

```
In [1]: import sympy as sp
import numpy as np
import matplotlib.pyplot as plt

g, m , rho, Vb, veq = sp.symbols('g m rho V_b V_eq')

v, vc, y, ym, th, th1, th2, B, p, a, b, e, gamma = sp.symbols('v v_c y y_m theta theta_1 theta_2 B p a b e gamma')
u, uc, zeta, omega = sp.symbols('u u_c zeta omega')

In [16]: y_u = B*u/(p**2+e*B)
y_m_uc = (omega**2)*uc/(p**2+zeta*omega*p+omega**2)

u_u = th1*(y-uc)-th2*p*y
y_uc = -B*th1*uc/(p**2+p*(B+B*th2))-B*th1
error = y_uc - ym_uc

uc_y = (p**2+p*(B+B*th2)-B*th1)*y/(-B*th1)
uc_ym = (p**2+zeta*omega*p+omega**2)*ym/(omega**2)

In [30]: de_th1 = sp.diff(error,th1).simplify()
de_th1

Out[30]: 
$$-\frac{Bpu_c(B(\theta_2+1)+p)}{(Bp(\theta_2+1)-B\theta_1+p^2)^2}$$


In [21]: de_th2 = sp.diff(error,th2).simplify()
de_th2

Out[21]: 
$$\frac{B^2p\theta_1u_c}{(Bp(\theta_2+1)-B\theta_1+p^2)^2}$$


In [25]: de_th1.subs(uc, uc_y).simplify()

Out[25]: 
$$\frac{py(B(\theta_2+1)+p)}{\theta_1(Bp(\theta_2+1)-B\theta_1+p^2)}$$


In [27]: de_th2.subs(uc, uc_y).simplify()

Out[27]: 
$$\frac{Bpy}{Bp(\theta_2+1)-B\theta_1+p^2}$$


In [28]: error

Out[28]: 
$$-\frac{B\theta_1u_c}{-B\theta_1+p^2+p(B\theta_2+B)}-\frac{\omega^2u_c}{\omega^2+2\omega p\zeta+p^2}$$


In [243]: #constants
ZETA = 1
OMEGA = 1

#
radius = 38/1000
g = 9.8
m = 4/7000
rho = 1.225
Vb = (4/3)*np.pi*(radius**3)
v_eq=2.8
B = 2*g*(m-rho*Vb)/(m*v_eq)
#B=2

#adaptation law
GAMMA = 0.5
alpha = 0.75
#simulation params
ts = 0.2
maxt = 90
n_steps = int(maxt/ts)

def u_signal(t):
    function_val = np.sin(np.pi*t/15)
    if function_val >= 0:
        result = 0.1
    else:
        result = 0.2
    return result

def system_model(uc_i, ym_pl_i, ym_i):
    ym_p2_o = (OMEGA**2)*uc_i - 2*ZETA*OMEGA*ym_pl_i - (OMEGA**2)*ym_i
    ym_pl_o = ym_pl_i + ts*ym_p2_o
    ym_o = ym_i + ts*ym_pl_o

    system_model_dict = {"ym_p2":ym_p2_o,
                        "ym_pl":ym_pl_o,
                        "ym":ym_o}

    return system_model_dict

def system_sim(y_pl_i,y_i,u_i):
    y_p2_o = B*u_i - B*y_pl_i

    y_pl_o = y_pl_i + ts*y_p2_o

    y_o = y_i + ts*y_pl_o + np.random.normal(0.0, 0.001)

    system_dict = {"y_p2":y_p2_o,
                  "y_pl":y_pl_o,
                  "y":y_o}

    return system_dict

def controller(y_i, uc_i, y_pl_i, th1, th2):
