2. Digital Filter Design (A)

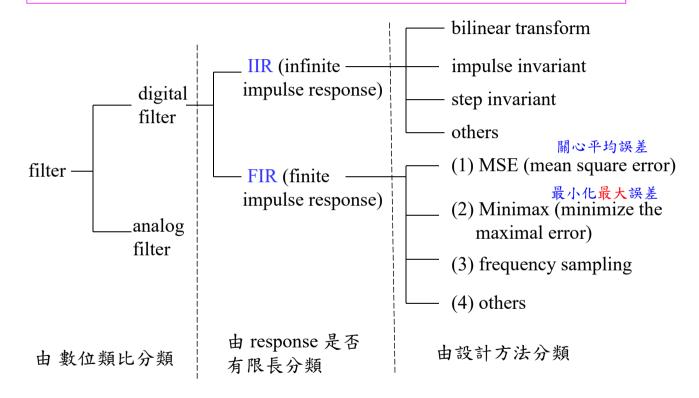
任何可以用來去除 noise 作用的 operation, 皆被稱為 filter

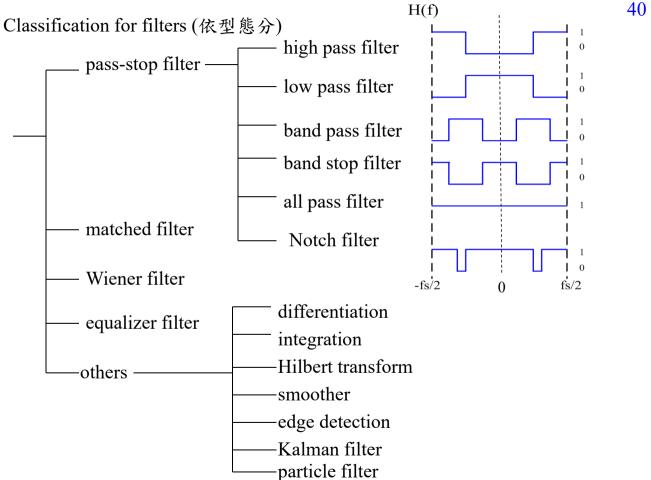
甚至有部分的 operation,雖然主要功用不是用來去除 noise,但是可以用 FT + multiplication + IFT 來表示,也被稱作是 filter convolution, LTI system

Reference

- [1] A. V. Oppenheim and R. W. Schafer, *Discrete-Time Signal Processing*, London: Prentice-Hall, 3rd ed., 2010.
- [2] D. G. Manolakis and V. K. Ingle, *Applied Digital Signal Processing*, Cambridge University Press, Cambridge, UK. 2011.
- [3] B. A. Shenoi, *Introduction to Digital Signal Processing and Filter Design*, Wiley-Interscience, N. J., 2006.
- [4] A. A. Khan, *Digital Signal Processing Fundamentals*, Da Vinci Engineering Press, Massachusetts, 2005.
- [5] S. Winder, Analog and Digital Filter Design, 2nd Ed., Amsterdam, 2002.

2-A Classification for Filters





• 2-B FIR Filter Design

FIR filter: impulse response is nonzero at finite number of points

$$h[n] = 0$$
 for $n < 0$ and $n \ge N$
 $(h[n] \text{ has } N \text{ points}, N \text{ is a finite number})$
 $h[n] \text{ is causal}$

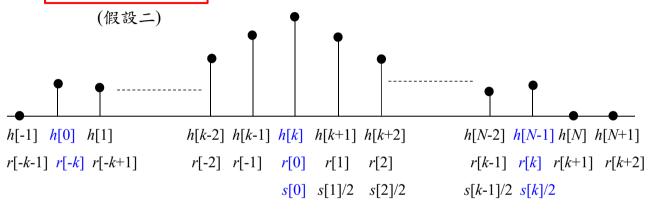


• FIR is more popular because its impulse response is finite.

