

```
% Auto Regressive Random Process computer exercise 4.1
%M.Tognolini 17.03.2022
% MSE statistical signal processing Lausanne
% rev 1.1 28.03.2022 Correction of the error computing in FIR filter
% design line 140 we must give all the sample to compute the xhat
% signal and the error over all the samples ! (Thanks M Mieville)

clear;

clc
```

Padé approximation

M.Tognolini HEIG-VD 2022 MA-StatDig

Example 4.3.2

Test Padé approximation with the signal $x(n)$ defined below.

Test with AR(2) model, AM(2) and ARMA(1,1).

```
N = 12;
n = 0:N-1;
xn = [1, 3* (0.5).^n(2:end)]
dn = [1,zeros(1,N-1)];
```

```
N = 12;
n = 0:N-1;
An = 0.2
```

```
An = 0.2000
```

```
rng default
x = [1, 3* (0.5).^n(2:end)] % noiseless
```

```
x = 1x12
    1.0000    1.5000    0.7500    0.3750    0.1875    0.0938    0.0469    0.0234 ...
```

```
x = x(:)
```

```
x = 12x1
    1.0000
    1.5000
    0.7500
    0.3750
    0.1875
    0.0938
    0.0469
    0.0234
    0.0117
    0.0059
```

⋮

```
xn = [1, 3* (0.5).^n(2:end)]+ An*randn(1,N) % noisy signal
```

```
xn = 1×12
    1.1075    1.8668    0.2982    0.5474    0.2513   -0.1678   -0.0398    0.0920 ...
```

```
dn = [1,zeros(1,N-1)];
```

a) Test with AR(2) : $p=2$ and $q = 0$

```
p=2
q=0
xn = xn(:)
```

```
p=2
```

```
p = 2
```

```
q=0
```

```
q = 0
```

```
xn = xn(:)
```

```
xn = 12×1
    1.1075
    1.8668
    0.2982
    0.5474
    0.2513
   -0.1678
   -0.0398
    0.0920
    0.7274
    0.5597
     ⋮
```

Compute the X_q matrix using the Matlab $X = \text{convmtx}(xn, \dots)$ function and chose the right index to extract the X_q as in theory eq (4.13)

```
X = convmtx(xn,p+1)
```

```
X = convmtx(xn,p+1)
```

```
X = 14×3
    1.1075         0         0
    1.8668    1.1075         0
    0.2982    1.8668    1.1075
    0.5474    0.2982    1.8668
    0.2513    0.5474    0.2982
   -0.1678    0.2513    0.5474
   -0.0398   -0.1678    0.2513
    0.0920   -0.0398   -0.1678
```

```

0.7274    0.0920   -0.0398
0.5597    0.7274    0.0920
⋮

```

```
Xq = X(q+2:q+p+1, 2:p+1)
```

```

Xq = 2x2
    1.1075    0
    1.8668    1.1075

```

Compute the vector x_{q+1} as well:

```
xq_1 = xn(q+2:q+p+1)
```

```

xq_1 = 2x1
    1.8668
    0.2982

```

```

%Xq(1,2) = 1.0 for test only
%Xq(2,1) = 1.0

```

And now compute the ap coefficients :

```

if (det(Xq)) %if det = 0 singular matrix
    ap = inv(Xq)*(-xq_1); % pinv is used in case
    ap = [1;ap]
else
    disp('Xq is a singular matrix !')
end

```

```

ap = 3x1
    1.0000
   -1.6855
    2.5717

```

Compute the numerator coefficients here is obvious that $bq(0) = xn(0)$

```
X0 = X(1: q+1, 1:p+1 )
```

```

X0 = 1x3
    1.1075    0    0

```

```
bq = X0*ap
```

```
bq = 1.1075
```

Now compute the impulse response that is $\hat{x}(n) = h(n)$

```
xhat = filter(bq',ap',dn')
```

```

xhat = 12x1
    1.1075
    1.8668
    0.2982
   -4.2982
   -8.0116

```

```

-2.4501
16.4740
34.0684
15.0566
-62.2361
⋮

```

Compute the error of the approximation and the least square error:

```

ep = x - xhat;
Els = norm(ep)

