

# EE3004 : Assignment 5 Report

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## 1 Install Instructions

We use *ipython* notebooks for the following assignment. The repository includes the relevant pip requirements.

Ensure python 3.6 is installed, and run:

```
pip3 install -r requirements.txt
jupyter lab
```

This should open a browser window with the code snippets.

## 2 Question 1

```
In []: import numpy as np
import math
import scipy
from control.matlab import *
import matplotlib.pyplot as plt
plt.style.use('ggplot')
pie=np.pi
```

We design for  $M_p = 17$ ,  $T_s = 3$

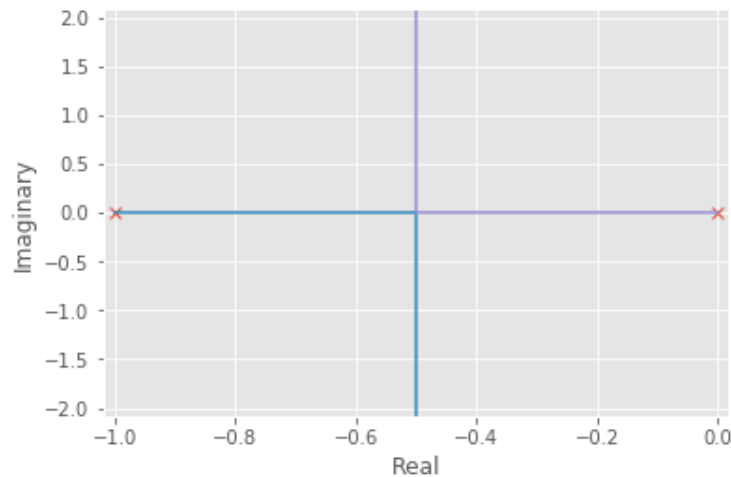
```
In []: real=-(4.0/Ts)
imag=4*pie/(Ts*np.log(100/Mp))
pole1=complex(real,imag)
pole2=complex(real,imag)
theta=180-np.angle(pole1,deg=True)
wn=np.abs(pole1)
print (pole1,pole2)
print(theta,wn)
```

```
(-1.3333333333333333+2.3639346657109117j) (-1.3333333333333333+2.3639346657109117j)
60.57563766283309 2.714031112851792
```

```
In []: (num,den)=zpk2tf([], [0,-1],1)
      H=tf(num,den)
      print(H)
```

```
1
-----
s^2 + s
```

```
In []: rlocus(H)
```



```
In []: anl=np.angle(evalfr(H,pole1),deg=True)
      print (anl)
```

```
142.54939989916585
```

## 2.1 Part 3

Clearly this doesnot lie on the root locus, hence we need a lead compensator. We calculate  $\phi$ , the necessary angle difference. In addition we calculate  $\theta$  the positive angle made by the dominant pole, by either  $\arccos(\zeta)$  or from the angle directly.

```
phi=-180+th
print(f"Phi {phi}")
theta=180-np.angle(pole1,deg=True)
print(f"Theta {theta}")
gamma=0.5*(180-theta-phi)
print(f"Gamma {gamma}")
alpha=np.sin(gamma*pie/180)*np.sin((theta+gamma+phi)*pie/180)/np.sin((theta+gamma)*pie/180)
print(f"Alpha {alpha}")
zc=wn*np.sin(gamma*pie/180)/np.sin((theta+gamma)*pie/180)
pc=zc/alpha
print (f"zc {zc} pc {pc}")
```

```

Phi 37.45060010083418
Theta 60.57563766283309
Gamma 40.986881118166366
Alpha 0.44819287447848166
zc 1.8169680716950904 pc 4.053986966681072

```

```

In []: Hnew = H * tf([1,zc],[1,pc])
      Hnew

```

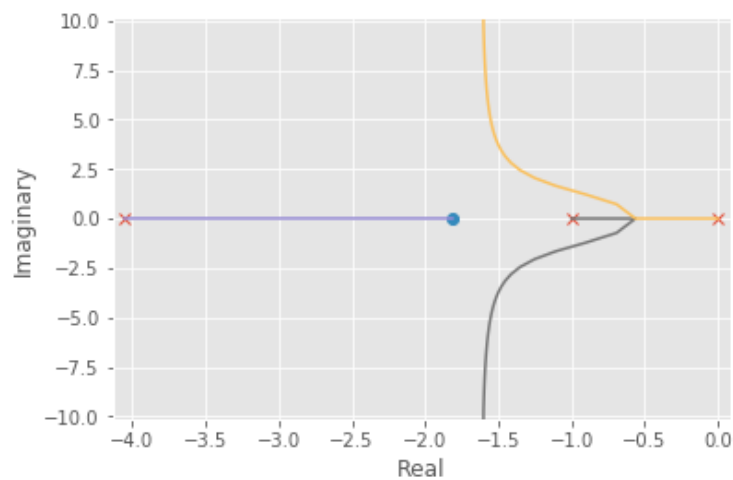
Out[49]:

$$\frac{s + 1.817}{s^3 + 5.054 s^2 + 4.054 s}$$

```

In []: rlocus(Hnew);