(假設一) 42

Specially, when h[n] is even symmetric h[n] = h[N-1-n]

and N is an odd number



(a)
$$r[n] = h[n + k]$$
, where $k = (N-1)/2$.

(b)
$$s[0] = r[0]$$
, $s[n] = 2r[n]$ for $0 < n \le k$.

Impulse Response of the FIR Filter:

$$h[n]$$
 $(h[n] \neq 0 \text{ for } 0 \le n \le N-1)$
 $r[n] = h[n+k], k = (N-1)/2 (r[n] \neq 0 \text{ for } -k \le n \le k, \text{ see page 42})$

Suppose that the filter is **even**, r[n] = r[-n].

Set
$$s[0] = r[0]$$
, $s[n] = 2r[n]$ for $n \neq 0$.

Then, the discrete-time Fourier transform of the filter is

$$H(F) = \sum_{n=-\infty}^{\infty} h[n]e^{-j2\pi F n} \qquad (F = f\Delta_t \text{ is the normalized frequency})$$
See page 22
$$H(F) = e^{-j2\pi F k}R(F) \qquad R(F) = \sum_{n=-k}^{k} r[n]e^{-j2\pi F n}$$

$$= \sum_{n=-k}^{-1} r[n]e^{-j2\pi F n} + r[0] + \sum_{n=1}^{k} r[n]e^{-j2\pi F n}$$

$$= \sum_{n=-k}^{k} r[-n]e^{j2\pi F n} + r[0] + \sum_{n=1}^{k} r[n]e^{-j2\pi F n}$$

2-C Least MSE Form and Minimax Form FIR Filters

(1) least MSE (mean square error) form

MSE =
$$f_s^{-1} \int_{-f_s/2}^{f_s/2} |H(f) - H_d(f)|^2 df$$
, f_s : sampling frequency

H(f): the spectrum of the filter we obtain

 $H_d(f)$: the spectrum of the desired filter

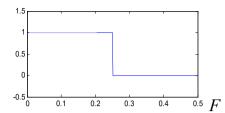
(2) mini-max (minimize the maximal error) form (關心 最大 誤差)

maximal error:
$$\underset{f}{\text{Max}} |H(f) - H_d(f)|$$

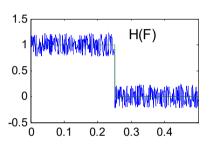
The transition band is always ignored

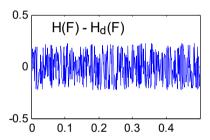
Example:

desired output $H_d(F)$

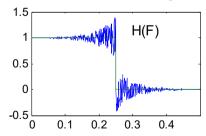


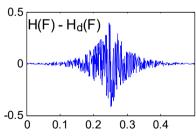
(A) larger MSE, but smaller maximal error





(B) smaller MSE, but larger maximal error





• 2-D Review: FIR Filter Design in the MSE Sense

$$R(F) = \sum_{n=0}^{k} s[n] \cos(2\pi n F)$$

$$MSE = \int_{s}^{-1} \int_{-f_{s}/2}^{f_{s}/2} |R(f) - H_{d}(f)|^{2} df = \int_{-1/2}^{1/2} |R(F) - H_{d}(F)|^{2} dF$$

$$= \int_{-1/2}^{1/2} |\sum_{n=0}^{k} s[n] \cos(2\pi nF) - H_{d}(F)|^{2} dF$$

$$= \int_{-1/2}^{1/2} \left(\sum_{n=0}^{k} s[n] \cos(2\pi nF) - H_{d}(F) \right) \left(\sum_{\tau=0}^{k} s[\tau] \cos(2\pi \tau F) - H_{d}(F) \right) dF$$

$$= \int_{-1/2}^{1/2} \sum_{n=0}^{k} s[n] \cos(2\pi nF) \sum_{\tau=0}^{k} s[\tau] \cos(2\pi \tau F) dF$$

