

1. Problem 6.13

$$V \frac{dC_R}{dt} = F(C_{R0} - C_R) - V k_1 C_R + V k_2 C_P - V k_3 C_R$$

$$V \frac{dC_P}{dt} = -F C_P + V k_1 C_R - V k_2 C_P$$

$$\frac{dC_R}{dt} = - \left(\frac{F}{V} + k_1 + k_3 \right) C_R + k_2 C_P + \frac{F}{V} C_{R0}$$

$$\frac{dC_P}{dt} = k_1 C_R - \left(\frac{F}{V} + k_2 \right) C_P$$

$$y = C_P$$

$$A = \begin{bmatrix} -\left(\frac{F}{V} + k_1 + k_3\right) & k_2 \\ k_1 & -\left(\frac{F}{V} + k_2\right) \end{bmatrix}$$

$$B = \begin{bmatrix} \frac{F}{V} \\ 0 \end{bmatrix}$$

$$C = \begin{bmatrix} 0 & 1 \end{bmatrix}$$

$$D = 0$$

$$G(s) = c(sI - A)^{-1} b + d$$

$$sI - A = \begin{bmatrix} s + \left(\frac{F}{V} + k_1 + k_3\right) & -k_2 \\ -k_1 & s + \left(\frac{F}{V} + k_2\right) \end{bmatrix}$$

$$\text{Adj}(sI - A) = \begin{bmatrix} s + \left(\frac{F}{V} + k_2\right) & k_2 \\ k_1 & s + \left(\frac{F}{V} + k_1 + k_3\right) \end{bmatrix}$$

$$\det(sI - A) = \left(s + \left(\frac{F}{V} + k_1 + k_3\right) \right) \left(s + \left(\frac{F}{V} + k_2\right) \right) - k_1 k_2$$

$$\det(sI - A) = \left(s + \frac{F}{V} + k_1 + k_3 \right) \left(s + \frac{F}{V} + k_2 \right) - k_1 k_2$$

$$\theta = \det(sI - A)$$

$$(sI - A)^{-1} = \theta^{-1} \begin{bmatrix} s + \left(\frac{F}{V} + k_2\right) & k_2 \\ k_1 & s + \left(\frac{F}{V} + k_1 + k_3\right) \end{bmatrix}$$

$$G(s) = c\theta^{-1} \begin{bmatrix} s + \left(\frac{F}{V} + k_2\right) & k_2 \\ k_1 & s + \left(\frac{F}{V} + k_1 + k_3\right) \end{bmatrix} \begin{bmatrix} \frac{F}{V} \\ 0 \end{bmatrix}$$

$$G(s) = c\theta^{-1} \begin{bmatrix} \frac{F}{V} \left(s + \left(\frac{F}{V} + k_2\right) \right) \\ \frac{F}{V} k_1 \end{bmatrix}$$

$$G(s) = \theta^{-1} \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{F}{V} \left(s + \left(\frac{F}{V} + k_2\right) \right) \\ \frac{F}{V} k_1 \end{bmatrix}$$

$$G(s) = \frac{\frac{F}{V} k_1}{\theta}$$

$$G(s) = \frac{\frac{F}{V} k_1}{\left(s + \frac{F}{V} + k_1 + k_3 \right) \left(s + \frac{F}{V} + k_2 \right) - k_1 k_2}$$

2. Problem 6.7

(a)

State space model:

$$\begin{aligned} A_1 \frac{dh_1}{dt} &= F_{in} - \frac{h_1 - h_2}{R_1} \\ A_2 \frac{dh_2}{dt} &= \frac{h_1 - h_2}{R_1} - \frac{h_2}{R_2} \end{aligned}$$

At steady state

$$\begin{aligned} 0 &= 1 - \frac{1.5 - 0.5}{R_1} \\ 0 &= \frac{1.5 - 0.5}{R_1} - \frac{0.5}{R_2} \\ R_1 &= 1 \\ R_2 &= 0.5 \\ \frac{dh}{dt} &= \begin{bmatrix} -\frac{1}{A_1 R_1} & \frac{1}{A_1 R_1} \\ \frac{1}{A_2 R_1} & -\left(\frac{1}{A_2 R_1} + \frac{1}{A_2 R_2}\right) \end{bmatrix} h + \begin{bmatrix} \frac{1}{A_1} \\ 0 \end{bmatrix} F_{in} \\ y &= \begin{bmatrix} 0 & 1 \end{bmatrix} h \\ A &= \begin{bmatrix} -1 & 1 \\ 0.5 & -1.5 \end{bmatrix} \\ B &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ h(0) &= \begin{bmatrix} 1.5 \\ 0.5 \end{bmatrix} \end{aligned}$$