```



```

In []: mag = np.abs(evalfr(Hnew,pole1))
      phase = np.angle(evalfr(Hnew,pole1),deg=True)
      print(f"Phase at pole {phase}")
      K=1.0/mag
      print(f"K {K}")
      Hlead=K*Hnew
      print(f"Hlead compensator only {Hlead}")

```

Phase at pole -180.0

K 9.678165381551548

Hlead compensator only

$$9.678 s + 17.58$$

$$\frac{9.678 s + 17.58}{s^3 + 5.054 s^2 + 4.054 s}$$

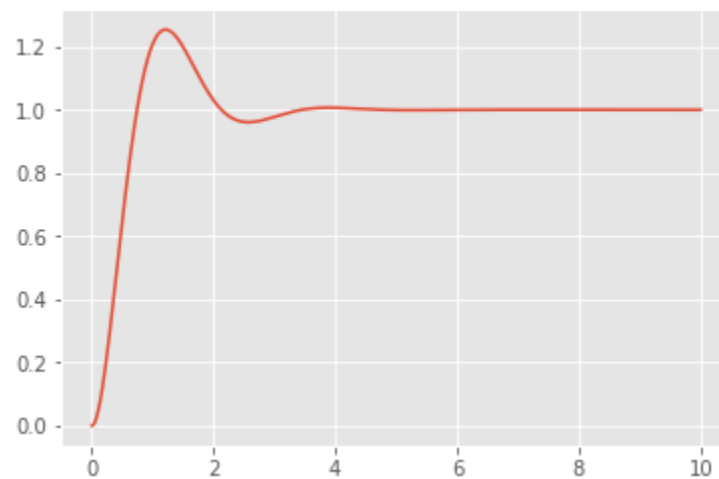
```
In []: Kv=K*zc/pc
       print (Kv)
```

4.337684762035719

## 2.2 Plot with only lead compensator

```
In []: T = linspace(0,10,1000)
       (y,T)=step(feedback(Hlead),T)
       plt.plot(T,y)
```

Out[54]: [<matplotlib.lines.Line2D at 0x1c15f30b00>]



```
In []: stepinfo(T,y)
```

```
OS 25.39881214713917 %
Tr 0.6706706706706707
Ts 2.032032032032032
```

## 2.3 $\alpha * \beta = 1$

```
In []: beta=1.0/alpha
       zg=0.05
       pg=zg/beta
       Hlag=tf([1,zg],[1,pg])
       Hleadlag=Hlead*Hlag
       Hleadlag
```

Out [56] :

$$\frac{9.678 s^2 + 18.07 s + 0.8792}{s^4 + 5.076 s^3 + 4.167 s^2 + 0.09085 s}$$

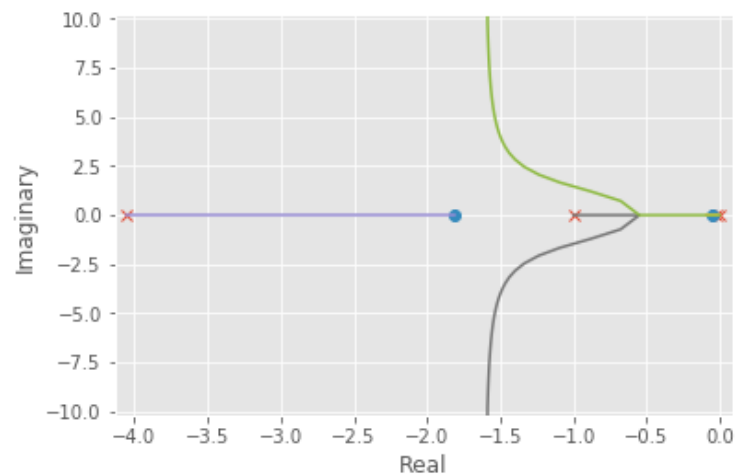
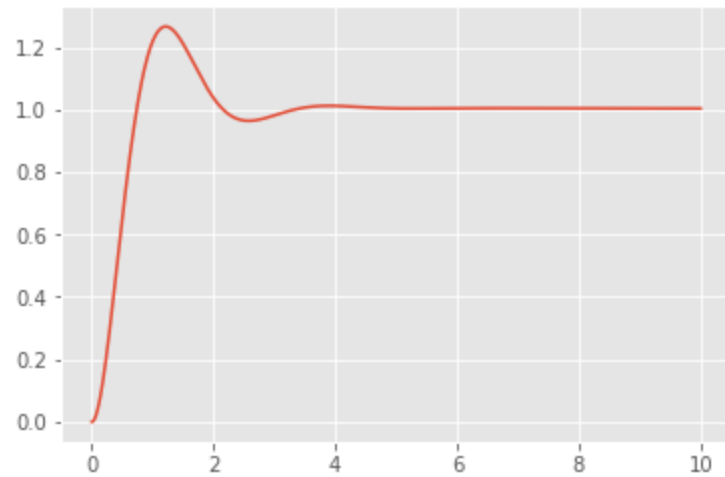
```
In []: T=linspace(0,10,1000)
(y,T)=step(Hleadlag/(1+Hleadlag),T)
plt.plot(T,y)
stepinfo(T,y);
print ("Kv:",Kv*beta)
rlocus(Hleadlag);
```