$$-2 \int_{-1/2}^{1/2} \sum_{s=0}^{k} s[n] \cos(2\pi nF) H_{d}(F) dF + \int_{-1/2}^{1/2} H_{d}^{2}(F) dF$$

$$\frac{\partial MSE}{\partial s[n]} = 2\sum_{\tau=0}^{k} s[\tau] \int_{-1/2}^{1/2} \cos(2\pi nF) \cos(2\pi \tau F) dF - 2\int_{-1/2}^{1/2} H_d(F) \cos(2\pi nF) dF = 0$$

$$\frac{\partial MSE}{\partial s[n]} = 2\sum_{\tau=0}^{k} s[\tau] \int_{-1/2}^{1/2} \cos(2\pi n F) \cos(2\pi \tau F) dF - 2\int_{-1/2}^{1/2} H_d(F) \cos(2\pi n F) dF = 0$$

From the facts that

$$\int_{-1/2}^{1/2} \cos(2\pi n F) \cos(2\pi \tau F) dF = 0 \quad \text{when } n \neq \tau,$$

$$\int_{-1/2}^{1/2} \cos(2\pi n F) \cos(2\pi \tau F) dF = 1/2 \quad \text{when } n = \tau, \ n \neq 0,$$

$$\int_{-1/2}^{1/2} \cos(2\pi n F) \cos(2\pi \tau F) dF = 1 \quad \text{when } n = \tau, \ n = 0.$$

Therefore,

$$\frac{\partial MSE}{\partial s[0]} = 2s[0] - 2\int_{-1/2}^{1/2} H_d(F) dF = 0$$

$$\frac{\partial MSE}{\partial s[n]} = s[n] - 2\int_{-1/2}^{1/2} \cos(2\pi nF) H_d(F) dF = 0 \qquad \text{for } n \neq 0.$$

Minimize MSE \rightarrow Make $\frac{\partial MSE}{\partial s[n]} = 0$ for all *n*'s

$$|s[0] = \int_{-1/2}^{1/2} H_d(F) dF |, \qquad |s[n] = 2 \int_{-1/2}^{1/2} \cos(2\pi n F) H_d(F) dF |.$$

Finally, set h[k] = s[0],

$$h[k+n] = s[n]/2$$
, $h[k-n] = s[n]/2$ for $n = 1, 2, 3, ..., k$,
 $h[n] = 0$ for $n < 0$ and $n \ge N$.

Then, h[n] is the impulse response of the designed filter.

O 2-E FIR Filter Design in the Mini-Max Sense

It is also called "Remez-exchange algorithm"

or "Parks-McClellan algorithm"

References

- [1] T. W. Parks and J. H. McClellan, "Chebychev approximation for nonrecursive digital filter with linear phase", *IEEE Trans. Circuit Theory*, vol. 19, no. 2, pp. 189-194, March 1972.
- [2] J. H. McClellan, T. W. Parks, and L. R. Rabiner "A computer program for designing optimum FIR linear phase digital filter", *IEEE Trans. Audio-Electroacoustics*, vol. 21, no. 6, Dec. 1973.
- [3] F. Mintz and B. Liu, "Practical design rules for optimum FIR bandpass digital filter", *IEEE Trans. ASSP*, vol. 27, no. 2, Apr. 1979.
- [4] E. Y. Remez, "General computational methods of Chebyshev approximation: The problems with linear real parameters," AEC-TR-4491. ERDA Div. Phys. Res., 1962.

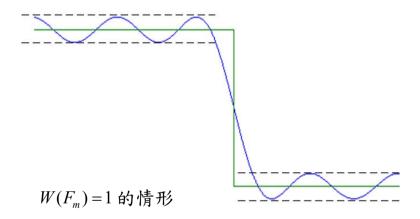
Suppose that:

- ① Filter length = N, N is odd, N = 2k+1.
- ② Frequency response of the **desired filter**: $H_d(F)$ is an even function (F is the normalized frequency)

Two constraints

用Mini-Max方法所設計出的filters,一定會滿足以下二個條件

- (1) 有k+2個以上的extreme points Error 的 local maximal local minimum
- (2) 在extreme points上, $W(F_m)|R(F_m)-H_d(F_m)|$ 是定值



證明可參考

T. W. Parks and J. H. McClellan, "Chebychev approximation for nonrecursive digital filter with linear phase", *IEEE Trans. Circuit Theory*, vol. 19, no. 2, pp. 189-194, March 1972.