Simulate response with lsim

$$D = 0$$

For h_1 response

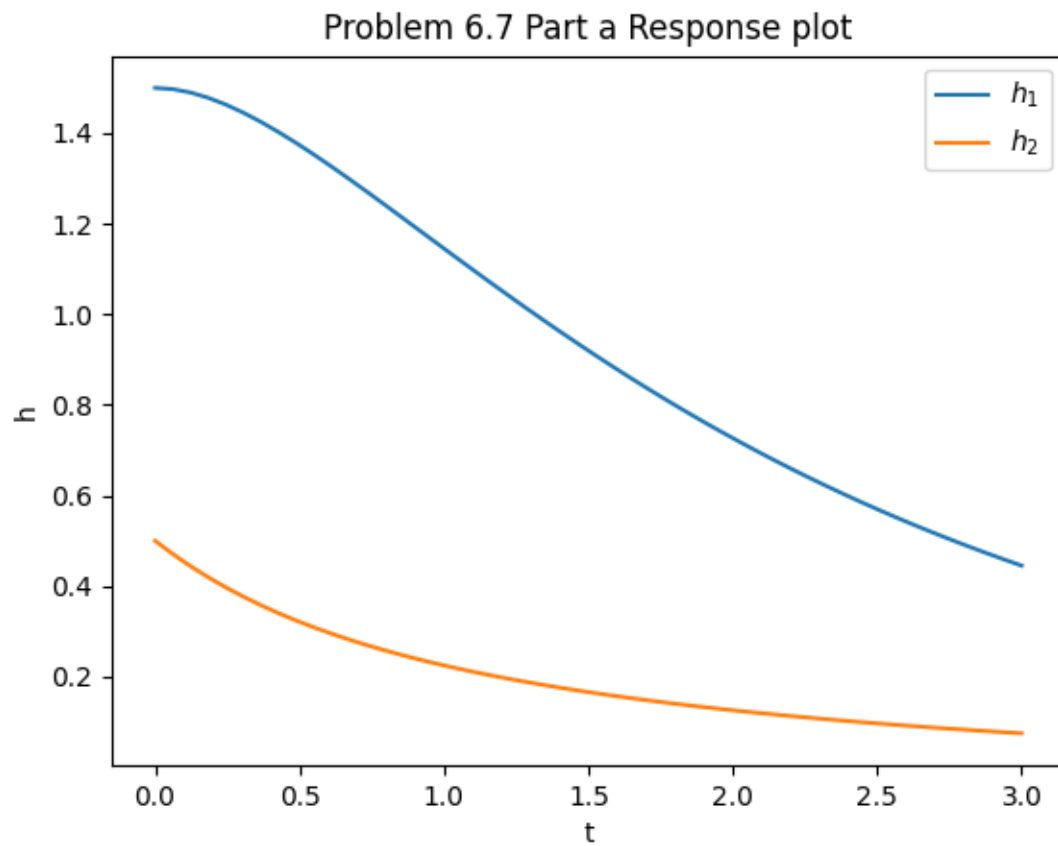
$$C = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

For h_2 response

$$C = \begin{bmatrix} 0 & 1 \end{bmatrix}$$

Simulate a response to an unforced input on the system. The initial heights are supplied as the initial condition.