OS 26.16900489320202 %

Tr 0.6706706706706707

Ts 2.042042042042042

Kv: 20.0



## 2.4 Part 4, 6 Independent $\alpha, \beta$

We set  $\beta = K v_{desired} / K v_{obtained}$ , when obtained from the lead compensator.

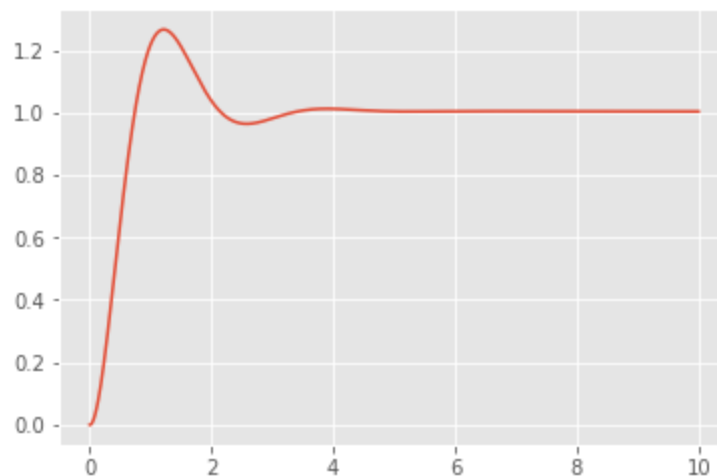
```
In []: Kv_desire=20
      beta=Kv_desire/Kv
      zg=0.05
      pg=zg/beta
      H_lag=tf([1,zg],[1,pg])
      Hlead_lag=Hlead*Hlag
      print(Hlead_lag)

          9.678 s^2 + 18.07 s + 0.8792
-----
s^4 + 5.076 s^3 + 4.167 s^2 + 0.09085 s
```

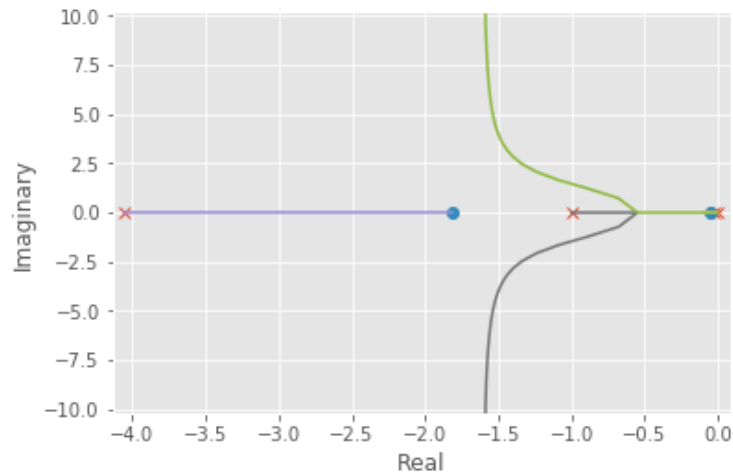
We plot the step response and the root locus of the above system.

```
In []: T=linspace(0,10,1000)
      (y,T)=step(feedback(Hlead_lag),T)
      plt.plot(T,y)
      stepinfo(T,y);
```

```
OS 26.16900489320171 %
Tr 0.6706706706706707
Ts 2.042042042042042
```



```
In []: rlocus(Hleadlag);
```



### 3 Question 3

```
In []: import numpy as np
import scipy
import matplotlib.pyplot as plt
import control
import math
plt.style.use('ggplot')
from sympy.solvers import solve
from sympy import Symbol
import sympy as sp
```