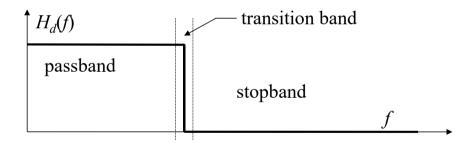
Generalization for Mini-Max Sense by weight function

maximal error:
$$Max |H(f) - H_d(f)|$$

weighted maximal error:
$$\underset{f}{\text{Max}}|W(f)[H(f)-H_d(f)]|$$

where W(f) is the weight function.

The weight function is designed according to which band is more important.



Q: How do we choose W(f) When SNR \uparrow ?

Example: If we treat the passband the same important as the stopband. W(f) = 1 in the passband, W(f) = 1 in the stopband

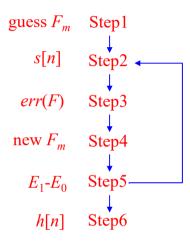
Q1: W(f) = 1 in the passband, W(f) < 1 in the stopband 代表什麼?

Q2: W(f) < 1 in the passband, W(f) = 1 in the stopband 代表什麼?

Q3: 如何用來壓縮特定區域 (如 transition band 附近) 的 error?

Q4: Weighting function 的概念可否用在 MSE sense ?

© 2-F Mini-Max Design Process



(Step 1): Choose arbitrary k+2 extreme frequencies in the range of

$$0 \le F \le 0.5$$
, (denoted by $F_0, F_1, F_2, \dots, F_{k+1}$)

Note: (1) Exclude the transition band.

(2) The extreme points cannot be all in the stop band.

Set
$$E_1$$
 (error) $\rightarrow \infty$

Extreme frequencies:

The locations where the error is maximal.

$$\begin{split} [R(F_0)-H_d(F_0)]W(F_0) &= -e & [R(F_1)-H_d(F_1)]W(F_1) = e \\ [R(F_2)-H_d(F_2)]W(F_2) &= -e & [R(F_3)-H_d(F_3)]W(F_3) = e \\ &\vdots \\ [R(F_{k+1})-H_d(F_{k+1})]W(F_{k+1}) &= (-1)^{k+2}e \end{split}$$

(Step 2): From page 43,
$$[R(F_m) - H_d(F_m)]W(F_m) = (-1)^{m+1}e$$
 (where $m = 0, 1, 2, \dots, k+1$) can be written as

$$\sum_{n=0}^{k} s[n]\cos(2\pi F_m n) + (-1)^m W^{-1}(F_m)e = H_d(F_m)$$
where $m = 0, 1, 2, \dots, k+1$.

Expressed by the matrix form:

$$\begin{bmatrix} 1 & \cos(2\pi F_0) & \cos(4\pi F_0) & \cdots & \cos(2\pi k F_0) & 1/W(F_0) \\ 1 & \cos(2\pi F_1) & \cos(4\pi F_1) & \cdots & \cos(2\pi k F_1) & -1/W(F_1) \\ 1 & \cos(2\pi F_2) & \cos(4\pi F_2) & \cdots & \cos(2\pi k F_2) & 1/W(F_2) \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & \cos(2\pi F_k) & \cos(4\pi F_k) & \cdots & \cos(2\pi k F_k) & (-1)^k/W(F_k) \\ 1 & \cos(2\pi F_{k+1}) & \cos(4\pi F_{k+1}) & \cdots & \cos(2\pi k F_{k+1}) & (-1)^{k+1}/W(F_{k+1}) \end{bmatrix} \begin{bmatrix} s[0] \\ s[1] \\ s[2] \\ \vdots \\ s[k] \\ e \end{bmatrix} = \begin{bmatrix} H_d[F_0] \\ H_d[F_1] \\ H_d[F_2] \\ \vdots \\ H_d[F_k] \\ H_d[F_{k+1}] \end{bmatrix}$$

Solve s[0], s[1], s[2],, s[k] from the above matrix

Square matrix

(performing the matrix inversion).

