

Final_HW3_MAE207

February 11, 2020

0.1 Homework 3 MAE-207

```
[1]: import time
from IPython.display import clear_output
import numpy as np
import matplotlib as mpl

# import sympy
from sympy import symbols, pprint
from sympy import sin, cos, asin, acos, atan, pi, sqrt
from sympy import Matrix
from IPython.core.interactiveshell import InteractiveShell
InteractiveShell.ast_node_interactivity = "all"
from sympy import lambdify
# import numpy as np

import matplotlib.pyplot as plt
import matplotlib.animation as animation
from matplotlib import style

from IPython import display # for the animation
# import matplotlib as mpl
# mpl.use('Qt5Agg')
# plt.ion()
import os
import matplotlib.pyplot as plt
# plt.ion()

# for the symbolic manipulation of jacobian
import sympy as sp
# from sympy import symbols
# from sympy import sin, cos, asin, acos, pi, atan2, sqrt
from sympy.utilities.lambdify import lambdify
# from sympy import Matrix

from scipy.optimize import minimize
from scipy.optimize import fsolve
```

```
import time

import odrive
from odrive.utils import *
import odrive.enums
%matplotlib qt
```

0.2 Connect to ODrive

```
[ ]: odrv0 = odrive.find_any()
if odrv0 is not None:
    print('Connected!')
    print('Odrive serial {}'.format(odrv0.serial_number))

    m0 = odrv0.axis0.motor.is_calibrated
    m1 = odrv0.axis1.motor.is_calibrated

    print('Motor 0 calibrated: {}'.format(m0))
    print('Motor 1 calibrated: {}'.format(m1))

else:
    print('Not connected')
```

```
[3]: m0 = odrv0.axis0.motor.is_calibrated
m1 = odrv0.axis1.motor.is_calibrated

print('Motor 0 calibrated: {}'.format(m0))
print('Motor 1 calibrated: {}'.format(m1))
```

Motor 0 calibrated: True
Motor 1 calibrated: True

Setting the velocity and calibration current limit

```
[5]: odrv0.axis0.controller.config.vel_limit=200000
odrv0.axis1.controller.config.vel_limit=200000
odrv0.axis0.motor.config.calibration_current=20.0
odrv0.axis1.motor.config.calibration_current=20.0

odrv0.axis0.controller.config.vel_gain = 16 / 10000.0
odrv0.axis1.controller.config.vel_gain = 16 / 10000.0
odrv0.axis1.controller.config.pos_gain=44
odrv0.axis1.controller.config.pos_gain=44

odrv0.axis0.controller.config.vel_integrator_gain = 0.001 #!/ 10000.0
```

```
odrv0.axis1.controller.config.vel_integrator_gain = 0.001 #5.0 / 10000.0
```

Calibrating the Motors

```
[33]: odrv0.axis0.requested_state = odrive.enums.AXIS_STATE_FULL_CALIBRATION_SEQUENCE
odrv0.axis1.requested_state = odrive.enums.AXIS_STATE_FULL_CALIBRATION_SEQUENCE

time.sleep(15)

print('\t Motor 0 calibration result: {} \r\n'.format(odrv0.axis0.motor.
    ↪is_calibrated),
      '\t Motor 1 calibration result: {}' .format(odrv0.axis1.motor.
    ↪is_calibrated))
```

```
Motor 0 calibration result: True
Motor 1 calibration result: True
```

Function to make the motors return to idle mode

```
[6]: # Function to make the motors return to idle mode
def motor_idle():
    odrv0.axis0.requested_state = odrive.enums.AXIS_STATE_IDLE
    odrv0.axis1.requested_state = odrive.enums.AXIS_STATE_IDLE
```

```
[2]: # Defining the Symbols
(theta_l,
 theta_r,
 l1,
 l2,
 w) = symbols(""" theta_l
                  theta_r
                  l1
                  l2
                  w """, real = True)
```

Forward Kinematics definition

```
[3]: A = 2*l2*l1*sin(theta_l)- 2*l1*l2*sin(theta_r)
B = 2*l2*w - 2*l1*l2*cos(theta_r) + 2*l2*l1*cos(theta_l)
C = 2*(l1**2) + w**2 - 2*l1*l1*sin(theta_l)*sin(theta_r) - 2*l1*w*cos(theta_r)
    ↪+ 2*l1*w*cos(theta_l) - 2*l1*l1*cos(theta_l)*cos(theta_r)

theta3_1 = 2*atan((A + sqrt(A**2 + B**2 - C**2))/(B-C))
theta3_2 = 2*atan((A - sqrt(A**2 + B**2 - C**2))/(B-C))

xc_1 = l1*cos(theta_l) + l2*cos(theta3_1) + w/2
xc_1
xc_2 = l1*cos(theta_l) + l2*cos(theta3_2) + w/2
```

```

xc_2

yc_1 = l1*sin(theta_1) + l2*sin(theta3_1)
yc_1
yc_2 = l1*sin(theta_1) + l2*sin(theta3_2)
yc_2
#sympy.simplify(theta3_1)
#sympy.simplify(theta3_2)

```

[3]:

$$l_1 \cos(\theta_l) + l_2 \cos \left(2 \operatorname{atan} \left(\frac{2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r) + \sqrt{(2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r))^2 + (2l_1 l_2 \cos(\theta_l) - 2l_1 l_2 \cos(\theta_r))^2}}{2l_1^2 \sin(\theta_l) \sin(\theta_r) + 2l_1^2 \cos(\theta_l) \cos(\theta_r) - 2l_1^2 + 2l_1 l_2} \right) \right) \frac{w}{2}$$

[3]:

$$l_1 \cos(\theta_l) + l_2 \cos \left(2 \operatorname{atan} \left(\frac{2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r) - \sqrt{(2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r))^2 + (2l_1 l_2 \cos(\theta_l) - 2l_1 l_2 \cos(\theta_r))^2}}{2l_1^2 \sin(\theta_l) \sin(\theta_r) + 2l_1^2 \cos(\theta_l) \cos(\theta_r) - 2l_1^2 + 2l_1 l_2} \right) \right) \frac{w}{2}$$

