

hw10

April 7, 2023

```
[ ]: import numpy as np
import matplotlib.pyplot as plt
import matplotlib_inline

from scipy.integrate import solve_ivp

%matplotlib inline
matplotlib_inline.backend_inline.set_matplotlib_formats('png', 'pdf')
```

1 Problem 12.12

$$T_1 = \frac{T_0}{\tau s + 1} + \frac{kG_c}{\tau s + 1} E$$

$$T_3 = \frac{T_1}{(\tau s + 1)^2}$$

$$E = T_{sp} - T_3$$

$$T_3 = \frac{\frac{T_0}{\tau s + 1} + \frac{kG_c}{\tau s + 1} E}{(\tau s + 1)^2}$$

$$T_3 = \frac{T_0}{(\tau s + 1)^3} + \frac{kG_c}{(\tau s + 1)^3} E$$

$$T_3 = \frac{T_0}{(\tau s + 1)^3} + \frac{kG_c}{(\tau s + 1)^3} T_{sp} - \frac{kG_c}{(\tau s + 1)^3} T_3$$

$$T_3 \left(1 + \frac{kG_c}{(\tau s + 1)^3} \right) = \frac{T_0}{(\tau s + 1)^3} + \frac{kG_c}{(\tau s + 1)^3} T_{sp}$$

$$T_3 \left((\tau s + 1)^3 + kG_c \right) = T_0 + kG_c T_{sp}$$

$$T_3 = \frac{1}{(\tau s + 1)^3 + kG_c} T_0 + \frac{kG_c}{(\tau s + 1)^3 + kG_c} T_{sp}$$

1.1 Part A

1.1.1 P controller

$$G_c = k_c$$

$$T_3 = \frac{1}{(\tau s + 1)^3 + k k_c} T_0 + \frac{k G_c}{(\tau s + 1)^3 + k k_c} T_{sp}$$

$$\text{Denominator} = \tau^3 s^3 + 3\tau^2 s^2 + 3\tau s + 1 + k k_c$$

Routh array;

$$a_0 = \tau^3 \quad a_2 = 3\tau$$

$$a_1 = 3\tau^2 \quad a_3 = 1 + kk_c$$

$$B_1 = \frac{9\tau^3 - \tau^3(1 + kk_c)}{3\tau^2}$$

$$C_1 = 1 + kk_c$$

$$\frac{9\tau^3 - \tau^3(1 + kk_c)}{3\tau^2} > 0$$

$$9 > 1 + kk_c$$

$$k_c < \frac{8}{k}$$

$$1 + kk_c > 0$$

$$k_c > -\frac{1}{k}$$

$$\boxed{-\frac{1}{k} < k_c < \frac{8}{k}}$$

1.1.2 PD controller

$$G_c = k_c (1 + \tau_D s)$$

$$\text{Denominator} = \tau^3 s^3 + 3\tau^2 s^2 + 3\tau s + 1 + kk_c (1 + \tau_D s)$$

Routh array;

$$a_0 = \tau^3 \quad a_2 = 3\tau + k_c \tau_D$$

$$a_1 = 3\tau^2 \quad a_3 = 1 + kk_c$$

$$B_1 = \frac{3\tau^2(3\tau + k_c \tau_D) - \tau^3(1 + kk_c)}{3\tau^2}$$

$$\frac{3\tau^2(3\tau + k_c \tau_D) - \tau^3(1 + kk_c)}{3\tau^2} > 0$$

$$3(3\tau + k_c \tau_D) - \tau(1 + kk_c) > 0$$

$$8\tau + k_c(3\tau_D - \tau k) > 0$$

$$k_c(\tau - \tau k) > -8\tau$$

$$k_c(1 - k) > -8$$

$$k_c > \frac{8}{k-1}$$

$$\boxed{k_c > -\frac{1}{k}}$$

Adding the derivative action to the proportional only controller has a stabilizing effect, and thus the PD controller can operate with a broader range of k_c values than the P only controller.

1.2 Part B

T_0 step change

$$T_0(s) = \frac{M}{s}$$

$$T_3 = \frac{1}{(\tau s + 1)^3 + k k_c} \cdot \frac{M}{s}$$

Final value theorem:

$$\lim_{s \rightarrow 0^+} sY(s) = \lim_{t \rightarrow \infty} y(t)$$

$$sT_3 = \frac{M}{(\tau s + 1)^3 + k k_c}$$

$$\lim_{s \rightarrow 0^+} sT_3 = \lim_{s \rightarrow 0^+} \frac{M}{(\tau s + 1)^3 + k k_c} = \frac{M}{1 + k k_c}$$

