

# A Design Study Approach to Classical Control

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## Homework F.13

The objective of this problem is to design an observer that estimates the state of the system and to use the estimated state in the controller designed in Homework [F.12](#).

- (a) For the sake of understanding the function of the observer, for this problem we will use exact parameters, without an input disturbance. Modify the VTOL dynamics so that the parameters known to the controller are the actual plant parameters (uncertainty parameter  $\alpha = 0$ ).
- (b) Verify that the state space system is observable by checking that  $\text{rank}(\mathcal{O}_{A,C}) = n$ .
- (c) In the control block, add an observer to estimate the state  $\hat{x}$ , and use the estimate of the state in your feedback controller. Tune the poles of the controller and observer to obtain good performance.
- (d) Modify the simulation files so that the controller outputs both  $u$  and  $\hat{x}$ . Add a plotting routine to plot both the state and the estimated state of the system on the same graph.
- (e) As motivation for the next chapter, add an input force disturbance to the system of 1.0 and an input torque disturbance of 0.1 and observe that there is steady state error in the response even though there is an integrator. This is caused by a steady state error in the observation error. In the next chapter we will show how to remove the steady state error in the observation error.

## Solution

Python code used to design the observer based controller is shown below:

```
1 # VTOL Parameter File
2 import numpy as np
3 from scipy import signal
4 import control as cnt
5 import sys
6 sys.path.append('.') # add parent directory
7 import VTOLParam as P
8
9
10 #####
11 # State Space
12 #####
13 # tuning parameters
14 wn_h = 1.0
15 zeta_h = 0.707
16 wn_z = 0.9905
17 zeta_z = 0.707
18 wn_th = 13.3803
19 zeta_th = 0.707
20 integrator_h = -3.0
21 integrator_z = -4.0
22
23 # observer gains
24 wn_h_obs = 10.0*wn_h
25 wn_z_obs = 10.0*wn_z
26 wn_th_obs = 5.0*wn_th
27
28
29 # State Space Equations
30 A_lon = np.array([[0.0, 1.0],
31                  [0.0, 0.0]])
32 B_lon = np.array([[0.0],
33                  [1.0/(P.mc+2.0*P.mr)]])
34 C_lon = np.array([[1.0, 0.0]])
35 A_lat = np.array([[0.0, 0.0, 1.0, 0.0],
36                  [0.0, 0.0, 0.0, 1.0],
37                  [0.0, -P.Fe/(P.mc+2.0*P.mr), -(P.mu/(P.mc+2.0*P.mr)), 0.0],
38                  [0.0, 0.0, 0.0, 0.0]])
39 B_lat = np.array([[0.0],
40                  [0.0],
```

```

41         [0.0],
42         [1.0/(P.Jc+2*P.mr*P.d**2)])])
43 C_lat = np.array([[1.0, 0.0, 0.0, 0.0],
44                  [0.0, 1.0, 0.0, 0.0]])
45
46 # form augmented system
47 A1_lon = np.array([[0.0, 1.0, 0.0],
48                  [0.0, 0.0, 0.0],
49                  [-1.0, 0.0, 0.0]])
50 B1_lon = np.array([[0.0],
51                  [1.0],
52                  [0.0]])
53 A1_lat = np.array([[0.0, 0.0, 1.0, 0.0, 0.0],
54                  [0.0, 0.0, 0.0, 1.0, 0.0],
55                  [0.0, -P.Fe/(P.mc+2.0*P.mr), -(P.mu/(P.mc+2.0*P.mr)), 0.0, 0.0],
56                  [0.0, 0.0, 0.0, 0.0, 0.0],
57                  [-1.0, 0.0, 0.0, 0.0, 0.0]])
58 B1_lat = np.array([[0.0],
59                  [0.0],
60                  [0.0],
61                  [1.0/(P.Jc+2*P.mr*P.d**2)],
62                  [0.0]])
63
64 # gain calculation
65 des_char_poly_lon = np.convolve([1.0, 2.0*zeta_h*wn_h, wn_h**2],
66                                np.poly(integrator_h))
67 des_poles_lon = np.roots(des_char_poly_lon)
68
69 des_char_poly_lat = np.convolve(
70     np.convolve([1.0, 2.0*zeta_z*wn_z, wn_z**2],
71                [1.0, 2.0*zeta_th*wn_th, wn_th**2]),
72     np.poly(integrator_z))
73 des_poles_lat = np.roots(des_char_poly_lat)
74
75
76 # Compute the gains if the system is controllable
77 if np.linalg.matrix_rank(cnt.ctrb(A1_lon, B1_lon)) != 3:
78     print("The longitudinal system is not controllable")
79 else:
80     K1_lon = cnt.acker(A1_lon, B1_lon, des_poles_lon)
81     K_lon = np.matrix([K1_lon.item(0), K1_lon.item(1)])
82     ki_lon = K1_lon.item(2)
83
84 if np.linalg.matrix_rank(cnt.ctrb(A1_lat, B1_lat)) != 5:
85     print("The lateral system is not controllable")

