

## Part II. Effect of component variations on pole locations in $H(s)$

Given: Circuit 1 has these nominal component values:  
(units are ohms, farads, and henries)

R1 = 1.0000  
R2 = 1.0000  
C1 = 0.1218  
C2 = 0.2940  
L1 = 0.2940  
L2 = 0.1218

This circuit's voltage magnitude-squared frequency response function  $|H(f)|^2$  may be expressed in terms of only  $\{p_1, p_2, SF\}$  instead of component values  $\{R_1, R_2, C_1, C_2, L_1, L_2\}$ :

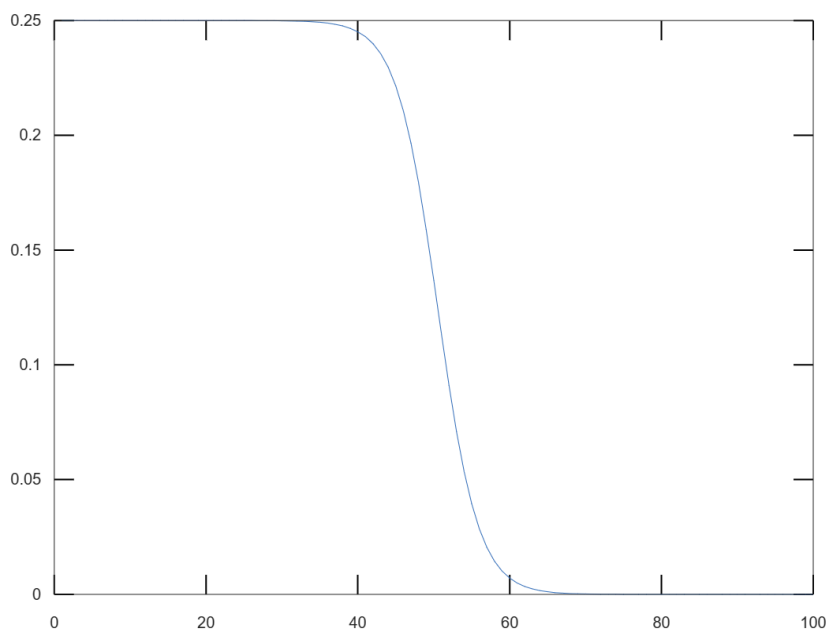
$$|H(s)|^2 = \left| \frac{SF}{(s - p_1)(s - p_1^*)(s - p_2)(s - p_2^*)} \right|^2, \text{ where } s = j2\pi f.$$

The goal of this part of Project 2 is to observe the effects of random variations in  $\{C_1, C_2, L_1, L_2\}$  on pole values  $\{p_1, p_2\}$ .

- a) Calculate  $|H(f)|^2$  for Circuit 1, using the provided function `Frequency_response1`, plot the result on a log frequency scale, using `f = logspace(-1,1,100)`.

**Hmag2 = Frequency\_response1(1.0000, 1.0000, 0.1218, 0.2940, 0.2940, 0.1218,  
f = logspace(-1,1,100))**

**plot(Hmag2);**



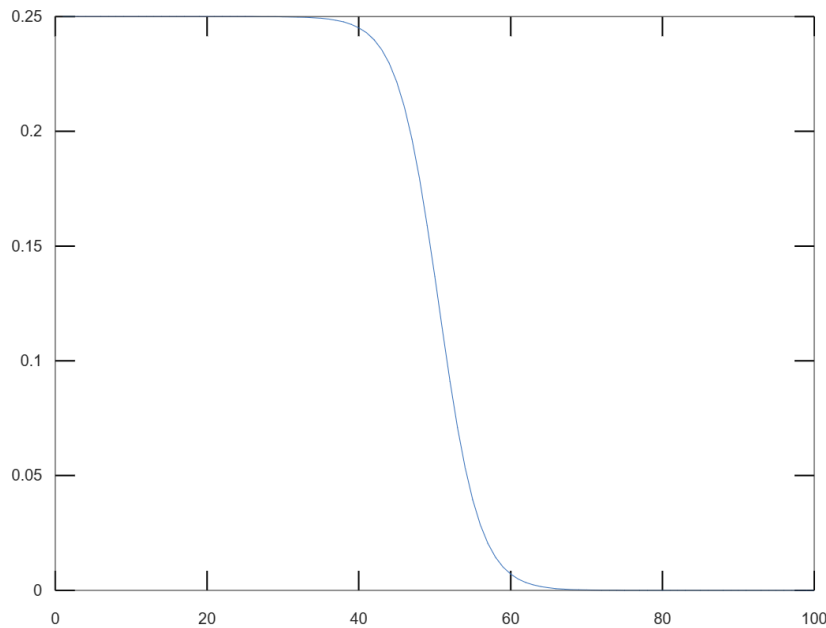
- b) Find pole values  $\{p_1, p_2\}$  and scale factor SF.

```
[p1,p2,SF] = Poles(1.0000, 1.0000, 0.1218, 0.2940, 0.2940, 0.1218);
```