[3]:

$$l_1 \sin(\theta_l) + l_2 \sin \left(2 \operatorname{atan} \left(\frac{2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r) + \sqrt{(2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r))^2 + (2l_1 l_2 \cos(\theta_l) - 2l_1 l_2 \cos(\theta_r))^2}}{2l_1^2 \sin(\theta_l) \sin(\theta_r) + 2l_1^2 \cos(\theta_l) \cos(\theta_r) - 2l_1^2 + 2l_1 l_2} \right) \right)$$

[3]:

$$l_1 \sin(\theta_l) + l_2 \sin \left(2 \operatorname{atan} \left(\frac{2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r) - \sqrt{(2l_1 l_2 \sin(\theta_l) - 2l_1 l_2 \sin(\theta_r))^2 + (2l_1 l_2 \cos(\theta_l) - 2l_1 l_2 \cos(\theta_r))^2}}{2l_1^2 \sin(\theta_l) \sin(\theta_r) + 2l_1^2 \cos(\theta_l) \cos(\theta_r) - 2l_1^2 + 2l_1 l_2} \right) \right)$$

[4]:

```

FK1 = Matrix([[xc_1],
               [yc_1]])
FK2 = Matrix([[xc_2],
               [yc_2]])

```

[5]:

```

FK1_fast = lambdify((theta_l, theta_r, l1, l2, w), FK1)
FK2_fast = lambdify((theta_l, theta_r, l1, l2, w), FK2)

```

Function to setup the Home Position

[6]:

```

# Function to setup home position
def autohome():
    print('Bring motors to home and press ENTER')
    input("<enter>")
    R_pos1=odrv0.axis0.encoder.pos_estimate
    L_pos1=odrv0.axis1.encoder.pos_estimate
    print('Bring motors to 90-90 and press ENTER')
    input("<enter>")
    R_pos2=odrv0.axis0.encoder.pos_estimate
    L_pos2=odrv0.axis1.encoder.pos_estimate
    ML_para=(90/(L_pos2-L_pos1))

```

```

MR_para=((90)/(R_pos2-R_pos1))
print(ML_para)
print(MR_para)
print(L_pos2, L_pos1, R_pos2, R_pos1)
print("Done")
return ML_para,MR_para,L_pos1,R_pos1

```

Code to check the home position and record angle data (if needed)

```

[ ]: theta=[]
cal_para=autohome()
while True:
    theta_R=(np.around(odrv0.axis0.encoder.
    ↪pos_estimate)-cal_para[3])*(cal_para[1])+180
    theta_L=((np.around(odrv0.axis1.encoder.
    ↪pos_estimate)-cal_para[2])*cal_para[0])
    print(theta_L, theta_R)
    clear_output()
#     os.system('cls')

    theta.append([theta_R,theta_L])

# Angle_value = np.asarray(theta)
# np.save('Angle_Workspace',Angle_value)

```

Function to implement Inverse Kinematics using Minimize function

```

[7]: l1 = 0.09;
l2 = 0.16;
w = 0.07;
def IK_5_link(x, y, l1 = l1, l2 = l2, w = w):

    def leg_wide(var):
        return np.linalg.norm([var[1] - np.pi, var[0]])

    def x_constraint_equation(var):
        return l1**2 - l2**2 + (x - w/2)**2 + y**2 - 2*l1*(y*np.sin(var[0]) +
    ↪(x - w/2)*np.cos(var[0]))

    def y_constraint_equation(var):
        return l1**2 - l2**2 + (x + w/2)**2 + y**2 - 2*l1*(y*np.sin(var[1]) +
    ↪(x + w/2)*np.cos(var[1]))

    res = minimize(leg_wide, (0.1, 9*np.pi/10), method="SLSQP", constraints=
    ↪({"type": "eq", "fun": x_constraint_equation},

```

```

→{"type": "eq", "fun": y_constraint_equation}))

    return (res, np.linalg.norm([x_constraint_equation(res.x),
→y_constraint_equation(res.x)]))

```

Code to implement the GAITs provided to us. This function returns a list of θ_L and θ_R for a giveb gait trajectory. These are later used to make the leg move in the desired trajectory

```

[8]: def get_trajectory(x,y,d,T,N):

    theta_points = []
    # Stance
    z = np.polyfit([x[0],x[1]], [y[0],y[1]],1)
    p = np.poly1d(z)
    n_stance = (np rint(N*d))
    n_stance = n_stance.astype(int)
    x_stancepoints = np.linspace(x[0],x[1],n_stance+1)
    y_stancepoints = p(x_stancepoints)

    # Swing
    z = np.polyfit(x,y,2)
    p = np.poly1d(z)
    n_swing = np rint((N*(1-d)))
    n_swing = n_swing.astype(int)
    x_swingpoints = np.linspace(x[1],x[0],n_swing+1)
    y_swingpoints = p(x_swingpoints)

    x_points = np.append(x_stancepoints,x_swingpoints[1:])
    y_points = np.append(y_stancepoints,y_swingpoints[1:])

    t_points = np.linspace(0,T,N+1)

    for i in range(x_points.shape[0]):

        res = IK_5_link(x_points[i],y_points[i])
        theta_points.append([res[0].x[1],res[0].x[0]])

    return theta_points,t_points

```

```

[131]: x = np.array([0.11,-0.11,0])
        y = np.array([0.12+0.05,0.12+0.05,0.05+0.05])

        data1, t1_points = get_trajectory(x,y,0.5,0.25,100)

        x = np.array([0.11-0.05,-0.08-0.05,0.05-0.05])

```

```

y = np.array([0.14+0.05,0.1+0.05,0.05+0.05])

data2, t2_points = get_trajectory(x,y,0.75,0.5,100)

x = np.array([0.08,-0.08,0])
y = np.array([0.1+0.05,0.1+0.05,0.04+0.05])

data3, t3_points = get_trajectory(x,y,0.5,1.5,200)

