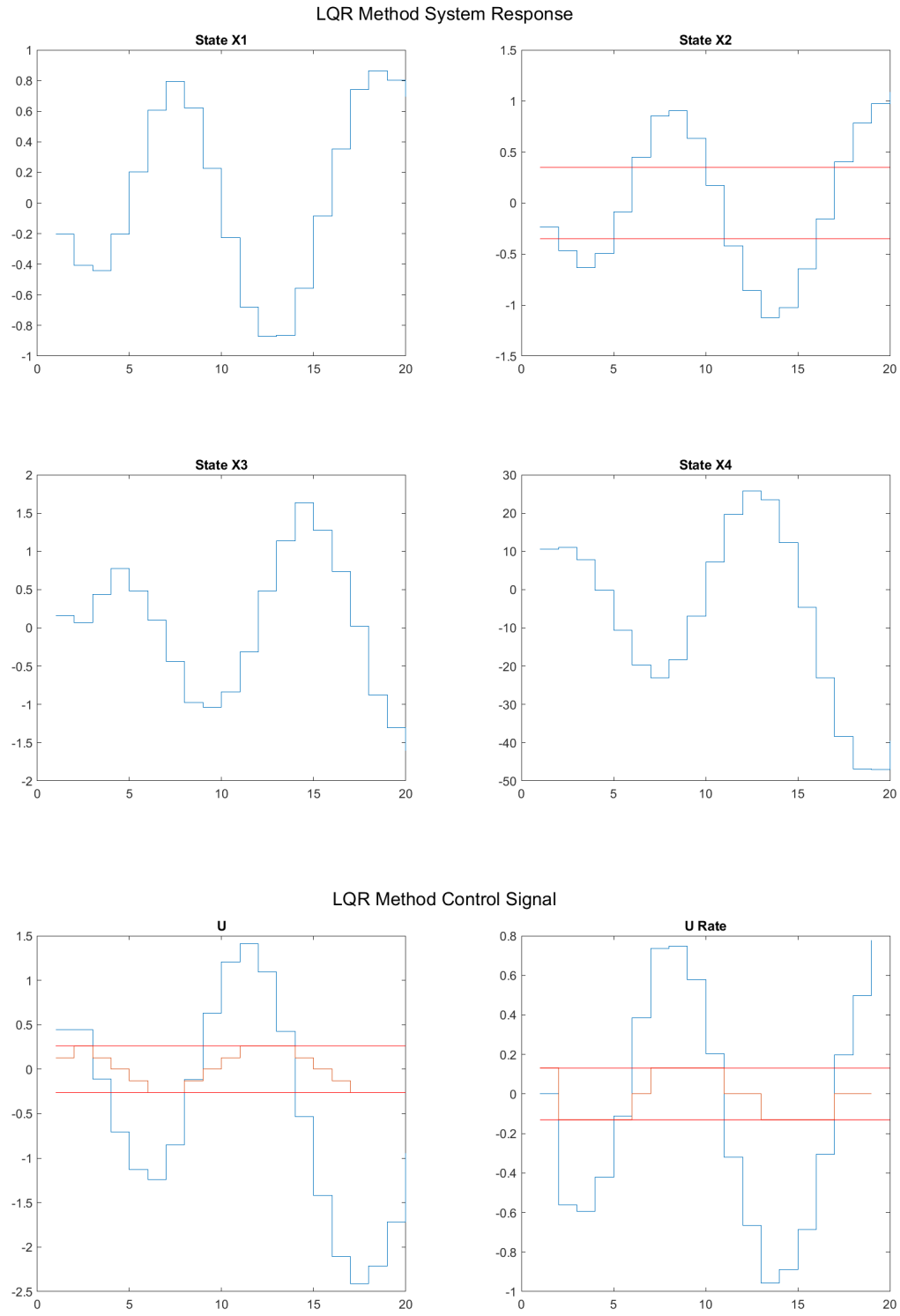


Figure 1: System response and control signal for the implementation of an infinite time LQR controller.

