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Signal Processing

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## An iterative method for the design of variable fractional-order FIR differintegrators

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where

$$h_e\left(\frac{N}{2}+n, m\right)=\frac{1}{2}\left[h\left(\frac{N}{2}+n, m\right)+h\left(\frac{N}{2}-n, m\right)\right], \\ -\frac{N}{2}\leq n\leq\frac{N}{2}, 0\leq m\leq M \quad (8a)$$

and

$$h_o\left(\frac{N}{2}+n, m\right)=\frac{1}{2}\left[h\left(\frac{N}{2}+n, m\right)-h\left(\frac{N}{2}-n, m\right)\right], \\ -\frac{N}{2}\leq n\leq\frac{N}{2}, 0\leq m\leq M. \quad (8b)$$

So, the frequency response of the designed filter can be formulated into

$$H(e^{j\omega}, p)=e^{-j(N/2)\omega}\left[\sum_{n=0}^{N/2}\sum_{m=0}^Ma(n, m)p^m\cos(n\omega)\right. \\ \left.+j\sum_{n=1}^{N/2}\sum_{m=0}^Mb(n, m)p^m\sin(n\omega)\right] \\ =e^{-j(N/2)\omega}\hat{H}(\omega, p), \quad (9)$$

where

$$a(n, m)=\begin{cases} h_e\left(\frac{N}{2}, m\right), & n=0, 0\leq m\leq M, \\ 2h_e\left(\frac{N}{2}-n, m\right), & 1\leq n\leq\frac{N}{2}, 0\leq m\leq M, \end{cases} \quad (10a)$$

$$b(n, m)=2h_o\left(\frac{N}{2}-n, m\right), \quad 1\leq n\leq\frac{N}{2}, 0\leq m\leq M \quad (10b)$$

and

$$\hat{H}(\omega, p)=\sum_{n=0}^{N/2}\sum_{m=0}^Ma(n, m)p^m\cos(n\omega) \\ +j\sum_{n=1}^{N/2}\sum_{m=0}^Mb(n, m)p^m\sin(n\omega). \quad (11)$$

Obviously,  $I=N/2$  in (1) and (3).Let **A** and **B** be  $(N/2+1)\times(M+1)$  and  $N/2\times(M+1)$  matrices defined by

$$\mathbf{A}=\begin{bmatrix} a(n, m), 0\leq n\leq\frac{N}{2}, 0\leq m\leq M \end{bmatrix} \quad (12a)$$

and

$$\mathbf{B}=\begin{bmatrix} b(n, m), 1\leq n\leq\frac{N}{2}, 0\leq m\leq M \end{bmatrix}, \quad (12b)$$

respectively; the following objective error function is used in the paper:

$$e(\mathbf{A}, \mathbf{B})=\sum_{i=0}^{K_\omega}\sum_{l=0}^{K_p}W(\omega_l)|D(\omega_i, p_l)-H(e^{j\omega_l}, p_l)|^2 \\ =\sum_{i=0}^{K_\omega}\sum_{l=0}^{K_p}W(\omega_l)|\hat{D}(\omega_i, p_l)-\hat{H}(\omega_i, p_l)|^2, \\ \omega_i=\omega_s+\frac{i(\omega_f-\omega_s)}{K_\omega}, p_l=p_s+\frac{l(p_f-p_s)}{K_p}, \quad (13)$$

where a  $(K_\omega+1)\times(K_p+1)$  grid is chosen for the error evaluation, and  $W(\omega)$  is a positive weighting function. In

## 2. Problem formulation

For designing a VFO differintegrator, the desired response is given by

$$D(\omega, p)=e^{-j\omega}j(\omega)^p, \quad p_s\leq p\leq p_f, \quad \omega_s\leq|\omega|\leq\omega_f, \quad (1)$$

where  $l$  is a prescribed delay and  $p$  is the variable order of the designed differintegrator. If a pure VFO differentiator is designed,  $p_s\geq 0$  and  $\omega_s\geq 0$ , while  $p_f\leq 0$  and  $\omega_s>0$  for designing a pure VFO integrator, and  $p_s<0<p_f$ ,  $\omega_s>0$  for the VFO differintegrator design. Let

$$\hat{D}(\omega, p)=j(\omega)^p \\ =|\omega|^p\left[\cos\left(\frac{p\pi}{2}\right)+j\operatorname{sgn}(\omega)\sin\left(\frac{p\pi}{2}\right)\right], \quad (2)$$