Simulated response plot:



The plot was generated with the code below:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.signal import lti, step, lsim

h_1_0 = 1.5
h_2_0 = 0.5

A = np.array([
    [-1, 1],
    [.5, -1.5]
])
B = np.array([
    [1],
    [0]
])
C_h1 = np.array([1, 0])
C_h2 = np.array([0, 1])
D = 0

sys_h1 = lti(A, B, C_h1, D)
sys_h2 = lti(A, B, C_h2, D)
```

```

# Part A
u = 0
t = np.linspace(0, 3, 50)

t_h1_a, h1_a, x1 = lsim(sys_h1, u, t, h_1_0)
t_h2_a, h2_a, x2 = lsim(sys_h2, u, t, h_2_0)

plt.plot(t_h1_a, h1_a, label=r"$h_1$")
plt.plot(t_h2_a, h2_a, label=r"$h_2$")
plt.xlabel(r"$t$")
plt.ylabel(r"$h$")
plt.title("Problem 6.7 Part a Response plot")
plt.legend()
plt.show()

# Part B
t_react = np.interp(0.2, np.flip(h2_a), np.flip(t_h2_a))
h_1_react = np.interp(t_react, t_h1_a, h1_a)
print(f"Time when h_2 = 0.2 m: {t_react} min")
print(f"h_1 at that time: {h_1_react}")

# Part C
t_h1_c, h1_c = step(sys_h1, X0=h_1_react)
t_h2_c, h2_c = step(sys_h2, X0=0.2)

plt.plot(t_h1_c, h1_c, label=r"$h_1$")
plt.plot(t_h2_c, h2_c, label=r"$h_2$")
plt.xlabel(r"$t$")
plt.ylabel(r"$h$")
plt.title("Problem 6.7 Part c Response plot")
plt.legend()
plt.show()