$$\text{Final } T_0 = 0$$

$$\text{Offset} = 0 - T_3 = 0 - \frac{M}{1 + k k_c}$$

$$\text{Largest } k_c = \frac{8}{k}$$

$$\text{Offset} = -\frac{M}{1 + k \frac{8}{k}}$$

$$\text{Smallest possible offset} = \boxed{-\frac{M}{9}}$$

2 Problem 12.13

2.1 Part A

$$C_P = \frac{G_c G_P}{1 + G_c G_P} C_{P_{sp}} + \frac{G'}{1 + G_c G_P} C_{in,R}$$

$$G' = \frac{2}{(s+1)(s+2)}$$

$$G_P = \frac{4-s}{(s+1)(s+2)}$$

$$C_P = \frac{G_c \frac{4-s}{(s+1)(s+2)}}{1 + G_c \frac{4-s}{(s+1)(s+2)}} C_{P_{sp}} + \frac{\frac{2}{(s+1)(s+2)}}{1 + G_c \frac{4-s}{(s+1)(s+2)}} C_{in,R}$$

$$C_P = \frac{G_c(4-s)}{s^2 + 3s + 2 + G_c(4-s)} C_{P_{sp}} + \frac{2}{s^2 + 3s + 2 + G_c(4-s)} C_{in,R}$$

2.1.1 P controller

$$G_c = k_c$$

$$\text{Denominator} = s^2 + 3s + 2 + k_c(4-s) = s^2 + (4k_c + 2)s + 4k_c + 2$$

Routh array:

$$a_0 = 1 \quad a_2 = 4k_c + 2$$

$$a_1 = 3 - k_c \quad a_3 = 0$$

$$B_1 = 4k_c + 2$$

$$k_c > -\frac{1}{2}$$

$$3 - k_c > 0$$

$$k_c < 3$$

$$\boxed{-\frac{1}{2} < k_c < 3}$$

2.1.2 PI controller

$$G_c = k_c \left(1 + \frac{1}{\tau_I s}\right) = k_c \left(1 + \frac{4}{s}\right)$$

$$\text{Denominator} = s^2 + 3s + 2 + k_c \left(1 + \frac{4}{s}\right) (4 - s) = s^3 + (3 - k_c) + 2s + 16k_c$$

Routh array:

$$a_0 = 1 \quad a_2 = 2$$

$$a_1 = 3 - k_c \quad a_3 = 16k_c$$

$$B_1 = \frac{6 - 2k_c - 16k_c}{3 - k_c}$$

$$\frac{6 - 2k_c - 16k_c}{3 - k_c} > 0$$

$$6 - 2k_c - 16k_c > 0$$

$$k_c < \frac{1}{3}$$

$$C_1 = 16k_c$$

$$16k_c > 0$$

$$k_c > 0$$

$$\boxed{0 < k_c < \frac{1}{3}}$$

Adding the integral action to the proportional only controller has a destabilizing effect on the system, and thus the PI controller requires a narrower range of k_c to operate with stability.

2.2 Part B

$$C_P = \frac{k_c(4-s)}{s^2+3s+2+k_c(4-s)} C_{P_{sp}} + \frac{2}{s^2+3s+2+k_c(4-s)} C_{in,R}$$

$$C_{P_{sp}} = \frac{M}{s}$$

$$C_P = \frac{k_c(4-s)}{s^2+3s+2+k_c(4-s)} \cdot \frac{M}{s}$$

Final value theorem:

$$\lim_{s \rightarrow 0^+} sY(s) = \lim_{t \rightarrow \infty} y(t)$$

$$sC_P = \frac{Mk_c(4-s)}{s^2+3s+2+k_c(4-s)}$$

$$\lim_{s \rightarrow 0^+} sC_P = \frac{2Mk_c}{2k_c+1}$$

$$\text{Offset} = c_{P_{sp}}(t) - c_P(t) = M - \lim_{t \rightarrow \infty} c_P(t)$$

$$\text{Offset} = M - \frac{2Mk_c}{2k_c+1} = M \left(\frac{2k_c+1}{2k_c+1} - \frac{2k_c}{2k_c+1} \right)$$

$$\text{Offset} = \frac{M}{2k_c+1}$$

$$\text{Maximum } k_c = 3$$

$$\text{Minimum offset} = \boxed{\frac{M}{7}}$$

3 Problem 13.4

3.0.1 State space models

$$A_1 \frac{dh_1}{dt} = -\frac{h_1}{R_1} + k_v u + F_w$$

$$A_2 \frac{dh_2}{dt} = \frac{h_1}{R_1} - \frac{h_2}{R_2}$$

$$\tau_m \frac{dh_{2,m}}{dt} + h_{2,m} = h_2$$

3.0.2 Real PID state space

$$\frac{de_I}{dt} = e$$

$$\frac{de_f}{dt} = -\frac{1}{\alpha\tau_D} e_f + \frac{1}{\alpha\tau_D} e$$

$$u = \frac{k_c}{\tau_I} e_I + k_c \left(1 - \frac{1}{\alpha}\right) e_f + \frac{k_c}{\alpha} e$$