```

Store the θ_L and θ_R to a numpy file to plot the data

```

[ ]: np.save('theta_values_Data_2',data1)
      np.save('theta_data3',data2)
      np.save('theta_data4',data3)

```

Code to implement the parabolic gait trajectories (using angle evaluted from IK module above) and store the real motor angles to plot later This code is ran 3 times for each θ_L and θ_R pertaining to each case i.e. 1,2 and 3.

```

[97]: theta_Gait=[]
      odrv0.axis0.requested_state = odrive.enums.AXIS_STATE_CLOSED_LOOP_CONTROL
      odrv0.axis1.requested_state = odrive.enums.AXIS_STATE_CLOSED_LOOP_CONTROL
      a=np.load('data3_new3.npy')
      start_time = time.time()
      # your code
      cpr= 2048
      t_step = 0.005
      w_r = 0
      w_l = 0
      for k in range(0,10):
          for i in range(0,len(a)):
              start_time2 = time.time()
              L_pos=(np.rad2deg(a[i][1] )/cal_para[0])+cal_para[2]
              R_pos=((np.rad2deg(a[i][0] )-180)/cal_para[1])+cal_para[3]
              if (i-1)>0:
                  w_r = ((a[i][0] - a[i-1][0])/(t_step*2*np.pi) )*cpr
                  w_l = ((a[i][1] - a[i-1][1])/(t_step*2*np.pi) )*cpr
                  odrv0.axis0.controller.set_pos_setpoint(R_pos,1*w_r,0)
                  odrv0.axis1.controller.set_pos_setpoint(L_pos,1*w_l,0)
                  while (time.time() - start_time2 < 0.2*t_step):
                      pass
                      #clear_output()
      #      time.sleep(0.1)
      theta_R=(np.around(odrv0.axis0.encoder.
      ↪pos_estimate)-cal_para[3])*(cal_para[1])+180
      theta_L=((np.around(odrv0.axis1.encoder.
      ↪pos_estimate)-cal_para[2])*cal_para[0])

```

```

        theta_Gait.append([theta_R,theta_L])
# motor_idle()
elapsed_time = time.time() - start_time
print(elapsed_time)

```

4.088146209716797

```
[98]: np.save('theta_data3.npy',theta_Gait)
```

```
[15]: # Actual theta
theta_Gait1=np.load('theta_values_Data_2.npy')
theta_Gait2=np.load('theta_data3.npy')
theta_Gait3=np.load('theta_data4.npy')
```

```
[16]: # Calculated from IK
theta_cal1=np.load('data2.npy')
theta_cal2=np.load('data3_new3.npy')
theta_cal3=np.load('data4.npy')
```

Plotting the variatio of θ_L and θ_R

```
[17]: a1=theta_cal1
b1=theta_cal2
c1=theta_cal3
for i in range(0,9):
    a1=np.concatenate((a1,theta_cal1),axis=0)
    b1=np.concatenate((b1,theta_cal2),axis=0)
    c1=np.concatenate((c1,theta_cal3),axis=0)
```

```
[18]: %matplotlib inline
plt.plot(theta_Gait1[:,0])
plt.plot(np.rad2deg(a1[:,0]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_R$')
plt.title(r'$\theta_R$ variation plot CASE1')
# plt.gca().set_aspect('equal')
plt.grid()
plt.legend(['Actual Theta','Commanded Theta'])
plt.show()

plt.plot(theta_Gait1[:,1])
plt.plot(np.rad2deg(a1[:,1]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_L$')
plt.title(r'$\theta_L$ variation plot CASE:1')
plt.legend(['Actual Theta','Commanded Theta'])
# plt.gca().set_aspect('equal')
```



```
plt.grid()
```

```
[18]: [<matplotlib.lines.Line2D at 0x1ae7d892948>]
```

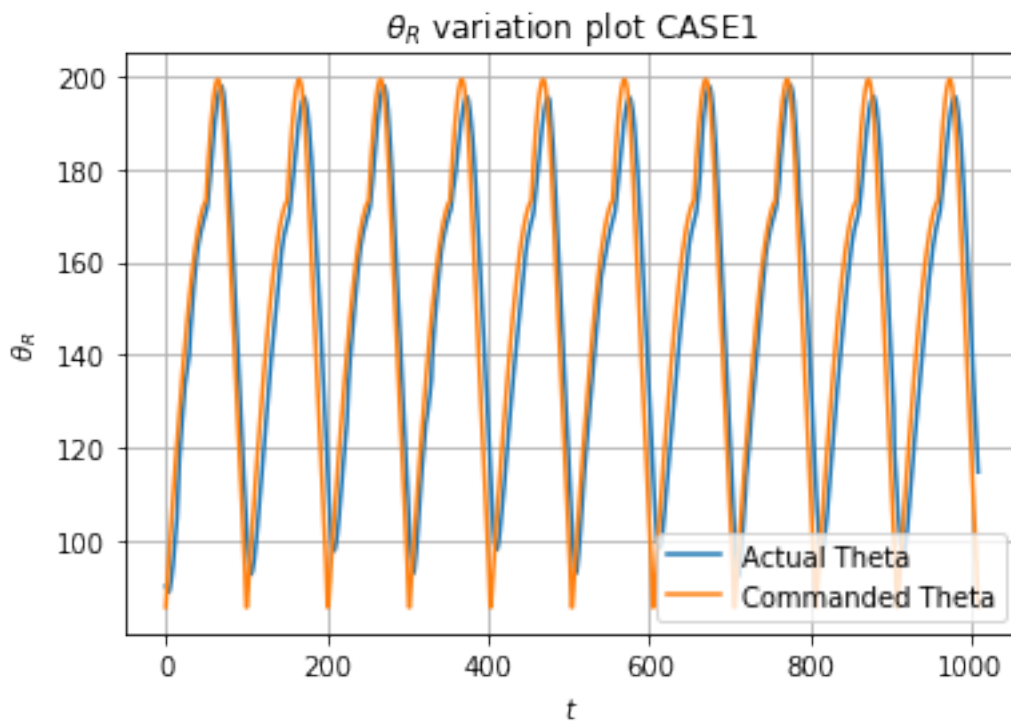
```
[18]: [<matplotlib.lines.Line2D at 0x1ae7d85c5c8>]
```

```
[18]: Text(0.5, 0, '$t$')
```

```
[18]: Text(0, 0.5, '$\\theta_R$')
```

```
[18]: Text(0.5, 1.0, '$\\theta_R$ variation plot CASE1')
```

```
[18]: <matplotlib.legend.Legend at 0x1ae7d2d9748>
```



```
[18]: [<matplotlib.lines.Line2D at 0x1ae7d912ec8>]
```

