

CONTROL THEORY ASSIGNMENT № 3

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1. Selecting Variant

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Variant: g

2. In this task use python for calculations

A. Design PD-controller that tracks time varying reference states i.e. $[x^*(t), \dot{x}^*(t)]$ as closely as possible. Test your controller on different trajectories, at least two.

System: $\ddot{x} + \mu\dot{x} + kx = u$, see variants below.

My system: $\ddot{x} + 7\dot{x} + 25x = u$

Code:

```
1 #importing needed libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import math
5 from scipy.integrate import odeint
6
7 time = np.linspace(0, 25, 10000) #defining in which bounds to plot
8 kp = 100 #proportional control coefficient
9 kd = 100 #derivative control coefficient
10 k = 7      #given constant
11 mu = 25    #given constant
12 init = [2.5, 3.5] #you can write here any initial conditions
13
14 def ref(t): #desired reference state
15     return math.sin(t) * math.sin(t) #function of desired reference state
16 def dref(t): #rate of change of desired reference state
17     return 2.0 * math.sin(t) * math.cos(t) #derivative of previous function
18 def no_control(x, t): #model without any control
19     y = x[0]
20     dy = x[1]
21     xdot = [[], []]
22     xdot[0] = dy
23     xdot[1] = -mu * dy - k * y
```

```

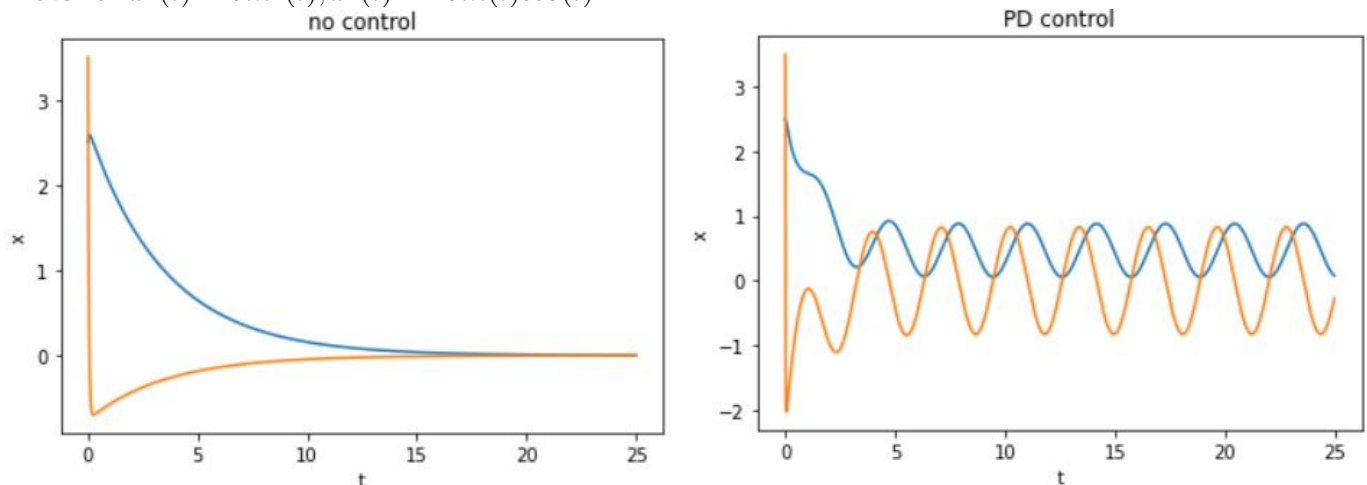
24     return xdot
25 def pd_control(x, t): #model with PD control
26     err = ref(t) - x[0] #error
27     derr = dref(t) - x[1] #rate of change of error
28     u = kd * derr + kp * err
29     y = x[0]
30     dy = x[1]
31     xdot = [[], []]
32     xdot[0] = dy
33     xdot[1] = u - mu * dy - k * y
34     return xdot
35
36 plt.subplot(221) #positioning the plotted graph
37 plt.title("no control") #title of graph
38 plt.xlabel("t") #what means x-axis (time)
39 plt.ylabel("x") #what means y-axis (x)
40 plt.plot(time, odeint(no_control, init, time)) #plotting ODE solution for ←
    model without control
41 plt.subplot(224)
42 plt.title("PD control")
43 plt.xlabel("t")
44 plt.ylabel("x")
45 plt.plot(time, odeint(pd_control, init, time)) #plotting ODE solution for ←
    model with PD control