We define a few helper functions ...

```
In []: def get_params(Mp,ts):
    '''
    Mp in percent
    '''

    x = np.log(Mp/100)**2
    zeta = x / (x + np.pi**2)
    zeta = np.sqrt(zeta)

    wn = 4/ts/zeta
```

```

        return zeta, wn

def polar(z):
    '''
    Get polar representation
    '''
    a= z.real
    b= z.imag
    r = math.hypot(a,b)
    theta = math.atan2(b,a)
    return r,theta # use return instead of print.

def get_poles(zeta,wn):
    '''
    Get poles from
    '''
    return (complex(-zeta*wn, wn*np.sqrt(1-zeta**2)), complex(-zeta*wn ,- wn*np.sqrt(1

def get_alpha(gamma, theta, phi):
    '''
    Returns alpha from gamma.

    All angles in radians
    '''
    num = np.sin(gamma) * np.sin(theta + gamma + phi)
    denom = np.sin(theta + gamma) * np.sin(gamma + phi)
    return num/denom

def get_zc(wn, gamma, theta):
    '''
    Get zc
    '''
    num = wn* np.sin(gamma)
    denom = np.sin(theta + gamma)
    return num/denom

```

We design for  $M_p = 5\%$ , and  $T_s$  is 0.02s,  $K_v$  is 1%. And we shall test if the design satisfies for  $T_r$ .

```

In []: Mp, ts = 5, 0.02
       zeta,wn = get_params(Mp,ts)
       p1,p2 = get_poles(zeta,wn)

```

Zeta 0.6901067305598216 wn 289.8102440440741

Required dominant poles (-200.00000000000003+209.73787820249777j), (-200.00000000000003-209.7378



### 3.1 Part a

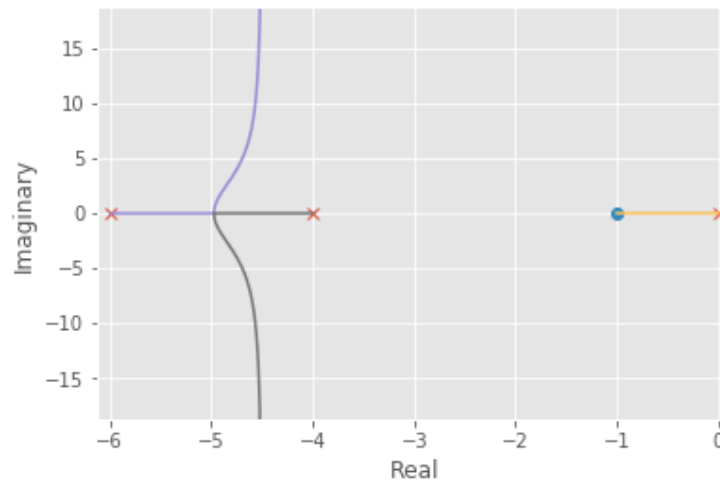
This is feasible if  $(-200.00 \pm j209.737j)$  lies on the root locus, as dominant poles.

```
In []: sys = control.TransferFunction([240,240],[1,10,24,0])
```

System open loop is

$$\frac{240s + 240}{s^3 + 10s^2 + 24s}$$

```
In []: control.root_locus(sys);
```



Hence choice 1 is not feasible.

### 3.2 Part b

We first analyse the existent system, understand if the transient parameters need improvement or the steady state. We then design appropriately, with the dominant pole assumption.

With [U+FFFD]  $G_c(s) = K_c \alpha * (1 + Ts) / (1 + \alpha * Ts)$ .

Needed  $K_v$  is 100 (1% error). Current  $K_v$  is 10.

Hence,  $K_c * \alpha = 10$ .

We determine  $\theta$ , angle of compensation needed at  $(-200, \pm j209.73)$

```
In []: polar(control.evalfr(sys, complex(200,-209.73)))[1]
```

```
Out[179]: 1.596108226418121
```

```
In []: np.pi - 1.596108226418121
```

```
Out[180]: 1.5454844271716721
```

We need -88.54 deg as compensation.  
 Besides this, we need to meet:

- Angle sum =  $\pi$
- Magnitude at roots = 1
- $K_c * \alpha = 10$

We treat this as a lead compensation case.

```
In []: phi = 85.97404456267036 * np.pi/180
        theta = np.arccos(zeta)
        gamma = (1.596108226418121 - theta)/2
        print(f"Gamma {gamma * 180/np.pi}")
        alpha = get_alpha(gamma = gamma, phi = phi, theta = theta)
        print(f"Alpha {alpha}")
        temp_sys = alpha * control.TransferFunction([T,1],[T*alpha, 1]) * sys
        Kc = 1/polar(control.evalfr(temp_sys,complex(-200,209.73)))[0]
        print(f"Kc {Kc}")

Gamma 22.54441156486525
Alpha 0.183975416417848
Kc 752.8216104925237
```