## 2.2 MPC Implementation

**Problem:** Implement a model predictive controller with horizon  $T_h = 10$  sample periods and constant quadratic cost parameters  $Q = I$  and  $R = 10$  that explicitly accounts for the elevator angle and pitch constraints. Simulate the closed-loop system from initial state  $x_0 = \begin{bmatrix} 0 & 0 & 0 & 10 \end{bmatrix}^T$  and compare to the LQR controller. Determine using your implementation how short the MPC planning horizon can be reduced before stability problems arise.

**Solution:** The MPC was implimentated in MATLAB (Appendix A.2) and was then tested for multiple time horizons.

### 2.2.1 $T_h = 10$

In the original implementation (Figure 2), the controller is very effective (especially in comparison to the LQR controller). It is clear that the MPC implementation was able to intrinsically limit itself to the control limitations while also stabilizing the system effectively and within the desired timeframe while also staying within the ranges of  $x_2$  dictated by the design goals.

### 2.2.2 $T_h = 5$

The MPC implementation for  $T_h = 5$  (Figure 3) was also effective. Compared to the longer time-horizon implementations, the system appears to have slightly larger control-signal magnitudes as well as a faster, yet more volatile, response for the individual state responses.

### 2.2.3 $T - h = 3$

The MPC implementation for  $T_h = 3$  (Figure 4) the system no longer stabalizable. Compared to the longer time-horizon implementations, the system has a troublesome appearance of uncontrolled ossilations that untimely cause state  $x_2$  to reach outside of the desired bounds.

### 2.2.4 $T - h = 20$

A very different result occurs when the time-horizon is increased, as opposed to decreased. The MPC implementation for  $T_h = 20$  (Figure 5) is even more effective then the original  $T_h = 10$  implementation. All of the states quickly stabalized and are then maintained with very little control effort.