- c) Find and plot  $|H(f)|^2$  from  $\{p_1, p_2, SF\}$  to demonstrate that all information is conveyed by these three parameters.

```
Hmag2 = Frequency_response_from_poles(-2.4047 + 5.8065i,-5.8055 +  
2.4051i,779.85,logspace(-1,1,100)) ;
```

```
plot(Hmag2);
```



- d) Keeping  $R_1 = 1$  and  $R_2 = 1$ , randomize the values of capacitors and inductors, then calculate the corresponding  $\{p_1, p_2\}$ . (Generate samples of  $\{C_1, C_2, L_1, L_2\}$  from continuous random variables uniformly-distributed over the range [nominal component values  $\pm 5\%$ ]). Repeat this 1000 times and store the resulting perturbed pole values. Plot the histograms of  $\text{Re}\{p_1\}$ ,  $\text{Im}\{p_1\}$ ,  $\text{Re}\{p_2\}$ , and  $\text{Im}\{p_2\}$ .

```

[p1, p2, SF] = Poles(R1, R2, C1, C2, L1, L2);

p = 5/100;
poles = [];
for n = 1:1000
    r1 = R1;
    r2 = R2;
    c1 = C1*(1+p*(2*rand-1));
    c2 = C2*(1+p*(2*rand-1));
    l1 = L1*(1+p*(2*rand-1));
    l2 = L2*(1+p*(2*rand-1));
    [P1, P2, SF] = Poles(r1, r2, c1, c2, l1, l2);
    poles = [poles; [P1, P2]];
end

```

end

```

rP1 = real(poles(:,1));
iP1 = imag(poles(:,1));
rP2 = real(poles(:,2));
iP2 = imag(poles(:,2));

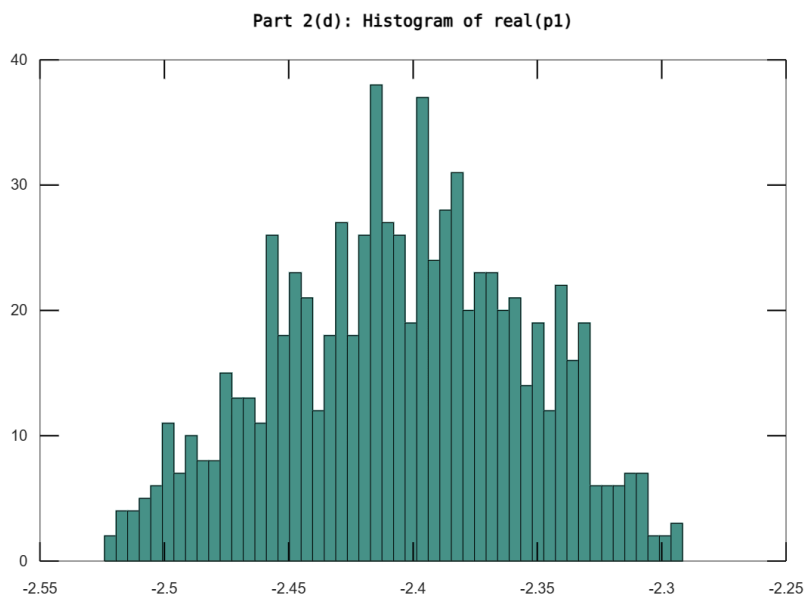
```