To determine  $T$ , we use the magnitude condition.

```
In []: cur_mag = polar(control.evalfr(sys, complex(-200,209.73)))[0]
        print(f"Current magnitude without compensator at requisite roots {cur_mag}")

Current magnitude without compensator at requisite roots 0.002919450489681809

Z_c 119.09350502020902
T 0.00839676353324482

Solved parameters :

Alpha 0.183975416417848
Kc 54.355088275967454
T 0.00839676353324482
```

### 3.3 Part 3

We perform a grid search, while looking for the following:

- Angle sum =  $\pi$
- Magnitude at roots = 1
- $K_c * \alpha = 10$

We keep varying  $\alpha$  and  $T$ .  
 We try:

- grid search
- gradient descent

```
In []: aa = np.arange(0.1,2.0,(0.2-0.1)/1000)
      tt = np.arange(0, 0.01, 0.01/100)
      ll = [] # loss array

      def loss1(a,t, pole = complex(-200,209.73)):
          '''
          Angle compensation
          '''
          diff = polar(1 + t*pole)[1] - polar(1 + a*t*pole)[1] - 1.5454844271716721
          diff = np.abs(diff)
          return diff

      def loss2(a,t, pole = complex(-200,209.73)):
          '''
          Magnitude compensation
          '''
          diff = polar(1 + t*pole)[0] / polar(1 + a*t*pole)[0] - 1/(10*cur_mag)
          return np.abs(diff)

      def loss(a,t, pole = complex(-200,209.73)):
          return loss1(a,t, pole) + 0*loss2(a,t,pole)

      l_min = None
      a_min = None
      t_min = None

      for a in aa:
          for t in tt:
              l = loss(a,t)
              if not l_min or (l_min > l):
                  l_min = l
                  a_min = a
                  t_min = t
              ll.append(loss)

      print(f"Loss {l_min} a_min {a_min} t_min {t_min}")
      print(f"Angle loss {loss1(a_min,t_min)}")
      #print(f"Mag loss {loss2(a_min, t_min)}")

Loss 1.884514472694221e-06 a_min 0.166900000000000194 t_min 0.0077
Angle loss 1.884514472694221e-06
Mag loss 32.09852385559125
```

### 3.4 Part 4,5 with Part 2

Clearly, we have used a lead compensator.

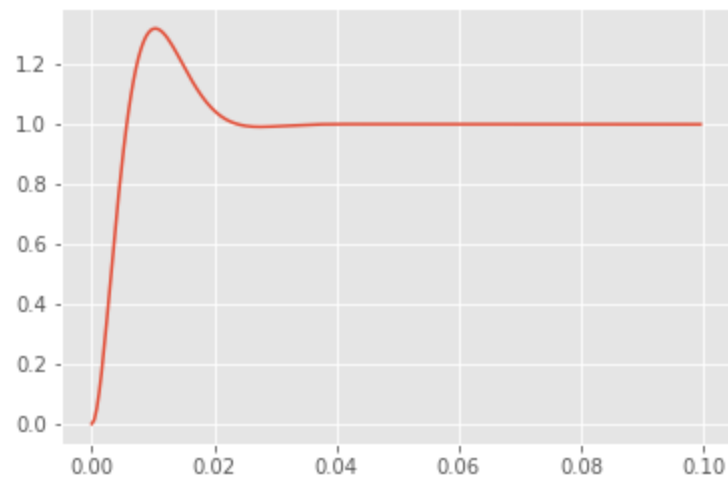
The compensated transfer function is:

Compensator  
$$\frac{1.163 s + 138.5}{0.001545 s + 1}$$

Final Sys  
$$\frac{279.1 s^2 + 3.352e+04 s + 3.324e+04}{0.001545 s^4 + 1.015 s^3 + 10.04 s^2 + 24 s}$$

The transient responses associated with this are:

```
In []: tt, yy = control.step_response(control.feedback(comp_sys), T = np.arange(0,0.1,0.04/100))  
       plt.plot(tt,yy)
```



### 3.5 Part 4,5 with Part 3

Clearly, we have used a lead compensator.

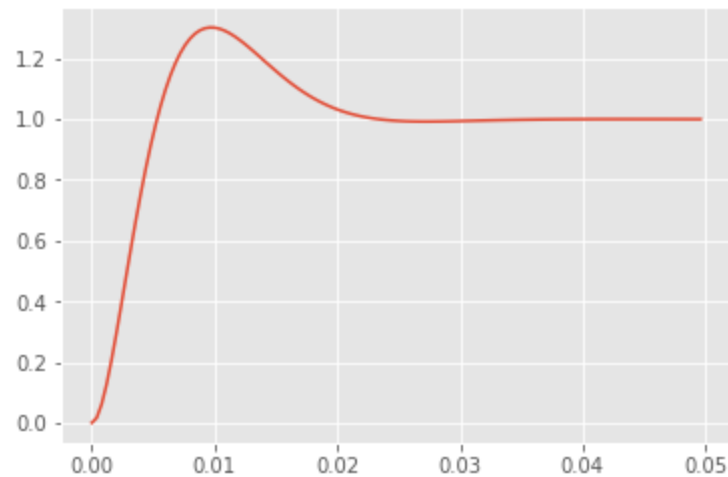
The compensated transfer function is:

Compensator  
$$\frac{1.224 s + 159}{0.001285 s + 1}$$

Final Sys  
$$293.8 s^2 + 3.845e+04 s + 3.816e+04$$

```
-----
0.001285 s^4 + 1.013 s^3 + 10.03 s^2 + 24 s
```

```
In []: tt, yy = control.step_response(control.feedback(comp_sys2), T = np.arange(0,0.05,0.04/100))
```



Which has a much smaller  $T_s$ . We also note that in the above two cases, no lag compensation is needed, since the  $K_v$  obtained is good enough.

## 4 Question 4

```
In [2]: import numpy as np
import math
import scipy
from control.matlab import *
import control as cont
import matplotlib.pyplot as plt
pie=np.pi
```

Stepinfo function which gives all the useful information like overshoot, rise time and settle time from the response function

```
In [23]: def stepinfo(t,yout):
    print ("Overshoot", (yout.max()/yout[-1]-1)*100, '%')
    print ("Time rise", t[next(i for i in range(0,len(yout)-1) if yout[i]>yout[-1]*.90)])
    print ("Time settle", t[next(len(yout)-i for i in range(2,len(yout)-1) if abs(yout[-1]-yout[i])<0.01]))
```

```
In [24]: M=10
ts=2
```

Calculating tau and wn from formulas

$$M = 100e^{\frac{-\zeta\pi}{\sqrt{1-\zeta^2}}}$$

$$ts = \frac{4}{\zeta\omega_n}$$

```
In [25]: tau=np.sqrt(np.log(M/100)**2/(np.log(M/100)**2+pie**2))
        wn=4/(tau*ts)
```

Calculating gamma and omega from the formulas

$$\gamma = \tan^{-1} \frac{2\zeta}{\sqrt{\sqrt{1+4\zeta^4}-2\zeta^2}}$$

$$\omega_{gc} = \omega_n \sqrt{\sqrt{1+4\zeta^4}-2\zeta^2}$$

```
In [26]: gamma=np.arctan(2*tau/np.sqrt(np.sqrt(1+4*tau**4)-2*tau**2))*180/pie
        wgc=wn*np.sqrt(np.sqrt(1+4*tau**4)-2*tau**2)
        print(f"Gamma {gamma} wgc {wgc}")
        print(f"w natural {wn}")
```

```
(num,den)=zpk2tf([-1,-0.01],[-10],100)
den1=tf([1],[1,2,2])
den2=tf([1],[1,0.02,0.0101])
H=tf(num,den)
H=H*den1*den2
print(f"Transfer function {H}")
```