where  $\operatorname{sgn}(\cdot)$  is a sign function, then Eq. (1) can be represented by

$$D(\omega, p)=e^{-j\omega}\hat{D}(\omega, p). \quad (3)$$

For approximating the desired response, the used transfer function is characterized by

$$H(z, p)=\sum_{n=0}^Nh_n(p)z^{-n}, \quad (4)$$

where the coefficients  $h_n(p)$  are expressed as the polynomials of  $p$ 

$$h_n(p)=\sum_{m=0}^Mh(n, m)p^m, \quad (5)$$

hence Eq. (4) becomes

$$H(z, p)=\sum_{n=0}^N\sum_{m=0}^Mh(n, m)p^mz^{-n}. \quad (6)$$

For simplicity, only even  $N$  is used in this section and the case for odd  $N$  will be given in Section 3. According to the symmetric and antisymmetric characteristics for the real part and imaginary part of (2), respectively, with respective to  $\omega$ , the coefficients  $h(n, m)$  in (6) are divided into even part and odd part by

$$h(n, m)=h_e(n, m)+h_o(n, m), \quad (7)$$

the paper,  $K_\omega = K_p = 200$  is used. By Pythagorean law,

$$\begin{aligned} e(\mathbf{A}, \mathbf{B}) &= \sum_{i=0}^{K_\omega} \sum_{l=0}^{K_p} W(\omega_i) \left[ \omega_i^{p_l} \cos\left(\frac{p_l \pi}{2}\right) - \sum_{n=0}^{N/2} \sum_{m=0}^M a(n, m) p_l^m \cos(n\omega_i) \right]^2 \\ &+ \sum_{i=0}^{K_\omega} \sum_{l=0}^{K_p} W(\omega_i) \left[ \omega_i^{p_l} \sin\left(\frac{p_l \pi}{2}\right) - \sum_{n=1}^{N/2} \sum_{m=0}^M b(n, m) p_l^m \right]^2 \\ &\times \sin(n\omega_i). \end{aligned} \quad (14)$$

Eq. (14) can be expressed in matrix form as

$$\begin{aligned} e(\mathbf{A}, \mathbf{B}) &= \text{tr}[(\mathbf{D}_A - \mathbf{CAP}^T)^T (\mathbf{D}_A - \mathbf{CAP}^T)] \\ &+ \text{tr}[(\mathbf{D}_B - \mathbf{SBP}^T)^T (\mathbf{D}_B - \mathbf{SBP}^T)] \\ &= e(\mathbf{A}) + e(\mathbf{B}), \end{aligned} \quad (15)$$

where  $\text{tr}(\cdot)$  denotes a trace operator, the superscript  $T$  denotes a transpose operator,

$$\mathbf{D}_A = [W^{1/2}(\omega_i) \omega_i^{p_l} \cos\left(\frac{p_l \pi}{2}\right), 0 \leq i \leq K_\omega, 0 \leq l \leq K_p], \quad (16a)$$

$$\mathbf{D}_B = [W^{1/2}(\omega_i) \omega_i^{p_l} \sin\left(\frac{p_l \pi}{2}\right), 0 \leq i \leq K_\omega, 0 \leq l \leq K_p], \quad (16b)$$

$$\mathbf{C} = [W^{1/2}(\omega_i) \cos(n\omega_i), 0 \leq i \leq K_\omega, 0 \leq n \leq \frac{N}{2}], \quad (16c)$$

$$\mathbf{S} = [W^{1/2}(\omega_i) \sin(n\omega_i), 0 \leq i \leq K_\omega, 1 \leq n \leq \frac{N}{2}], \quad (16d)$$

$$\mathbf{P} = [p_l^m, 0 \leq l \leq K_p, 0 \leq m \leq M] \quad (16e)$$

and

$$\begin{aligned} e(\mathbf{A}) &= \text{tr}[(\mathbf{D}_A - \mathbf{CAP}^T)^T (\mathbf{D}_A - \mathbf{CAP}^T)] \\ &= \text{tr}[\mathbf{D}_A^T \mathbf{D}_A - \mathbf{D}_A^T \mathbf{CAP}^T - (\mathbf{CAP}^T)^T \mathbf{D}_A + (\mathbf{CAP}^T)^T (\mathbf{CAP}^T)], \end{aligned} \quad (17)$$

$$\begin{aligned} e(\mathbf{B}) &= \text{tr}[(\mathbf{D}_B - \mathbf{SBP}^T)^T (\mathbf{D}_B - \mathbf{SBP}^T)] \\ &= \text{tr}[\mathbf{D}_B^T \mathbf{D}_B - \mathbf{D}_B^T \mathbf{SBP}^T - (\mathbf{SBP}^T)^T \mathbf{D}_B + (\mathbf{SBP}^T)^T (\mathbf{SBP}^T)]. \end{aligned} \quad (18)$$

