

# A Design Study Approach to Classical Control

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## Homework E.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 [E.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 ball beam 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 disturbance to the system of 0.5 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 # ballbeam Parameter File
2 import numpy as np
3 import control as cnt
4 import sys
5 sys.path.append('.') # add parent directory
6 import ballbeamParam as P
7 import numpy as np
8 from scipy import signal
9 import control as cnt
10
11 #####
12 # State Space
13 #####
14 # tuning parameters
15 tr_z = 1.2 # rise time for position
16 tr_theta = 0.25 # rise time for angle
17 zeta_z = 0.707 # damping ratio position
18 zeta_th = 0.707 # damping ratio angle
19 integrator_pole = -5.0
20 # pick observer poles
21 wn_th_obs = 5.0*2.2/tr_theta
22 wn_z_obs = 5.0*2.2/tr_z
23
24 # State Space Equations
25 # xdot = A*x + B*u
26 # y = C*x
27 A = np.array([[0.0, 0.0, 1.0, 0.0],
28               [0.0, 0.0, 0.0, 1.0],
29               [0.0, -P.g, 0.0, 0.0],
30               [-P.m1*P.g/((P.m2*P.length**2)/3.0+P.m1*(P.length/2.0)**2), 0.0, 0.0, 0.0])
31
32 B = np.array([[0.0],
33               [0.0],
34               [0.0],
35               [P.length / (P.m2 * P.length ** 2 / 3.0 + P.m1 * P.length ** 2 / 4.0)])
36
37 C = np.array([[1.0, 0.0, 0.0, 0.0],
38               [0.0, 1.0, 0.0, 0.0]])
39
40 # form augmented system
```

```

41 A1 = np.matrix([[0.0, 0.0, 1.0, 0.0, 0.0],
42                 [0.0, 0.0, 0.0, 1.0, 0.0],
43                 [0.0, -P.g, 0.0, 0.0, 0.0],
44                 [-P.m1*P.g/((P.m2*P.length**2)/3.0+P.m1*(P.length/2.0)**2), 0.0, 0.0, 0.0, 0.0],
45                 [-1.0, 0.0, 0.0, 0.0, 0.0]])
46
47 B1 = np.matrix([[0.0],
48                 [0.0],
49                 [0.0],
50                 [P.length/(P.m2*P.length**2/3.0+P.m1*P.length**2/4.0)],
51                 [0.0]])
52
53 # gain calculation
54 wn_th = 2.2/tr_theta # natural frequency for angle
55 wn_z = 2.2/tr_z # natural frequency for position
56 des_char_poly = np.convolve(
57     np.convolve([1, 2*zeta_z*wn_z, wn_z**2],
58                 [1, 2*zeta_th*wn_th, wn_th**2]),
59     np.poly(integrator_pole))
60 des_poles = np.roots(des_char_poly)
61
62 # Compute the gains if the system is controllable
63 if np.linalg.matrix_rank(cnt.ctrb(A1, B1)) != 5:
64     print("The system is not controllable")
65 else:
66     K1 = cnt.acker(A1, B1, des_poles)
67     K = np.array([K1.item(0), K1.item(1), K1.item(2), K1.item(3)])
68     ki = K1.item(4)
69
70 # compute observer gains
71 des_obs_char_poly = np.convolve([1, 2*zeta_z*wn_z_obs, wn_z_obs**2],
72                                 [1, 2*zeta_th*wn_th_obs, wn_th_obs**2])
73 des_obs_poles = np.roots(des_obs_char_poly)
74
75 # Compute the gains if the system is observable
76 if np.linalg.matrix_rank(cnt.ctrb(A.T, C.T)) != 4:
77     print("The system is not observable")
78 else:
79     # place_poles returns an object with various properties. The gains are acces
80     # .T transposes the matrix
81     L = signal.place_poles(A.T, C.T, des_obs_poles).gain_matrix.T
82
83 print('K: ', K)
84 print('ki: ', ki)
85 print('L^T: ', L.T)

```

Python code for the observer based control is shown below:

```

1 import numpy as np
2 import ballbeamParam as P
3 import ballbeamParamHW13 as P13
4
5 class ballbeamController:
6     def __init__(self):
7         self.x_hat = np.array([
8             [0.0], # initial estimate for z_hat
9             [0.0], # initial estimate for theta_hat
10            [0.0], # initial estimate for z_hat_dot
11            [0.0]]) # initial estimate for theta_hat_dot
12        self.F_d1 = 0.0 # Computed Force, delayed by one sample
13        self.integrator = 0.0 # integrator
14        self.error_d1 = 0.0 # error signal delayed by 1 sample
15        self.K = P13.K # state feedback gain
16        self.ki = P13.ki # Integral gain
17        self.L = P13.L # observer gain
18        self.A = P13.A # system model
19        self.B = P13.B
20        self.C = P13.C
21        self.limit = P.Fmax # Maximum force
22        self.Ts = P.Ts # sample rate of controller
23
24    def update(self, z_r, y):
25        # update the observer and extract z_hat
26        x_hat = self.update_observer(y)
27        z_hat = x_hat.item(0)
28
29        # integrate error
30        error = z_r - z_hat
31        self.integrateError(error)
32
33        # Construct the state
34        xe = np.array([[P.ze], [0.0], [0.0], [0.0]])
35        x_tilde = x_hat - xe
36
37        # equilibrium force
38        F_e = P.m1*P.g*P.ze/P.length + P.m2*P.g/2.0
39        # Compute the state feedback controller
40        F_tilde = -self.K @ x_tilde \
41            - self.ki * self.integrator
42        F_unsat = F_e + F_tilde.item(0)

```

```

43         F = self.saturate(F_unsat)
44         self.integratorAntiWindup(F, F_unsat)
45         self.F_d1 = F
46         return F, x_hat
47
48     def update_observer(self, y):
49         # update the observer using RK4 integration
50         F1 = self.observer_f(self.x_hat, y)
51         F2 = self.observer_f(self.x_hat + self.Ts / 2 * F1, y)
52         F3 = self.observer_f(self.x_hat + self.Ts / 2 * F2, y)
53         F4 = self.observer_f(self.x_hat + self.Ts * F3, y)
54         self.x_hat += self.Ts / 6 * (F1 + 2 * F2 + 2 * F3 + F4)
55         return self.x_hat
56
57     def observer_f(self, x_hat, y):
58         # xhatdot = A*(xhat-xe) + B*u + L(y-C*xhat)
59         xe = np.array([[P.ze], [0.0], [0.0], [0.0]])
60
61         # equilibrium force
62         F_e = P.m1*P.g*P.ze/P.length + P.m2*P.g/2.0
63         xhat_dot = self.A @ (x_hat - xe) \
64             + self.B * (self.F_d1 - F_e) \
65             + self.L @ (y - self.C @ x_hat)
66         return xhat_dot
67
68     def integrateError(self, error):
69         self.integrator = self.integrator \
70             + (self.Ts/2.0)*(error + self.error_d1)
71         self.error_d1 = error
72
73     def integratorAntiWindup(self, F, F_unsat):
74         # integrator anti - windup
75         if self.ki != 0.0:
76             self.integrator = self.integrator \
77                 + P.Ts/self.ki*(F-F_unsat)
78
79     def saturate(self, u):
80         if abs(u) > self.limit:
81             u = self.limit*np.sign(u)
82         return u

```

See the wiki for the complete solution.