print(f"Final h_1 = {h1_c[-1]}")
print(f"Final h_2 = {h2_c[-1]}")

```

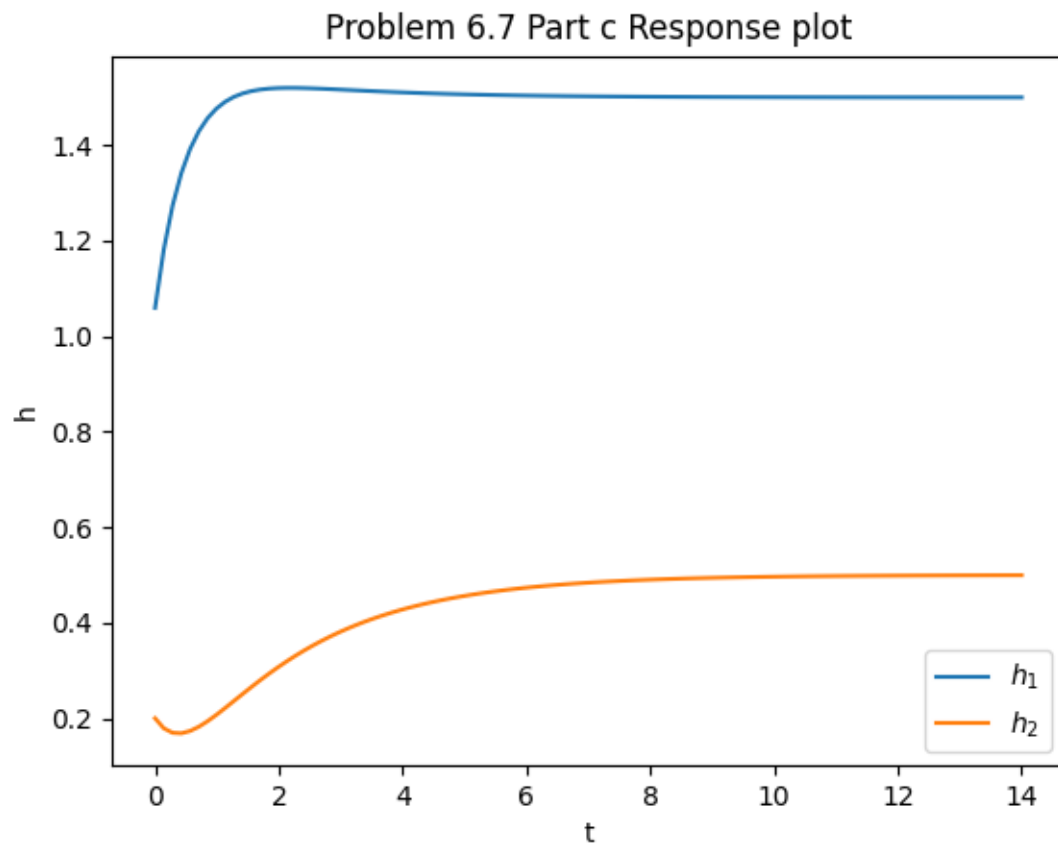
- (b) To find the time that $h_2 = 0.2$ m, the output arrays for the h_2 response were interpolated. With the time value, h_1 at that time was also interpolated from the h_1 response arrays. The code above shows the interpolation and output.

h_2 reaches 0.2 m at 1.1845min.

At that time, h_1 is 1.0594m.

- (c) To find the response of the heights when the flow is restored, another response simulation was made. The flow switching on is equivalent to a step response, and so a step response to the system will be simulated. The code above also contains the code for this simulation.

Response plot:



h_1 ends up at 1.5 m, and h_2 ends up at 0.5 m. Both heights return to their steady state values.

3. Problem 6.8

(a)

$$\begin{aligned}
 \tau \frac{dx_1}{dt} + x_1 &= ku \\
 \tau \frac{dx_2}{dt} + x_2 &= kx_1 \\
 \frac{dx_1}{dt} &= -\frac{x_1}{\tau} + \frac{k}{\tau}u \\
 \frac{dx_2}{dt} &= \frac{k}{\tau}x_1 - \frac{x_2}{\tau} \\
