A Design Study Approach to Classical Control

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Homework F.14

(a) Modify your solution from HW F.13 so that the uncertainty parameter is $\alpha=0.2$, representing 20% inaccuracy in the knowledge of the system parameters. Modify VTOL_dynamics to add an altitude disturbance of 1.0 and a wind disturbance of 1.0 m/s. A Python code snippet that implements these disturbances is given by

Before adding the disturbance observer, run the simulation and note that the controller is not robust to the input disturbance.

(b) Add a disturbance observer to both controllers, and verify that the steady state error in the estimator has been removed and that the

disturbances have been adequately compensated. Tune the system to get good response.

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
  State Space
13 # tuning parameters
14 wn_h
        = 1.0
15 \text{ zeta\_h} = 0.707
       = 0.9905
16 WN_Z
zeta_z = 0.707
18 wn th = 13.3803
19 zeta th = 0.707
integrator_h = np.array([-3.0])
integrator_z = np.array([-2.0])
23 # observer gains
v_{4} wn_{h_obs} = 10.0 * wn_{h_obs}
           = 10.0 \times wn_z
25 wn_z_obs
wn_th_obs = 5.0*wn_th
dist_obsv_pole_lon = np.array([-10.0])
dist_obsv_pole_lat = np.array([-10.0])
29
30
31 # State Space Equations
A_{lon} = np.array([[0.0, 1.0],
                   [0.0, 0.0]])
B_{\text{lon}} = np.array([[0.0],
                   [1.0/(P.mc+2.0*P.mr)])
C_{lon} = np.array([[1.0, 0.0]])
```

```
A_{\text{lat}} = \text{np.array}([[0.0, 0.0, 1.0, 0.0],
                       [0.0, 0.0, 0.0, 1.0],
                       [0.0, -P.Fe/(P.mc+2.0*P.mr), -(P.mu/(P.mc+2.0*P.mr)), 0.0],
39
                       [0.0, 0.0, 0.0, 0.0]])
  B_lat = np.array([[0.0],
41
                       [0.0],
42
                       [0.0],
43
                       [1.0/(P.Jc+2*P.mr*P.d**2)])
44
   C_{lat} = np.array([[1.0, 0.0, 0.0, 0.0],
45
                       [0.0, 1.0, 0.0, 0.0]]
46
47
48 # form augmented system
49 A1_lon = np.array([[0.0, 1.0, 0.0],
                        [0.0, 0.0, 0.0],
50
                        [-1.0, 0.0, 0.0]
51
B1_{lon} = np.array([[0.0],
                        [1.0/(P.mc+2.0*P.mr)],
53
                        [0.011)
54
55 Al_lat = np.array([[0.0, 0.0, 1.0, 0.0, 0.0],
                        [0.0, 0.0, 0.0, 1.0, 0.0],
56
                        [0.0, -P.Fe/(P.mc+2.0*P.mr), -(P.mu/(P.mc+2.0*P.mr)), 0.0, 0.
57
                        [0.0, 0.0, 0.0, 0.0, 0.0],
58
                        [-1.0, 0.0, 0.0, 0.0, 0.0]
  B1_lat = np.array([[0.0],
60
                        [0.0],
61
                        [0.0],
62
                        [1.0/(P.Jc+2*P.mr*P.d**2)],
63
                        [0.0]])
64
65
66 # gain calculation
67 des_char_poly_lon = np.convolve([1.0, 2.0*zeta_h*wn_h, wn_h**2],
                                     np.poly(integrator_h))
69 des_poles_lon = np.roots(des_char_poly_lon)
70
  des_char_poly_lat = np.convolve(
71
                            np.convolve([1.0, 2.0*zeta_z*wn_z, wn_z**2],
72
73
                                         [1.0, 2.0*zeta_th*wn_th, wn_th**2]),
                            np.poly(integrator_z))
75 des_poles_lat = np.roots(des_char_poly_lat)
77
78 # Compute the gains if the system is controllable
79 if np.linalg.matrix_rank(cnt.ctrb(A1_lon, B1_lon)) != 3:
       print("The longitudinal system is not controllable")
80
81 else:
```