Gamma 58.59306826496366 wgc 2.4422747101330673

w natural 3.38320725639016

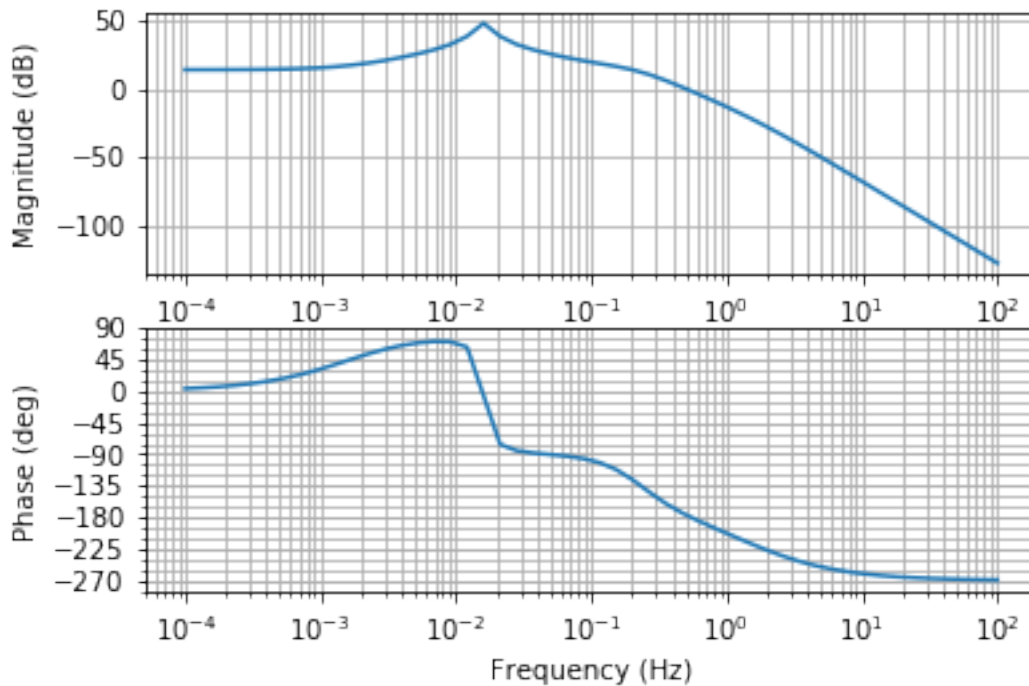
Transfer function

$$100 s^2 + 101 s + 1$$

$$\frac{100 s^2 + 101 s + 1}{s^5 + 12.02 s^4 + 22.25 s^3 + 20.56 s^2 + 0.6222 s + 0.202}$$

## 4.1 Bode Plot of H

```
In [27]: bode(H);
```



```
In [28]: [gm,pm,wg,wm]=margin(H)
print (f"gain-margin {gm} phase-margin {pm} \ngain cross over freq {wg} phase cross over

phi=gamma-pm
wm=wm
print(f"\nPhi {phi} w_m {wm}")

alpha=(1-np.sin(phi*pie/180))/(1+np.sin(phi*pie/180))
print(f"\nAlpha {alpha}")

T=1/np.sqrt(alpha)/wm
print(f"T {T}")

Gc=tf([T,1],[alpha*T,1])
Hnew=H*Gc
print(f"\nHnew {Hnew}")

gain-margin 1.214662566961802 phase-margin 3.7857102810716015
gain cross over freq 3.4359780521729917 phase cross over freq 3.1344441272838512

Phi 54.80735798389206 w_m 3.1344441272838512

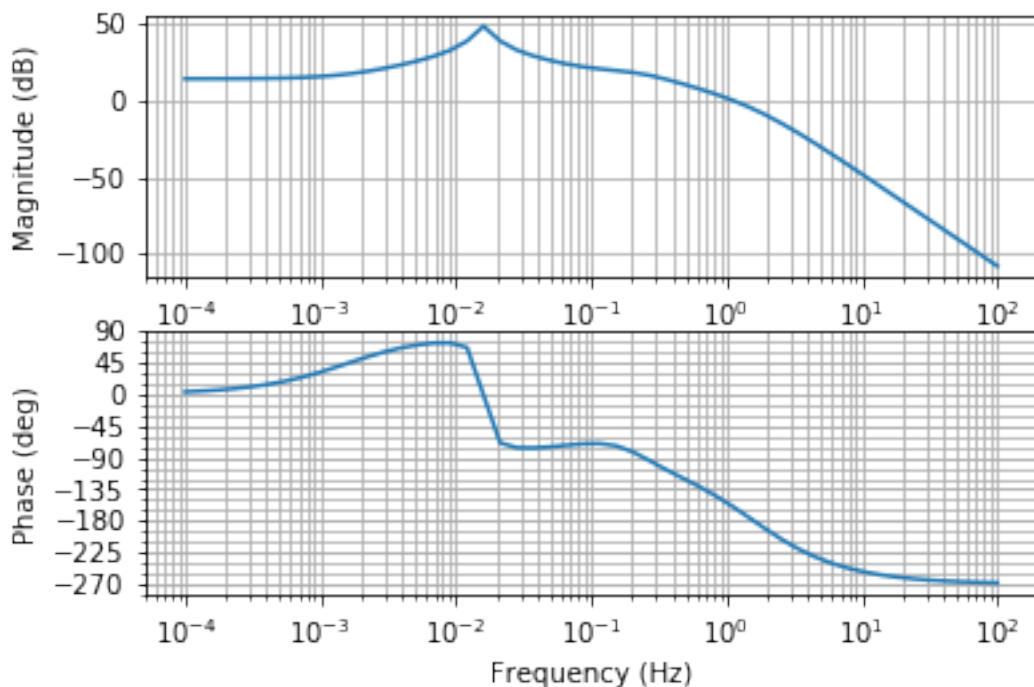
Alpha 0.1005828635112053
T 1.0059524889079905
```

Hnew

$$100.6 s^3 + 201.6 s^2 + 102 s + 1$$

$$0.1012 s^6 + 2.216 s^5 + 14.27 s^4 + 24.33 s^3 + 20.62 s^2 + 0.6426 s + 0.202$$

```
In [45]: bode(Hnew)
plt.show()
[gm,pm2,wg,wm2]=margin(Hnew)
print(f"Phase margin {pm2} Gain margin {wg2}")
print(f"Phi new {phi}")
```



```
Phase margin 20.823595702607207 Gain margin 6.246114286537816
Phi new 54.80735798389206
```