 A &= \begin{bmatrix} -\frac{1}{\tau} & 0 \\ \frac{k}{\tau} & -\frac{1}{\tau} \end{bmatrix} \\
 B &= \begin{bmatrix} \frac{k}{\tau} \\ 0 \end{bmatrix} \\
 sI - A &= \begin{bmatrix} s + \frac{1}{\tau} & 0 \\ -\frac{k}{\tau} & s + \frac{1}{\tau} \end{bmatrix} \\
 \text{Adj}(sI - A) &= \begin{bmatrix} s + \frac{1}{\tau} & 0 \\ \frac{k}{\tau} & s + \frac{1}{\tau} \end{bmatrix} \\
 \det(sI - A) &= (s + \frac{1}{\tau})^2 \\
 (sI - A)^{-1} &= \begin{bmatrix} \frac{1}{s + \frac{1}{\tau}} & 0 \\ \frac{\tau k}{\tau^2 s^2 + 2\tau s + 1} & \frac{1}{s + \frac{1}{\tau}} \end{bmatrix} \\
 \mathcal{L}^{-1} \{ (sI - A)^{-1} \} &= \begin{bmatrix} e^{-t/\tau} & 0 \\ \frac{k}{\tau} t e^{-t/\tau} & e^{-t/\tau} \end{bmatrix} \\
 e^{At} &= \begin{bmatrix} e^{-t/\tau} & 0 \\ \frac{k}{\tau} t e^{-t/\tau} & e^{-t/\tau} \end{bmatrix}
 \end{aligned}$$

(b)

$$\begin{aligned}
 x(t) &= \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} e^{-t/\tau} & 0 \\ \frac{k}{\tau} t e^{-t/\tau} & e^{-t/\tau} \end{bmatrix} \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} \\