Differentiating  $e(\mathbf{A}, \mathbf{B})$  with respect to  $\mathbf{A}$  [35],

$$\begin{aligned} \frac{\partial e(\mathbf{A}, \mathbf{B})}{\partial \mathbf{A}} &= \frac{\partial e(\mathbf{A})}{\partial \mathbf{A}} = -(\mathbf{D}_A^T \mathbf{C})^T (\mathbf{P}^T)^T - \mathbf{C}^T \mathbf{D}_A \mathbf{P} \\ &+ (\mathbf{P} \mathbf{A}^T \mathbf{C})^T (\mathbf{P}^T)^T + \mathbf{C}^T \mathbf{CAP}^T \mathbf{P}, \end{aligned} \quad (19)$$

which is then set to zero, and the coefficient matrix  $\mathbf{A}$  can be obtained as

$$\mathbf{A} = (\mathbf{C}^T \mathbf{C})^{-1} \mathbf{C}^T \mathbf{D}_A \mathbf{P} (\mathbf{P}^T \mathbf{P})^{-1}. \quad (20)$$

Similarly, the coefficient matrix  $\mathbf{B}$  can be achieved by differentiating  $e(\mathbf{A}, \mathbf{B})$  with respect to  $\mathbf{B}$  and setting the result to zero, which yields

$$\mathbf{B} = (\mathbf{S}^T \mathbf{S})^{-1} \mathbf{S}^T \mathbf{D}_B \mathbf{P} (\mathbf{P}^T \mathbf{P})^{-1}. \quad (21)$$

Notice that the weighting function  $W(\omega)$  has been incorporated in the relevant matrices, so that the peak absolute error of variable frequency response can be

reduced by a proper iterative method, which will be shown in Section 3.

### 3. Numerical examples and discussions

To demonstrate the effectiveness and flexibility of the proposed method, several examples including a VFO differentiator, two pure VFO differentiators and a pure VFO integrator are presented in this section. To evaluate the performance, the normalized root-mean-squared error of variable frequency response and the maximum absolute error of variable frequency response are defined by

$$\varepsilon_{rms} = \left[ \frac{\int_{p_s}^{p_f} \int_{\omega_s}^{\omega_f} |D(\omega, p) - H(e^{j\omega}, p)|^2 d\omega dp}{\int_{p_s}^{p_f} \int_{\omega_s}^{\omega_f} |D(\omega, p)|^2 d\omega dp} \right]^{1/2} \times 100\% \quad (22a)$$

and

$$\varepsilon_m = \max(|D(\omega, p) - H(e^{j\omega}, p)|), \quad \omega_s \leq \omega \leq \omega_f, \quad p_s \leq p \leq p_f, \quad (22b)$$

respectively. To compute the error  $\varepsilon_{rms}$ , the general trapezoidal rule is used [34] with step sizes  $(\omega_f - \omega_s)/200$  and  $(p_f - p_s)/200$  for  $\omega$ -axis and  $p$ -axis, respectively. Also, the error  $\varepsilon_m$  is computed with the same sampling sizes as above.

**Example 1.** This example deals with the least-squares design of a VFO differentiator with  $N = 40$ ,  $M = 5$ ,  $\omega_s = (0.05)\pi$ ,  $\omega_f = (0.95)\pi$ ,  $p_s = -0.5$ ,  $p_f = 0.5$  and  $W(\omega) = 1$ . Fig. 1(a) and (b) present the obtained magnitude response and the absolute error of variable frequency response, respectively, and the errors  $\varepsilon_{rms} \approx 0.60277728\%$  and  $\varepsilon_m = 0.1369375$ .

It is noted that the phase difference between  $\omega = \pi$  and  $\omega = -\pi$  is  $p\pi$ , which is not an integer multiple of  $2\pi$  for all  $p$  in the range  $[p_s, p_f]$ , so it is not recommended to set  $\omega_f = \pi$ . However, for comparing with the results of [22], the differentiator is designed again with  $\omega_s = (0.01)\pi$ ,  $\omega_f = \pi$ . If the computation of integration in [22] is implemented by using the trapezoidal rule with step sizes  $(\omega_f - \omega_s)/200$  and  $(p_f - p_s)/200$  for  $\omega$ -axis and  $p$ -axis, respectively, both the method of [22] and the proposed method induce the exactly same results:  $\varepsilon_{rms} \approx 10.01497046\%$  and  $\varepsilon_m = 3.11763459$ .