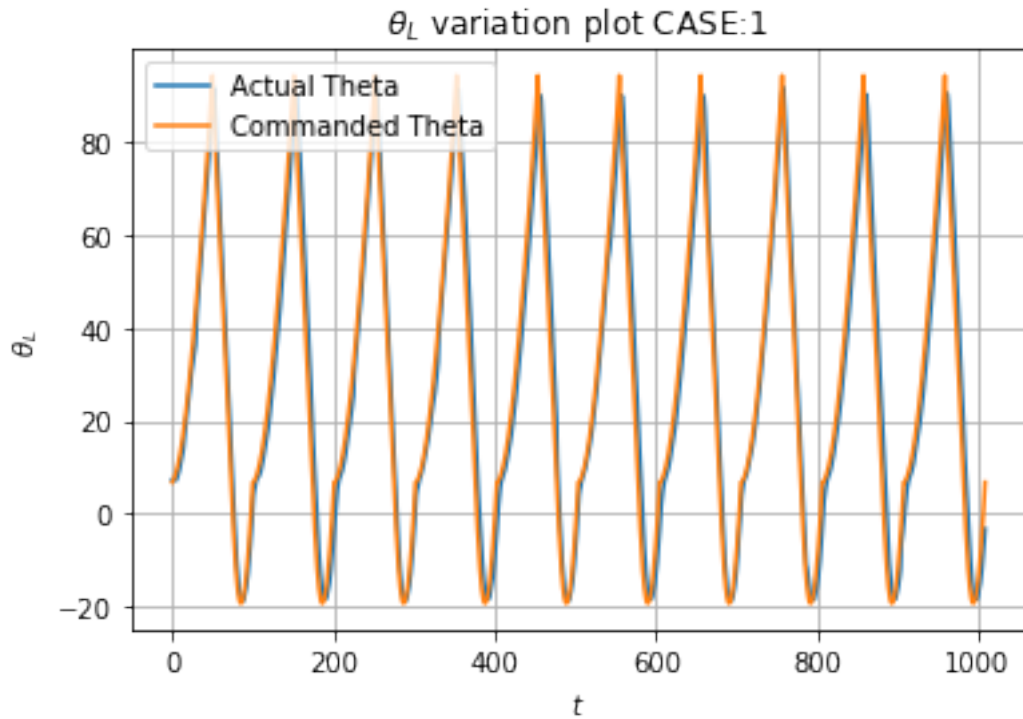
```
[18]: [<matplotlib.lines.Line2D at 0x1ae7d898c48>]
```

```
[18]: Text(0.5, 0, '$t$')
```

```
[18]: Text(0, 0.5, '$\\theta_L$')
```

```
[18]: Text(0.5, 1.0, '$\\theta_L$ variation plot CASE:1')
```

[18]: <matplotlib.legend.Legend at 0x1ae7d8a5308>



```
[19]: %matplotlib inline
plt.plot(theta_Gait2[:,0])
plt.plot(np.rad2deg(b1[:,0]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_R$')
plt.title(r'$\theta_R$ variation plot CASE:2')
plt.legend(['Actual Theta', 'Commanded Theta'])
# plt.gca().set_aspect('equal')
plt.grid()
plt.show()

plt.plot(theta_Gait2[:,1])
plt.plot(np.rad2deg(b1[:,1]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_L$')
plt.title(r'$\theta_L$ variation plot CASE:2')
plt.legend(['Actual Theta', 'Commanded Theta'])
# plt.gca().set_aspect('equal')
plt.grid()
```

[19]: [<matplotlib.lines.Line2D at 0x1ae7d99d9c8>]

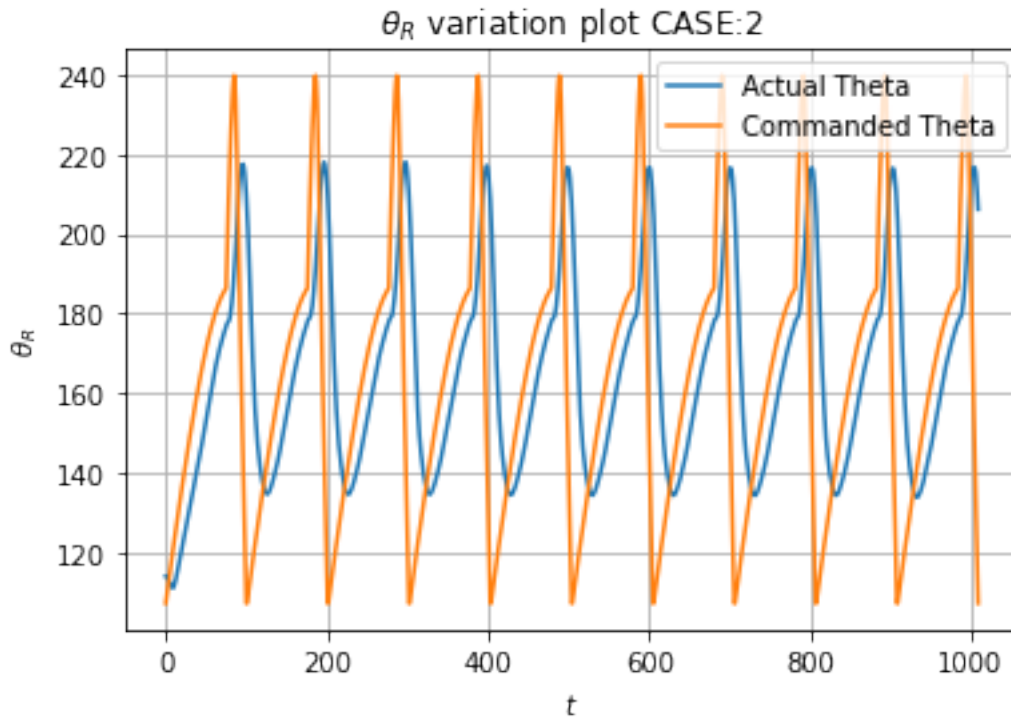
[19]: [<matplotlib.lines.Line2D at 0x1ae7d9dd908>]