(Step 3): Compute err(F) for $0 \le F \le 0.5$, exclude the transition band.

$$err(F) = [R(F) - H_d(F)]W(F) = \{\sum_{n=0}^{k} s[n]\cos(2\pi nF) - H_d(F)\}W(F)$$

(Step 4): Find k+2 local maximal (or minimal) points of err(F)

local maximal point: if $q(\tau) > q(\tau + \Delta_F)$ and $q(\tau) > q(\tau - \Delta_F)$,

then τ is a local maximal of q(x).

local minimal point: if $q(\tau) < q(\tau + \Delta_F)$ and $q(\tau) < q(\tau - \Delta_F)$,

then τ is a local minimal of q(x).

Other rules: Page 60

Denote the local maximal (or minimal) points by $P_0, P_1, \dots, P_k, P_{k+1}$

These k+2 extreme points could include the boundary points of the transition band

Then h[n] is the impulse response of the designed filter.

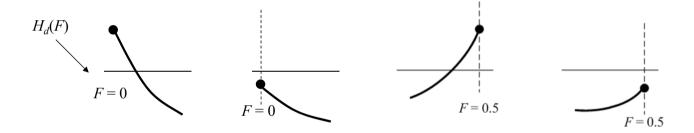
● 2-G Mini-Max FIR Filter 設計時需注意的地方

(1) Extreme points 不要選在 transition band

Initial guess的extreme points只要注意別取在transition band裡,即能保證 converge,不同的guess會影響converge的速度但不影響結果

- (2) E_1 (error of the previous iteration) < E_0 (present error) 時,亦不為收斂
- (3) 思考: page 56 中的 matrix operation,如何用一行的指令寫出?

- (4) Extreme points 判斷的規則:
- (a) The local peaks or local dips that are not at boundaries must be extreme points. boundaries: F = 0, F = 0.5, 以及 transition band 的雨端
- (b) For boundary points



Add a zero to the outside and conclude whether the point is a local maximum or a local minimum.

- (5) 有時,會找到多於 k+2 個 extreme points, 該如何選 $P_0, P_1, \dots, P_k, P_{k+1}$
 - (a) 優先選擇不在 boundaries 的 extreme points
 - (b) 其次選擇 boundary extreme points 當中 |err(F)| 較大的, 直到湊足 k+2 個 extreme points 為止

2-H Examples for Mini-Max FIR Filter Design

• Example 1: Design a 9-length highpass filter in the mini-max sense

ideal filter: $H_d(F) = 0$ for $0 \le F < 0.25$,

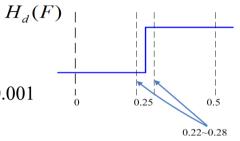
$$H_d(F) = 1$$
 for $0.25 < F \le 0.5$,

transition band: 0.22 < F < 0.28

$$\Delta = 0.001$$

weighting function: W(F) = 0.25 for $0 \le F \le 0.22$,

$$W(F) = 1$$
 for $0.28 \le F \le 0.5$,



(Step 1) Since
$$N = 9$$
, $k = (N-1)/2 = 4$, $k+2 = 6$,

→ Choose 6 extreme frequencies

(e.g.,
$$F_0 = 0$$
, $F_1 = 0.1$, $F_2 = 0.2$, $F_3 = 0.3$, $F_4 = 0.4$, $F_5 = 0.5$)

$$[R(F_n) - H_d(F_n)]W(F_n) = (-1)^{n+1}e, \quad n = 0, 1, 2, 3, 4, 5.$$

(**Step 2**)