Since there is a huge difference between the phase margin and the desired value, we increment phi by 60 degrees

```
In [43]: phi_one=phi+60
print(f"Phi altered {phi_one}")

alpha_one=(1-np.sin(phi_one*pie/180))/(1+np.sin(phi_one*pie/180))
```



```

print(f"Alpha {alpha_one}")

mag10 = 20*np.log10(1/np.sqrt(alpha_one))
print(f"decibels magnitude {mag10}")

mag_one=np.sqrt(alpha_one)
print(f"Magnitude descent needed to offset {mag_one}")

```

```

Phi altered 114.80735798389206
Alpha 0.04836989761311832
decibels magnitude 13.154248319025266
Magnitude descent needed to offset 0.21993157484344608

```

To calculate the frequency at which magnitude is square root of alpha, we find the gain crossover frequency of the transfer function H/square root of alpha

```

In [40]: [gm1,pm1,wm1,wg1]=margin(H/mag_one)
print(f"Cross over {wg1}")

T_1=1/np.sqrt(alpha_one)/wg1
print(f"T_1 {T_1}")

Gc_one=tf([T_1,1],[alpha_one*T_1,1])
Hnew_one=Gc_one*H
print(Hnew_one)

```

```

Cross over 6.246114286537809
T_1 0.7279515743095443

```

```

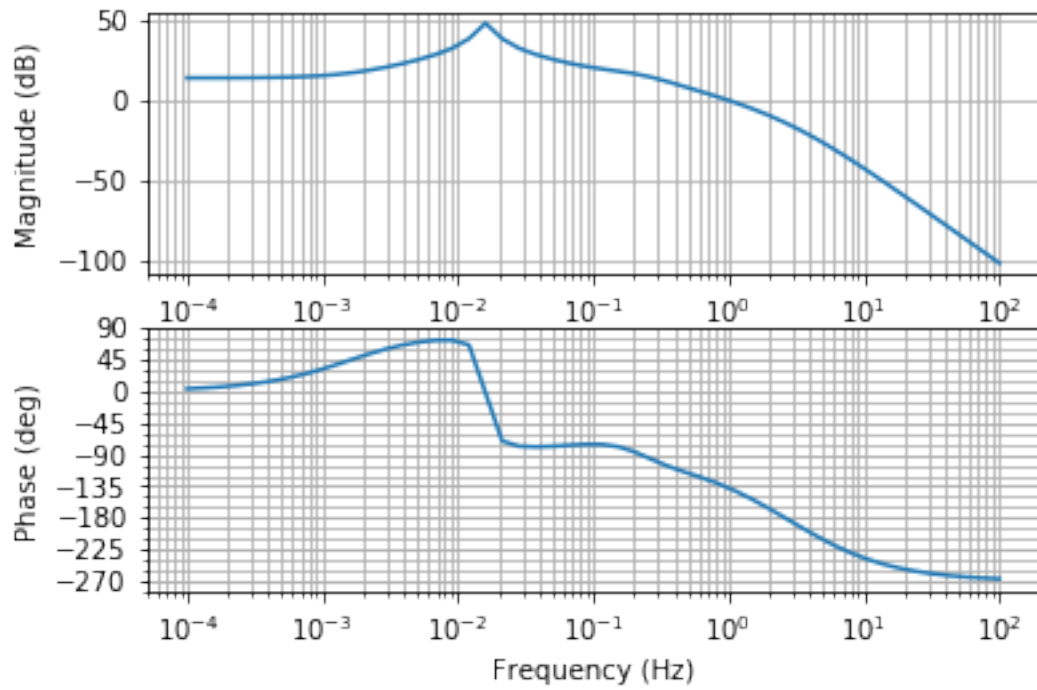
              72.8 s^3 + 173.5 s^2 + 101.7 s + 1
-----
0.03521 s^6 + 1.423 s^5 + 12.8 s^4 + 22.97 s^3 + 20.58 s^2 + 0.6293 s + 0.202

```

```

In [41]: bode(Hnew_one)
plt.show()
[gm2,pm2,wm2,wg2]=margin(Hnew_one)
print(f"Phase margin {pm2} Gain margin {wg2}")

```



Phase margin 42.84878119640814 Gain margin 6.246114286537816

## 4.2 Step Response of Closed Loop System

```
In [42]: T, yout = cont.step_response(feedback(Hnew_one,1), np.linspace(0,20,10000))
         plt.plot(T,yout*10)
         stepinfo(T,yout*10)
```

Overshoot 29.495260712314074 %

Time rise 0.266026602660266

Time settle 8.586858685868586

