

Complementary Filters Shaping Using \mathcal{H}_∞ Synthesis - Matlab Computation

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In this document, the design of complementary filters is studied.

One use of complementary filter is described below:

The basic idea of a complementary filter involves taking two or more sensors, filtering out unreliable frequencies for each sensor, and combining the filtered outputs to get a better estimate throughout the entire bandwidth of the system. To achieve this, the sensors included in the filter should complement one another by performing better over specific parts of the system bandwidth.

This document is divided into several sections:

- in section 1, the \mathcal{H}_∞ synthesis is used for generating two complementary filters
- in section 2, a method using the \mathcal{H}_∞ synthesis is proposed to shape three of more complementary

filters

- in section 3, the \mathcal{H}_∞ synthesis is used and compared with FIR complementary filters used for LIGO

Note

Add the Matlab code use to obtain the results presented in the paper are accessible [here](#) and presented below.

1 H-Infinity synthesis of complementary filters

Note

The Matlab file corresponding to this section is accessible [here](#).

1.1 Synthesis Architecture

We here synthesize two complementary filters using the \mathcal{H}_∞ synthesis. The goal is to specify upper bounds on the norms of the two complementary filters $H_1(s)$ and $H_2(s)$ while ensuring their complementary property ($H_1(s) + H_2(s) = 1$).

In order to do so, we use the generalized plant shown on figure 1.1 where $W_1(s)$ and $W_2(s)$ are weighting transfer functions that will be used to shape $H_1(s)$ and $H_2(s)$ respectively.

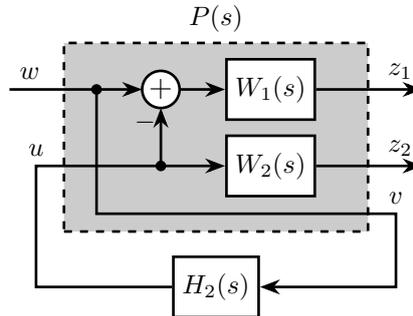


Figure 1.1: \mathcal{H}_∞ synthesis of the complementary filters

The \mathcal{H}_∞ synthesis applied on this generalized plant will give a transfer function H_2 (figure 1.1) such that the \mathcal{H}_∞ norm of the transfer function from w to $[z_1, z_2]$ is less than one:

$$\left\| \begin{array}{c} (1 - H_2(s))W_1(s) \\ H_2(s)W_2(s) \end{array} \right\|_\infty < 1$$

Thus, if the above condition is verified, we can define $H_1(s) = 1 - H_2(s)$ and we have that:

$$\left\| \begin{array}{c} H_1(s)W_1(s) \\ H_2(s)W_2(s) \end{array} \right\|_\infty < 1$$

Which is almost (with an maximum error of $\sqrt{2}$) equivalent to:

$$|H_1(j\omega)| < \frac{1}{|W_1(j\omega)|}, \quad \forall \omega$$

$$|H_2(j\omega)| < \frac{1}{|W_2(j\omega)|}, \quad \forall \omega$$

We then see that $W_1(s)$ and $W_2(s)$ can be used to shape both $H_1(s)$ and $H_2(s)$ while ensuring their complementary property by the definition of $H_1(s) = 1 - H_2(s)$.

1.2 Design of Weighting Function

A formula is proposed to help the design of the weighting functions:

$$W(s) = \left(\frac{\frac{1}{\omega_0} \sqrt{\frac{1 - \left(\frac{G_0}{G_c}\right)^{\frac{2}{n}}}{1 - \left(\frac{G_c}{G_\infty}\right)^{\frac{2}{n}}} s + \left(\frac{G_0}{G_c}\right)^{\frac{1}{n}}}}{\left(\frac{1}{G_\infty}\right)^{\frac{1}{n}} \frac{1}{\omega_0} \sqrt{\frac{1 - \left(\frac{G_0}{G_c}\right)^{\frac{2}{n}}}{1 - \left(\frac{G_c}{G_\infty}\right)^{\frac{2}{n}}} s + \left(\frac{1}{G_c}\right)^{\frac{1}{n}}}} \right)^n \quad (1.1)$$

The parameters permits to specify:

- the low frequency gain: $G_0 = \lim_{\omega \rightarrow 0} |W(j\omega)|$
- the high frequency gain: $G_\infty = \lim_{\omega \rightarrow \infty} |W(j\omega)|$
- the absolute gain at ω_0 : $G_c = |W(j\omega_0)|$
- the absolute slope between high and low frequency: n

The general shape of a weighting function generated using the formula is shown in figure 1.2.

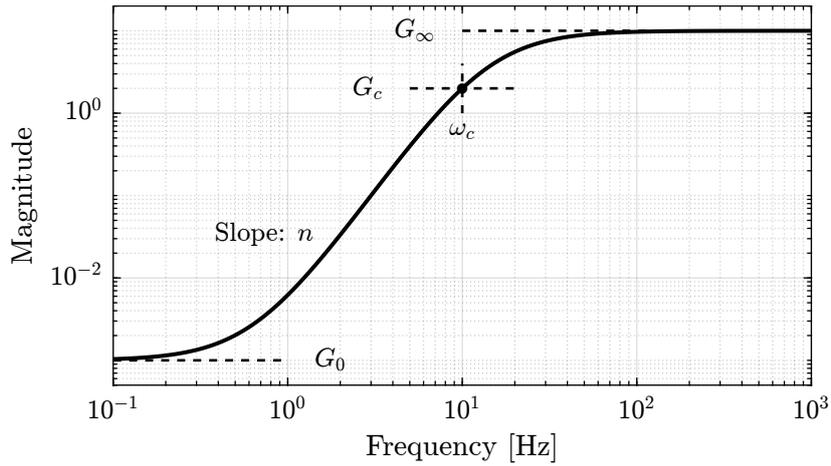


Figure 1.2: Gain of the Weighting Function formula

```

Matlab
n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
→ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 3; w0 = 2*pi*10; G0 = 1000; G1 = 0.1; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
→ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

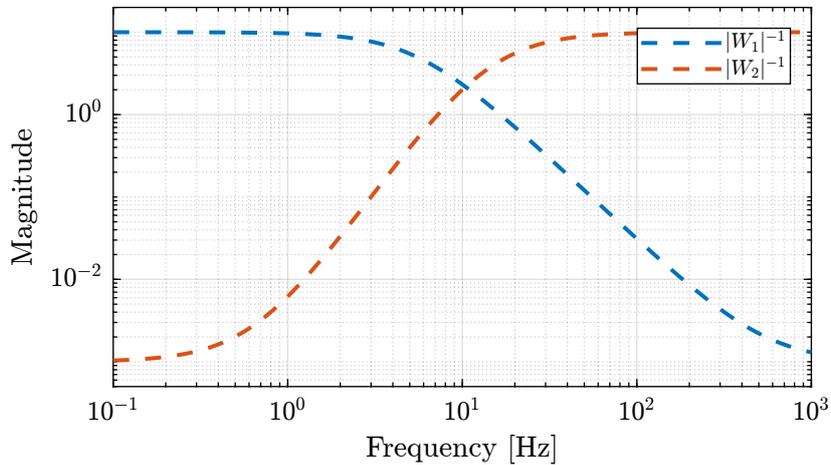


Figure 1.3: Weights on the complementary filters W_1 and W_2 and the associated performance weights

1.3 H-Infinity Synthesis

We define the generalized plant P on matlab.

```

Matlab
P = [W1 -W1;
     0  W2;
     1  0];

```

And we do the \mathcal{H}_∞ synthesis using the `hinfsyn` command.