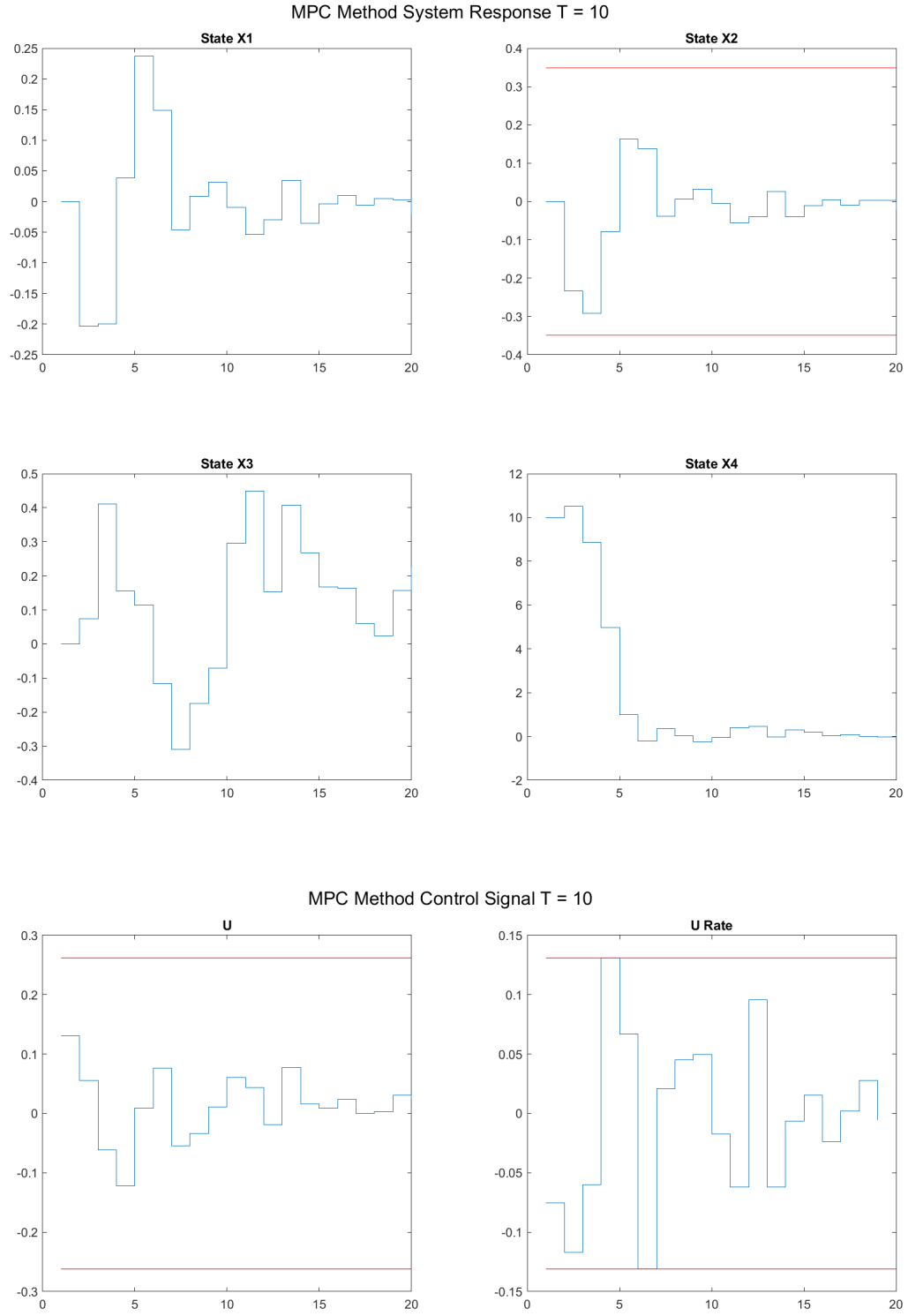


Figure 2: System response and control signal for the implementation of an MPC with  $T_h = 10$ .