$$e = h_{2,sp} - h_{2,m}$$

$$\frac{de_I}{dt} = h_{2,sp} - h_{2,m}$$

$$\frac{de_f}{dt} = -\frac{1}{\alpha\tau_D} e_f + \frac{1}{\alpha\tau_D} (h_{2,sp} - h_{2,m})$$

$$u = \frac{k_c}{\tau_I} e_I + k_c \left(1 - \frac{1}{\alpha}\right) e_f + \frac{k_c}{\alpha} (h_{2,sp} - h_{2,m})$$

$$A_1 \frac{dh_1}{dt} = -\frac{h_1}{R_1} + k_v \left(\frac{k_c}{\tau_I} e_I + k_c \left(1 - \frac{1}{\alpha}\right) e_f + \frac{k_c}{\alpha} (h_{2,sp} - h_{2,m}) \right) + F_w$$

3.1 State space model that describes the system

$$\frac{dh_1}{dt} = -\frac{1}{A_1 R_1} h_1 - \frac{k_c k_v}{A_1 \alpha} h_{2,m} + \frac{k_c k_v}{A_1 \tau_I} e_I + \left(\frac{k_c k_v}{A_1} - \frac{k_c k_v}{A_1 \alpha} \right) e_f + \frac{1}{A_1} F_w + \frac{k_c k_v}{A_1 \alpha} h_{2,sp}$$

$$\frac{dh_2}{dt} = \frac{1}{A_2 R_1} h_1 - \frac{1}{A_2 R_2} h_2$$

$$\frac{dh_{2,m}}{dt} = \frac{1}{\tau_m} h_2 - \frac{1}{\tau_m} h_{2,m}$$

$$\frac{de_I}{dt} = -h_{2,m} + h_{2,sp}$$

$$\frac{de_f}{dt} = -\frac{1}{\alpha\tau_D} h_{2,m} - \frac{1}{\alpha\tau_D} e_f + \frac{1}{\alpha\tau_D} h_{2,sp}$$

3.2 System Simulation

```
[ ]: # 13.4 simulation
ode_kwargs = {
    'method': 'Radau',
    'rtol': 1e-8,
    'atol': 1e-8,
}

t_range = [0, 10]

initial_cond = [0, 0, 0, 0, 0]

def real_pid_ode(t, y):
    f = y*0

    h_1 = y[0]
    h_2 = y[1]
    h_2m = y[2]
    e_I = y[3]
    e_f = y[4]

    h_2sp = 1
    F_w = 0

    k_c = 5
    tau_I = 2
    tau_D = 0.25
    alpha = 0.1
    tau_m = 0.1
    A_1, A_2, R_1, R_2, k_v = 1, 1, 1, 1, 1

    e = h_2sp - h_2m

    u = k_c * (e_I / tau_I + (1 - 1 / alpha) * e_f + e / alpha)

    f[0] = (-h_1 / R_1 + k_v * u + F_w) / A_1
    f[1] = (h_1 / R_1 - h_2 / R_2) / A_2
    f[2] = (h_2 - h_2m) / tau_m
    f[3] = e
    f[4] = (e - e_f) / (alpha * tau_D)

    return f

def pi_ode(t, y):
    f = y*0
```

```

h_1 = y[0]
h_2 = y[1]
h_2m = y[2]
e_I = y[3]

h_2sp = 1
F_w = 0

k_c = 5
tau_I = 2
tau_m = 0.1
A_1, A_2, R_1, R_2, k_v = 1, 1, 1, 1, 1

e = h_2sp - h_2m

u = k_c * (e_I / tau_I + e)

f[0] = (-h_1 / R_1 + k_v * u + F_w) / A_1
f[1] = (h_1 / R_1 - h_2 / R_2) / A_2
f[2] = (h_2 - h_2m) / tau_m
f[3] = e

return f

real_pid_sol = solve_ivp(real_pid_ode, t_range, initial_cond, **ode_kwargs)

pi_sol = solve_ivp(pi_ode, t_range, initial_cond[:4], **ode_kwargs)

plt.plot(real_pid_sol.t, np.ones(real_pid_sol.t.shape[0]), '--', label='Set_
↳Point')
plt.plot(real_pid_sol.t, real_pid_sol.y[1], label='Real PID Response')
plt.plot(pi_sol.t, pi_sol.y[1], label="PI Response")
plt.xlabel('Time')
plt.ylabel('Response')
plt.title('Problem 13.4 Closed Loop Response Simulation')
plt.legend()

```

```
[ ]: <matplotlib.legend.Legend at 0x1b8ffb041d0>
```