    control_signal = th1*(y_i - uc_i) - th2*y_pl_i
    return control_signal

def Adaptation_Law_Model(y_i, ym_i, ym_pl_i, ym_p2_i,
                        thetal_p2_i, thetal_pl_i, thetal_i,
                        theta2_p2_i, theta2_pl_i, theta2_i):

    error = y_i - ym_i

    thetal_p3_o = (GAMMA*error*B/OMEGA**2)*(2*ZETA*OMEGA*ym_pl_i+ym_p2_i) - 2*ZETA*OMEGA*thetal_p2_i - (OMEGA**2)*thetal_p2_o
    thetal_p2_o = thetal_p2_i + ts*thetal_p3_o
    thetal_pl_o = thetal_pl_i + ts*thetal_p2_o
    thetal_o = thetal_i + ts*thetal_pl_o

    theta2_p3_o = GAMMA*error*B*ym_pl_i - 2*ZETA*OMEGA*theta2_p2_i - (OMEGA**2)*theta2_p2_i
    theta2_p2_o = theta2_p2_i + ts*theta2_p3_o
    theta2_pl_o = theta2_pl_i + ts*theta2_p2_o
    theta2_o = theta2_i + ts*theta2_pl_o

    result_dict = {
        "thetal_p2": thetal_p2_o,
        "thetal_pl": thetal_pl_o,
        "thetal": thetal_o,
        "thetal_p1": thetal_p2_o,
        "thetal2_p1": theta2_p2_o,
        "thetal2": theta2_o
    }

    return result_dict

def Adaptation_Law_Normalized(y_i, ym_i, ym_pl_i, ym_p2_i,
                        thetal_p2_i, thetal_pl_i, thetal_i,
                        theta2_p2_i, theta2_pl_i, theta2_i,
                        thetal_pl_n, thetal_n_i,
                        theta2_pl_n, theta2_n_i):

    error_val = y_i - ym_i

    if error_val !=0:
        error = error_val
    else:
        error = 1/(10**100)

    thetal_p3_o = (GAMMA*error/OMEGA**2)*(2*ZETA*OMEGA*ym_pl_i+ym_p2_i) - 2*ZETA*OMEGA*thetal_p2_i - (OMEGA**2)*thetal_p2_o
    thetal_p2_o = thetal_p2_i + ts*thetal_p3_o
    thetal_pl_o = thetal_pl_i + ts*thetal_p2_o
    thetal_o = thetal_i + ts*thetal_pl_o

    thetal_pl_n = thetal_pl_o/(alpha+(thetal_pl_o/(GAMMA*error))**2)
    thetal_n = thetal_n_i + ts*thetal_pl_n

    theta2_p3_o = GAMMA*error*ym_pl_i - 2*ZETA*OMEGA*theta2_p2_i - (OMEGA**2)*theta2_p2_i
    theta2_p2_o = theta2_p2_i + ts*theta2_p3_o
    theta2_pl_o = theta2_pl_i + ts*theta2_p2_o
    theta2_o = theta2_n_i + ts*theta2_pl_o

    theta2_pl_n = theta2_pl_o/(alpha+(theta2_pl_o/(GAMMA*error))**2)
    theta2_n = theta2_n_i + ts*theta2_pl_n

    result_dict = {
        "thetal_p2": thetal_p2_o,
        "thetal_pl": thetal_pl_o,
        "thetal": thetal_o,