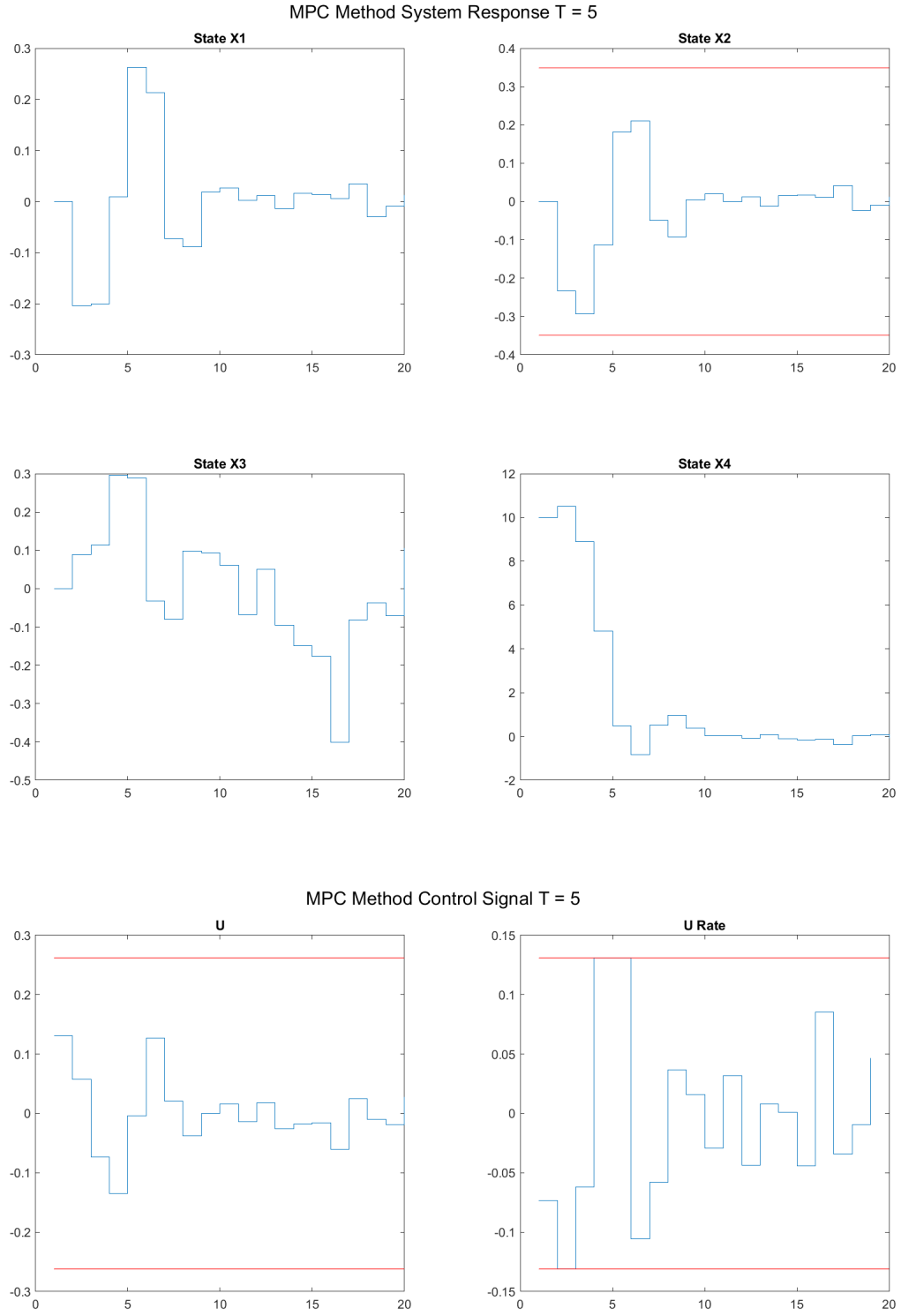


Figure 3: System response and control signal for the implementation of an MPC with  $T_h = 5$ .