```
K1_lon = cnt.acker(A1_lon, B1_lon, des_poles_lon)
                K_lon = np.array([K1_lon.item(0), K1_lon.item(1)])
               ki_lon = K1_lon.item(2)
 84
 s6 if np.linalq.matrix_rank(cnt.ctrb(A1_lat, B1_lat)) != 5:
                print("The lateral system is not controllable")
 87
       else:
 88
               K1_lat = cnt.acker(A1_lat, B1_lat, des_poles_lat)
 89
                K_{lat} = np.array([K1_lat.item(0), K1_lat.item(1), K1_lat.item(2), K1_lat.item(3), K1_lat.item(3), K1_lat.item(3), K1_lat.item(4), K1_lat.item(4), K1_lat.item(4), K1_lat.item(5), K1_lat.item(6), K1_lat.i
 90
 91
                ki_lat = K1_lat.item(4)
 92
 93 # form augmented system for observer design
 94 # form augmented system
 95 A2_lon = np.array([[0.0, 1.0, 0.0],
 96
                                                   [0.0, 0.0, 1.0/(P.mc+2.0*P.mr)],
                                                   [0.0, 0.0, 0.0]
 97
 98 C2\_lon = np.array([[1.0, 0.0, 0.0]])
 99 A2_lat = np.array([[0.0, 0.0, 1.0, 0.0, 0.0],
100
                                                   [0.0, 0.0, 0.0, 1.0, 0.0],
                                                   [0.0, -P.Fe/(P.mc+2.0*P.mr), -(P.mu/(P.mc+2.0*P.mr)), 0.0, 0.
101
102
                                                   [0.0, 0.0, 0.0, 0.0, 1.0/(P.Jc+2*P.mr*P.d**2)],
                                                   [0.0, 0.0, 0.0, 0.0, 0.0]
103
       C2_{lat} = np.array([[1.0, 0.0, 0.0, 0.0, 0.0],
104
                                                   [0.0, 1.0, 0.0, 0.0, 0.0]]
105
106
107 # compute observer poles
los des_obs_char_poly_lon = np.convolve([1.0, 2.0*zeta_h*wn_h_obs, |wn_h_obs**2],
                                                                                     np.poly(dist_obsv_pole_lon))
109
des_obs_poles_lon = np.roots(des_obs_char_poly_lon)
111 des_obs_char_poly_lat = np.convolve(
                                                                    np.convolve([1.0, 2.0*zeta_z*wn_z_dbs, wn_z_obs**2],
112
113
                                                                                              [1.0, 2.0*zeta_th*wn_th_obs, wn_th_obs**2
114
                                                                    np.poly(dist_obsv_pole_lat))
115 des_obs_poles_lat = np.roots(des_obs_char_poly_lat)
116
if np.linalg.matrix_rank(cnt.ctrb(A2_lon.T, C2_lon.T)) != 3:
1118
               print("The longitudinal system is not observable")
119 else:
                # place_poles returns an object with various properties. The gains are acces
120
                # .T transposes the matrix
121
               L2_lon = signal.place_poles(A2_lon.T, C2_lon.T, des_obs_poles_lon).gain_matri
122
123
               L_lon = L2_lon[0:2, 0:1]
124
               Ld_lon = L2_lon[2:3, 0:1]
|126 if np.linalg.matrix_rank(cnt.ctrb(A2_lat.T, C2_lat.T)) != 5:
```

```
print("The lateral system is not observable")
127
128 else:
       # place_poles returns an object with various properties. The gains are acces
129
       # .T transposes the matrix
       L2_lat = signal.place_poles(A2_lat.T, C2_lat.T, des_obs_poles_lat).gain_matri
131
       L_lat = L2_lat[0:4, 0:2]
132
       Ld_lat = L2_lat[4:5, 0:2]
133
134
135 print('K_lon: ', K_lon)
136 print('ki_lon: ', ki_lon)
137 print('L_lon^T: ', L_lon.T)
138 print('Ld_lon: ', Ld_lon)
139 print('K_lat: ', K_lat)
140 print('ki_lat: ', ki_lat)
141 print('L_lat^T: ', L_lat.T)
142 print('Ld_lat: ', Ld_lat)
```

Python code for the observer based control is shown below:

```
1 import numpy as np
2 import VTOLParam as P
3 import VTOLParamHW14 as P14
 class VTOLController:
6
       def __init__(self):
           self.xhat_lon = np.array([[0.0], [0.0], [0.0]])
           self.xhat_lat = np.array([[0.0], [0.0], [0.0], [0.0],
           self.integrator_z = 0.0
                                         # integrator on position z
9
                                         # error signal delayed by 1 sample
           self.error_z_d1 = 0.0
10
                                         # integrator on altitude h
           self.integrator_h = 0.0
11
           self.error_h_d1 = 0.0
                                         # error signal delayed by 1 sample
12
           self.F_d1 = 0.0 # Force signal delayed by 1 sample
13
           self.tau_d1 = 0.0 # torque signal delayed by 1 sample
14
           self.limit = P.fmax
           self.Ts = P.Ts
16
17
       def update(self, r, y):
18
           z_r = r.item(0)
           h_r = r.item(1)
20
           y_lat = np.array([[y.item(0)],[y.item(2)]])
^{21}
           y_{lon} = y.item(1)
22
23
           # update the observers
           xhat_lat, dhat_lat = self.update_lat_observer(y_lat)
^{24}
25
           xhat_lon, dhat_lon = self.update_lon_observer(y_lon)
```