$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 4 \\ 1 & 0.809 & 0.309 & -0.309 & -0.809 & -4 \\ 1 & 0.309 & -0.809 & -0.809 & 0.309 & 4 \\ 1 & -0.309 & -0.809 & 0.809 & 0.309 & -1 \\ 1 & -0.809 & 0.309 & 0.309 & -0.809 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 \\ \end{bmatrix} \begin{bmatrix} s[0] \\ s[1] \\ s[2] \\ s[3] \\ s[4] \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} s[0] \\ s[1] \\ s[2] \\ s[3] \\ s[4] \\ e \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & 4 \\ 1 & 0.809 & 0.309 & -0.309 & -0.809 & -4 \\ 1 & 0.309 & -0.809 & -0.809 & 0.309 & 4 \\ 1 & -0.309 & -0.809 & 0.809 & 0.309 & -1 \\ 1 & -0.809 & 0.309 & 0.309 & -0.809 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.5120 \\ -0.6472 \\ -0.0297 \\ 0.2472 \\ 0.0777 \\ -0.040 \end{bmatrix}$$

(Step 4) Extreme points:

$$F_0 = 0, F_1 = 0.125, F_2 = 0.22, F_3 = 0.28, F_4 = 0.356, F_5 = 0.5$$

(Step 5) $E_0 = \text{Max}[|\text{err}(F)|] = \underline{0.1501}$, return to Step 2.

Second iteration

(Step 2) Using
$$F_0 = 0$$
, $F_1 = 0.125$, $F_2 = 0.22$, $F_3 = 0.28$, $F_4 = 0.356$, $F_5 = 0.5$
 $\Rightarrow s[0] = 0.5018$, $s[1] = -0.6341$, $s[2] = -0.0194$, $s[3] = 0.3355$, $s[4] = 0.1385$

(Step 3)
$$err(F) = [R(F) - H_d(F)]W(F),$$

(Step 4) extreme points: 0, 0.132, 0.22, 0.28, 0.336, 0.5

(Step 5)
$$E_0 = \text{Max}[|\text{err}(F)|] = \underline{0.0951}$$
, return to Step 2.

Third iteration

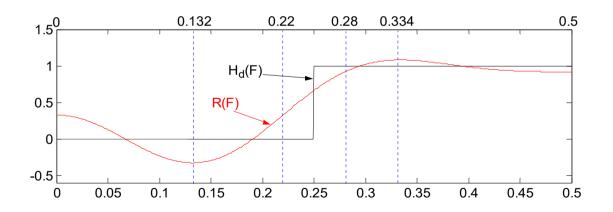
(Step 2), (Step 3), (Step 4), peaks : 0, 0.132, 0.22, 0.28, 0.334, 0.5 (Step 5) $E_0 = 0.0821$, return to Step 2.

Fourth iteration

(Step 2), (Step 3), (Step 4), peaks : 0, 0.132, 0.22, 0.28, 0.334, 0.5 (Step 5) $E_0 = \underline{0.0820}$, $E_1 - E_0 = 0.0001 \le \Delta$, continues to Step 6.

(Step 6) From
$$s[0] = 0.4990$$
, $s[1] = -0.6267$, $s[2] = -0.0203$, $s[3] = 0.3316$, $s[4] = 0.1442$

$$h[4] = s[0] = 0.4990,$$
 $h[3] = h[5] = s[1]/2 = -0.3134,$
 $h[2] = h[6] = s[2]/2 = -0.0101,$ $h[1] = h[7] = s[3]/2 = 0.1658,$
 $h[0] = h[8] = s[4]/2 = 0.0721.$