        "thetaln_pl":thetal_pl_n,
        "thetal_n":thetal_n,
        "thetal2_p2": theta2_p2_o,
        "thetal2n_pl":theta2_pl_n,
        "thetal2n":theta2_n,
        "thetal2_n":theta2_n_i
    }

    return result_dict
```

## Regular MIT Rule

```
In [325]: i = 0

#initial state of system
y_pl_i = 0
y_i = 0
u_i = 0

#initial state of model
ym_pl_i = 0
ym_i = 0

thetal_p2_i, thetal_pl_i, theta2_p2_i, theta2_pl_i = 0,0,0,0
thetal_pl_n, theta2_pl_n = 0,0
thetal_i = -(OMEGA**2)/B
#theta2_i = (2*ZETA*OMEGA)/(1+B)
theta2_i = (2*ZETA*OMEGA-B)/B
thetal_n_i = thetal_i/(alpha + thetal_i**2)
theta2_n_i = theta2_i/(alpha + theta2_i**2)

#thetal_i = 0.1
#theta2_i = 0.1

y_list = []
ym_list = []
u_list = []
tc= []
error_list = []

uc = []
for i in range(int(maxt/ts)+1):
    uc.append(u_signal(i*ts))

thetal_list = []
theta2_list = []

i=0
for k in range(n_steps):
    #system
    system_results = system_sim(y_pl_i,y_i,u_i)
    y_i = system_results["y"]
    y_pl_i = system_results["y_pl"]

    #model
    model_results = system_model(uc_i, ym_pl_i, ym_i)
    ym_p2_i = model_results["ym_p2"]
    ym_pl_i = model_results["ym_pl"]
    ym_i = model_results["ym"]

    adapt_val = Adaptation_Law_Model(y_i, ym_i, ym_pl_i, ym_p2_i,
                                    thetal_p2_i, thetal_pl_i, thetal_i,
                                    theta2_p2_i, theta2_pl_i, theta2_i)

    thetal_p2_i, thetal_pl_i, thetal_i = adapt_val["thetal_p2"], adapt_val["thetal_pl"], adapt_val["thetal"]
    theta2_p2_i, theta2_pl_i, theta2_i = adapt_val["theta2_p2"], adapt_val["theta2_pl"], adapt_val["theta2"]

    uc_i = uc[i]
    #uc_i = 70
    u_i = controller(y_i, uc_i, y_pl_i, thetal_i, theta2_i)

    thetal_list.append(thetal_i)
    theta2_list.append(theta2_i)

    y_list.append(y_i)
    u_list.append(u_i)
    ym_list.append(ym_i)
    error_list.append(y_i-ym_i)

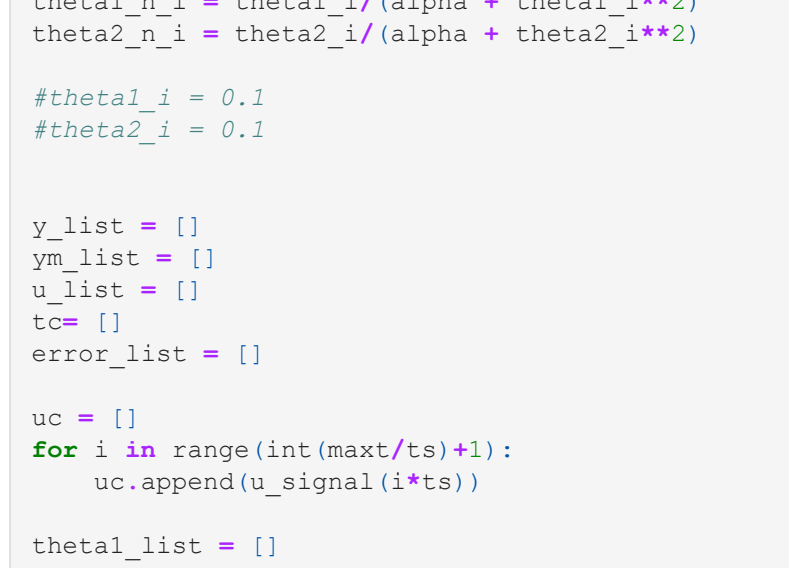
    tc.append(i)
    i = i+1

print("regular MIT rule simulation complete")

regular MIT rule simulation complete

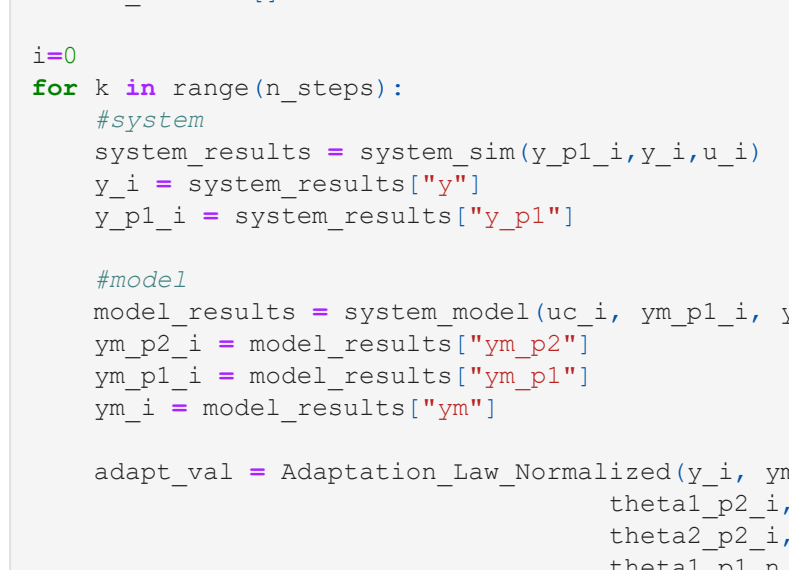
In [326]: start_idx = 0
end_idx = int(maxt/ts)
plt.plot(tc[start_idx:end_idx], y_list[start_idx:end_idx], label = "y")
plt.plot(tc[start_idx:end_idx], ym_list[start_idx:end_idx], label = "ym")