```

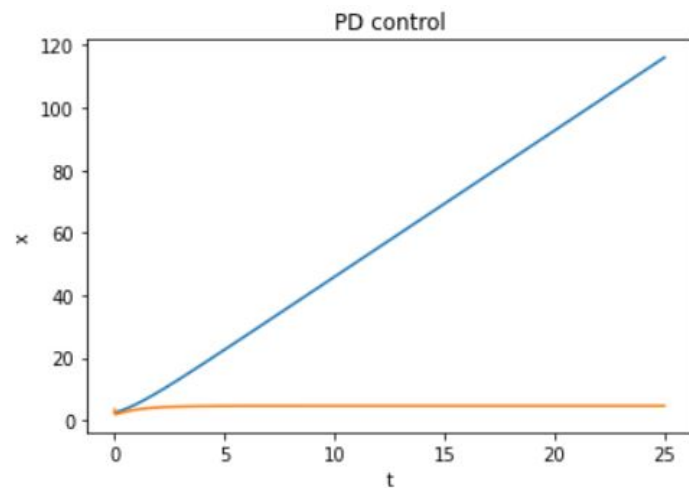
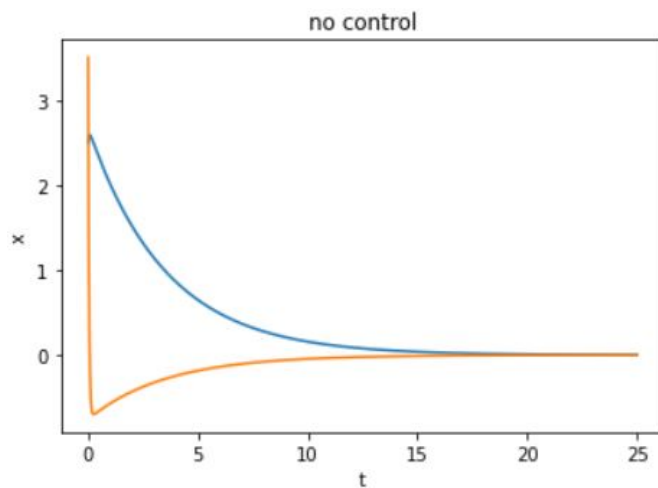
Notes

I chose $kp = 100$, $kd = 100$.

Plots for $x^*(t) = \sin^2(t)$, $\dot{x}^*(t) = 2\sin(t)\cos(t)$



Plots for $x^*(t) = 5t$, $\dot{x}^*(t) = 5$



B. Tune controller gains k_p and k_d . Find gains that provide no oscillations and no overshoot. Prove it with step input.