[19]: Text(0.5, 0, '\$t\$')

[19]: Text(0, 0.5, '\$\\theta_R\$')

[19]: Text(0.5, 1.0, '\$\\theta_R\$ variation plot CASE:2')

[19]: <matplotlib.legend.Legend at 0x1ae7d9afac8>



[19]: [<matplotlib.lines.Line2D at 0x1ae7da2bf88>]

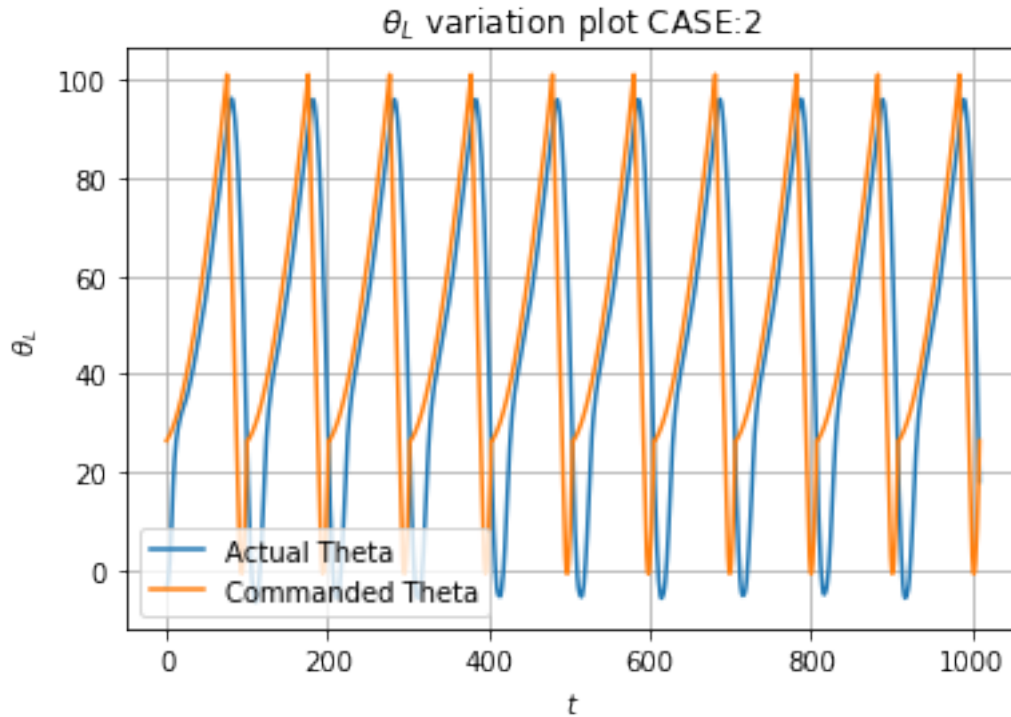
[19]: [<matplotlib.lines.Line2D at 0x1ae7d9c1d88>]

[19]: Text(0.5, 0, '\$t\$')

[19]: Text(0, 0.5, '\$\\theta_L\$')

[19]: Text(0.5, 1.0, '\$\\theta_L\$ variation plot CASE:2')

[19]: <matplotlib.legend.Legend at 0x1ae7da1c188>



```
[20]: %matplotlib inline
plt.plot(theta_Gait3[:,0])
plt.plot(np.rad2deg(c1[:,0]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_R$')
plt.title(r'$\theta_R$ variation plot CASE:3')
plt.legend(['Actual Theta', 'Commanded Theta'])
# plt.gca().set_aspect('equal')
plt.grid()
plt.show()

plt.plot(theta_Gait3[:,1])
plt.plot(np.rad2deg(c1[:,1]))
plt.xlabel(r'$t$')
plt.ylabel(r'$\theta_L$')
plt.title(r'$\theta_L$ variation plot CASE:3')
plt.legend(['Actual Theta', 'Commanded Theta'])
# plt.gca().set_aspect('equal')
plt.grid()
```

[20]: [<matplotlib.lines.Line2D at 0x1ae7da9b988>]