Histogram of  $\text{Re}\{p_1\}$

```

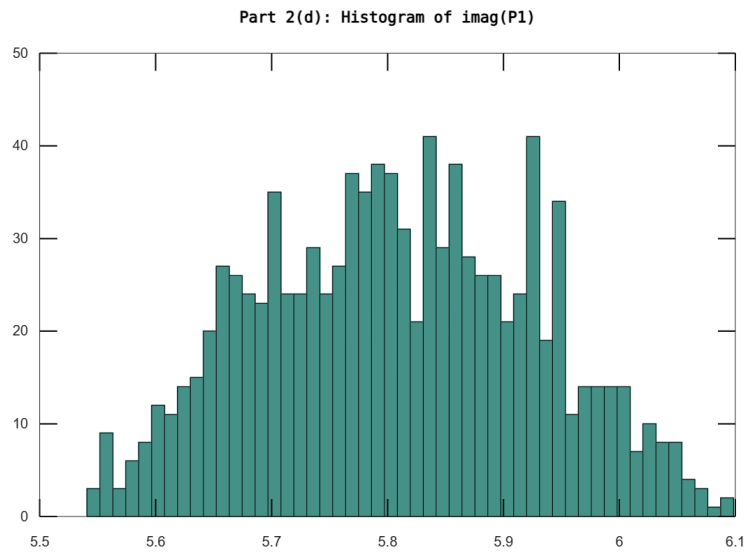
hist(rP1, 50)
mrP1 = mean(rP1);
srP1 = std(rP1);
title('Part 2(d): Histogram of real(p1)')

```



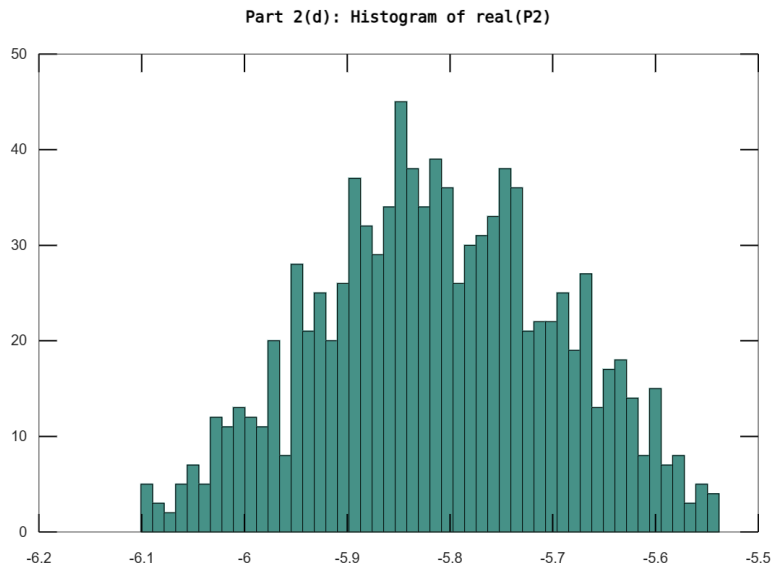
### Histogram of $\text{Im}\{p1\}$

```
hist(iP1, 50)
miP1 = mean(iP1);
siP1 = std(iP1);
title('Part 2(d): Histogram of imag(P1))
```



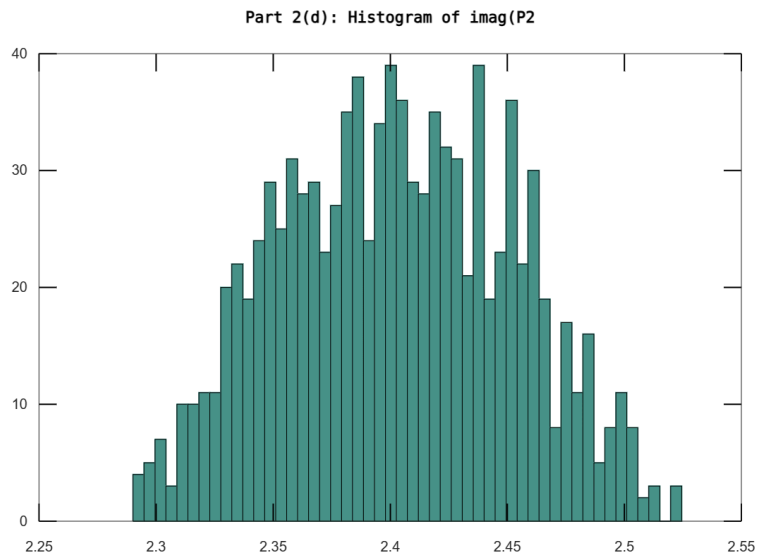
### Histogram of $\text{Re}\{p2\}$

```
hist(rP2, 50)
mrP2 = mean(rP2);
srP2 = std(rP2);
title('Part 2(d): Histogram of real(P2)')
```