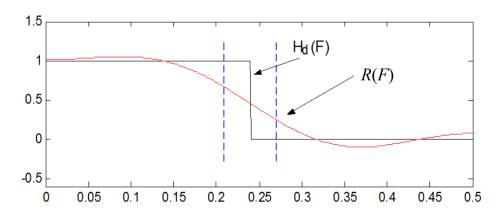
• Example 2: Design a 7-length digital filter in the mini-max sense

ideal filter:
$$H_d(F) = 1$$
 for $0 \le F < 0.24$, $H_d(F) = 0$ for $0.24 < F \le 0.5$, transition band: $0.21 < F < 0.27$ weighting function: $W(F) = 1$ for $0 \le F \le 0.21$, $W(F) = 0.5$ for $0.27 \le F \le 0.5$,

(Step 1) Since
$$N = 7$$
, $k = (N-1)/2 = 3$, $k+2 = 5$,
 \rightarrow Choose 5 extreme frequencies
(e.g., $F_0 = 0.05$, $F_1 = 0.15$, $F_2 = 0.3$, $F_3 = 0.4$, $F_4 = 0.5$)

(Step 2)
$$\begin{bmatrix} 1 & 0.9511 & 0.8090 & 0.5878 & 1 \\ 1 & 0.5878 & -0.309 & -0.9511 & -1 \\ 1 & -0.309 & -0.809 & 0.809 & 2 \\ 1 & -0.809 & 0.309 & 0.309 & -2 \\ 1 & -1 & 1 & -1 & 2 \end{bmatrix} \begin{bmatrix} s[0] \\ s[1] \\ s[2] \\ e \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\rightarrow s[0] = 0.4638, \ s[1] = 0.6327, \ s[2] = 0.0809, \ s[3] = -0.1608, \ e = -0.0364$$



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After Step 2,

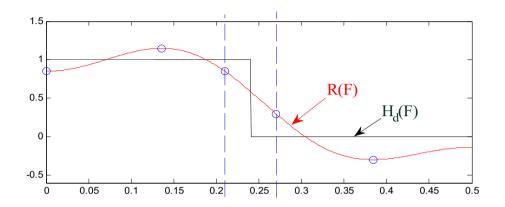
(Step 3)
$$\operatorname{err}(F) = [0.4638 + 0.6327 \cos(2\pi F) + 0.0809 \cos(4\pi F) - 0.1608 \cos(6\pi F) - H_d(F)]W(F)$$

(Step 4) extreme points: 0.089, 0.21, 0.27, 0.369, 0.5.

(Step 5) $E_0 = \text{Max}[|\text{err}(F)|] = 0.3396$, return to Step 2.

Iteration	1	2	3	4	5	6	7
Max[err(F)]	0.3396	0.2371	0.3090	0.1944	0.1523	0.1493	0.1493

After 7 times of iteration



$$s[0] = 0.4243, \ s[1] = 0.7559, \ s[2] = -0.0676, \ s[3] = -0.2619, \ e = 0.1493$$

(Step 6):

$$h[3] = 0.4243, \quad h[2] = h[4] = s[1]/2 = 0.3780,$$

 $h[1] = h[5] = s[2]/2 = -0.0338,$
 $h[0] = h[6] = s[3]/2 = -0.1309, \quad h[n] = 0 \text{ for } n < 0 \text{ and } n > 6$

附錄二: Spectrum Analysis for Sampled Signals

(學信號處理的人一定要會的基本常識)

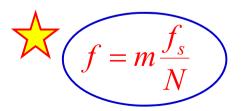
已知x[n] 是由一個 continuous signal y(t) 取樣而得

$$x[n] = y(n\Delta_t)$$

DFT:
$$X[m] = \sum_{n=0}^{N-1} x[n] e^{-j2\pi nm/N}$$
 FT: $Y(f) = \int_{-\infty}^{\infty} e^{-j2\pi f t} y(t) dt$

Q: x[n] 的 DFT 和 y(t) 的 Fourier transform 之間有什麼關係?