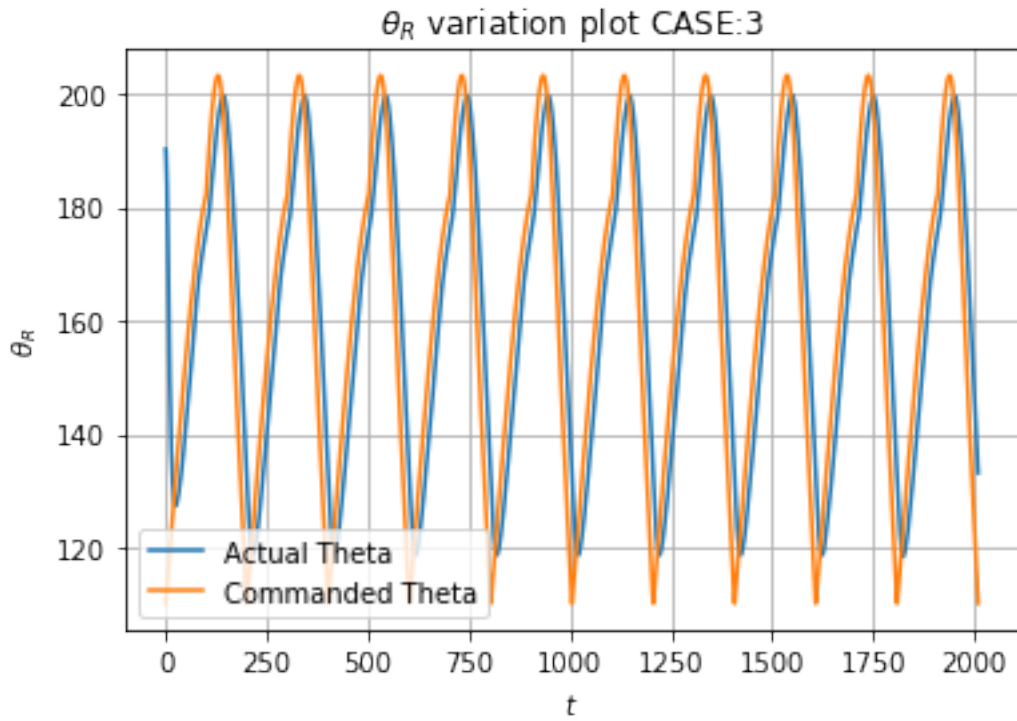
[20]: [<matplotlib.lines.Line2D at 0x1ae7dae7488>]

[20]: Text(0.5, 0, '\$t\$')

[20]: Text(0, 0.5, '\$\\theta_R\$')

[20]: Text(0.5, 1.0, '\$\\theta_R\$ variation plot CASE:3')

[20]: <matplotlib.legend.Legend at 0x1ae7da51a08>



[20]: [<matplotlib.lines.Line2D at 0x1ae7db36388>]

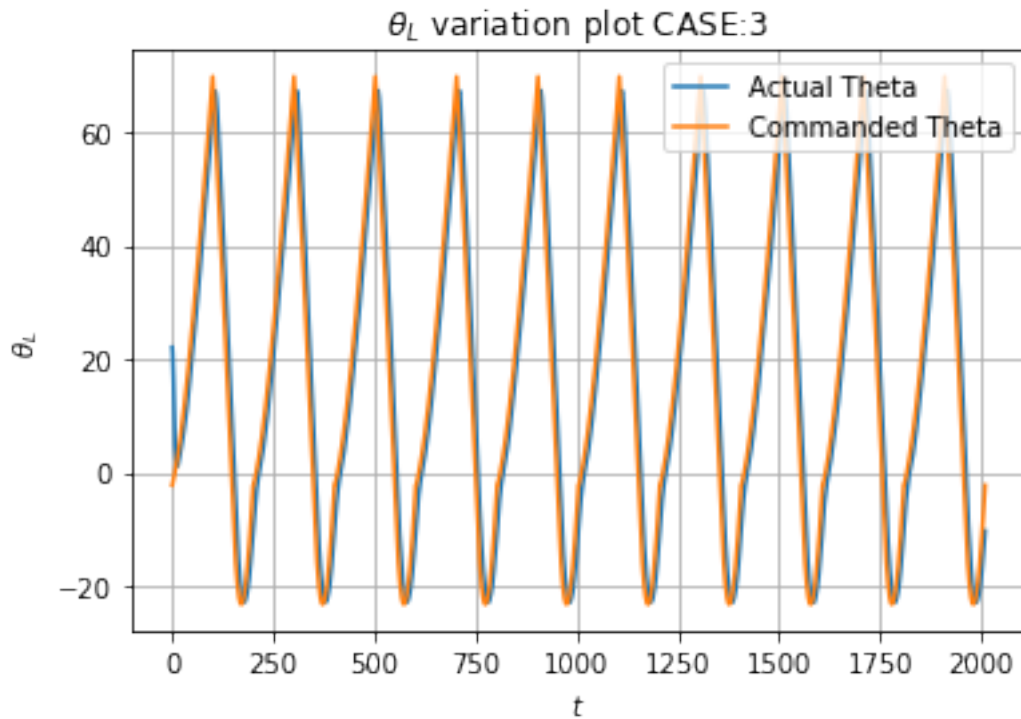
[20]: [<matplotlib.lines.Line2D at 0x1ae7db54188>]

[20]: Text(0.5, 0, '\$t\$')

[20]: Text(0, 0.5, '\$\\theta_L\$')

[20]: Text(0.5, 1.0, '\$\\theta_L\$ variation plot CASE:3')

[20]: <matplotlib.legend.Legend at 0x1ae7db16c08>



Plotting the (θ_L , θ_R) Workspace.

```
[139]: plt.plot(*zip(*theta_Gait1))
plt.xlabel(r'$\theta_R$')
plt.ylabel(r'$\theta_L$')
plt.title('Angle workspace plot CASE:1')
plt.gca().set_aspect('equal')
plt.grid()
plt.show()

plt.plot(*zip(*theta_Gait2))
plt.xlabel(r'$\theta_R$')
plt.ylabel(r'$\theta_L$')
plt.title('Angle workspace plot CASE:2')
plt.gca().set_aspect('equal')
plt.grid()
plt.show()

plt.plot(*zip(*theta_Gait3))
plt.xlabel(r'$\theta_R$')
plt.ylabel(r'$\theta_L$')
plt.title('Angle workspace plot CASE:3')
plt.gca().set_aspect('equal')
```

```
plt.grid()
```

```
[139]: [<matplotlib.lines.Line2D at 0x190ba1e8788>]
```

```
[139]: Text(0.5, 0, '$\\theta_R$')
```

```
[139]: Text(0, 0.5, '$\\theta_L$')
```

```
[139]: Text(0.5, 1.0, 'Angle workspace plot CASE:1')
```