**Example 2.** For designing a pure VFO differentiator,  $0 \leq p_s < p_f$ . For example, a VFO differentiator is designed with  $N = 30$ ,  $M = 6$ ,  $\omega_s = 0$ ,  $\omega_f = 0.9\pi$ ,  $p_s = 1$ ,  $p_f = 2$  and  $W(\omega) = 1$ , the variable magnitude response and the absolute error of variable frequency response are shown in Fig. 2(a) and (b), respectively, and the errors  $\varepsilon_{rms} \approx 0.166372\%$  and  $\varepsilon_m = 0.03382684$  which are better than  $\varepsilon_{rms} \approx 1.17212338\%$  and  $\varepsilon_m = 0.11866149$  obtained with the method of [22] where the ill-conditioned problem will occur for the relevant matrix.

說明：學過微積分後自然對微分及積分不陌生，對多階（整數）微分亦了然於胸。但大部份同學可能就沒聽過  $\frac{1}{2}$  階微分，或  $\frac{3}{4}$  階積分了。此即分數階的微分器或積分器，其廣泛應用於自然及商學科學。本單元即要帶入此觀念，並運用前一單元之技術設計可調式分數階微積分器。相關公式推導如次。



desired frequency response:

$$H_d(\omega, p) = e^{-j\omega p} (j\omega)^p \quad p \leq p_2 \quad \omega \leq |\omega| \leq \omega_2$$

$$= e^{-j\omega p} (\omega)^p \left[ \cos\left(\frac{p\pi}{2}\right) + j \sin\left(\frac{p\pi}{2}\right) \right]$$

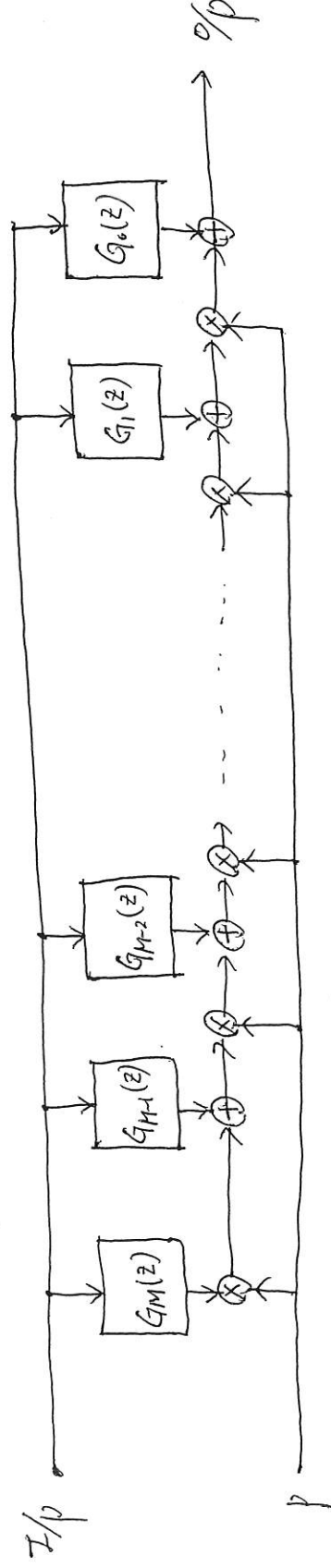
variable FIR digital filter:

$$H(z, p) = \sum_{n=0}^N h_n(p) z^{-n}$$

$$h_n(p) = \sum_{m=0}^M h(n, m) p^m$$

$$H(z, p) = \sum_{n=0}^N \sum_{m=0}^M h(n, m) p^m z^{-n} = \sum_{n=0}^N \left( \sum_{m=0}^M h(n, m) z^{-n} \right) p^m = \sum_{m=0}^M G_m(z) p^m$$

$$G_m(z) = \sum_{n=0}^N h(n, m) z^{-n}$$



$$h(n, m) = h_e(n, m) + h_o(n, m)$$

$$h_e\left(\frac{N}{2} + n, m\right) = \frac{1}{2} \left[ h\left(\frac{N}{2} + n, m\right) + h\left(\frac{N}{2} - n, m\right) \right] \quad -\frac{N}{2} \leq n \leq \frac{N}{2} \quad 0 \leq m \leq M$$

$$h_o\left(\frac{N}{2} + n, m\right) = \frac{1}{2} \left[ h\left(\frac{N}{2} + n, m\right) - h\left(\frac{N}{2} - n, m\right) \right] \quad -\frac{N}{2} \leq n \leq \frac{N}{2} \quad 0 \leq m \leq M$$