 x(t) &= \begin{bmatrix} e^{-t/\tau} x_1(0) \\ \frac{k}{\tau} t e^{-t/\tau} x_1(0) + e^{-t/\tau} x_2(0) \end{bmatrix}
 \end{aligned}$$

(c)

$$\begin{aligned}A_d &= \begin{bmatrix} e^{-T_s/\tau} & 0 \\ \frac{k}{\tau}T_s e^{-T_s/\tau} & e^{-T_s/\tau} \end{bmatrix} \\B_d &= \int_0^{T_s} e^{At} B dt \\e^{At} B &= \begin{bmatrix} e^{-t/\tau} & 0 \\ \frac{k}{\tau}t e^{-t/\tau} & e^{-t/\tau} \end{bmatrix} \begin{bmatrix} \frac{k}{\tau} \\ 0 \end{bmatrix} \\e^{At} B &= \begin{bmatrix} \frac{k}{\tau} e^{-t/\tau} \\ \frac{k^2}{\tau^2} t e^{-t/\tau} \end{bmatrix} \\B_d &= \int_0^{T_s} \begin{bmatrix} \frac{k}{\tau} e^{-t/\tau} \\ \frac{k^2}{\tau^2} t e^{-t/\tau} \end{bmatrix} dt \\B_d &= \begin{bmatrix} -k e^{-t/\tau} \\ \frac{k^2}{\tau} (-\tau t e^{-t/\tau} - \tau^2 e^{-t/\tau}) \end{bmatrix}\end{aligned}$$

4. MATLAB Problem

(a)

$$\frac{d}{dt} \begin{bmatrix} C_R \\ C_I \\ C_P \end{bmatrix} = \begin{bmatrix} -\left(\frac{F}{V} + k_1\right) & 0 & 0 \\ k_1 & -\left(\frac{F}{V} + k_2\right) & k_3 \\ 0 & k_2 & -\left(\frac{F}{V} + k_3\right) \end{bmatrix} \begin{bmatrix} C_R \\ C_I \\ C_P \end{bmatrix} + \begin{bmatrix} \frac{F}{V} \\ 0 \\ 0 \end{bmatrix} C_{R0}$$