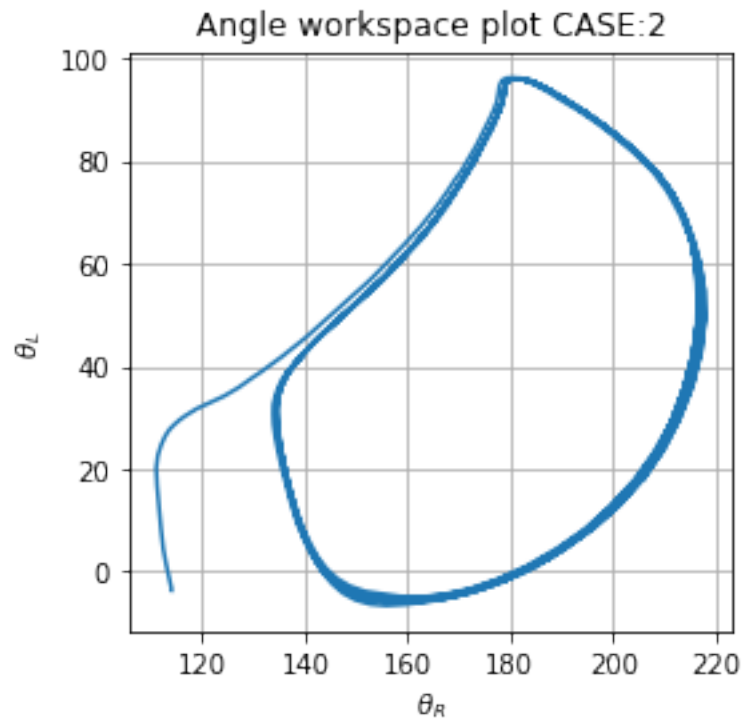


```
[139]: [<matplotlib.lines.Line2D at 0x190ba276048>]
```

```
[139]: Text(0.5, 0, '$\\theta_R$')
```

```
[139]: Text(0, 0.5, '$\\theta_L$')
```

```
[139]: Text(0.5, 1.0, 'Angle workspace plot CASE:2')
```

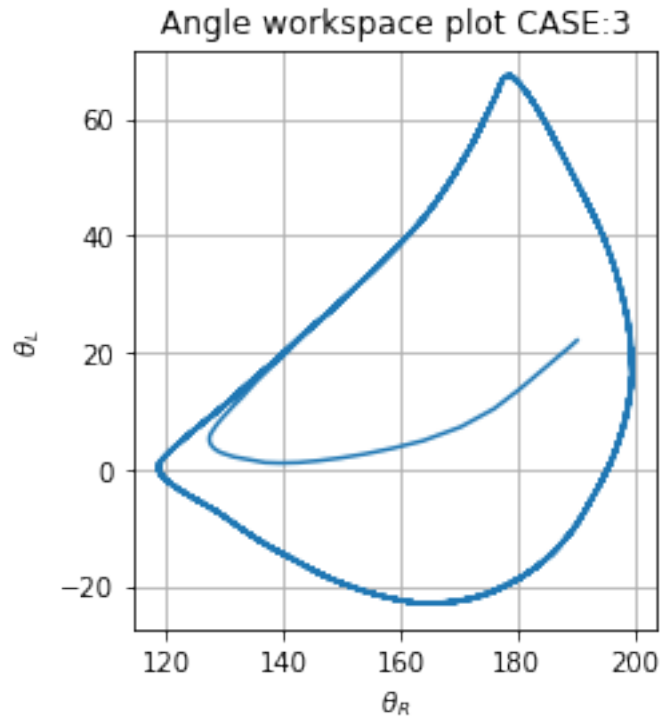


```
[139]: [<matplotlib.lines.Line2D at 0x190b4eeb408>]
```

```
[139]: Text(0.5, 0, '$\\theta_R$')
```

```
[139]: Text(0, 0.5, '$\\theta_L$')
```

```
[139]: Text(0.5, 1.0, 'Angle workspace plot CASE:3')
```

Plotting the (x, y) coordinate Workspace

```
[11]: data_pointsFK1 = np.zeros([1,2])
data_pointsFK2 = np.zeros([1,2])
# for i in [1,2,3]

theta=theta_Gait1
for k in range(0,len(theta)):

    data_pointsFK1 = np.append(data_pointsFK1,FK1_fast(np.
    ↪radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
#     data_pointsFK2 = np.append(data_pointsFK2,FK2_fast(np.
    ↪radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
# data_pointsFK1.shape
data_pointsFK1 = data_pointsFK1[1:data_pointsFK1.shape[0],:]
# data_pointsFK2 = data_pointsFK2[1:data_pointsFK2.shape[0],:]

plt.clf()
plt.plot(data_pointsFK1[:,0],data_pointsFK1[:,1])
# plt.plot(data_pointsFK2[:,0],data_pointsFK2[:,1])
plt.xlabel(r'$X$')
plt.ylabel(r'$Y$')
plt.title('(X,Y) workspace plot CASE:1')
```