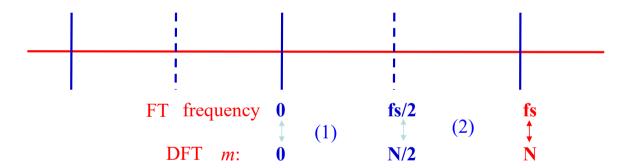
Basic rule:把間隔由 1 換成 f_s/N where $f_s = 1/\Delta_t$



(Very important)

(1)
$$X[m]\Delta_t \cong Y\left(m\frac{f_s}{N}\right)$$
 $f_s = 1/\Delta_t$ for $m \le N/2$

(2)
$$X[m]\Delta_t \cong Y\left((m-N)\frac{f_s}{N}\right) = Y\left(m\frac{f_s}{N} - f_s\right) \quad \text{for } m > N/2$$



If the sampling frequency is f_s , the FT output has the period of f_s . The DFT output has the period of N

Proof:
$$Y(f) = \int_{-\infty}^{\infty} e^{-j2\pi f t} y(t) dt$$
 $\exists t = n\Delta_t, f = m\Delta_f$ 代入

 $Y(m\Delta_f) \cong \sum_n e^{-j2\pi m\Delta_f n\Delta_t} y(n\Delta_t) \Delta_t = \Delta_t \sum_n e^{-j2\pi m\Delta_f n\Delta_t} x[n]$
 $\Delta_t \Delta_f = \frac{1}{N}$ i.e., $\Delta_f = \frac{1}{N\Delta_t} = \frac{f_s}{N}$
 $Y(m\frac{f_s}{N}) \cong \Delta_t \sum_n e^{-j2\pi \frac{mn}{N}} x[n]$
 $= \Delta_t DFT\{x[n]\}$

Example:已知

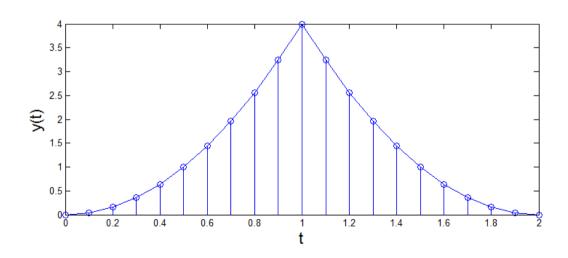
$$y(t) = (2t)^2$$
 for $0 \le t \le 1$

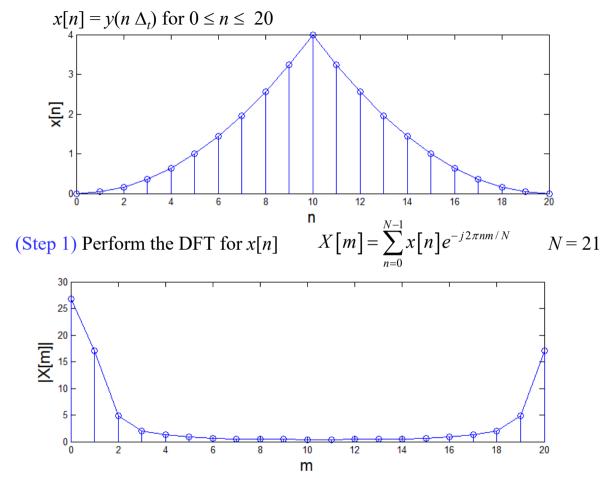
$$y(t) = (2t)^2$$
 for $0 \le t \le 1$ $y(t) = (4-2t)^2$ for $1 \le t \le 2$

取樣間隔: $\Delta_t = 0.1$

$$x[n] = y(n \Delta_t)$$
 for $0 \le n \le 20$

如何用 DFT 來正確的畫出 y(t) 的頻譜?





(Step 2-1)
$$Y\left(m\frac{f_s}{N}\right) \cong X[m]\Delta_t$$
 for $m \le N/2$

(Step 2-2)
$$Y\left((m-N)\frac{f_s}{N}\right) \cong X[m]\Delta_t \quad \text{for } m > N/2$$

In this example,
$$\frac{f_s}{N} = \frac{1}{N\Delta_s} = \frac{1}{21 \cdot 0.1} = 0.4762$$