plt.plot(tc[start_idx:end_idx], uc[start_idx:end_idx], label = "uc")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[326]: <matplotlib.legend.Legend at 0x2318aa52fd0>



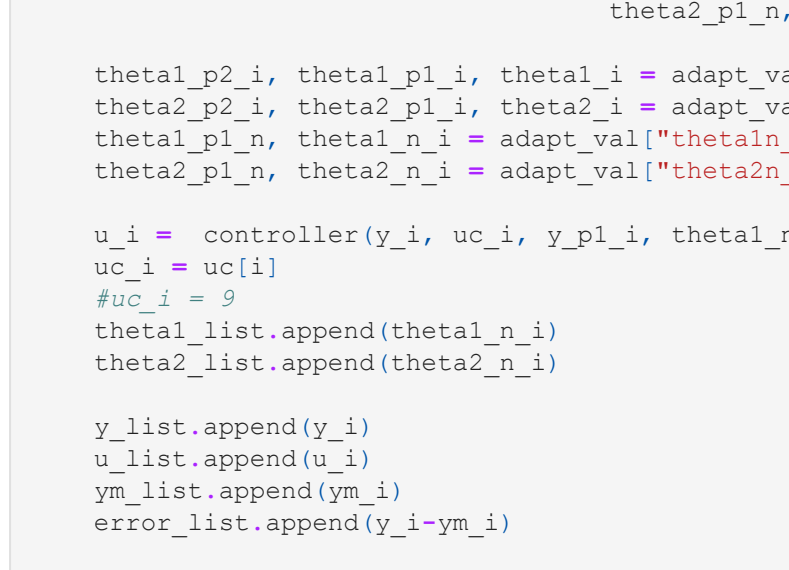
In [327]: start_idx = 0
end_idx = int(maxt/ts)
#plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.plot(tc[start_idx:end_idx], ym_list[start_idx:end_idx], label = "ym")
plt.plot(tc[start_idx:end_idx], error_list[start_idx:end_idx], label = "y - ym")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[327]: <matplotlib.legend.Legend at 0x2318a99b2e0>



In [328]: start_idx = 0
end_idx = int(maxt/ts)
plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.plot(tc[start_idx:end_idx], thetal_list[start_idx:end_idx], label = "thetal")
plt.plot(tc[start_idx:end_idx], theta2_list[start_idx:end_idx], label = "theta2")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()


Out[328]: <matplotlib.legend.Legend at 0x2318ab28580>



In [329]: start_idx = 0
end_idx = int(maxt/ts)

plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[329]: <matplotlib.legend.Legend at 0x2318ab496a0>


```

## normalized MIT Rule

```
In [320]: i = 0

#system input
uc_i = 1

#initial state of system
y_pl_i = 0
y_i = 0
u_i = 0

#initial state of model
ym_pl_i = 0
ym_i = 0

thetal_p2_i, thetal_pl_i, theta2_p2_i, theta2_pl_i = 0,0,0,0
thetal_pl_n, theta2_pl_n = 0,0
thetal_i = -(OMEGA**2)/B
#theta2_i = (2*ZETA*OMEGA)/(1+B)
theta2_i = (2*ZETA*OMEGA-B)/B
thetal_n_i = thetal_i/(alpha + thetal_i**2)
theta2_n_i = theta2_i/(alpha + theta2_i**2)

#thetal_i = 0.1
#theta2_i = 0.1

y_list = []
ym_list = []
u_list = []
tc= []
error_list = []

uc = []
for i in range(int(maxt/ts)+1):
    uc.append(u_signal(i*ts))

thetal_list = []
theta2_list = []

i=0
for k in range(n_steps):
    #system
    system_results = system_sim(y_pl_i,y_i,u_i)
    y_i = system_results["y"]
    y_pl_i = system_results["y_pl"]

    #model
    model_results = system_model(uc_i, ym_pl_i, ym_i)
    ym_p2_i = model_results["ym_p2"]
    ym_pl_i = model_results["ym_pl"]
    ym_i = model_results["ym"]

    adapt_val = Adaptation_Law_Normalized(y_i, ym_i, ym_pl_i, ym_p2_i,
                                    thetal_p2_i, thetal_pl_i, thetal_i,