```
z_hat = xhat_lat.item(0)
26
           h_hat = xhat_lon.item(0)
27
           theta_hat = xhat_lat.item(1)
28
           # integrate error
30
           error_z = z_r - z_hat
31
           self.integrateErrorZ(error_z)
32
           error_h = h_r - h_hat
33
           self.integrateErrorH(error_h)
34
35
           # Construct the states
36
           # Compute the state feedback controllers
37
           F_tilde = -P14.K_lon @ xhat_lon \
38
                      - P14.ki_lon * self.integrator_h
39
           F = P.Fe/np.cos(theta_hat) \
               + F_tilde.item(0) \
41
               - dhat_lon
42
           tau = -P14.K_lat @ xhat_lat \
43
                 - P14.ki_lat*self.integrator_z \
                 - dhat_lat
45
           u = np.array([[F], [tau.item(0)]])
           self.F_d1 = F
47
           self.tau_d1 = tau.item(0)
48
           return u, xhat_lat, xhat_lon
49
50
       def update_lat_observer(self, y_m):
51
           # update the observer using RK4 integration
52
           F1 = self.observer_f_lat(self.xhat_lat, y_m)
53
           F2 = self.observer_f_lat(self.xhat_lat + self.Ts / 2 * F1, y_m)
54
           F3 = self.observer_f_lat(self.xhat_lat + self.Ts / 2 * F2, y_m)
55
           F4 = self.observer_f_lat(self.xhat_lat + self.Ts * F3, |y_m)
56
           self.xhat_lat += self.Ts / 6 * (F1 + 2 * F2 + 2 * F3 + F4)
57
           xhat = np.array([[self.xhat_lat.item(0)],
58
                             [self.xhat_lat.item(1)],
59
                             [self.xhat_lat.item(2)],
60
                             [self.xhat_lat.item(3)],
61
62
                            1)
           dhat = self.xhat_lat.item(4)
           return xhat, dhat
64
65
       def observer_f_lat(self, x_hat, y_m):
66
           \# xhatdot = A*xhat + B*u + L(y-C*xhat)
           xhat_dot = P14.A2_lat @ x_hat \
68
                       + P14.B1_lat * self.tau_d1 \
69
                       + P14.L2_lat @ (y_m - P14.C2_lat @ x_hat)
70
```

```
return xhat_dot
71
72
       def update_lon_observer(self, y_m):
73
            # update the observer using RK4 integration
           F1 = self.observer_f_lon(self.xhat_lon, y_m)
75
           F2 = self.observer_f_lon(self.xhat_lon + self.Ts / 2 * F1, y_m)
76
           F3 = self.observer_f_lon(self.xhat_lon + self.Ts / 2 * F2, y_m)
77
           F4 = self.observer_f_lon(self.xhat_lon + self.Ts * F3, |y_m)
78
           self.xhat_lon += self.Ts / 6 * (F1 + 2 * F2 + 2 * F3 + F4)
79
            xhat = np.array([[self.xhat_lon.item(0)],
80
                             [self.xhat lon.item(1)],
81
82
            dhat = self.xhat_lon.item(2)
83
            return xhat, dhat
84
       def observer_f_lon(self, x_hat, y_m):
86
            \# xhatdot = A*xhat + B*u + L(y-C*xhat)
87
           xhat_dot = P14.A2_lon @ x_hat \
88
                       + P14.B1_lon * (self.F_d1 - P.Fe) \
                       + P14.L2_lon @ (y_m - P14.C2_lon @ x_hat)
90
91
            return xhat_dot
92
       def integrateErrorZ(self, error_z):
            self.integrator_z = self.integrator_z \
94
                                 + (P.Ts/2.0) * (error_z + self.error_z_d1)
            self.error_z_d1 = error_z
96
       def integrateErrorH(self, error_h):
98
            self.integrator h = self.integrator h \
                                 + (P.Ts/2.0) * (error_h + self.error_h_d1)
100
101
            self.error_h_d1 = error_h
102
       def saturate(self,u):
103
            if abs(u) > self.limit:
104
                u = self.limit*np.sign(u)
105
106
            return u
```

See the wiki for the complete solution.