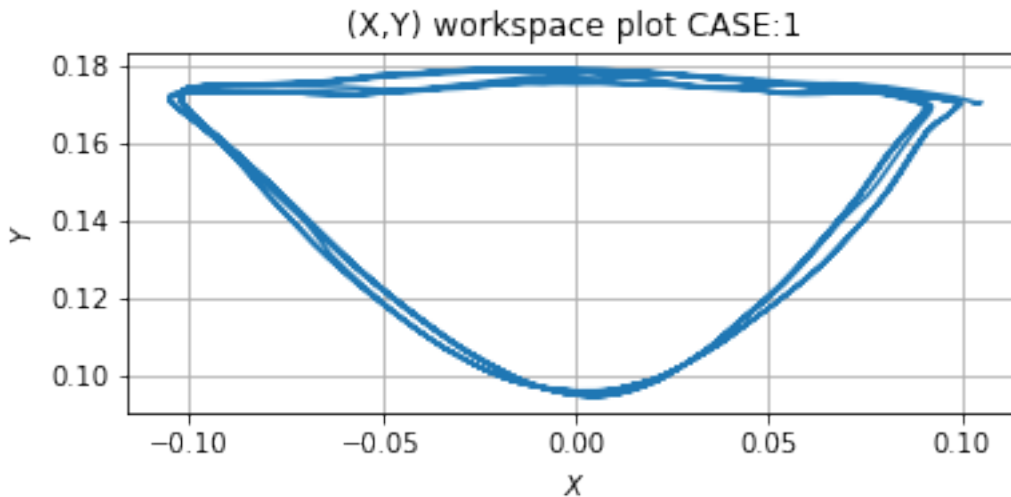
```
plt.grid()
plt.gca().set_aspect('equal')
```

```
[11]: [<matplotlib.lines.Line2D at 0x1ae7d576dc8>]
```

```
[11]: Text(0.5, 0, '$X$')
```

```
[11]: Text(0, 0.5, '$Y$')
```

```
[11]: Text(0.5, 1.0, '(X,Y) workspace plot CASE:1')
```



```
[12]: data_pointsFK1 = np.zeros([1,2])
data_pointsFK2 = np.zeros([1,2])
# for i in [1,2,3]

theta=theta_Gait2
for k in range(0,len(theta)):

    data_pointsFK1 = np.append(data_pointsFK1,FK1_fast(np.
→radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
#     data_pointsFK2 = np.append(data_pointsFK2,FK2_fast(np.
→radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
# data_pointsFK1.shape
data_pointsFK1 = data_pointsFK1[1:data_pointsFK1.shape[0],:]
# data_pointsFK2 = data_pointsFK2[1:data_pointsFK2.shape[0],:]

plt.clf()
plt.plot(data_pointsFK1[:,0],data_pointsFK1[:,1])
# plt.plot(data_pointsFK2[:,0],data_pointsFK2[:,1])
```

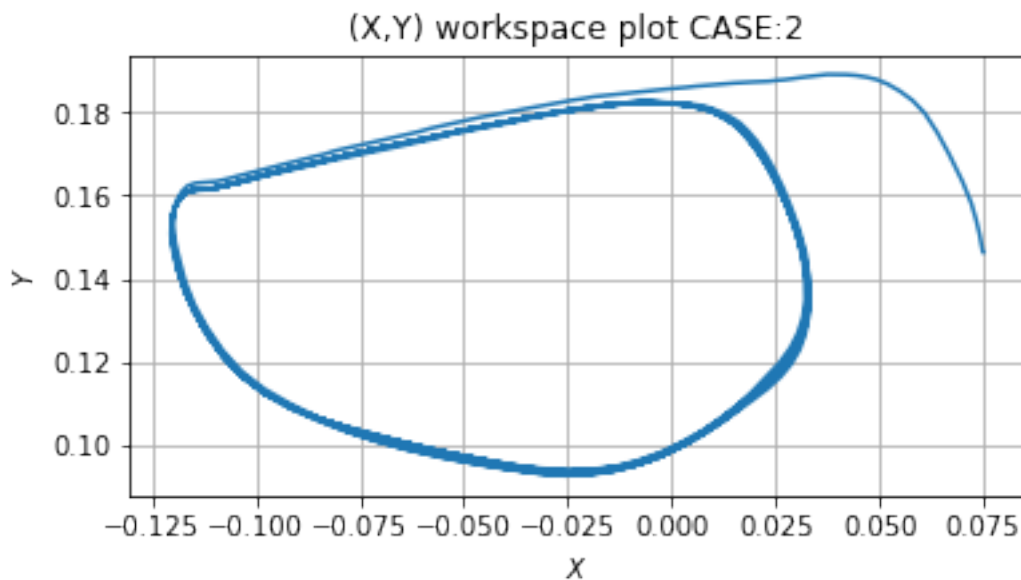
```
plt.xlabel(r'$X$')
plt.ylabel(r'$Y$')
plt.title('(X,Y) workspace plot CASE:2')
plt.grid()
plt.gca().set_aspect('equal')
```

[12]: [<matplotlib.lines.Line2D at 0x1ae7d5bb208>]

[12]: Text(0.5, 0, '\$X\$')

[12]: Text(0, 0.5, '\$Y\$')

[12]: Text(0.5, 1.0, '(X,Y) workspace plot CASE:2')



```
[13]: data_pointsFK1 = np.zeros([1,2])
data_pointsFK2 = np.zeros([1,2])
# for i in [1,2,3]

theta=theta_Gait3
for k in range(0,len(theta)):

    data_pointsFK1 = np.append(data_pointsFK1,FK1_fast(np.
→radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
#     data_pointsFK2 = np.append(data_pointsFK2,FK2_fast(np.
→radians(theta[k][1]),np.radians(theta[k][0]),0.09,0.16,0.07).T,axis=0)
# data_pointsFK1.shape
data_pointsFK1 = data_pointsFK1[1:data_pointsFK1.shape[0],:]
```

```

# data_pointsFK2 = data_pointsFK2[1:data_pointsFK2.shape[0],:]

plt.clf()
plt.plot(data_pointsFK1[:,0],data_pointsFK1[:,1])
# plt.plot(data_pointsFK2[:,0],data_pointsFK2[:,1])
plt.xlabel(r'$X$')
plt.ylabel(r'$Y$')
plt.title('(X,Y) workspace plot CASE:3')
plt.grid()
plt.gca().set_aspect('equal')

```

[13]: [<matplotlib.lines.Line2D at 0x1ae7d6c6488>]

[13]: Text(0.5, 0, '\$X\$')

[13]: Text(0, 0.5, '\$Y\$')

[13]: Text(0.5, 1.0, '(X,Y) workspace plot CASE:3')