### Histogram of $\text{Im}\{p_2\}$

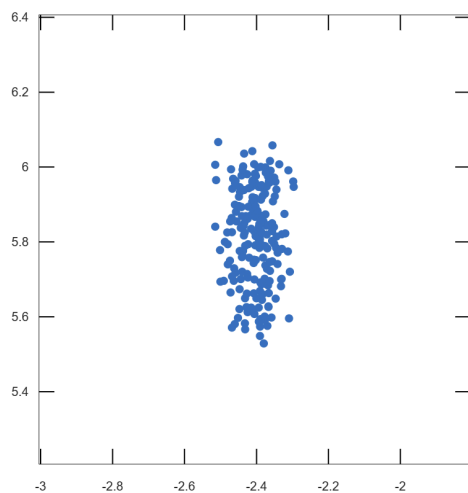
```
hist(iP2, 50)
miP2 = mean(iP2);
siP2 = std(iP2);
title('Part 2(d): Histogram of imag(P2)')
```



e) Plot the first 200 of the 1000 calculated values of  $p_1$  as dots on the complex plane:

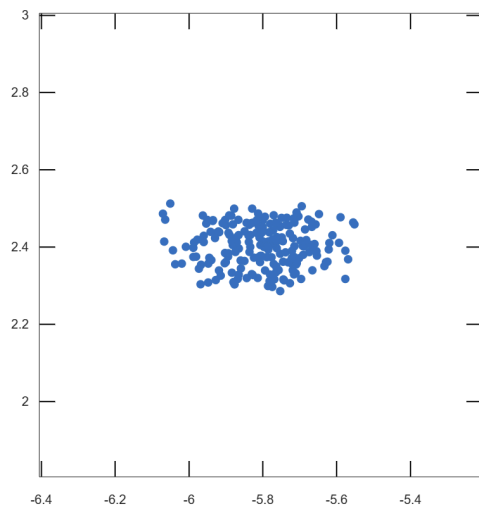
```
P1 = poles(:,1);
P2 = poles(:,2);

plot(P1(1:200),'.')
axis([real(p1)+[-0.6,0.6],imag(p1)+[-0.6,0.6]])
axis square
```



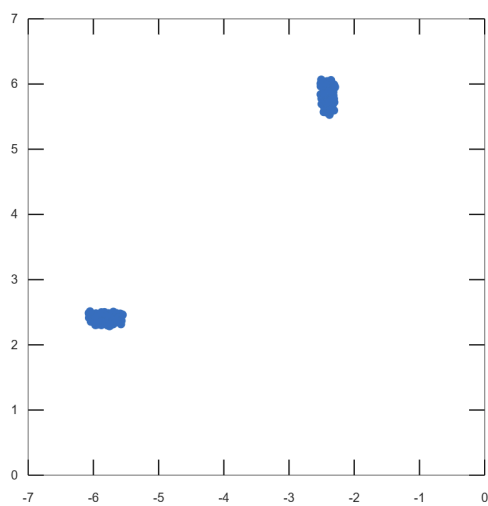
f) Plot the first 200 of the 1000 calculated values of  $p_2$  as dots on the complex plane:

```
plot(P2(1:200),'.')  
axis([real(p2)+[-0.6,0.6],imag(p2)+[-0.6,0.6])]  
axis square
```



g) Plot the first 200 of the 1000 calculated  $\{p_1, p_2\}$  pairs on the complex plane:

```
Poles1 = P1(1:200);  
Poles2 = P2(1:200);  
plot([Poles1(:);Poles2(:)],'.')  
axis([-7,0,0,7])  
axis square
```



- h) From the list of perturbed pole values calculated in (d), estimate the mean values and standard deviations of  $\text{Re}(p_1)$ ,  $\text{Im}(p_1)$ ,  $\text{Re}(p_2)$ , and  $\text{Im}(p_2)$ .

```
f = logspace(-1,1,1e2);  
[p1,p2,SF] = Poles(R1,R2,C1,C2,L1,L2);  
  
fc = 5/100;  
poles = [];  
for n = 1:1000  
    r1 = R1;  
    r2 = R2;  
    c1 = C1*(1+fc*(2*rand-1));  
    c2 = c2*(1+fc*(2*rand-1));  
    l1 = L1*(1+fc*(2*rand-1));  
    l2 = L2*(1+fc*(2*rand-1));  
    [P1,P2,SF] = Poles(r1,r2,c1,c2,l1,l2);  
    poles = [poles; [P1,P2]];  
  
    end  
  
rP1 = real(poles(:,1));  
iP1 = imag(poles(:,1));  
rP2 = real(poles(:,2));  
iP2 = imag(poles(:,2));  
  
mean_real_P1 = mean(rP1)  
sdev_real_P1 = std(rP1)  
  
mean_imag_P1 = mean(iP1)  
sdev_imag_P1 = std(iP1)  
  
mean_real_P2 = mean(rP2)  
sdev_real_P2 = std(rP2)  
  
mean_imag_P2 = mean(iP2)  
sdev_imag_P2 = std(iP2)
```

**Answers:**

**mean\_real\_P1 = -2.4943**

**sdev\_real\_P1 = 0.3987**

**mean\_imag\_P1 = 5.5685**

**sdev\_imag\_P1 = 0.9440**

**mean\_real\_P2 = -5.7355**

**sdev\_real\_P2 = 0.3968**

**mean\_imag\_P2 = 2.2254**

**sdev\_imag\_P2 = 0.5813**

- i) From the statistics calculated in (h), generate 200 samples of the following "synthetic perturbed poles" whose PDF's are intended to mimic those of the actual perturbed poles found in (d)

**X1 = randn(200, 1);**

**X2 = randn(200, 1);**

**X3 = randn(200, 1);**

**X4 = randn(200, 1);**

**pertp1 = (mrP1 + stdrP1 + X1) + j\*(miP1 + stdrP1 + X2);**

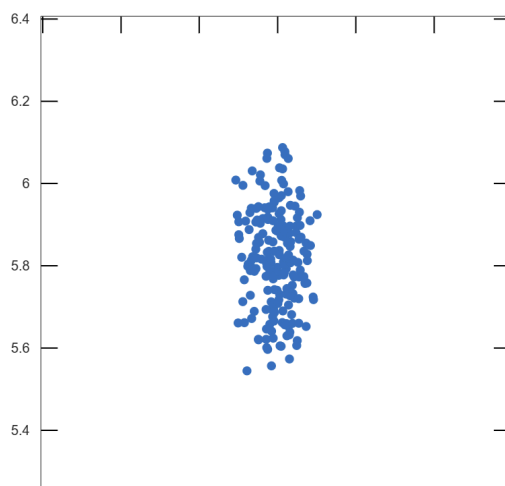
**pertp2 = (mrP2 + stdrP2 + X4) + j\*(miP2 + stdrP2 + X4);**

- j) Repeat pole plots (e)-(g) for the 1000 synthetic perturbed poles  $\{\tilde{p}_1, \tilde{p}_2\}$  that were generated in (i).

**plot(pertpl(1:200), 'b')**

**axis([real(p1)+[-0.6,0.6],imag(p1)+[-0.6,0.6]])**

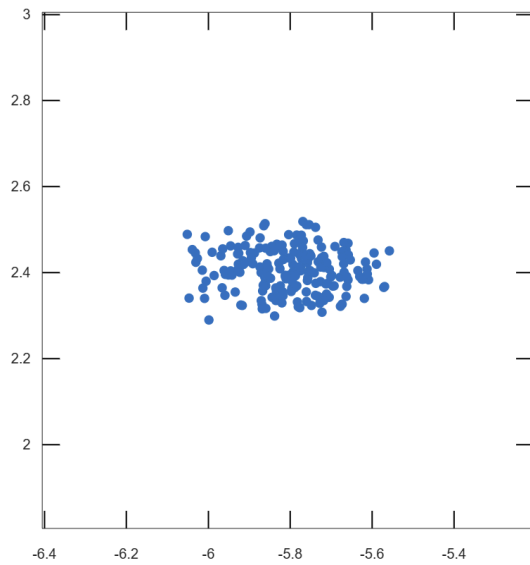
**axis square**





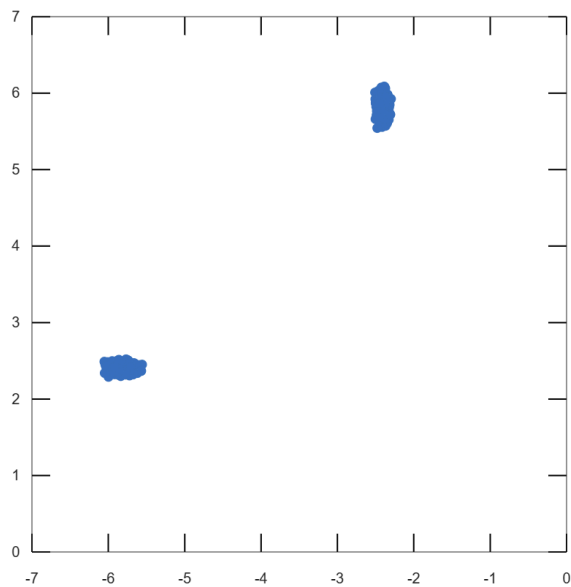
k)

```
plot(pertpl(1:200), 'b.')  
axis([real(p2)+[-0.6,0.6],imag(p2)+[-0.6,0.6]])  
axis square
```



l)

```
plot([pertp1(:); pertp2(:)], 'b.')  
axis([-7,0,0,7])  
axis square
```



- m) Create 25 superimposed plots of  $|H(f)|^2$  vs.  $f$  on a single graph, using the provided function `Frequency_response_from_poles` and the first 25 out of the 1000 perturbed pole values that you found in (d).

```
[p1, p2, SF] = Poles(R1, R2, C1, C2, L1, L2);
```

```
percent = 25/1000;
```

```
poles = [];
```

```
for n = 1:25
```

```
    r1 = R1;
```

```
    r2 = R2;
```

```
    c1 = 0.1218*(1+percent*(2*rand-1));
```

```
    c2 = 0.2940*(1+percent*(2*rand-1));
```

```
    l1 = 0.2940*(1+percent*(2*rand-1));
```

```
    l2 = 0.1218*(1+percent*(2*rand-1));
```

```
    [P1, P2, SF] = Poles(r1, r2, c1, c2, l1, l2);
```

```
    Hmag2m = Frequency_response_from_poles(P1, P2, SF, f);
```

```
    semilogx(f, Hmag2m);
```

```
    hold on
```

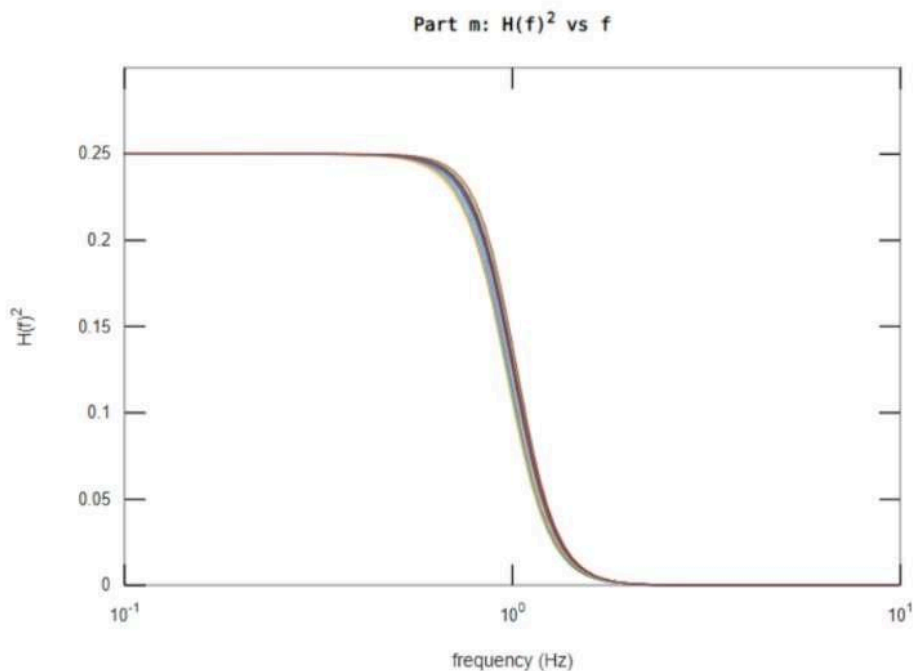
```
end
```

```
title('Part m:  $H(f)^2$  vs  $f$ ')
```

```
xlabel('frequency (Hz)')
```

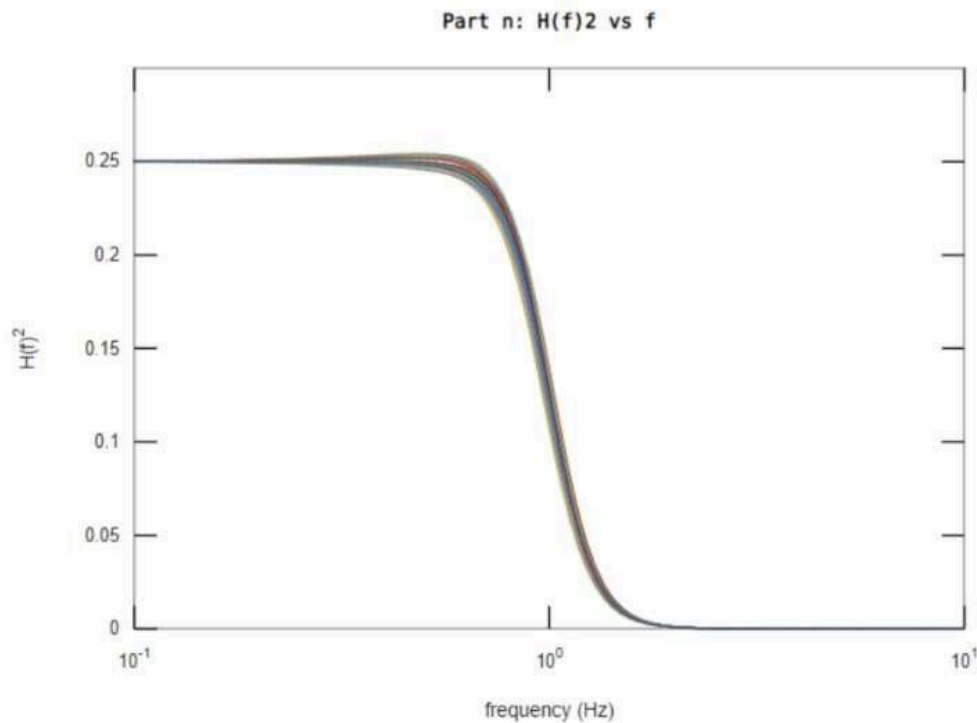
```
ylabel('H(f)^2')
```

```
axis([0.1, 10.0, 0.3])
```



- n) Create 25 superimposed plots of  $|H(f)|^2$  vs.  $f$  on a single graph, using the provided function `Frequency_response_from_poles` and the first 25 out of the 1000 synthetic pole values that you found in (i). (Since the scale factor was not calculated for the synthetic poles, use  $SF = 1/2 |\tilde{p}_1 \tilde{p}_2|^2$  in each case.)

```
SF = 0.5*(abs(pertp1(1:25).*pertp2(1:25))).^2;  
Hmag2n = Frequency_response_from_poles(pertp1(1:25), pertp2(1:25), SF, f);  
semilogx(f, Hmag2n)  
title('Part n: H(f)2 vs f')
```



Conclusion:

In parts m and n, the two methods resulted in slightly different  $|H(f)|^2$  vs  $f$  graphs. For part m, there is very little difference due to it being a uniform distribution, which each outcome is equally likely. However in part n, there is a larger difference in the graph due to a normal distribution being used. In a normal distribution, each outcome has its own probability occurrence, which causes more variation in the graph. Pole values are closer together when using a uniform distribution rather than a normal distribution.