```

```

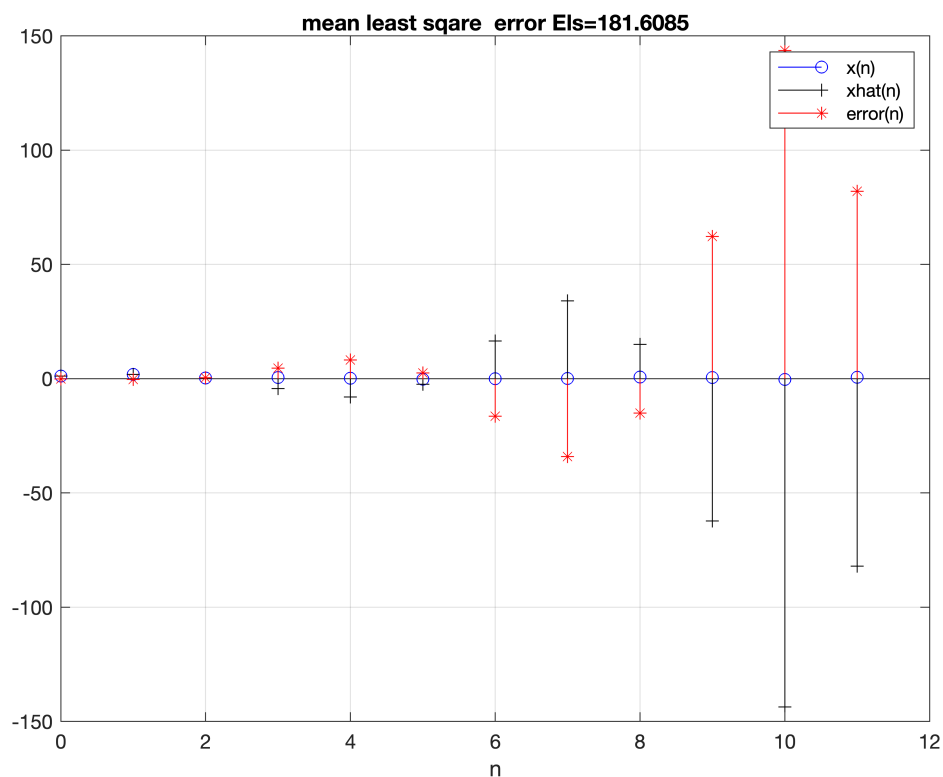
Els = 181.6085

```

```

figure(1)
stem(n,xn,'ob')
hold on
stem(n,xhat,'+k')
stem(n,ep,'*r')
hold off
xlabel('n')
legend('x(n)', 'xhat(n)', 'error(n)')
grid;
title(['mean least sqare error Els=', num2str(Els)])

```



b) Test with MA(2) : $p=0$ and $q = 2$

In this case the solution for $bq(n)$ is obvious because $h(n) = bq(n)$ so $bq(n) = x(n)$:

```
q= 2
```

```
q = 2
```

```
p= 0
```

```
p = 0
```

```
bqma = xn(1:q+1)
```

```
bqma = 3x1
    1.1075
    1.8668
    0.2982
```

```
apma = 1
```

```
apma = 1
```

```
xhatma = filter(bqma',apma',dn')
```

```
xhatma = 12x1
    1.1075
    1.8668
    0.2982
         0
         0
         0
         0
         0
         0
         0
         0
         0
         0
         0
         0
```

```
epma = x - xhatma;
Elsma = norm(epma)
```

```
Elsma = 0.7333
```

c) Test with ARMA(1,1) : $p=1$ and $q = 1$

In this case we need to solve with 2 step algorithm first we compute the ap coeff as in point a)

```
q= 1
```

```
q = 1
```

```
p= 1
```

```
p = 1
```

```
Xq = X(q+2:q+p+1, 2:p+1)
```

```
Xq = 1.8668
```

Compute the vector x_{q+1} as well:

```
xq_1 = xn(q+2:q+p+1)
```

```
xq_1 = 0.2982
```

```
if (det(Xq)) %if det = 0 singular matrix
    ap_arma = inv(Xq)*(-xq_1); % pinv is used in case
    ap_arma = [1;ap_arma]
else
    disp('Xq is a singular matrix !')
end
```

```
ap_arma = 2x1
    1.0000
   -0.1598
```