N: even

$$H(e^{j\omega}, p) = e^{-j\frac{N}{2}\omega} \left[ \sum_{n=0}^{N/2} \sum_{m=0}^M a(n, m) p^m \cos(n\omega) + j \sum_{n=1}^{N/2} \sum_{m=0}^M b(n, m) p^m \sin(n\omega) \right]$$

$$= e^{-j\frac{N}{2}\omega} \left[ a^T c(\omega, p) + j b^T s(\omega, p) \right]$$

$$a(n, m) = \begin{cases} h_e\left(\frac{N}{2}, m\right) & n=0 \quad 0 \leq m \leq M \\ 2h_e\left(\frac{N}{2} - n, m\right) & 1 \leq n \leq \frac{N}{2} \quad 0 \leq m \leq M \end{cases}$$

$$b(n, m) = 2h_o\left(\frac{N}{2} - n, m\right) \quad 1 \leq n \leq \frac{N}{2}, \quad 0 \leq m \leq M$$

$$a = \begin{bmatrix} a(n, m) \end{bmatrix} \quad 0 \leq n \leq \frac{N}{2}, \quad 0 \leq m \leq M \quad ]^T$$

$$b = \begin{bmatrix} b(n, m) \end{bmatrix} \quad 1 \leq n \leq \frac{N}{2}, \quad 0 \leq m \leq M \quad ]^T$$

$$c(\omega, p) = \begin{bmatrix} p^m \cos(n\omega) \end{bmatrix} \quad 0 \leq n \leq \frac{N}{2}, \quad 0 \leq m \leq M \quad ]^T$$

$$s(\omega, p) = \begin{bmatrix} p^m \sin(n\omega) \end{bmatrix} \quad 1 \leq n \leq \frac{N}{2}, \quad 0 \leq m \leq M \quad ]^T$$

p 代表微积分器之阶数,  
p=1 代表一阶微分器, 即对  
输入信号做一求微分。  
p=-1 代表一阶积分器, 即  
以此类推。

$$\text{let } I = \frac{N}{2}$$

objective error function:

$$\begin{aligned} e(a, b) &= \int_{p_1}^{p_2} \int_{\omega_1}^{\omega_2} |H(\omega, p) - H(e^{j\omega}, p)|^2 d\omega dp \\ &= \int_{p_1}^{p_2} \int_{\omega_1}^{\omega_2} \left| \omega^p \cos\left(\frac{p\pi}{2}\right) + j\omega^p \sin\left(\frac{p\pi}{2}\right) - a^T C(\omega, p) - j b^T S(\omega, p) \right|^2 d\omega dp \\ &= \int_{p_1}^{p_2} \int_{\omega_1}^{\omega_2} \left[ \omega^p \cos\left(\frac{p\pi}{2}\right) - a^T C(\omega, p) \right]^2 d\omega dp + \int_{p_1}^{p_2} \int_{\omega_1}^{\omega_2} \left[ \omega^p \sin\left(\frac{p\pi}{2}\right) - b^T S(\omega, p) \right]^2 d\omega dp \\ &= e(a) + e(b) \end{aligned}$$

$$e(a) = S_a + r_a^T a + a^T Q_a a$$

$$e(b) = S_b + r_b^T b + b^T Q_b b$$

$$S_a = \iint \left( \omega^p \cos\left(\frac{p\pi}{2}\right) \right)^2 d\omega dp$$

$$S_b = \iint \left( \omega^p \sin\left(\frac{p\pi}{2}\right) \right)^2 d\omega dp$$

$$r_a = -2 \iint \omega^p \cos\left(\frac{p\pi}{2}\right) C(\omega, p) d\omega dp$$

$$r_b = -2 \iint \omega^p \sin\left(\frac{p\pi}{2}\right) S(\omega, p) d\omega dp$$

$$Q_a = \iint C(\omega, p) C^T(\omega, p) d\omega dp$$

$$Q_b = \iint S(\omega, p) S^T(\omega, p) d\omega dp$$

$$\frac{\partial e(a, b)}{\partial a} = r_a + 2Q_a a = 0 \Rightarrow a = -\frac{1}{2} Q_a^{-1} r_a$$

$$\frac{\partial e(a, b)}{\partial b} = r_b + 2Q_b b = 0 \Rightarrow b = -\frac{1}{2} Q_b^{-1} r_b$$