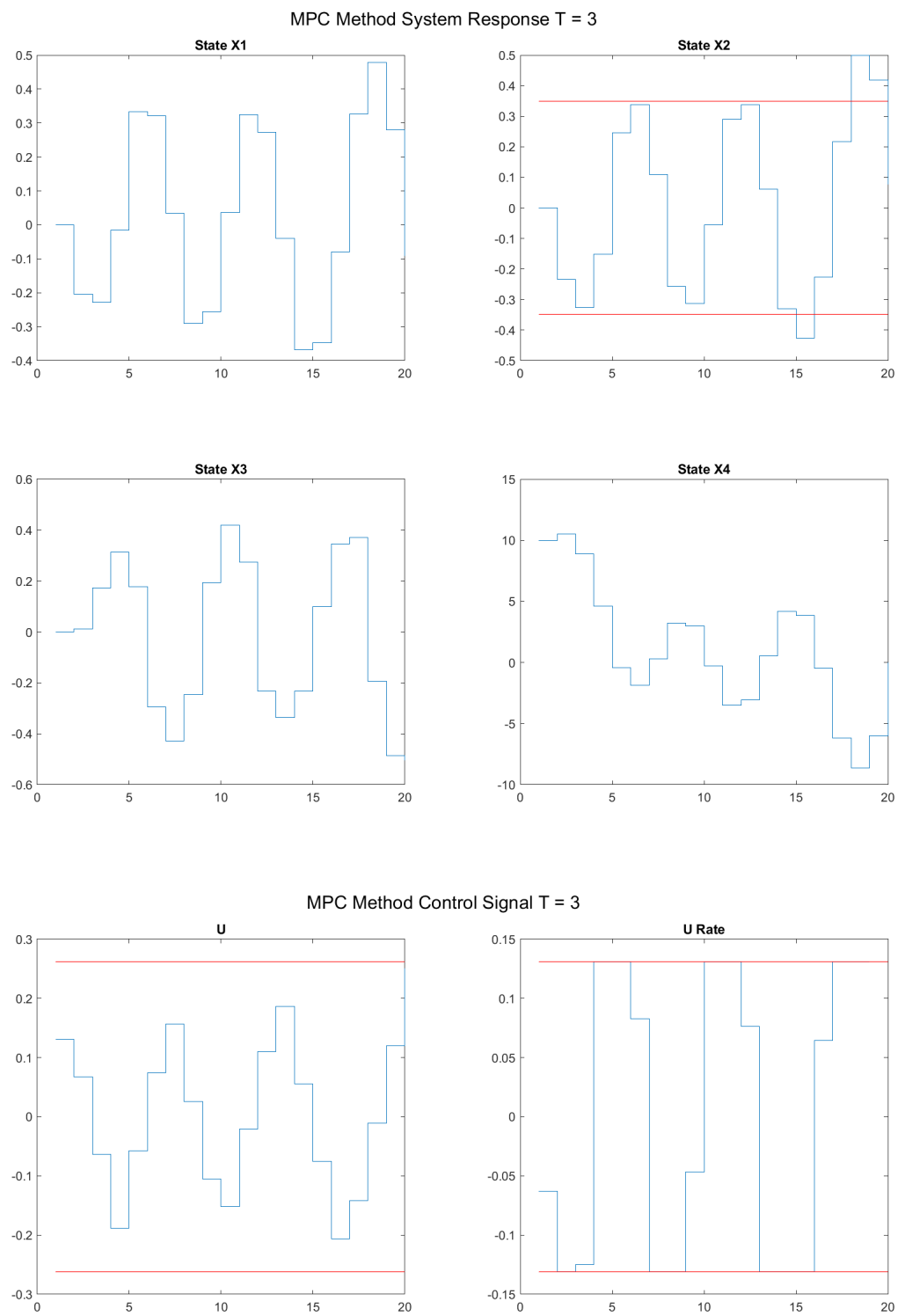


Figure 4: System response and control signal for the implementation of an MPC with  $T_h = 3$ .



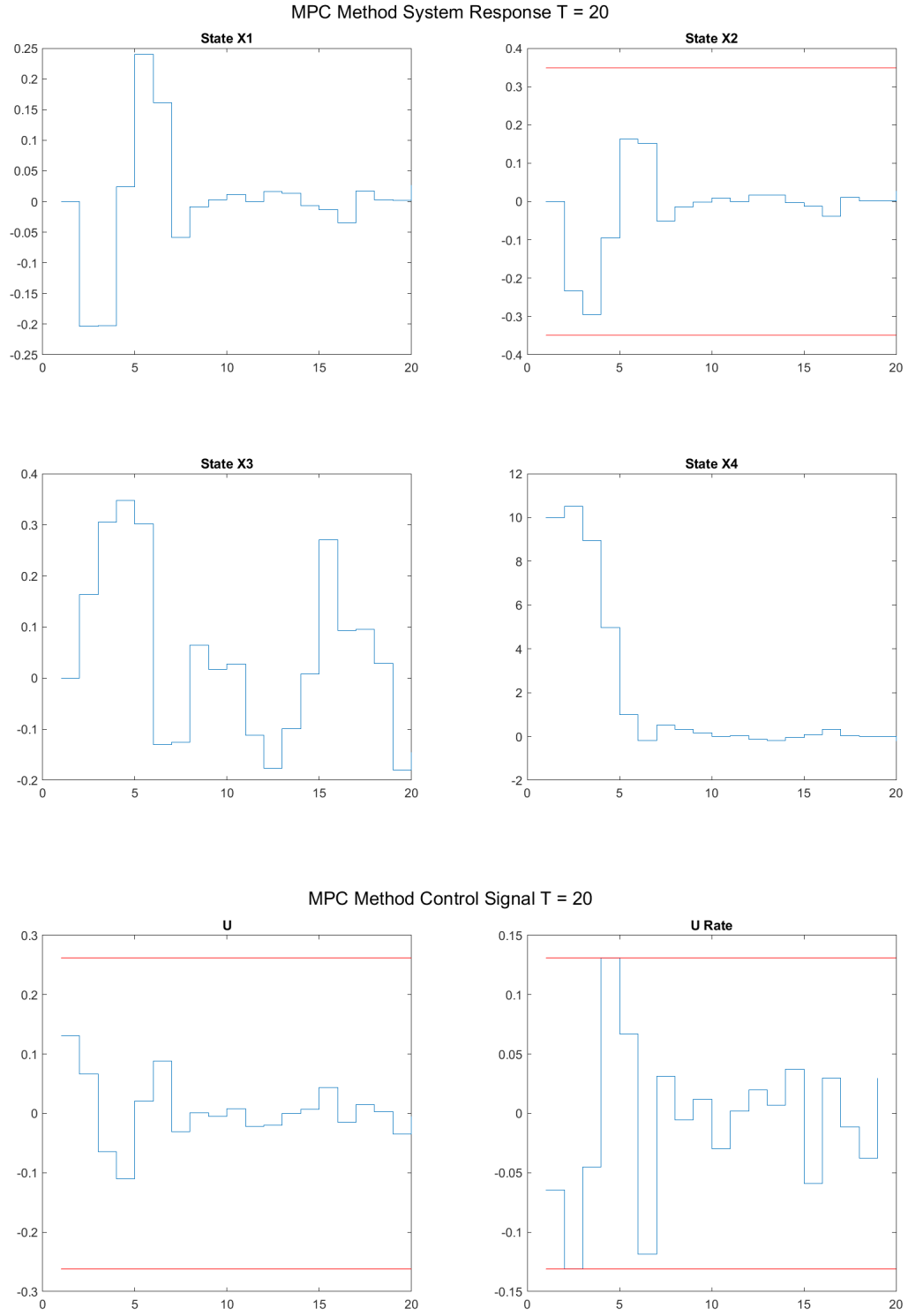


Figure 5: System response and control signal for the implementation of an MPC with  $T_h = 20$ .

## A MATLAB Code:

All code I write in this course can be found on my GitHub repository:

<https://github.com/jonaswagner2826/MECH6327>

Script 1: MECH6327\_HW5

```
1 % MECH 6327 - Homework 5
2 % Author: Jonas Wagner
3 % Date: 2020-05-02
4
5 clc;
6 clear
7 close all
8
9 % System Definition
10 MECH6327_HW5_sys_def
11
12 % Problem 1
13 MECH6327_HW5_pblm1
14
15 % Problem 2
16 MECH6327_HW5_pblm2
17
18 close all
```

## A.1 Problem 1 Code

Script 2: MECH6327\_HW5\_pblm1

```
1 % MECH 6327 - Homework 5 - Problem 1
2 % Author: Jonas Wagner
3 % Date: 2020-05-02
4
5 clear;
6 close all;
7
8 h2norm = true;
9 hinftynorm = true;
10
11 %% System Setup
12 MECH6327_HW5_sys_def
13 n = size(A,1);
14 p = size(B,2);
15 q = size(C,1);
16
17 if h2norm
18 %% H-2 Norm Control
19 tol = 1e-6;
20 cvx_begin sdp
21     variable X(n,n) symmetric
22     variable W(q,q) symmetric
23     variable L(1,n)
24     variable Gamma
25     minimize Gamma
26     subject to
27         [W, C*X+D*L;
28          X*C'+L'*D', X] >= 0
29         A*X + X*A' + B*L + L'*B' + F*F' <= 0
30         trace(W) <= Gamma - tol
31         X >= tol * eye(n)
32         W >= tol * eye(q)
33 cvx_end
34
35 K_H2 = L*inv(X)
36 Norm_H2 = sqrt(Gamma)
37 end
38
39
40 if hinftynorm
```

```

41 %% H-infty Norm Control
42 tol = 1e-6;
43 cvx_begin sdp
44     variable Q(n,n) symmetric
45     variable L(p,n)
46     variable eta
47
48     minimize eta
49     subject to
50         [Q*A'+L'*B'+A*Q+B*L F Q*C'+L'*D';
51          F' -eta zeros(p,q);
52          C*Q+D*L zeros(q,p) -eye(q)] <= -tol*eye(n+q+1)
53         Q >= tol*eye(n)
54 cvx_end
55
56
57 K_Hinfty = L*inv(Q)
58 Norm_Hinfty = sqrt(eta)
59 end