I chose $k_p = 100$, $k_d = 100$.

```

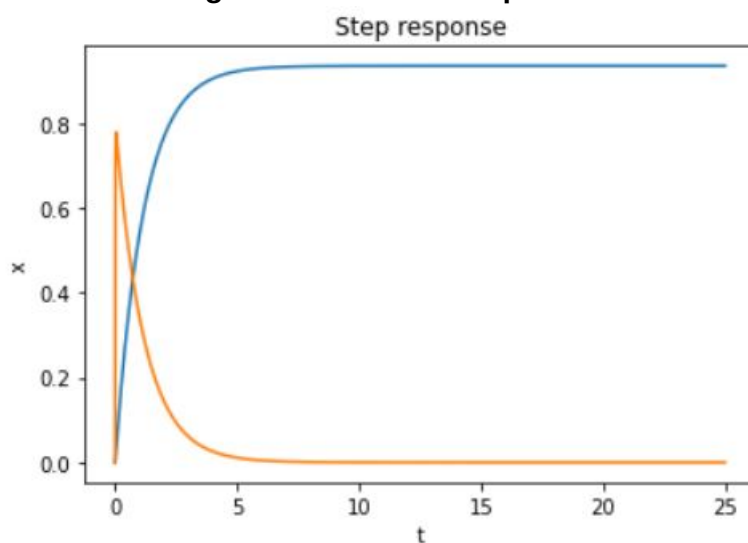
1  #importing needed libraries
2  import numpy as np
3  import matplotlib.pyplot as plt
4  import math
5  from scipy.integrate import odeint
6
7  time = np.linspace(0, 25, 10000) #defining in which bounds to plot
8  kp = 100 #proportional control coefficient
9  kd = 100 #derivative control coefficient
10 k = 7    #given constant
11 mu = 25  #given constant
12 init = [0, 0] #zero initial conditions
13
14 def ref(t): #desired reference state (step function)
15     if t == 0:
16         return 0.5
17     elif t < 0:
18         return 0
19     else:
20         return 1
21 def dref(t): #rate of change of desired reference state
22     return 0 #derivative of previous function
23 def step(x, t): #function for defining ODE for step response
24     err = ref(t) - x[0] #error
25     derr = dref(t) - x[1] #rate of change of error
26     u = kd * derr + kp * err #control
27     y = x[0]
```

```

28     dy = x[1]
29     xdot = [[], []]
30     xdot[0] = dy
31     xdot[1] = u - mu * dy - k * y
32     return xdot
33 plt.subplot(111)
34 plt.title("Step response")
35 plt.xlabel("t")
36 plt.ylabel("x")
37 plt.plot(time, odeint(step, init, time)) #solving ODE and plotting step ↵
      response

```

No overshooting and no oscillations proof:



C. Prove that controlled oscillator dynamics is stable for your choice of k_p and k_d .

$$\begin{aligned}
 \begin{bmatrix} \ddot{x} \\ \dot{x} \end{bmatrix} &= \begin{bmatrix} -7 & -25 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{x} \\ x \end{bmatrix} + \begin{bmatrix} k_d & k_p \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}^* - \dot{x} \\ x^* - x \end{bmatrix} \\
 \begin{bmatrix} \ddot{x} \\ \dot{x} \end{bmatrix} &= \begin{bmatrix} -7 & -25 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{x} \\ x \end{bmatrix} + \begin{bmatrix} k_d & k_p \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}^* \\ x^* \end{bmatrix} - \begin{bmatrix} k_d & k_p \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x} \\ x \end{bmatrix} \\
 \begin{bmatrix} \ddot{x} \\ \dot{x} \end{bmatrix} &= \begin{bmatrix} -7 - k_d & -25 - k_p \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{x} \\ x \end{bmatrix} + \begin{bmatrix} k_d & k_p \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}^* \\ x^* \end{bmatrix} \\
 A &= \begin{bmatrix} -7 - k_d & -25 - k_p \\ 1 & 0 \end{bmatrix}
 \end{aligned}$$

Let's find eigenvalues of A matrix

$$\begin{aligned}
 \det(A - \lambda I) &= 0 \\
 (-7 - k_d - \lambda)(-\lambda) + 25 + k_p &= 0 \\
 7\lambda + k_d\lambda + \lambda^2 + 25 + k_p &= 0 \\
 \lambda^2 + \lambda(7 + k_d) + 25 + k_p &= 0
 \end{aligned}$$