Example:  $N=40$   $M=5$   $\omega_1=0.05\pi$   $\omega_2=0.95\pi$   $p_1=-0.5$   $p_2=0.5$

Example:  $N=30$   $M=6$   $\omega_1=0$   $\omega_2=0.9\pi$   $p_1=1$   $p_2=2$

Example:  $N=60$   $M=6$   $\omega_1=0.05\pi$   $\omega_2=0.9\pi$   $p_1=-1.5$   $p_2=-0.5$

```

##
## Design of variable fractional-order (VFO) differentiators
##
import numpy as np
import math
import matplotlib.pyplot as plt
from scipy import signal

N=40
M=5
w1=0.05*math.pi
w2=0.95*math.pi
p1=-0.5
p2=1
Nw=200
Np=60
#
NH=N//2
nma=(NH+1)*(M+1)
nmb=NH*(M+1)
deltaw=(w2-w1)/Nw
deltap=(p2-p1)/Np
Nwp=(Nw+1)*(Np+1)
NVa=np.arange(0,NH+1); NVa=NVa[:,np.newaxis]
NVb=np.arange(1,NH+1); NVb=NVb[:,np.newaxis]
##
ra=np.zeros((nma,1))
Qa=np.zeros((nma,nma))
for ip in range(0,Np+1):
    p=p1+ip*deltap
    for iw in range(0,Nw+1):
        w=w1+iw*deltaw
        cwp=np.zeros((nma,1))
        for im in range(0,M+1):
            cwp[im*(NH+1):(im+1)*(NH+1),0]=p**(im)*np.cos(w*NVa[:,0])
            ra=ra-2*w**p*np.cos(p*math.pi/2)*cwp
            Qa=Qa+cwp@np.transpose(cwp)
        ra=(w2-w1)*(p2-p1)*ra/Nwp
        Qa=(w2-w1)*(p2-p1)*Qa/Nwp
        a=-0.5*np.linalg.inv(Qa)@ra
    ##
    rb=np.zeros((nmb,1))
    Qb=np.zeros((nmb,nmb))
    for ip in range(0,Np+1):
        p=p1+ip*deltap
        for iw in range(0,Nw+1):
            w=w1+iw*deltaw
            swp=np.zeros((nmb,1))
            for im in range(0,M+1):
                swp[im*NH:(im+1)*NH,0]=p**im*np.sin(w*NVb[:,0])
                rb=rb-2*w**p*np.sin(p*math.pi/2)*swp
                Qb=Qb+swp@np.transpose(swp)
            rb=(w2-w1)*(p2-p1)*rb/Nwp
            Qb=(w2-w1)*(p2-p1)*Qb/Nwp
            b=-0.5*np.linalg.inv(Qb)@rb
    ##
    a2=np.reshape(a,(M+1,NH+1)); a2=np.transpose(a2)
    he=np.zeros((N+1,M+1))
    he[NH,:]=a2[0,:]
    he[0:NH,:]=0.5*np.flipud(a2[1:NH+1,:])
    he[NH+1:N+1,:]=0.5*a2[1:NH+1,:]
    ##
    b2=np.reshape(b,(M+1,NH)); b2=np.transpose(b2)
    ho=np.zeros((N+1,M+1))

```

```

ho[0:NH,:]=0.5*np.flipud(b2)
ho[NH+1:N+1,:]=-0.5*b2
##
h2=he+ho
MR=np.zeros((Nw+1,Np+1,1))
for ip in range (0,Np+1):
    p=pl+ip*deltap
    h=h2[:,0]
    for im in range (1,M+1): h=h+h2[:,im]*p**im
    rr=np.linspace(wl,w2,num=Nw+1); rr=rr[:,np.newaxis]
    MRR=np.absolute(signal.freqz(h,1,rr))
    MR[:,ip]=MRR[1]

for i in range (0,Np+1): plt.plot(rr/math.pi,MR[:,i])
plt.axis([wl/math.pi,w2/math.pi,0,3])
plt.xlabel('Normalized frequency')
plt.ylabel('Amplitude response')
plt.title('VFO')
plt.show()

```