Next we apply the step 2 and compute the bq coefficients:

```
X0 = X(1: q+1, 1:p+1 )
```

```
X0 = 2x2
    1.1075      0
    1.8668    1.1075
```

```
bq_arma = X0*ap_arma
```

```
bq_arma = 2x1
    1.1075
    1.6898
```

The transfer function $H(z)$ is:

```
printsys(bq_arma',ap_arma','z')
```

num/den =

$$\frac{1.1075 z + 1.6898}{z - 0.15976}$$

```
xhat_arma = filter(bq_arma',ap_arma',dn')
```

```
xhat_arma = 12x1
    1.1075
    1.8668
    0.2982
    0.0476
    0.0076
    0.0012
    0.0002
    0.0000
    0.0000
    0.0000
```

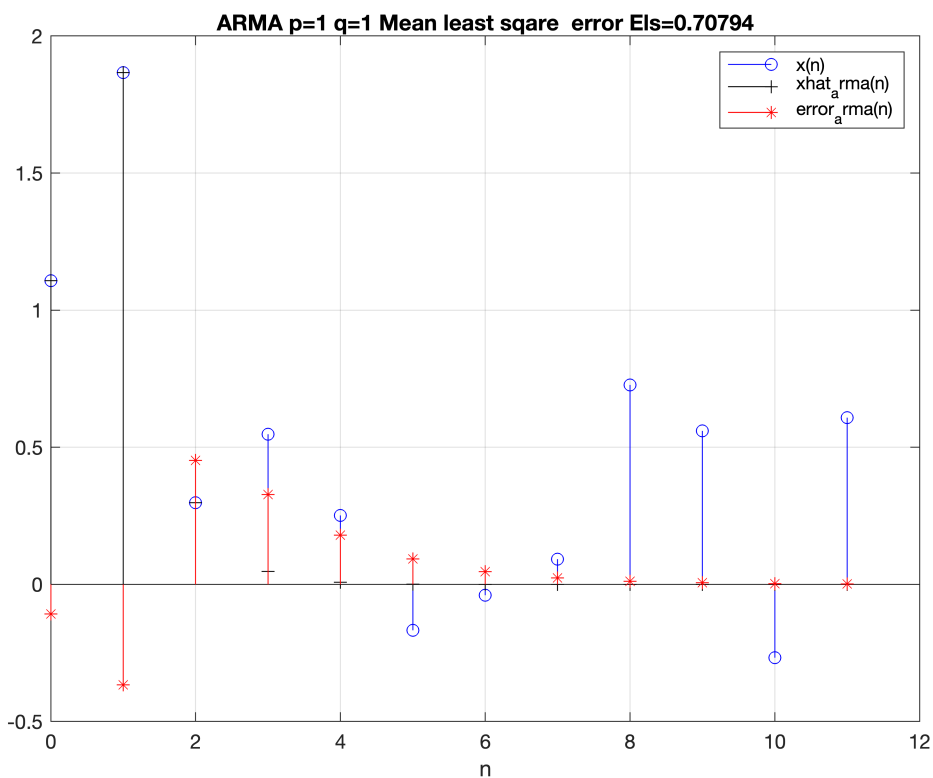
0.0000

⋮

```
ep_arma = x - xhat_arma;  
Els_arma = norm(ep_arma)
```

Els_arma = 0.7079

```
figure(2)  
stem(n,xn,'ob')  
hold on  
stem(n,xhat_arma,'+k')  
stem(n,ep_arma,'*r')  
hold off  
xlabel('n')  
legend('x(n)', 'xhat_arma(n)', 'error_arma(n)')  
grid;  
title(['ARMA p=', num2str(p), ' q=', num2str(q), ' Mean least square error Els=', num2str(Els_arma)])
```



Build a function to compute the coefficients a_p and b_q by Padé approximation

The function header would be :

help **pade**

pade compute the pade approximation of the coeff of ARMA system with p poles and q zeros usage: [ap,bq,Els,xhat] = pade(x,p,q)
Els is optional and is the L2 norm of the approx error x-xhat
xhat is optional and it is the approximated signal.
(c) M.Tognolini HEIG-VD 2022 r1.0

Other functions named pade

test it with the previous example.