$$k_1 = 0.1$$

$$k_2 = 0.05$$

$$k_3 = 0.02$$

$$F = 10$$

$$V = 100$$

$$\frac{F}{V} = 0.1$$

$$C_{R0,s} = 2$$

$$A = \begin{bmatrix} -(0.1 + 0.1) & 0 & 0 \\ 0.1 & -(0.1 + 0.05) & 0.02 \\ 0 & 0.5 & -(0.1 + 0.02) \end{bmatrix}$$

$$A = \begin{bmatrix} -0.2 & 0 & 0 \\ 0.1 & -0.15 & 0.02 \\ 0 & 0.5 & -0.12 \end{bmatrix}$$

$$Bu = \begin{bmatrix} 0.1 \\ 0 \\ 0 \end{bmatrix} \cdot 2$$

$$Bu = \begin{bmatrix} 0.2 \\ 0 \\ 0 \end{bmatrix}$$

At steady state:

$$0 = Ax + Bu$$

$$x = -A^{-1}Bu$$

$$x = - \begin{bmatrix} -0.2 & 0 & 0 \\ 0.1 & -0.15 & 0.02 \\ 0 & 0.5 & -0.12 \end{bmatrix}^{-1} \begin{bmatrix} 0.2 \\ 0 \\ 0 \end{bmatrix}$$

$$x = \begin{bmatrix} 1 \\ 0.7059 \\ 0.2941 \end{bmatrix}$$

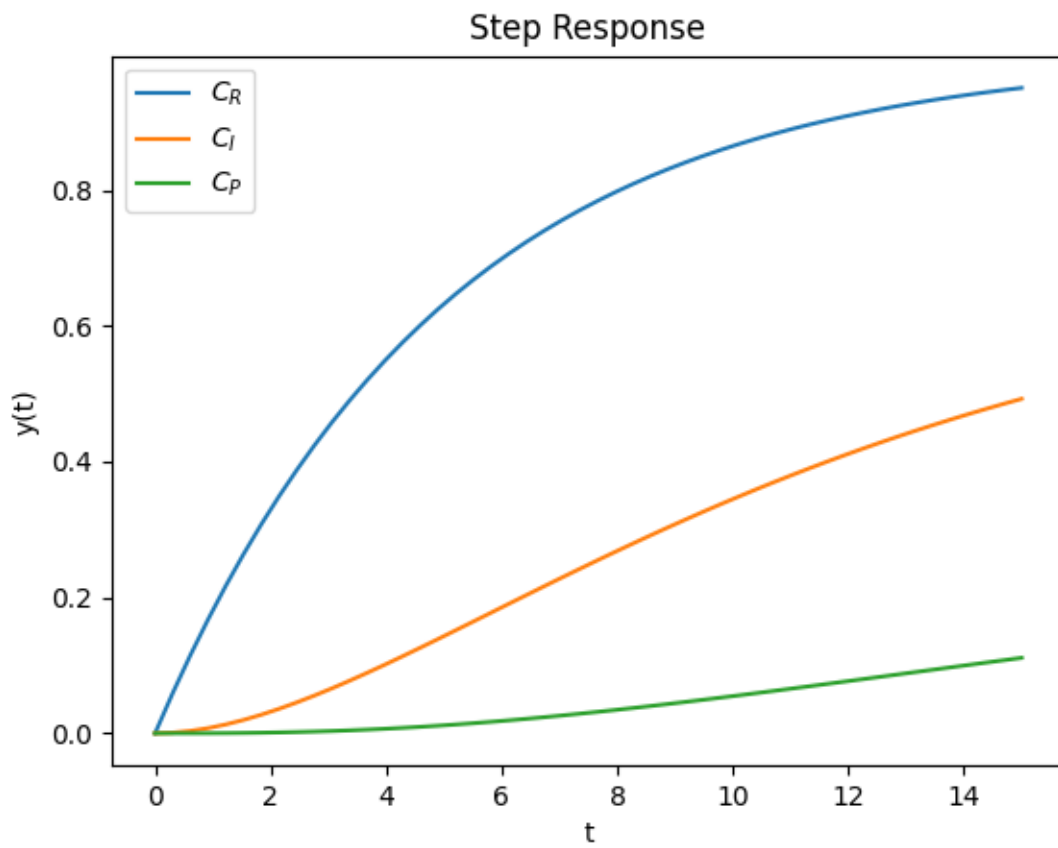
$$C_R = 1$$

$$C_I = 0.7059$$

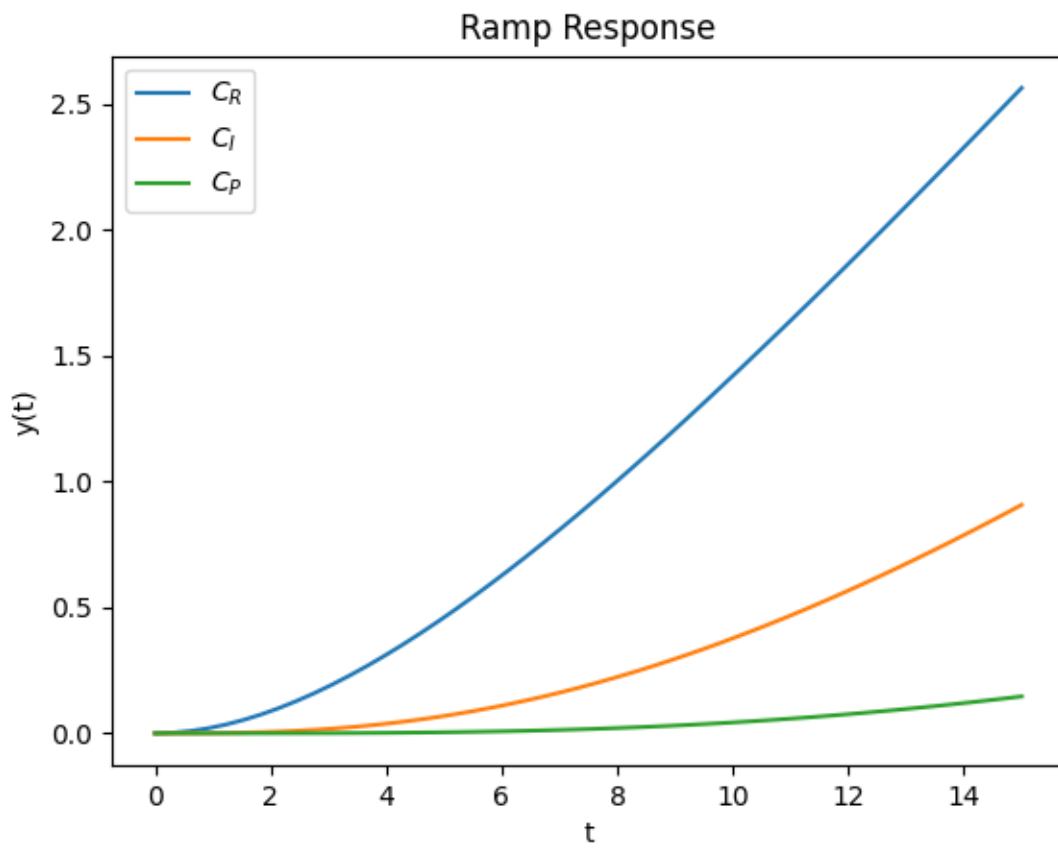
$$C_P = 0.2941$$

(b) Response plots:

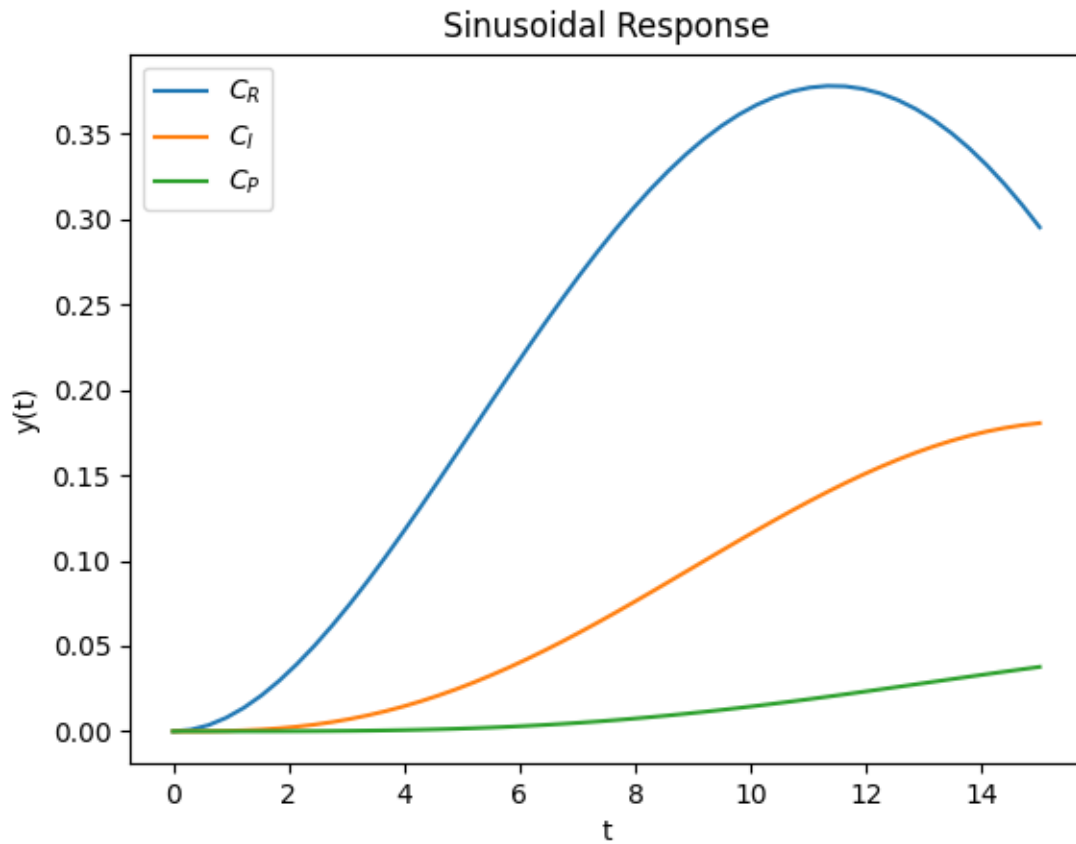
Step response:



Ramp response:



Sinusoidal response:



Code for creating the plots and calculating zero hold discretization:

```
import numpy as np
from scipy import linalg, signal
import matplotlib.pyplot as plt
import math

import pprint

def steady_state_composition(A, B) -> None:
    C_R0s = 2

    Bu = B * C_R0s

    steady_state_solution = -linalg.solve(A, Bu)

    print(steady_state_composition)

def step_response(t, M, A, B, c):
    step_response_formula = lambda t: M * (c @ (linalg.expm(A * t) - np.eye(3)) @
    ↪ linalg.inv(A) @ B)
    response = np.zeros(len(t))
```

```

for i, val in enumerate(t):
    response[i] = step_response_formula(val)[0]

return response

def ramp_response(t, M, A, B, c):
    ramp_response_formula = lambda t: M * (c @ ((linalg.expm(A * t) - np.eye(3)) @
    ↪ linalg.inv(A) - t * np.eye(3)) @ linalg.inv(A) @ B)
    response = np.zeros(len(t))

    for i, val in enumerate(t):
        response[i] = ramp_response_formula(val)[0]

    return response

def sinusoidal_response(t, M, omega, A, B, c):
    sinusoidal_response_formula = lambda t: c @ (omega * linalg.expm(A * t) - A *
    ↪ math.sin(omega * t) - omega * np.eye(3) * math.cos(omega * t)) @
    ↪ linalg.inv(np.linalg.matrix_power(A, 2) + omega**2 * np.eye(3)) @ B * M
    response = np.zeros(len(t))

    for i, val in enumerate(t):
        response[i] = sinusoidal_response_formula(val)[0]

    return response

def plot_step_response(A, B):
    M = 2

    c_C_R = np.array([1, 0, 0])
    c_C_I = np.array([0, 1, 0])
    c_C_P = np.array([0, 0, 1])

    t_plot = np.linspace(0, 15, 50)

    plt.plot(t_plot, step_response(t_plot, M, A, B, c_C_R))
    plt.plot(t_plot, step_response(t_plot, M, A, B, c_C_I))
    plt.plot(t_plot, step_response(t_plot, M, A, B, c_C_P))
    plt.title("Step Response")
    plt.xlabel("t")
    plt.ylabel("y(t)")
    plt.legend([r"$C_R$", r"$C_I$", r"$C_P$"])
    plt.show()

```

```

def plot_ramp_response(A, B):
    M = 0.5

    c_C_R = np.array([1, 0, 0])
    c_C_I = np.array([0, 1, 0])
    c_C_P = np.array([0, 0, 1])

    t_plot = np.linspace(0, 15, 50)

    plt.plot(t_plot, ramp_response(t_plot, M, A, B, c_C_R))
    plt.plot(t_plot, ramp_response(t_plot, M, A, B, c_C_I))
    plt.plot(t_plot, ramp_response(t_plot, M, A, B, c_C_P))
    plt.title("Ramp Response")
    plt.xlabel("t")
    plt.ylabel("y(t)")
    plt.legend([r"$C_R$", r"$C_I$", r"$C_P$"])
    plt.show()

def plot_sinusoidal_response(A, B):
    M = 1
    omega = 0.2

    c_C_R = np.array([1, 0, 0])
    c_C_I = np.array([0, 1, 0])
    c_C_P = np.array([0, 0, 1])

    t_plot = np.linspace(0, 15, 50)

    plt.plot(t_plot, sinusoidal_response(t_plot, M, omega, A, B, c_C_R))
    plt.plot(t_plot, sinusoidal_response(t_plot, M, omega, A, B, c_C_I))
    plt.plot(t_plot, sinusoidal_response(t_plot, M, omega, A, B, c_C_P))
    plt.title("Sinusoidal Response")
    plt.xlabel("t")
    plt.ylabel("y(t)")
    plt.legend([r"$C_R$", r"$C_I$", r"$C_P$"])
    plt.show()

def main():
    k_1 = 0.1
    k_2 = 0.05
    k_3 = 0.02
    F = 10
    V = 100

    A = np.array([
        -(F / V + k_1), 0, 0],

```

```

    [k_1, -(F / V + k_2), k_3],
    [0, k_2, -(F / V + k_3)],
])

B = np.array([
    [F/V],
    [0],
    [0],
])

steady_state_composition(A, B)

plot_step_response(A, B)

plot_ramp_response(A, B)

plot_sinusoidal_response(A, B)

C = np.array([1, 1, 1])
D = 0
sys = signal.lti(A, B, C, D)

pp = pprint.PrettyPrinter(indent=4)

pp.pprint(signal.cont2discrete((A, B, C, D), 0.2))

if __name__ == "__main__":
    main()

```

(c) Zero hold discretization was calculated with the code in the previous part.

$$A_d = \begin{bmatrix} 0.960789439 & 0 & 0 \\ 0.0193123181 & 0.970464981 & 3.89347676 \cdot 10^{-5} \\ 9.69160980 \cdot 10^{-5} & 9.73369191 \cdot 10^{-3} & 0.976305197 \end{bmatrix}$$

$$B_d = \begin{bmatrix} 0.0196052804 \\ 1.95395071 \cdot 10^{-4} \\ 6.51198080 \cdot 10^{-7} \end{bmatrix}$$