```

## A.2 Problem 2 Code

Script 3: MECH6327\_HW5\_pblm2

```
1 % MECH 6327 - Homework 5 - Problem 1
2 % Author: Jonas Wagner
3 % Date: 2020-05-02
4
5 clear;
6 close all;
7
8
9 lqr = true;
10 mpc = true;
11 runModels = true;
12 plotResults = true;
13
14 if runModels
15 %% System Setup
16 MECH6327_HW5_sys_def
17 n = size(A,1);
18 p = size(B,2);
19 q = size(C,1);
20
21 A_c = A;
22 B_c = B;
23 C_c = C;
24 D_c = D;
25 F_c = F;
26
27
28 % Discretization
29 dt = 0.25;
30 A = expm(A_c * dt);
31 B = dt*B_c + (1/2)*dt*A_c*B_c + (1/6)*(dt^2)*(A_c^2)*B_c;
32 C = C_c;
33 D = D_c;
34 sys = ss(A,B,C,D,dt)
35
36 % System Limitations
37 u_limit = 0.262; % rad (input limit)
38 u_rate_limit = 0.524 * dt; % rad/s (Input Change Max
39 x2_limit = 0.349; % rad (objective)
40
```

```

41 %% Simulation Setup
42 N = 20;
43
44 w_power = 0.1;
45 W = w_power * randn(size(F,2),N);
46
47 x0 = [0;0;0;10];
48
49 if lqr
50 %% LQR Control Design
51 % LQR Design Parameters
52 Q = eye(n);
53 R = 10;
54
55 % Feedback gain calculation
56 [~,K_LQR,~] = idare(A,B,Q,R) % Prints out K
57
58 %% Simulate LQR Methods
59 % Feedback Gain
60 K = K_LQR;
61 % Simulation
62 X = zeros(n,N); % States
63 U = zeros(p,N); % Inputs
64 Y = zeros(q,N); % Outputs
65 U_sat = U;
66 % Initialization
67 x = x0;
68 U(:,1) = - K*x;
69 U_sat(:,1) = min(u_rate_limit, max(-u_rate_limit, U(:,1)));
70 %Assuming U(:,0) = 0
71 X(:,1) = A * x + B * U_sat(:,1) + F * W(:,1);
72 Y(:,1) = C * x + D * U_sat(:,1);
73 for i = 2:N
74     % Control Calculation and saturation
75     U(:,i) = - K*x;
76     U_sat(:,i) = min(u_limit, max(-u_limit, U(:,i)));
77     %Apparently this isn't what the limitation meant...
78 % % Applied Control and Limitation
79 % Bu = B * U_sat(:,i);
80 % Bu(3) = min(x3_limit, max(-x3_limit, Bu(3)));
81 % Elevator Angle Rate Max
82 if (U_sat(:,i) - U_sat(:,i-1) >= u_rate_limit)
83     U_sat(:,i) = U_sat(:,i-1) + u_rate_limit;

```

```

84     elseif (U_sat(:,i) - U_sat(:,i-1) <= -u_rate_limit)
85         U_sat(:,i) = U_sat(:,i-1)-u_rate_limit;
86     end
87     % Time Update
88     x = A * x + B * U_sat(:,i) + F * W(:,i);
89     y = C * x + D * U_sat(:,i);
90     % Save Values
91     X(:,i) = x;
92     Y(:,i) = y;
93 end
94
95 % Save to LQR Values
96 X_LQR = X;
97 U_LQR = U;
98 U_sat_LQR = U_sat;
99
100 end
101
102 if mpc
103 %% MPC Control Design
104 % MPC Parameters
105 Q = eye(n);
106 R = 10;
107 T = 10; % Time Horizon
108 umax = u_limit;
109
110 % History Matrices
111 Xhist = zeros(n,N); % States
112 Uhist = zeros(p,N); % Inputs
113 Yhist = zeros(q,N); % Outputs
114 Jhist = zeros(1,N); % Stage costs
115
116 % Simulation setup
117 X = zeros(n,T);
118 U = zeros(p,T);
119 x = x0;
120
121 % Dynamics Matrices
122 G = zeros(n*T,n);
123 for i=1:T
124     G((i-1)*n+1:n*i,:) = A^i;
125 end
126