p=2

p = 2

q=0

q = 0

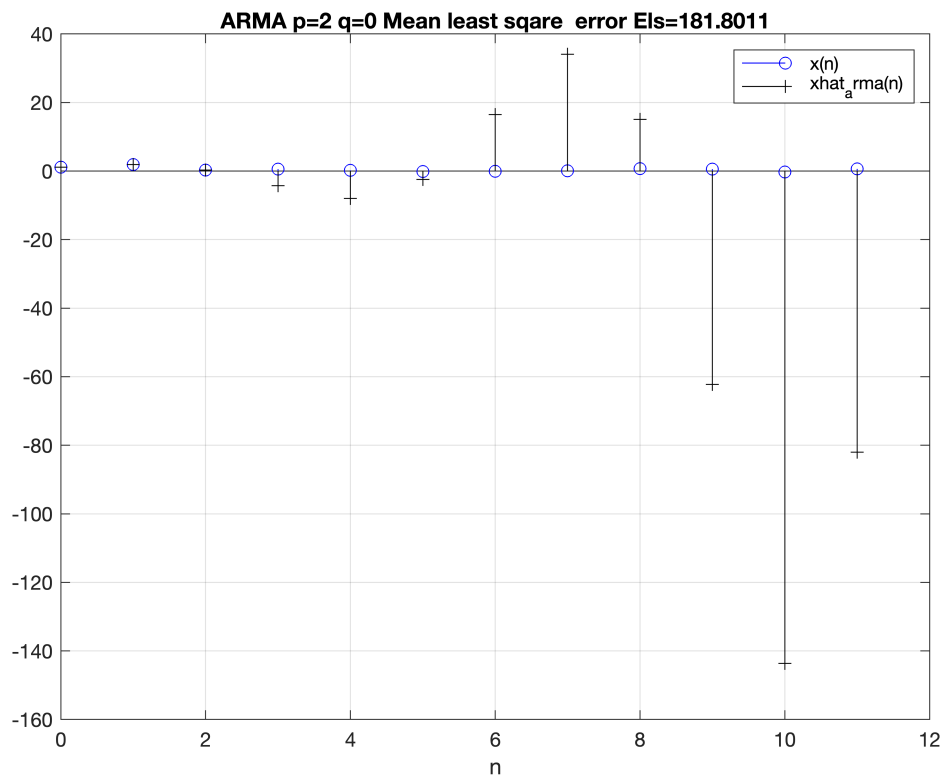
```
[ap_test,bq_test,Els_test,xhat_test] = pade(xn,p,q)
```

Calculating the squared error of the approximation
and the approximation signal xhat

```
ap_test = 3x1
    1.0000
   -1.6855
    2.5717
bq_test = 1.1075
Els_test = 181.8011
xhat_test = 12x1
    1.1075
    1.8668
    0.2982
   -4.2982
   -8.0116
   -2.4501
   16.4740
   34.0684
   15.0566
  -62.2361
      ⋮
      ⋮
```

```
figure(3)
stem(n,xn,'ob')
hold on
stem(n,xhat_test,'+k')

hold off
xlabel('n')
legend('x(n)','xhat_arma(n)')
grid;
title(['ARMA p=',num2str(p),' q=',num2str(q),' Mean least square error Els=',num2str(Els)])
```

Example 4.3.3 : Singular exemple

Digital filter approximation

```
x2 = [1,4,2,1,3]
```

```
x2 = 1x5
      1   4   2   1   3
```

```
p=2
```

```
p = 2
```

```
q=2
```

```
q = 2
```

Compute the pade approximation with the pade function:

```
[ap_2,bq_2] = pade(x2,p,q)
```

```
Error using pade
Xq is a singular matrix !
```

The Xq matrix is singular in this case !

That means the assumption that $a_p(0) = 1$ is incorrect for this model so we put it to 0.

Example 4.3.4: Filter design using the Padé approximation

The problem of the filter approximation is to find the impulse response $h(n)$ of the filter from a frequency constraints.