$$\lambda_1 + \lambda_2 = -(7 + k_d), \lambda_1 \lambda_2 = 25 + k_p$$

If system is stable, then

$$\lambda_1 < 0 \text{ and } \lambda_2 < 0$$

$$-7 - k_d < 0, k_d > -7$$

$$25 + k_p > 0, k_p > -25$$

As I chose $k_d = 100$, $k_p = 100$, inequalities hold, and system is stable

E. Implement a PI/PID controller for the system: $\ddot{x} + \mu\dot{x} + kx + 9.8 = u$. Test your controller on different trajectories, at least two.

```

1  #importing needed libraries
2  import numpy as np
3  import matplotlib.pyplot as plt
4  import scipy.integrate as integrate
5  import math
6  from scipy.integrate import odeint
7
8  time = np.linspace(0, 25, 10000) #defining in which bounds to plot
9  kp = 1000 #proportional control coefficient
10 ki = 1 #integral control coefficient
11 k = 7 #given constant
12 mu = 25 #given constant
13 init = [2.5, 3.5] #you can write here any initial conditions
14
15 def ref(t): #desired reference state
16     return math.sin(t)*math.sin(t) #function of desired reference state
17 def no_control(x, t): #model without any control
18     y = x[0]
19     dy = x[1]
20     xdot = [[], []]
21     xdot[0] = dy
22     xdot[1] = -mu * dy - k * y - 9.8
23     return xdot
24 def pi_control(x, t): #model with PD control
25     err = ref(t) - x[0] #error
26     diff = lambda t: ref(t) - x[0]
27     ierr = integrate.quad(diff, 0, t)[0] #integral of error
28     u = ki * ierr + kp * err #control
29     y = x[0]
30     dy = x[1]
31     xdot = [[], []]
32     xdot[0] = dy
33     xdot[1] = u - mu * dy - k * y - 9.8

```

```

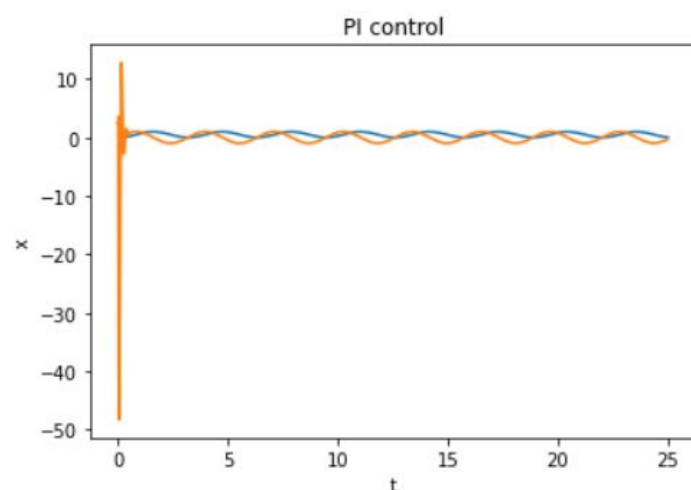
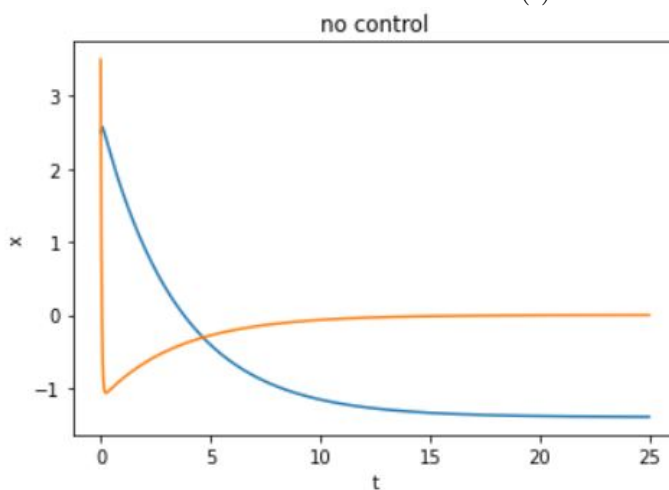
34     return xdot
35
36 plt.subplot(221) #positioning the plotted graph
37 plt.title("no control") #title of graph
38 plt.xlabel("t") #what means x-axis (time)
39 plt.ylabel("x") #what means y-axis (x)
40 plt.plot(time, odeint(no_control, init, time)) #plotting ODE solution for ←
    model without control
41
42 plt.subplot(224)
43 plt.title("PI control")
44 plt.xlabel("t")
45 plt.ylabel("x")
46 plt.plot(time, odeint(pi_control, init, time)) #plotting ODE solution for ←
    model with PI control

```

Notes

I chose $k_p = 1000$, $k_i = 1$

Plots with desired reference state $\sin^2(t)$



Plots with desired reference state $5t$