```

```

127 H = eye(n*T);
128 for i=1:T
129     for j=1:T
130         if i > j
131             H((i-1)*n+1:n*i,(j-1)*n+1:n*j) = A^(i-j);
132         end
133     end
134 end
135 BB = kron(eye(T),B);
136 H = H * BB;
137
138 % Simulation
139 tic;
140 disp('----- MPC Method: -----')
141 for i = 1:N
142     u_last = U(:,1); %assumes at 0 at start
143     disp(['Iteration: ', num2str(i)]);
144     % solve open-loop optimization problem
145     cvx_begin quiet
146         variable X(n,T) % predicted state trajectory
147         variable U(p,T) % planned control actions
148         minimize ((vec(X)' * kron(eye(T), Q)...
149                 * vec(X) + vec(U)'*kron(eye(T), R)...
150                 *vec(U)))
151         subject to
152             vec(X) == G * x + H * vec(U); % System Dynamics
153             norms(U,inf) <= umax; % Input Limitationsn
154             norm(U(:,1)- u_last ,inf) <= u_rate_limit;
155             for j = 2:T
156                 norm(U(:,j)-U(:,j-1),inf) <= u_rate_limit;
157             end
158     cvx_end
159
160 % Current Input and Output
161 u = U(:,1);
162 y = C * x + D * u;
163
164 % Store Data
165 Xhist(:,i) = x;
166 Uhist(:,i) = u;
167 Yhist(:,i) = y;
168 Jhist(i) = x'*Q*x+u'*R*u;
169

```



```

170     % State Update
171     x = A*x + B*u + F * W(:,i);
172 end
173 MPC_runtime = toc
174
175 X_MPC = Xhist;
176 U_MPC = Uhist;
177 Y_MPC = Yhist;
178 J_MPC = Jhist;
179
180 end
181 end
182
183
184
185 if plotResults
186 %% Plotting
187
188 if lqr
189 % Plot LQR Plots
190 % Plot MPC Plots
191 figure('position',[0,0,1200,1000])
192 sgtitle('LQR Method System Response')
193 for i = 1:4
194     subplot(2,2,i)
195     stairs(X_LQR(i,:));
196     hold on
197     if i == 2
198         plot(x2_limit*ones(N,1),'r')
199         plot(-x2_limit*ones(N,1),'r')
200     end
201     title(['State X',num2str(i)])
202 end
203 saveas(gcf,[pwd,'\Homework\HW5\fig\pblm2_LQR_sys_response.png'])
204
205 figure('position',[0,0,1200,500])
206 sgtitle('LQR Method Control Signal')
207 subplot(1,2,1)
208 stairs(U_LQR')
209 hold on
210 stairs(U_sat_LQR')
211 plot(u_limit*ones(N,1),'r')
212 plot(-u_limit*ones(N,1),'r')

```

```

213 title('U')
214
215 subplot(1,2,2)
216 stairs(diff(U_LQR'))
217 hold on
218 stairs(diff(U_sat_LQR'))
219 plot(u_rate_limit*ones(N,1), 'r')
220 plot(-u_rate_limit*ones(N,1), 'r')
221 title('U Rate')
222 saveas(gcf,[pwd,'\Homework\HW5\fig\pblm2_LQR_ctrl_signal.png'])
223 end
224
225 if mpc
226 % Plot MPC Plots
227 figure('position',[0,0,1200,1000])
228 sgtitle(['MPC Method System Response T = ',num2str(T)])
229 for i = 1:4
230     subplot(2,2,i)
231     stairs(X_MPC(i,:));
232     hold on
233     if i == 2
234         plot(x2_limit*ones(N,1),'r')
235         plot(-x2_limit*ones(N,1),'r')
236     end
237     title(['State X',num2str(i)])
238 end
239 saveas(gcf,[pwd,'\Homework\HW5\fig\pblm2_MPC_T',num2str(T),...
240     '_sys_response.png'])
241
242 figure('position',[0,0,1200,500])
243 sgtitle(['MPC Method Control Signal T = ',num2str(T)])
244 subplot(1,2,1)
245 stairs(U_MPC')
246 hold on
247 plot(u_limit*ones(N,1),'r')
248 plot(-u_limit*ones(N,1),'r')
249 title('U')
250
251
252 subplot(1,2,2)
253 stairs(diff(U_MPC'))
254 hold on
255 plot(u_rate_limit*ones(N,1), 'r')

```

```
256 plot(-u_rate_limit*ones(N,1), 'r')
257 title('U Rate')
258 saveas(gcf,[pwd,'\Homework\HW5\fig\pblm2_MPC_T',num2str(T),...
259     '_ctrl_signal.png'])
260
261 end
262
263
264 end
```