For this example to be as simple as possible we want to synthesize a filter with a frequency response $|H(jf)| = 1$ for $f < F_p$ and 0 for $F_p < f < 0.5$,

additionally we want a linear phase response then $H(jf) = e^{-jn_d 2\pi f}$ for $f < F_p$ and 0 for $F_p < f < 0.5$, the constant n_d is the time delay of the filter to assure the causality of the filter.

$$H(jf) = \begin{cases} e^{jn_d 2\pi f} & ; |f| < 0.25 \\ 0 & ; \text{otherwise} \end{cases}$$

Since the frequency response is a rectangular function the inverse transform is a sinc function of the variable n , the impulse response is :

$$h(n) = \frac{\sin[(n - n_d)\pi/2]}{(n - n_d)\pi}$$

With the function sinc we write it as: $h_n = 1/2 * \text{sinc}((n - n_d) * 2 * F_p)$

```
N = 100
Fp = 0.25
n = 0:N-1;
nd = 5

hn = 1/2 * sinc((n-nd)*2*Fp)

H = fftshift(fft(hn));
f = -0.5:1/N:0.5-0.5/N;
```

```
figure;
subplot(2,1,1)
stem(n,hn)
axis([0,20,-0.2,0.6])
grid;
xlabel('n')
legend('h(n)')
subplot(2,1,2)
plot(f,20*log10(abs(H)))
axis([0,0.5,-40,1])
grid;
xlabel('f [-]')
legend('|H(f)|')
```

N = 100

```
N = 100
```

```
Fp = 0.25
```

```
Fp = 0.2500
```

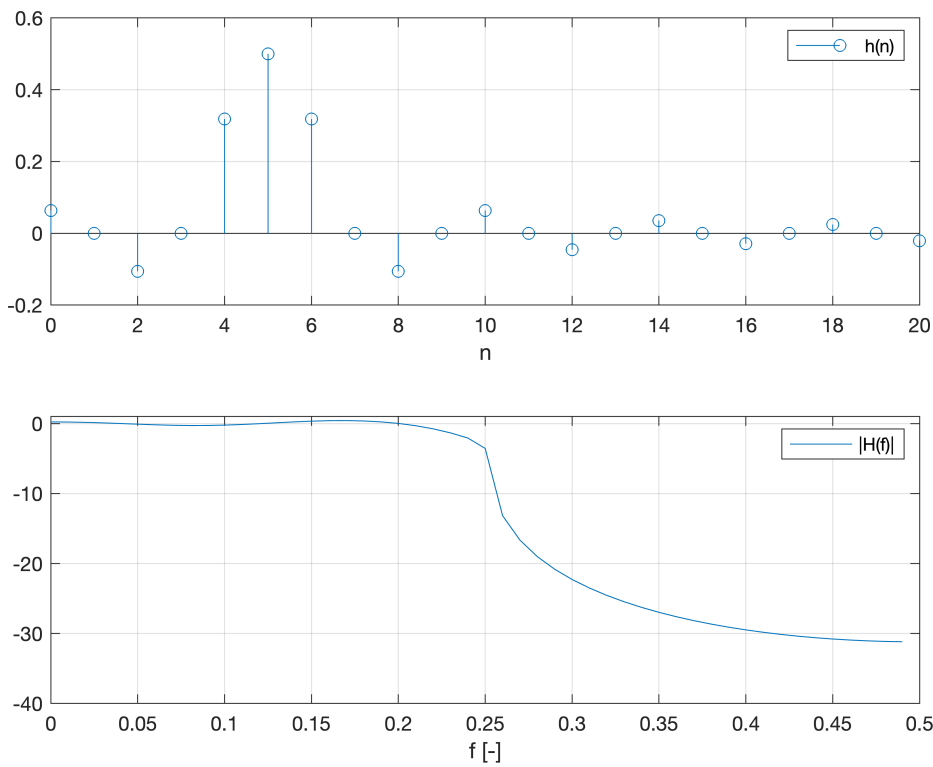
```
n= 0:N-1;  
nd = 5
```

```
nd = 5
```

```
hn = 1/2 * sinc((n-nd)*2*Fp)
```

```
hn = 1×100  
    0.0637    -0.0000    -0.1061     0.0000     0.3183     0.5000     0.3183     0.0000 ...
```

```
H = fftshift(fft(hn));  
f = -0.5:1/N:0.5-0.5/N;  
figure;  
subplot(2,1,1)  
stem(n,hn)  
axis([0,20,-0.2,0.6])  
grid;  
xlabel('n')  
legend('h(n)')  
subplot(2,1,2)  
plot(f,20*log10(abs(H)))  
axis([0,0.5,-40,1])  
grid;  
xlabel('f [-]')  
legend('|H(f)|')
```



If we design a FIR filter with $q=10$ and $p=0$ (MA) the Padé approximation wil give the 11 first values of the $h(n)$. In this case $bq(n) = h(n)$ for $n = 0 \dots 10$.