```

```

86 else:
87     K1_lat = cnt.acker(A1_lat, B1_lat, des_poles_lat)
88     K_lat = np.matrix([K1_lat.item(0), K1_lat.item(1), K1_lat.item(2), K1_lat.item(3)])
89     ki_lat = K1_lat.item(4)
90
91 # compute observer poles
92 des_obs_poles_lon = np.roots([1.0, 2.0*zeta_h*wn_h_obs, wn_h_obs**2])
93 des_char_poly_lat = np.convolve([1.0, 2.0*zeta_z*wn_z_obs, wn_z_obs**2],
94                                 [1.0, 2.0*zeta_th*wn_th_obs, wn_th_obs**2])
95 des_obs_poles_lat = np.roots(des_char_poly_lat)
96
97 if np.linalg.matrix_rank(cnt.ctrb(A_lon.T, C_lon.T)) != 2:
98     print("The longitudinal system is not observable")
99 else:
100     # place_poles returns an object with various properties. The gains are accessible
101     # .T transposes the matrix
102     L_lon = signal.place_poles(A_lon.T, C_lon.T, des_obs_poles_lon).gain_matrix.T
103
104 if np.linalg.matrix_rank(cnt.ctrb(A_lat.T, C_lat.T)) != 4:
105     print("The lateral system is not observable")
106 else:
107     # place_poles returns an object with various properties. The gains are accessible
108     # .T transposes the matrix
109     L_lat = signal.place_poles(A_lat.T, C_lat.T, des_obs_poles_lat).gain_matrix.T
110
111 print('K_lon: ', K_lon)
112 print('ki_lon: ', ki_lon)
113 print('L_lon^T: ', L_lon.T)
114 print('K_lat: ', K_lat)
115 print('ki_lat: ', ki_lat)
116 print('L_lat^T: ', L_lat.T)

```

Python code for the observer based control is shown below:

```

1 import numpy as np
2 import VTOLParam as P
3 import VTOLParamHW13 as P13
4
5 class VTOLController:
6     def __init__(self):
7         self.xhat_lon = np.array([[0.0], [0.0]])
8         self.xhat_lat = np.array([[0.0], [0.0], [0.0], [0.0]])
9         self.integrator_z = 0.0 # integrator on position z
10        self.error_z_d1 = 0.0 # error signal delayed by 1 sample

```

```

11         self.integrator_h = 0.0          # integrator on altitude h
12         self.error_h_d1 = 0.0            # error signal delayed by 1 sample
13         self.F_d1 = 0.0 # Force signal delayed by 1 sample
14         self.tau_d1 = 0.0 # torque signal delayed by 1 sample
15         self.limit = P.fmax
16         self.Ts = P.Ts
17
18     def update(self, r, y):
19         z_r = r.item(0)
20         h_r = r.item(1)
21         y_lat = np.array([[y.item(0)], [y.item(2)]])
22         y_lon = y.item(1)
23         # update the observers
24         xhat_lat = self.update_lat_observer(y_lat)
25         xhat_lon = self.update_lon_observer(y_lon)
26         z_hat = xhat_lat.item(0)
27         h_hat = xhat_lon.item(0)
28         theta_hat = xhat_lat.item(1)
29
30         # integrate error
31         error_z = z_r - z_hat
32         self.integrateErrorZ(error_z)
33         error_h = h_r - h_hat
34         self.integrateErrorH(error_h)
35
36         # Construct the states
37         # Compute the state feedback controllers
38         F_tilde = -P13.K_lon @ xhat_lon \
39                 - P13.ki_lon * self.integrator_h
40         F = P.Fe/np.cos(theta_hat) + F_tilde.item(0)
41         tau = -P13.K_lat @ xhat_lat \
42              - P13.ki_lat*self.integrator_z
43         u = np.array([[F], [tau.item(0)]])
44         self.F_d1 = F
45         self.tau_d1 = tau.item(0)
46         return u, xhat_lat, xhat_lon
47
48     def update_lat_observer(self, y_m):
49         # update the observer using RK4 integration
50         F1 = self.observer_f_lat(self.xhat_lat, y_m)
51         F2 = self.observer_f_lat(self.xhat_lat + self.Ts / 2 * F1, y_m)
52         F3 = self.observer_f_lat(self.xhat_lat + self.Ts / 2 * F2, y_m)
53         F4 = self.observer_f_lat(self.xhat_lat + self.Ts * F3, y_m)
54         self.xhat_lat += self.Ts / 6 * (F1 + 2 * F2 + 2 * F3 + F4)
55         return self.xhat_lat

```

```

56
57 def observer_f_lat(self, x_hat, y_m):
58     # xhatdot = A*xhat + B*u + L(y-C*xhat)
59     xhat_dot = P13.A_lat @ x_hat \
60         + P13.B_lat * self.tau_d1 \
61         + P13.L_lat @ (y_m - P13.C_lat @ x_hat)
62     return xhat_dot
63
64 def update_lon_observer(self, y_m):
65     # update the observer using RK4 integration
66     F1 = self.observer_f_lon(self.xhat_lon, y_m)
67     F2 = self.observer_f_lon(self.xhat_lon + self.Ts / 2 * F1, y_m)
68     F3 = self.observer_f_lon(self.xhat_lon + self.Ts / 2 * F2, y_m)
69     F4 = self.observer_f_lon(self.xhat_lon + self.Ts * F3, y_m)
70     self.xhat_lon += self.Ts / 6 * (F1 + 2 * F2 + 2 * F3 + F4)
71     return self.xhat_lon
72
73 def observer_f_lon(self, x_hat, y_m):
74     # xhatdot = A*xhat + B*u + L(y-C*xhat)
75     xhat_dot = P13.A_lon @ x_hat \
76         + P13.B_lon * (self.F_d1 - P.Fe) \
77         + P13.L_lon @ (y_m - P13.C_lon @ x_hat)
78     return xhat_dot
79
80 def integrateErrorZ(self, error_z):
81     self.integrator_z = self.integrator_z \
82         + (P.Ts/2.0)*(error_z + self.error_z_d1)
83     self.error_z_d1 = error_z
84
85 def integrateErrorH(self, error_h):
86     self.integrator_h = self.integrator_h \
87         + (P.Ts/2.0)*(error_h + self.error_h_d1)
88     self.error_h_d1 = error_h
89
90 def saturate(self, u):
91     if abs(u) > self.limit:
92         u = self.limit*np.sign(u)
93     return u

```

See the wiki for the complete solution.