```

Matlab
[H2, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

```

```

Results
[H2, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
Resetting value of Gamma min based on D_11, D_12, D_21 terms

Test bounds:      0.1000 < gamma <= 1050.0000

   gamma   hamx_eig  xinf_eig  hamy_eig  yinf_eig  nrho_xy  p/f
1.050e+03  2.8e+01  2.4e-07  4.1e+00  0.0e+00  0.0000  p
525.050    2.8e+01  2.4e-07  4.1e+00  0.0e+00  0.0000  p
262.575    2.8e+01  2.4e-07  4.1e+00  0.0e+00  0.0000  p
131.337    2.8e+01  2.4e-07  4.1e+00  -1.0e-13 0.0000  p
65.719     2.8e+01  2.4e-07  4.1e+00  -9.5e-14 0.0000  p
32.909     2.8e+01  2.4e-07  4.1e+00  0.0e+00  0.0000  p

```

```
16.505 2.8e+01 2.4e-07 4.1e+00 -1.0e-13 0.0000 p
8.302 2.8e+01 2.4e-07 4.1e+00 -7.2e-14 0.0000 p
4.201 2.8e+01 2.4e-07 4.1e+00 -2.5e-25 0.0000 p
2.151 2.7e+01 2.4e-07 4.1e+00 -3.8e-14 0.0000 p
1.125 2.6e+01 2.4e-07 4.1e+00 -5.4e-24 0.0000 p
0.613 2.3e+01 -3.7e+01# 4.1e+00 0.0e+00 0.0000 f
0.869 2.6e+01 -3.7e+02# 4.1e+00 0.0e+00 0.0000 f
0.997 2.6e+01 -1.1e+04# 4.1e+00 0.0e+00 0.0000 f
1.061 2.6e+01 2.4e-07 4.1e+00 0.0e+00 0.0000 p
1.029 2.6e+01 2.4e-07 4.1e+00 0.0e+00 0.0000 p
1.013 2.6e+01 2.4e-07 4.1e+00 0.0e+00 0.0000 p
1.005 2.6e+01 2.4e-07 4.1e+00 0.0e+00 0.0000 p
1.001 2.6e+01 -3.1e+04# 4.1e+00 -3.8e-14 0.0000 f
1.003 2.6e+01 -2.8e+05# 4.1e+00 0.0e+00 0.0000 f
1.004 2.6e+01 2.4e-07 4.1e+00 -5.8e-24 0.0000 p
1.004 2.6e+01 2.4e-07 4.1e+00 0.0e+00 0.0000 p

Gamma value achieved: 1.0036
```

We then define the high pass filter $H_1 = 1 - H_2$. The bode plot of both H_1 and H_2 is shown on figure 1.4.

```
H1 = 1 - H2; Matlab
```

1.4 Obtained Complementary Filters

The obtained complementary filters are shown on figure 1.4.

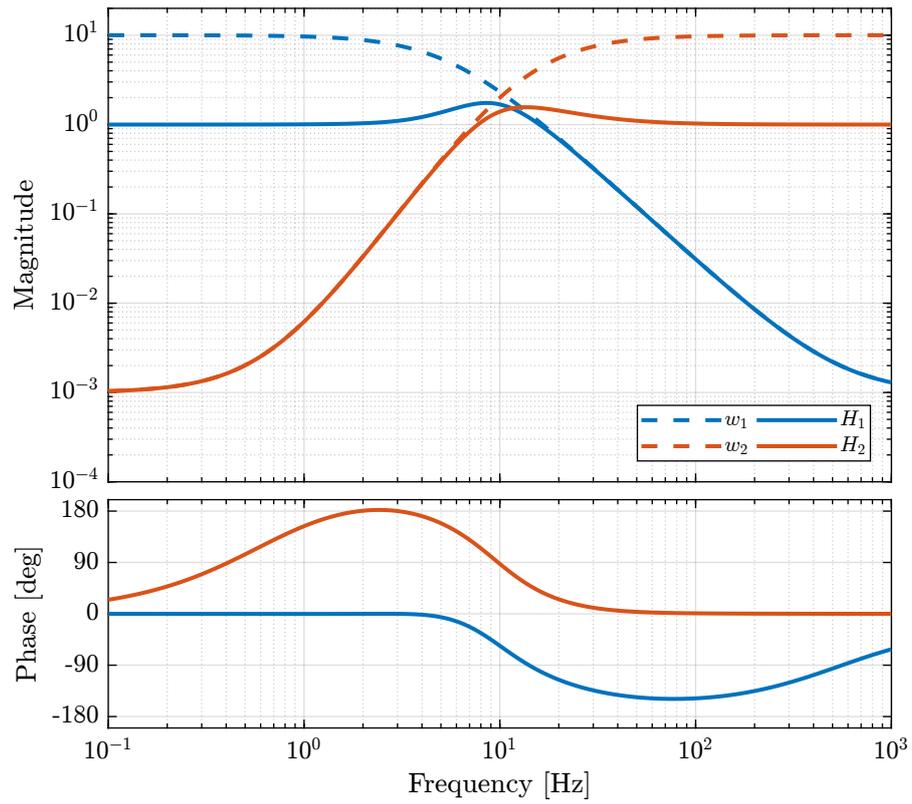


Figure 1.4: Obtained complementary filters using \mathcal{H}_∞ synthesis

2 Generating 3 complementary filters

Note

The Matlab file corresponding to this section is accessible [here](#).

2.1 Theory

We want:

$$\begin{aligned} |H_1(j\omega)| &< 1/|W_1(j\omega)|, \quad \forall\omega \\ |H_2(j\omega)| &< 1/|W_2(j\omega)|, \quad \forall\omega \\ |H_3(j\omega)| &< 1/|W_3(j\omega)|, \quad \forall\omega \\ H_1(s) + H_2(s) + H_3(s) &= 1 \end{aligned}$$

For that, we use the \mathcal{H}_∞ synthesis with the architecture shown on figure 2.1.

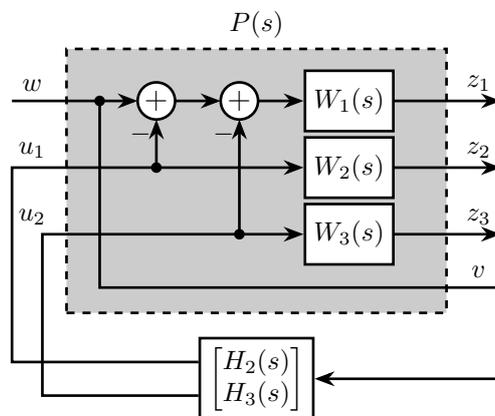


Figure 2.1: Generalized architecture for generating 3 complementary filters

The \mathcal{H}_∞ objective is:

$$\begin{aligned} |(1 - H_2(j\omega) - H_3(j\omega))W_1(j\omega)| &< 1, \quad \forall\omega \\ |H_2(j\omega)W_2(j\omega)| &< 1, \quad \forall\omega \\ |H_3(j\omega)W_3(j\omega)| &< 1, \quad \forall\omega \end{aligned}$$

And thus if we choose $H_1 = 1 - H_2 - H_3$ we have solved the problem.

2.2 Weights

First we define the weights.

```

Matlab
n = 2; w0 = 2*pi*1; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
↪ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

W2 = 0.22*(1 + s/2/pi/1)^2/(sqrt(1e-4) + s/2/pi/1)^2*(1 + s/2/pi/10)^2/(1 + s/2/pi/1000)^2;

n = 3; w0 = 2*pi*10; G0 = 1000; G1 = 0.1; Gc = 1/2;
W3 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
↪ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

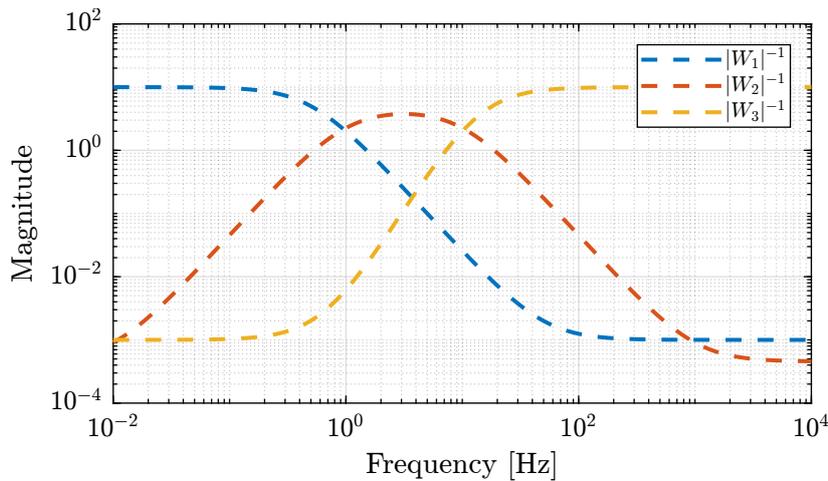


Figure 2.2: Three weighting functions used for the \mathcal{H}_∞ synthesis of the complementary filters

2.3 H-Infinity Synthesis

Then we create the generalized plant P .

```

Matlab
P = [W1 -W1 -W1;
     0 W2 0;
     0 0 W3;
     1 0 0];

```

And we do the \mathcal{H}_∞ synthesis.

```

Matlab
[H, ~, gamma, ~] = hinfyn(P, 1, 2, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

```

```

Results
[H, ~, gamma, ~] = hinfyn(P, 1, 2, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
Resetting value of Gamma min based on D_11, D_12, D_21 terms

```

```

Test bounds:      0.1000 < gamma <= 1050.0000

```

gamma	hamx_eig	xinf_eig	hamy_eig	yinf_eig	nrho_xy	p/f
1.050e+03	3.2e+00	4.5e-13	6.3e-02	-1.2e-11	0.0000	p
525.050	3.2e+00	1.3e-13	6.3e-02	0.0e+00	0.0000	p
262.575	3.2e+00	2.1e-12	6.3e-02	-1.5e-13	0.0000	p
131.337	3.2e+00	1.1e-12	6.3e-02	-7.2e-29	0.0000	p
65.719	3.2e+00	2.0e-12	6.3e-02	0.0e+00	0.0000	p
32.909	3.2e+00	7.4e-13	6.3e-02	-5.9e-13	0.0000	p
16.505	3.2e+00	1.4e-12	6.3e-02	0.0e+00	0.0000	p
8.302	3.2e+00	1.6e-12	6.3e-02	0.0e+00	0.0000	p
4.201	3.2e+00	1.6e-12	6.3e-02	0.0e+00	0.0000	p
2.151	3.2e+00	1.6e-12	6.3e-02	0.0e+00	0.0000	p
1.125	3.2e+00	2.8e-12	6.3e-02	0.0e+00	0.0000	p
0.613	3.0e+00	-2.5e+03#	6.3e-02	0.0e+00	0.0000	f
0.869	3.1e+00	-2.9e+01#	6.3e-02	0.0e+00	0.0000	f
0.997	3.2e+00	1.9e-12	6.3e-02	0.0e+00	0.0000	p
0.933	3.1e+00	-6.9e+02#	6.3e-02	0.0e+00	0.0000	f
0.965	3.1e+00	-3.0e+03#	6.3e-02	0.0e+00	0.0000	f
0.981	3.1e+00	-8.6e+03#	6.3e-02	0.0e+00	0.0000	f
0.989	3.2e+00	-2.7e+04#	6.3e-02	0.0e+00	0.0000	f
0.993	3.2e+00	-5.7e+05#	6.3e-02	0.0e+00	0.0000	f
0.995	3.2e+00	2.2e-12	6.3e-02	0.0e+00	0.0000	p
0.994	3.2e+00	1.6e-12	6.3e-02	0.0e+00	0.0000	p
0.994	3.2e+00	1.0e-12	6.3e-02	0.0e+00	0.0000	p

```

Gamma value achieved:      0.9936

```

2.4 Obtained Complementary Filters

The obtained filters are:

```

Matlab
H2 = tf(H(1));
H3 = tf(H(2));
H1 = 1 - H2 - H3;

```

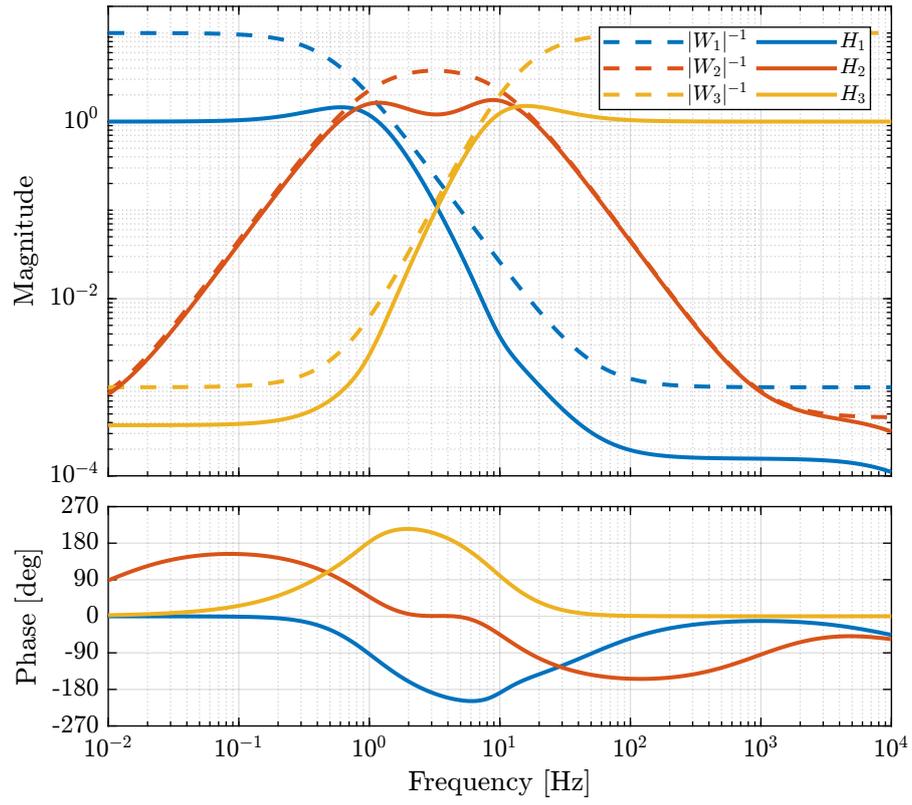


Figure 2.3: The three complementary filters obtained after \mathcal{H}_∞ synthesis

3 Implement complementary filters for LIGO

Note

The Matlab file corresponding to this section is accessible [here](#).

Let's try to design complementary filters that are corresponding to the complementary filters design for the LIGO and described in [hua05_low_ligo].

The FIR complementary filters designed in [hua05_low_ligo] are of order 512.

3.1 Specifications

The specifications for the filters are:

1. From 0 to 0.008 Hz, the magnitude of the filter's transfer function should be less than or equal to 8×10^{-3}
2. From 0.008 Hz to 0.04 Hz, it attenuates the input signal proportional to frequency cubed
3. Between 0.04 Hz and 0.1 Hz, the magnitude of the transfer function should be less than 3
4. Above 0.1 Hz, the maximum of the magnitude of the complement filter should be as close to zero as possible. In our system, we would like to have the magnitude of the complementary filter to be less than 0.1. As the filters obtained in [hua05_low_ligo] have a magnitude of 0.045, we will set that as our requirement

The specifications are translated in upper bounds of the complementary filters are shown on figure 3.1.

3.2 FIR Filter

We here try to implement the FIR complementary filter synthesis as explained in [hua05_low_ligo]. For that, we use the [CVX matlab Toolbox](#).

We setup the CVX toolbox and use the [SeDuMi](#) solver.

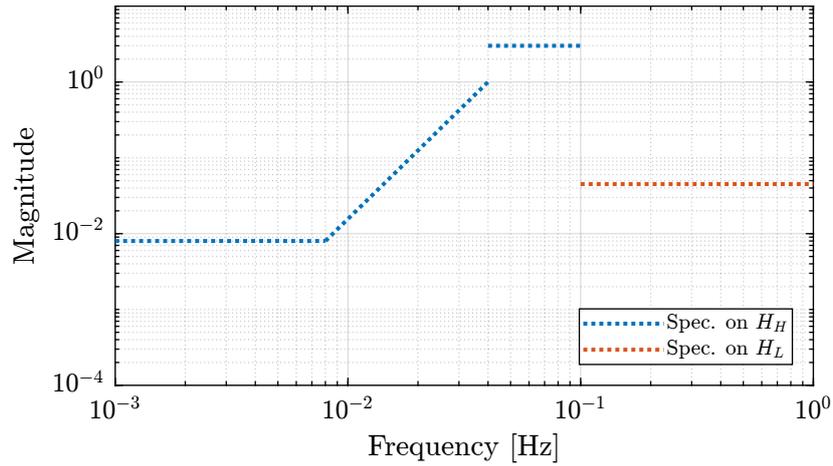


Figure 3.1: Specification for the LIGO complementary filters

```

Matlab
cvx_startup;
cvx_solver sedumi;

```

We define the frequency vectors on which we will constrain the norm of the FIR filter.

```

Matlab
w1 = 0:4.06e-4:0.008;
w2 = 0.008:4.06e-4:0.04;
w3 = 0.04:8.12e-4:0.1;
w4 = 0.1:8.12e-4:0.83;

```

We then define the order of the FIR filter.

```

Matlab
n = 512;

```

```

Matlab
A1 = [ones(length(w1),1), cos(kron(w1'.*(2*pi),[1:n-1]))];
A2 = [ones(length(w2),1), cos(kron(w2'.*(2*pi),[1:n-1]))];
A3 = [ones(length(w3),1), cos(kron(w3'.*(2*pi),[1:n-1]))];
A4 = [ones(length(w4),1), cos(kron(w4'.*(2*pi),[1:n-1]))];

B1 = [zeros(length(w1),1), sin(kron(w1'.*(2*pi),[1:n-1]))];
B2 = [zeros(length(w2),1), sin(kron(w2'.*(2*pi),[1:n-1]))];
B3 = [zeros(length(w3),1), sin(kron(w3'.*(2*pi),[1:n-1]))];
B4 = [zeros(length(w4),1), sin(kron(w4'.*(2*pi),[1:n-1]))];

```

We run the convex optimization.

```

Matlab
cvx_begin
variable y(n+1,1)
% t

```

```

maximize(-y(1))

for i = 1:length(w1)
    norm([0 A1(i,:); 0 B1(i,:)]*y) <= 8e-3;
end

for i = 1:length(w2)
    norm([0 A2(i,:); 0 B2(i,:)]*y) <= 8e-3*(2*pi*w2(i)/(0.008*2*pi))^3;
end

for i = 1:length(w3)
    norm([0 A3(i,:); 0 B3(i,:)]*y) <= 3;
end

for i = 1:length(w4)
    norm([1 0]' - [0 A4(i,:); 0 B4(i,:)]*y) <= y(1);
end

cvx_end

h = y(2:end);

```

Results

```

cvx_begin
variable y(n+1,1)
% t
maximize(-y(1))
for i = 1:length(w1)
    norm([0 A1(i,:); 0 B1(i,:)]*y) <= 8e-3;
end
for i = 1:length(w2)
    norm([0 A2(i,:); 0 B2(i,:)]*y) <= 8e-3*(2*pi*w2(i)/(0.008*2*pi))^3;
end
for i = 1:length(w3)
    norm([0 A3(i,:); 0 B3(i,:)]*y) <= 3;
end
for i = 1:length(w4)
    norm([1 0]' - [0 A4(i,:); 0 B4(i,:)]*y) <= y(1);
end
cvx_end

Calling SeDuMi 1.34: 4291 variables, 1586 equality constraints
For improved efficiency, SeDuMi is solving the dual problem.
-----
SeDuMi 1.34 (beta) by AdvOL, 2005-2008 and Jos F. Sturm, 1998-2003.
Alg = 2: xz-corrector, Adaptive Step-Differentiation, theta = 0.250, beta = 0.500
eqs m = 1586, order n = 3220, dim = 4292, blocks = 1073
nnz(A) = 1100727 + 0, nnz(ADA) = 1364794, nnz(L) = 683190
it :    b*y      gap  delta  rate  t/tP*  t/tD*  feas cg cg  prec
0 :      4.11E+02  0.000
1 :  -2.58E+00  1.25E+02  0.000  0.3049  0.9000  0.9000  4.87  1  1  3.0E+02
2 :  -2.36E+00  3.90E+01  0.000  0.3118  0.9000  0.9000  1.83  1  1  6.6E+01
3 :  -1.69E+00  1.31E+01  0.000  0.3354  0.9000  0.9000  1.76  1  1  1.5E+01
4 :  -8.60E-01  7.10E+00  0.000  0.5424  0.9000  0.9000  2.48  1  1  4.8E+00
5 :  -4.91E-01  5.44E+00  0.000  0.7661  0.9000  0.9000  3.12  1  1  2.5E+00
6 :  -2.96E-01  3.88E+00  0.000  0.7140  0.9000  0.9000  2.62  1  1  1.4E+00
7 :  -1.98E-01  2.82E+00  0.000  0.7271  0.9000  0.9000  2.14  1  1  8.5E-01
8 :  -1.39E-01  2.00E+00  0.000  0.7092  0.9000  0.9000  1.78  1  1  5.4E-01
9 :  -9.99E-02  1.30E+00  0.000  0.6494  0.9000  0.9000  1.51  1  1  3.3E-01
10 : -7.57E-02  8.03E-01  0.000  0.6175  0.9000  0.9000  1.31  1  1  2.0E-01
11 : -5.99E-02  4.22E-01  0.000  0.5257  0.9000  0.9000  1.17  1  1  1.0E-01
12 : -5.28E-02  2.45E-01  0.000  0.5808  0.9000  0.9000  1.08  1  1  5.9E-02
13 : -4.82E-02  1.28E-01  0.000  0.5218  0.9000  0.9000  1.05  1  1  3.1E-02
14 : -4.56E-02  5.65E-02  0.000  0.4417  0.9045  0.9000  1.02  1  1  1.4E-02
15 : -4.43E-02  2.41E-02  0.000  0.4265  0.9004  0.9000  1.01  1  1  6.0E-03
16 : -4.37E-02  8.90E-03  0.000  0.3690  0.9070  0.9000  1.00  1  1  2.3E-03
17 : -4.35E-02  3.24E-03  0.000  0.3641  0.9164  0.9000  1.00  1  1  9.5E-04
18 : -4.34E-02  1.55E-03  0.000  0.4788  0.9086  0.9000  1.00  1  1  4.7E-04
19 : -4.34E-02  8.77E-04  0.000  0.5653  0.9169  0.9000  1.00  1  1  2.8E-04
20 : -4.34E-02  5.05E-04  0.000  0.5754  0.9034  0.9000  1.00  1  1  1.6E-04
21 : -4.34E-02  2.94E-04  0.000  0.5829  0.9136  0.9000  1.00  1  1  9.9E-05
22 : -4.34E-02  1.63E-04  0.015  0.5548  0.9000  0.0000  1.00  1  1  6.6E-05
23 : -4.33E-02  9.42E-05  0.000  0.5774  0.9053  0.9000  1.00  1  1  3.9E-05
24 : -4.33E-02  6.27E-05  0.000  0.6658  0.9148  0.9000  1.00  1  1  2.6E-05
25 : -4.33E-02  3.75E-05  0.000  0.5972  0.9187  0.9000  1.00  1  1  1.6E-05

```

```

26 : -4.33E-02 1.89E-05 0.000 0.5041 0.9117 0.9000 1.00 1 1 8.6E-06
27 : -4.33E-02 9.72E-06 0.000 0.5149 0.9050 0.9000 1.00 1 1 4.5E-06
28 : -4.33E-02 2.94E-06 0.000 0.3021 0.9194 0.9000 1.00 1 1 1.5E-06
29 : -4.33E-02 9.73E-07 0.000 0.3312 0.9189 0.9000 1.00 2 2 5.3E-07
30 : -4.33E-02 2.82E-07 0.000 0.2895 0.9063 0.9000 1.00 2 2 1.6E-07
31 : -4.33E-02 8.05E-08 0.000 0.2859 0.9049 0.9000 1.00 2 2 4.7E-08
32 : -4.33E-02 1.43E-08 0.000 0.1772 0.9059 0.9000 1.00 2 2 8.8E-09

iter seconds digits      c*x          b*y
32    49.4    6.8 -4.3334083581e-02 -4.3334090214e-02
|Ax-b| = 3.7e-09, [Ay-c]_+ = 1.1E-10, |x|= 1.0e+00, |y|= 2.6e+00

Detailed timing (sec)
Pre      IPM      Post
3.902E+00 4.576E+01 1.035E-02
Max-norms: ||b||=1, ||c|| = 3,
Cholesky |add|=0, |skip| = 0, ||L.L|| = 4.26267.
-----
Status: Solved
Optimal value (cvx_optval): -0.0433341
h = y(2:end);

```

Finally, we compute the filter response over the frequency vector defined and the result is shown on figure 3.2 which is very close to the filters obtain in [hua05_low_ligo].

```

----- Matlab -----
w = [w1 w2 w3 w4];
H = [exp(-j*kron(w'.*2*pi,[0:n-1]))]*h;

```

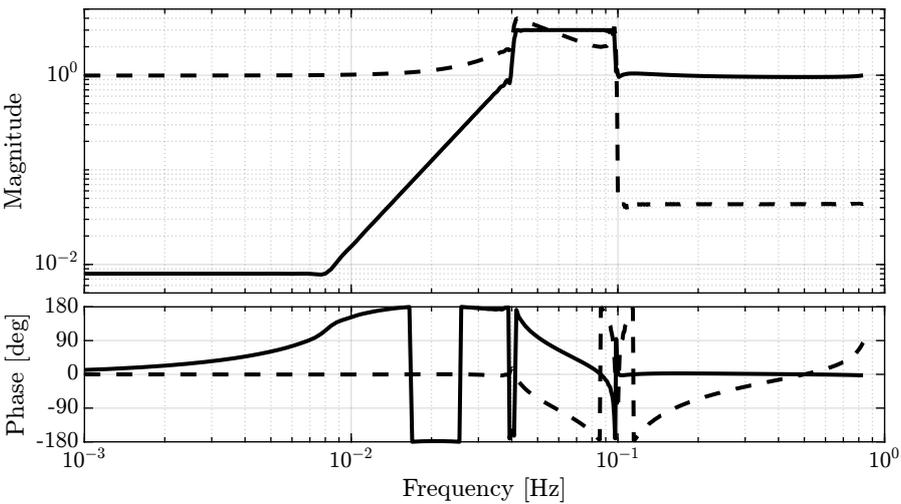


Figure 3.2: FIR Complementary filters obtain after convex optimization

3.3 Weights

We design weights that will be used for the \mathcal{H}_∞ synthesis of the complementary filters. These weights will determine the order of the obtained filters. Here are the requirements on the filters:

- reasonable order

- to be as close as possible to the specified upper bounds
- stable minimum phase

The bode plot of the weights is shown on figure 3.3.

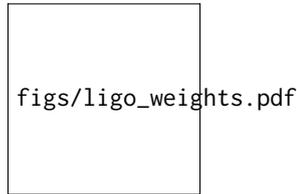


Figure 3.3: Weights for the \mathcal{H}_∞ synthesis

3.4 H-Infinity Synthesis

We define the generalized plant as shown on figure 1.1.

```
Matlab
P = [0 wL;
     wH -wH;
     1 0];
```

And we do the \mathcal{H}_∞ synthesis using the `hinfsyn` command.

```
Matlab
[H1, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
```

```
Results
[H1, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
Resetting value of Gamma min based on D_11, D_12, D_21 terms

Test bounds:      0.3276 < gamma <=      1.8063

gamma  hamx_eig  xinf_eig  hamy_eig  yinf_eig  nrho_xy  p/f
1.806  1.4e-02  -1.7e-16  3.6e-03  -4.8e-12  0.0000  p
1.067  1.3e-02  -4.2e-14  3.6e-03  -1.9e-12  0.0000  p
0.697  1.3e-02  -3.0e-01# 3.6e-03  -3.5e-11  0.0000  f
0.882  1.3e-02  -9.5e-01# 3.6e-03  -1.2e-34  0.0000  f
0.975  1.3e-02  -2.7e+00# 3.6e-03  -1.6e-12  0.0000  f
1.021  1.3e-02  -8.7e+00# 3.6e-03  -4.5e-16  0.0000  f
1.044  1.3e-02  -6.5e-14  3.6e-03  -3.0e-15  0.0000  p
1.032  1.3e-02  -1.8e+01# 3.6e-03  0.0e+00  0.0000  f
1.038  1.3e-02  -3.8e+01# 3.6e-03  0.0e+00  0.0000  f
1.041  1.3e-02  -8.3e+01# 3.6e-03  -2.9e-33  0.0000  f
1.042  1.3e-02  -1.9e+02# 3.6e-03  -3.4e-11  0.0000  f
1.043  1.3e-02  -5.3e+02# 3.6e-03  -7.5e-13  0.0000  f

Gamma value achieved:      1.0439
```

The high pass filter is defined as $H_H = 1 - H_L$.

```
Matlab
Hh = 1 - Hl;
```

The size of the filters is shown below.

```
Results
size(Hh), size(Hl)
State-space model with 1 outputs, 1 inputs, and 27 states.
State-space model with 1 outputs, 1 inputs, and 27 states.
```

The bode plot of the obtained filters as shown on figure 3.4.

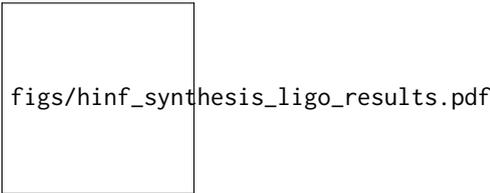


Figure 3.4: Obtained complementary filters using the \mathcal{H}_∞ synthesis

3.5 Compare FIR and H-Infinity Filters

Let's now compare the FIR filters designed in [hua05_low_ligo] and the one obtained with the \mathcal{H}_∞ synthesis on figure 3.5.

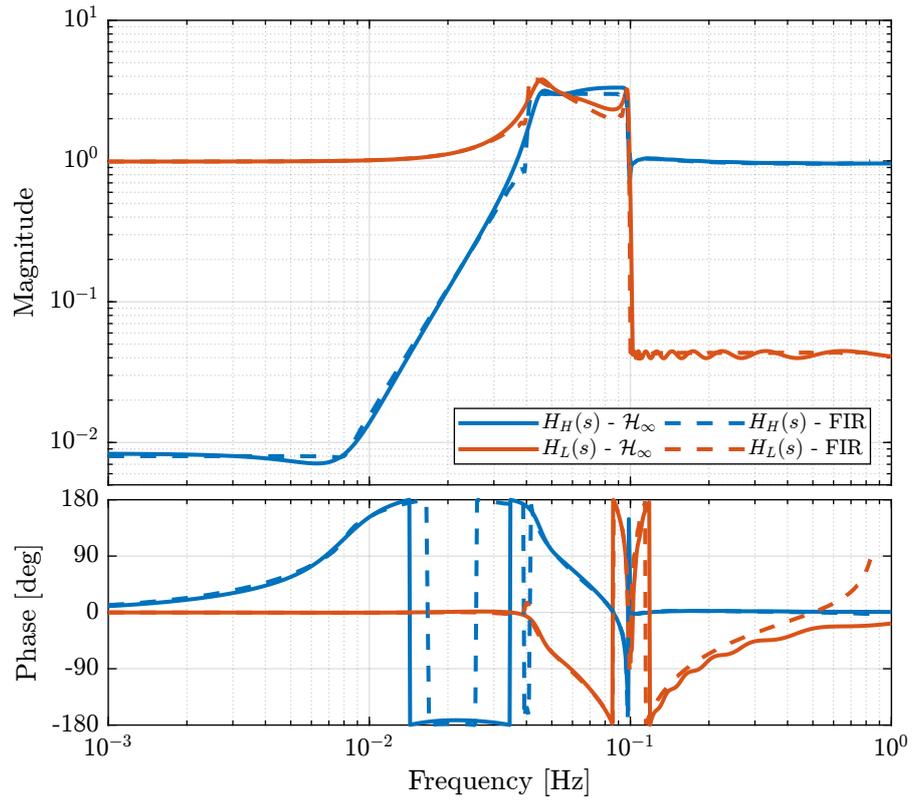


Figure 3.5: Comparison between the FIR filters developed for LIGO and the \mathcal{H}_∞ complementary filters

4 Alternative Synthesis

4.1 Two generalized plants

In order to synthesize the complementary filter using the proposed method, we can use two alternative generalized plant as shown in Figures 4.1 and 4.2.

$$P_1 = \begin{bmatrix} W_1 & -W_1 \\ 0 & W_2 \\ 1 & 0 \end{bmatrix} \quad (4.1)$$

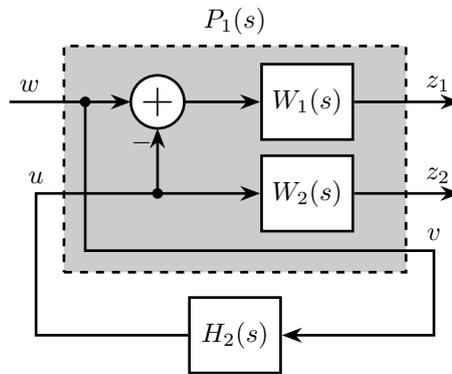


Figure 4.1: Complementary Filter Synthesis - Conf 1

$$P_2 = \begin{bmatrix} 0 & W_1 & 1 \\ W_2 & -W_1 & 0 \end{bmatrix} \quad (4.2)$$

Let's run the \mathcal{H}_∞ synthesis for both generalized plant using the same weights and see if the obtained filters are the same:

```

Matlab
n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
-> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 3; w0 = 2*pi*10; G0 = 1000; G1 = 0.1; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
-> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

```

Matlab
P1 = [W1 -W1;
      0 W2;
      1 0];

```

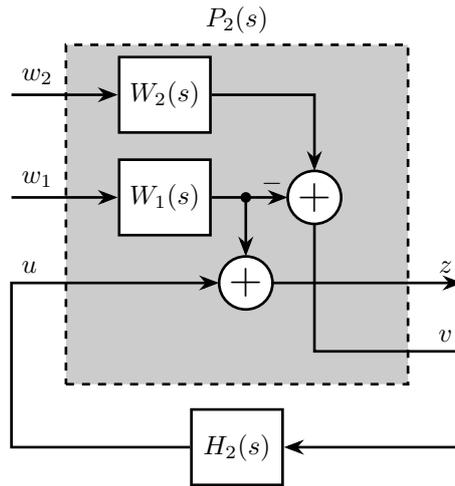


Figure 4.2: Complementary Filter Synthesis - Conf 2

```
Matlab
[H2, ~, gamma, ~] = hinfsyn(P1, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
```

```
Results
[H2, ~, gamma, ~] = hinfsyn(P1, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
```

Test bounds: 0.3263 <= gamma <= 1000

gamma	X>=0	Y>=0	rho(XY)<1	p/f
1.807e+01	1.4e-07	0.0e+00	1.185e-18	p
2.428e+00	1.5e-07	0.0e+00	1.285e-18	p
8.902e-01	-2.9e+02 #	-7.1e-17	5.168e-19	f
1.470e+00	1.5e-07	0.0e+00	1.462e-14	p
1.144e+00	1.5e-07	0.0e+00	1.260e-14	p
1.009e+00	1.5e-07	0.0e+00	4.120e-13	p
9.478e-01	-6.8e+02 #	-2.4e-17	1.449e-14	f
9.780e-01	-1.6e+03 #	-7.3e-17	6.791e-14	f
9.934e-01	-4.2e+03 #	-1.2e-16	3.524e-14	f
1.001e+00	-2.0e+04 #	-2.3e-17	5.717e-20	f
1.005e+00	1.5e-07	0.0e+00	8.953e-18	p
1.003e+00	-2.2e+05 #	-1.8e-17	3.225e-12	f
1.004e+00	1.5e-07	0.0e+00	2.445e-12	p
Limiting gains...				
1.004e+00	1.6e-07	0.0e+00	5.811e-18	p

Best performance (actual): 1.004

```
Matlab
P2 = [0 W1 1;
      W2 -W1 0];
```

```
Matlab
[H2b, ~, gamma, ~] = hinfsyn(P2, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
```

```
Results
[H2b, ~, gamma, ~] = hinfsyn(P2, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');
```

Test bounds: 0.3263 <= gamma <= 1000

gamma	X>=0	Y>=0	rho(XY)<1	p/f
1.807e+01	0.0e+00	1.4e-07	2.055e-16	p
2.428e+00	0.0e+00	1.4e-07	1.894e-18	p
8.902e-01	-2.1e-16	-2.7e+02 #	1.466e-16	f
1.470e+00	0.0e+00	1.4e-07	4.118e-16	p
1.144e+00	0.0e+00	1.5e-07	2.105e-18	p
1.009e+00	0.0e+00	1.5e-07	2.590e-13	p
9.478e-01	-9.5e-17	-6.3e+02 #	1.663e-19	f
9.780e-01	-1.1e-16	-1.5e+03 #	1.546e-14	f
9.934e-01	-2.8e-17	-4.0e+03 #	3.934e-14	f
1.001e+00	-3.1e-17	-1.9e+04 #	1.191e-19	f
1.005e+00	0.0e+00	1.5e-07	1.443e-12	p
1.003e+00	-8.3e-17	-2.1e+05 #	8.807e-13	f
1.004e+00	0.0e+00	1.5e-07	1.459e-15	p
Limiting gains...				
1.004e+00	0.0e+00	1.5e-07	9.086e-19	p

Best performance (actual): 1.004

And indeed, we can see that the exact same filters are obtained (Figure 4.3).

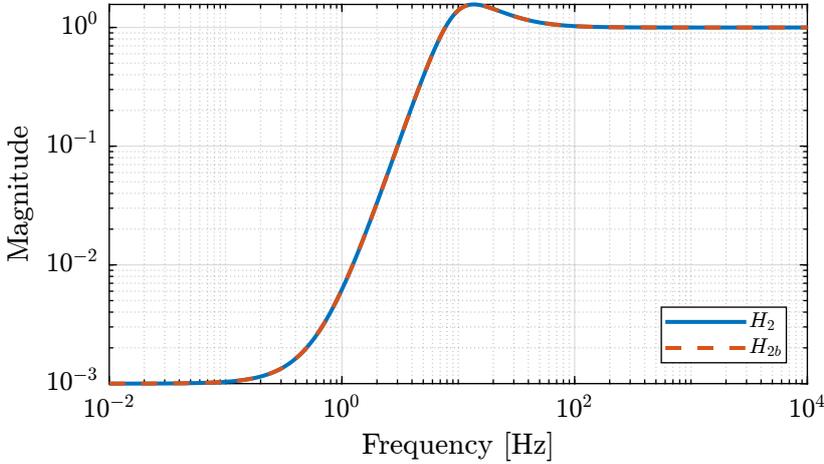


Figure 4.3: Comparison of $H_2(s)$ when using $P_1(s)$ or $P_2(s)$

4.2 Shaping the Low pass filter or the high pass filter?

Let's see if there is a difference by explicitly shaping $H_1(s)$ or $H_2(s)$.

```

Matlab
n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
-> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 3; w0 = 2*pi*10; G0 = 1000; G1 = 0.1; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
-> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

Let's first synthesize $H_1(s)$:

```

Matlab
P1 = [W2 -W2;
      0  W1;
      1  0];

```

```

Matlab
[H1, ~, gamma, ~] = hinfsyn(P1, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

```

```

Results
[H1, ~, gamma, ~] = hinfsyn(P1, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

Test bounds: 0.3263 <= gamma <= 1.712

   gamma      X>=0      Y>=0      rho(XY)<1      p/f
7.476e-01    -2.5e+01 #    -8.3e-18    4.938e-20    f
1.131e+00     1.9e-07     0.0e+00    1.566e-16    p
9.197e-01    -1.4e+02 #    -7.9e-17    4.241e-17    f
1.020e+00     1.9e-07     0.0e+00    2.095e-16    p
9.686e-01    -3.8e+02 #    -7.0e-17    1.463e-23    f
9.940e-01    -1.5e+03 #    -1.3e-17    3.168e-19    f
1.007e+00     1.9e-07     0.0e+00    1.696e-15    p
1.000e+00    -4.8e+03 #    -7.1e-18    7.203e-20    f
1.004e+00     1.9e-07     0.0e+00    1.491e-14    p
1.002e+00    -1.1e+04 #    -2.6e-16    2.579e-14    f
1.003e+00    -2.8e+04 #    -6.0e-18    8.558e-20    f
Limiting gains...
1.004e+00     2.0e-07     0.0e+00    5.647e-18    p
1.004e+00     1.0e-06     0.0e+00    5.648e-18    p

Best performance (actual): 1.004

```

And now $H_2(s)$:

```

Matlab
P2 = [W1 -W1;
      0  W2;
      1  0];

```

```

Matlab
[H2, ~, gamma, ~] = hinfsyn(P2, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

```

```

Results
[H2b, ~, gamma, ~] = hinfsyn(P2, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'ric', 'DISPLAY', 'on');

Test bounds: 0.3263 <= gamma <= 1000

   gamma      X>=0      Y>=0      rho(XY)<1      p/f
1.807e+01    1.4e-07     0.0e+00    1.185e-18    p
2.428e+00    1.5e-07     0.0e+00    1.285e-18    p
8.902e-01    -2.9e+02 #    -7.1e-17    5.168e-19    f
1.470e+00    1.5e-07     0.0e+00    1.462e-14    p
1.144e+00    1.5e-07     0.0e+00    1.260e-14    p
1.009e+00    1.5e-07     0.0e+00    4.120e-13    p
9.478e-01    -6.8e+02 #    -2.4e-17    1.449e-14    f
9.780e-01    -1.6e+03 #    -7.3e-17    6.791e-14    f
9.934e-01    -4.2e+03 #    -1.2e-16    3.524e-14    f
1.001e+00    -2.0e+04 #    -2.3e-17    5.717e-20    f
1.005e+00    1.5e-07     0.0e+00    8.953e-18    p
1.003e+00    -2.2e+05 #    -1.8e-17    3.225e-12    f
1.004e+00    1.5e-07     0.0e+00    2.445e-12    p
Limiting gains...
1.004e+00    1.6e-07     0.0e+00    5.811e-18    p

```

And compare $H_1(s)$ with $1 - H_2(s)$ and $H_2(s)$ with $1 - H_1(s)$ in Figure 4.4.

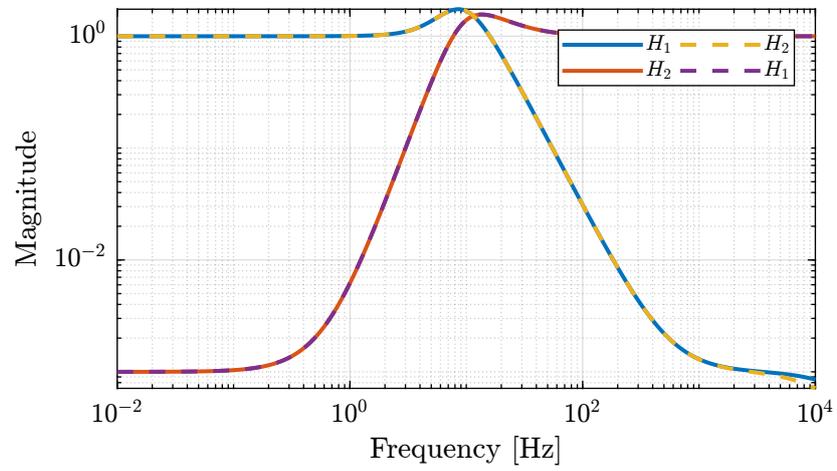


Figure 4.4: Comparison of $H_1(s)$ with $1 - H_2(s)$, and $H_2(s)$ with $1 - H_1(s)$

5 Impose a positive slope at DC or a negative slope at infinite frequency

5.1 Manually shift zeros to the origin after synthesis

Suppose we want $H_2(s)$ to be an high pass filter with a slope of +2 at low frequency (from 0Hz).

We cannot impose that using the weight $W_2(s)$ as it would be improper.

However, we may manually shift 2 of the low frequency zeros to the origin.

```
Matlab
n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
→ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 3; w0 = 2*pi*10; G0 = 1e4; G1 = 0.1; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
→ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;
```

```
Matlab
P = [W1 -W1;
     0  W2;
     1  0];
```

And we do the \mathcal{H}_∞ synthesis using the `hinfsyn` command:

```
Matlab
[H2, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'lmi', 'DISPLAY', 'on');
```

```
Matlab
[z,p,k] = zpndata(H2)
```

Looking at the zeros, we see two low frequency complex conjugate zeros.

```
Results
z{1}
ans =
    -4690930.24283199 + 0i
    -163.420524657426 + 0i
    -0.8531922261081498 + 0.713416012479897i
    -0.8531922261081498 - 0.713416012479897i
    -3.15812268762265 + 0i
```

We manually put these zeros at the origin:

```
Matlab  
z{1}([3,4]) = 0;
```

And we create a modified filter $H_{2z}(s)$:

```
Matlab  
H2z = zp(z,p,k);
```

And as usual, $H_{1z}(s)$ is defined as the complementary of $H_{2z}(s)$:

```
Matlab  
H1z = 1 - H2z;
```

The bode plots of $H_1(s)$, $H_2(s)$, $H_{1z}(s)$ and $H_{2z}(s)$ are shown in Figure 5.1. And we see that $H_{1z}(s)$ is slightly modified when setting the zeros at the origin for $H_{2z}(s)$.

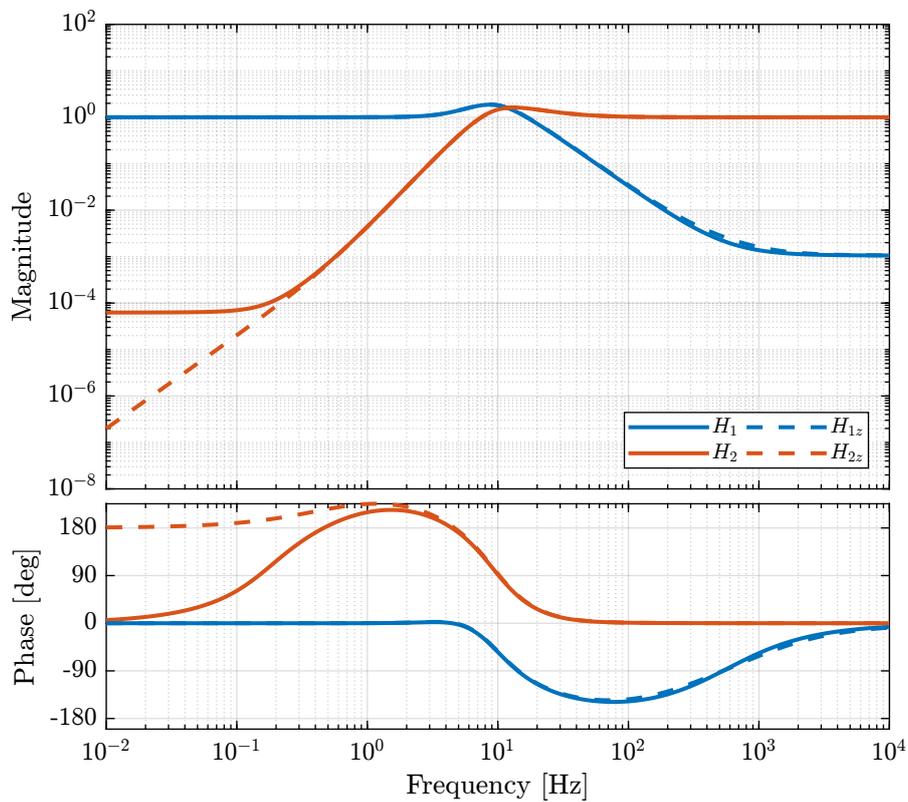


Figure 5.1: Bode plots of $H_1(s)$, $H_2(s)$, $H_{1z}(s)$ and $H_{2z}(s)$

5.2 Imposing a positive slope at DC during the synthesis phase

Suppose we want to synthesize $H_2(s)$ such that it has a slope of +2 from DC. We can include this “feature” in the generalized plant as shown in Figure 5.2.

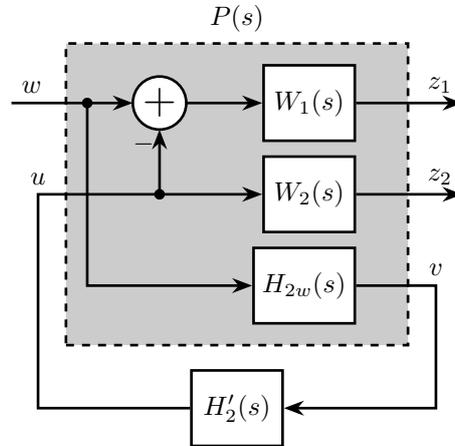


Figure 5.2: Generalized plant with included wanted feature represented by $H_{2w}(s)$

After synthesis, the obtained filter will be:

$$H_2(s) = H_2'(s)H_{2w}(s) \quad (5.1)$$

and therefore the “feature” will be included in the filter.

For $H_1(s)$ nothing is changed: $H_1(s) = 1 - H_2(s)$.

The weighting functions are defined as usual:

```

Matlab
n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
  -> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 3; w0 = 2*pi*10; G0 = 1e4; G1 = 0.1; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
  -> (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

The wanted feature here is a +2 slope at low frequency. For that, we use an high pass filter with a slope of +2 at low frequency.

```

Matlab
w0 = 2*pi*50;
H2w = (s/w0/(s/w0+1))^2;

```

We define the generalized plant as shown in Figure 5.2.

```

Matlab
P = [W1 -W1;
     0 W2;
     H2w 0];

```

And we do the \mathcal{H}_∞ synthesis using the `hinfsyn` command.

```
[H2p, ~, gamma, ~] = hinfsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'lmi', 'DISPLAY', 'on');
```

Finally, we define $H_2(s)$ as the product of the synthesized filter and the wanted “feature”:

```
H2 = H2p*H2w;
```

And we define $H_1(s)$ to be the complementary of $H_2(s)$:

```
H1 = 1 - H2;
```

The obtained complementary filters are shown in Figure 5.3.

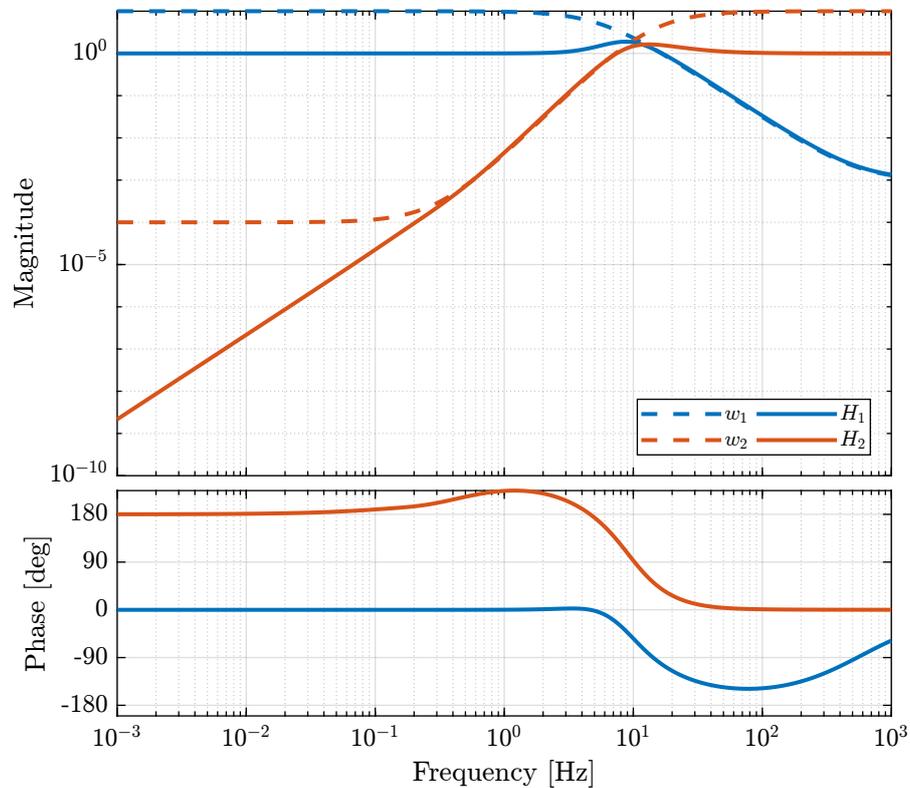


Figure 5.3: Obtained complementary filters

5.3 Imposing a negative slope at infinity frequency during the synthesis phase

Let's suppose we now want to shape a low pass filter that as a negative slope until infinite frequency.

The used technique is the same as in the previous section, and the generalized plant is shown in Figure 5.2.

The weights are defined as usual.

```

Matlab
n = 3; w0 = 2*pi*10; G0 = 1000; G1 = 0.1; Gc = 1/2;
W1 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
  ↪ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

n = 2; w0 = 2*pi*11; G0 = 1/10; G1 = 1000; Gc = 1/2;
W2 = (((1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s +
  ↪ (G0/Gc)^(1/n))/((1/G1)^(1/n)*(1/w0)*sqrt((1-(G0/Gc)^(2/n))/(1-(Gc/G1)^(2/n))))*s + (1/Gc)^(1/n))^n;

```

This time, the feature is a low pass filter with a slope of -2 at high frequency.

```

Matlab
H2w = 1/(s/(2*pi*10) + 1)^2;

```

The generalized plant is defined:

```

Matlab
P = [W1 -W1;
     0 W2;
     H2w 0];

```

And we do the \mathcal{H}_∞ synthesis using the `hifsyn` command.

```

Matlab
[H2p, ~, gamma, ~] = hifsyn(P, 1, 1, 'TOLGAM', 0.001, 'METHOD', 'lmi', 'DISPLAY', 'on');

```

The feature is added to the synthesized filter:

```

Matlab
H2 = H2p*H2w;

```

And $H_1(s)$ is defined as follows:

```

Matlab
H1 = 1 - H2;

```

The obtained complementary filters are shown in Figure 5.4.

ref

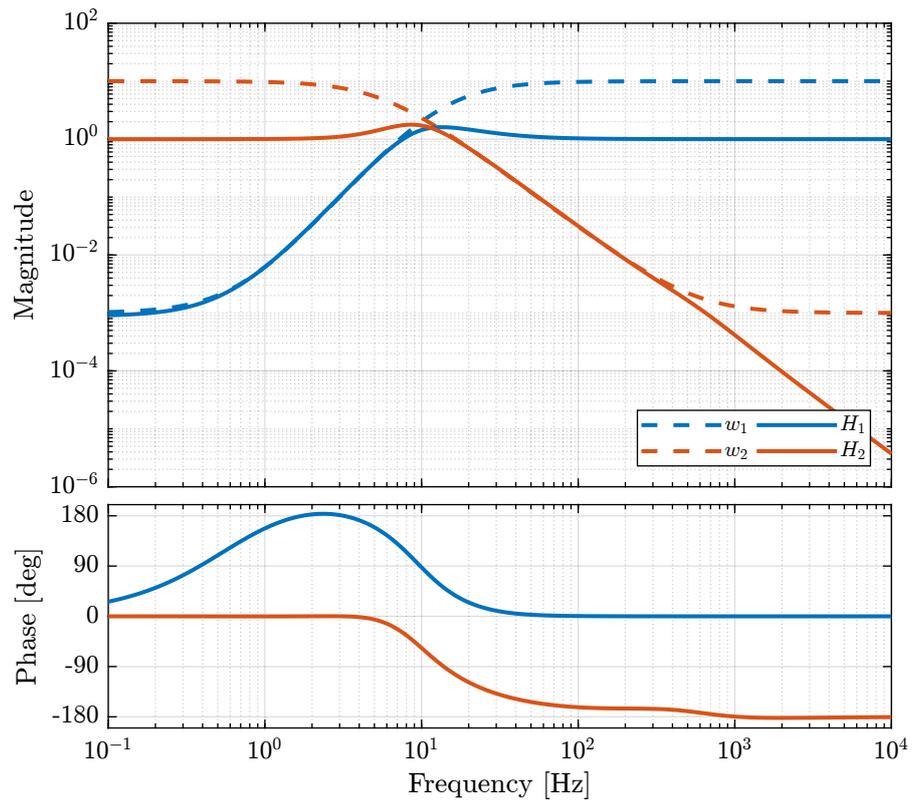


Figure 5.4: Obtained complementary filters