$p=0$

$p = 0$

$q=10$

$q = 10$

$bqFIR = hn(1:p+q+1)$

```
bqFIR = 1x11
0.0637 -0.0000 -0.1061 0.0000 0.3183 0.5000 0.3183 0.0000 ...
```

$apFIR = [1, zeros(1,10)]$

```
apFIR = 1x11
1 0 0 0 0 0 0 0 0 0
```

$Hfir = tf(bqFIR,apFIR,1)$

$Hfir =$

$$0.06366 z^{10} - 1.949e-17 z^9 - 0.1061 z^8 + 1.949e-17 z^7 + 0.3183 z^6 + 0.5 z^5 + 0.3183 z^4 + 1.949e-17 z^3 - 0.1061 z^2 - 1.949e-17 z - 0.06366$$

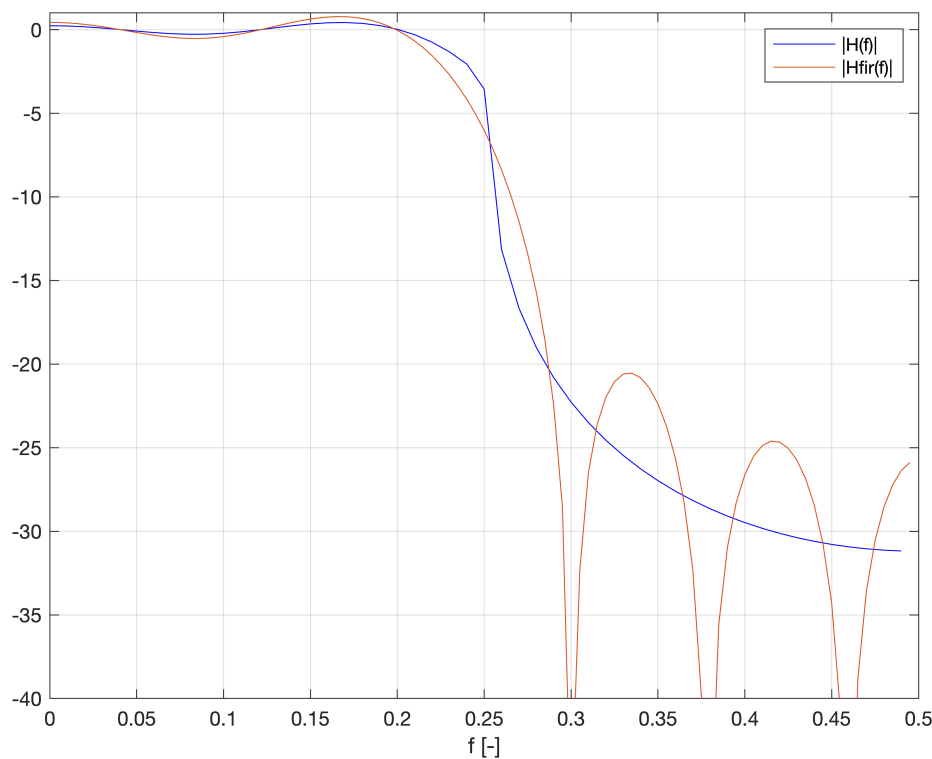
$$z^{10}$$

Sample time: 1 seconds
Discrete-time transfer function.

We could plot the frequency response of this filter with the function `freqz()`:

```
[HFIR, f2]=freqz(bqFIR,apFIR ,N,1);

figure;
plot(f,20*log10(abs(H)), 'b', f2,20*log10(abs(HFIR)))
axis([0,0.5,-40,1])
grid;
xlabel('f [-]')
legend('|H(f)|', '|Hfir(f)|')
```



We want to find the transfer function of an ARMA system with $p=6$ and $q=6$

```
p=6
```

```
p = 6
```

```
q=6
```

```
q = 6
```

```
[ap_IIR,bq_IIR,Els_IIR] = pade(hn,p,q)
```

Calculating the squared error of the approximation

```
ap_IIR = 7×1
```

```
1.0000
```

```
-2.5171
```

```
3.9055
```

```
-4.0872
```

```
2.9647
```

```
-1.4088
```

```
0.3500
```

```
bq_IIR = 7×1
```

```
0.0637
```

```
-0.1602
```

```
0.1425
```

```
0.0069
```

```
0.0927
```

```
0.0428
```

```
0.0106
```

```
Els_IIR = 0.1365
```

```
Hiir = tf(bq_IIR',ap_IIR',1)
```

```
Hiir =
```

$$\frac{0.06366 z^6 - 0.1602 z^5 + 0.1425 z^4 + 0.006869 z^3 + 0.09267 z^2 + 0.04277 z + 0.01064}{z^6 - 2.517 z^5 + 3.905 z^4 - 4.087 z^3 + 2.965 z^2 - 1.409 z + 0.35}$$

Sample time: 1 seconds

Discrete-time transfer function.

As before we will plot the frequency response compared to the desired frequency response $H(f)$:

```
[HIIR,f2]=freqz(bq_IIR,ap_IIR ,N,1);
```

```
figure;
```

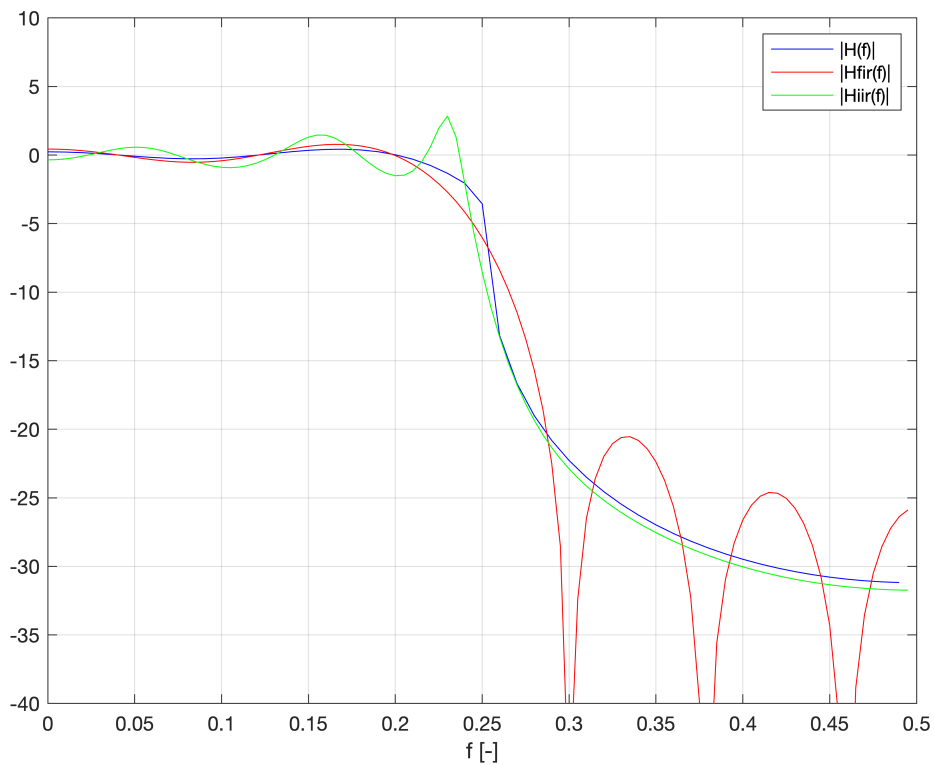
```
plot(f,20*log10(abs(H)), 'b', f2,20*log10(abs(HFIR)), 'r', f2,20*log10(abs(HIIR)), 'g')
```

```
axis([0,0.5,-40,10])
```

```
grid;
```

```
xlabel('f [-]')
```

```
legend('|H(f)|', '|Hfir(f)|', '|Hiir(f)|')
```



Compute the impulse response of this filter over the N samples and compare to the original impulse response $h(n)$:

```
dn=[1,zeros(1,N-1)];
hiir = filter(bq_IIR,ap_IIR,dn)
```

```
hiir = 1×100
    0.0637   -0.0000   -0.1061         0    0.3183    0.5000    0.3183   -0.0000 ...
```

```
figure
stem(n,hn,'ob')
hold on
stem(n,hiir,'*r')
hold off
axis([0,20,-0.2,0.6])
grid;
xlabel('n')
legend('h(n)', 'hFIR(n)')
```