                                    theta2_p2_i, theta2_pl_i, theta2_i,
                                    thetal_pl_n, thetal_n_i,
                                    theta2_pl_n, theta2_n_i)

    thetal_p2_i, thetal_pl_i, thetal_i = adapt_val["thetal_p2"], adapt_val["thetal_pl"], adapt_val["thetal"]
    theta2_p2_i, theta2_pl_i, theta2_i = adapt_val["theta2_p2"], adapt_val["theta2_pl"], adapt_val["theta2"]
    thetal_pl_n, theta2_pl_n = adapt_val["thetaln_pl"], adapt_val["thetaln"]
    theta2_pl_n, theta2_n_i = adapt_val["theta2n_pl"], adapt_val["theta2n"]

    u_i = controller(y_i, uc_i, y_pl_i, thetal_n_i, theta2_n_i)
    uc_i = uc[i]
    #uc_i = 9
    thetal_list.append(thetal_n_i)
    theta2_list.append(theta2_n_i)

    y_list.append(y_i)
    u_list.append(u_i)
    ym_list.append(ym_i)
    error_list.append(y_i-ym_i)

    tc.append(i*ts)
    i = i+1

print("normalized MIT Rule Simulation Complete")

normalized MIT Rule Simulation Complete

In [321]: start_idx = 0
end_idx = int(maxt/ts)
plt.plot(tc[start_idx:end_idx], y_list[start_idx:end_idx], label = "y")
plt.plot(tc[start_idx:end_idx], ym_list[start_idx:end_idx], label = "ym")
plt.plot(tc[start_idx:end_idx], uc[start_idx:end_idx], label = "uc")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[321]: <matplotlib.legend.Legend at 0x2318a8a0fa0>



In [322]: start_idx = 0
end_idx = int(maxt/ts)
plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[322]: <matplotlib.legend.Legend at 0x2318a912f10>



In [323]: start_idx = 0
end_idx = int(maxt/ts)
#plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.plot(tc[start_idx:end_idx], ym_list[start_idx:end_idx], label = "ym")
plt.plot(tc[start_idx:end_idx], error_list[start_idx:end_idx], label = "y - ym")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[323]: <matplotlib.legend.Legend at 0x2318a7f17c0>



In [324]: start_idx = 0
end_idx = int(maxt/ts)
#plt.plot(tc[start_idx:end_idx], u_list[start_idx:end_idx], label = "u")
plt.plot(tc[start_idx:end_idx], thetal_list[start_idx:end_idx], label = "thetal")
plt.plot(tc[start_idx:end_idx], theta2_list[start_idx:end_idx], label = "theta2")
plt.xlabel("time (s)")
plt.ylabel("parameter")
plt.legend()

Out[324]: <matplotlib.legend.Legend at 0x2318a9d8790>


```

In [13]:

In [ ]: