



Fourier Applications

傅立葉表示法對混合訊號的應用

Lecture 4-2



Sampling (取樣)

連續時間訊號的取樣

取樣操作從連續時間訊號獲得離散時間訊號

離散時間訊號的取樣

取樣操作改變資料傳輸速率與儲存樣本數目



Sampling Continuous-Time Signals

連續時間訊號的取樣

- 取樣操作從連續時間訊號 $x(t)$ 獲得離散時間訊號 $x[n]$ ，在取樣間距 T_s 的整數倍之處 $x(t)$ 所取的值亦稱為樣本 (sample)。

- $x[n] = x(nT_s)$, n is integer。

- $x[n]$ 的連續時間表示法：
$$x_\delta(t) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s)$$

$$x_\delta(t) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s) = \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s)$$



$$\begin{aligned}
 x_{\delta}(t) &= \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s) = \sum_{n=-\infty}^{\infty} x(t) \delta(t - nT_s) \\
 &= x(t) \cdot \sum_{n=-\infty}^{\infty} \delta(t - nT_s) = x(t) \cdot p(t),
 \end{aligned}$$

where $p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$ 單位脈衝列

應用乘積和褶積性質：

$$X_{\delta}(j\omega) = \frac{1}{2\pi} X(j\omega) * P(j\omega)$$

$$X_{\delta}(j\omega) = \frac{1}{2\pi} X(j\omega) * \frac{2\pi}{T_s} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s)$$

where $\omega_s = 2\pi / T_s$ sampling frequency

$$X_{\delta}(j\omega) = \frac{1}{2\pi} X(j\omega) * \frac{2\pi}{T_s} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$



$$X_{\delta}(j\omega) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

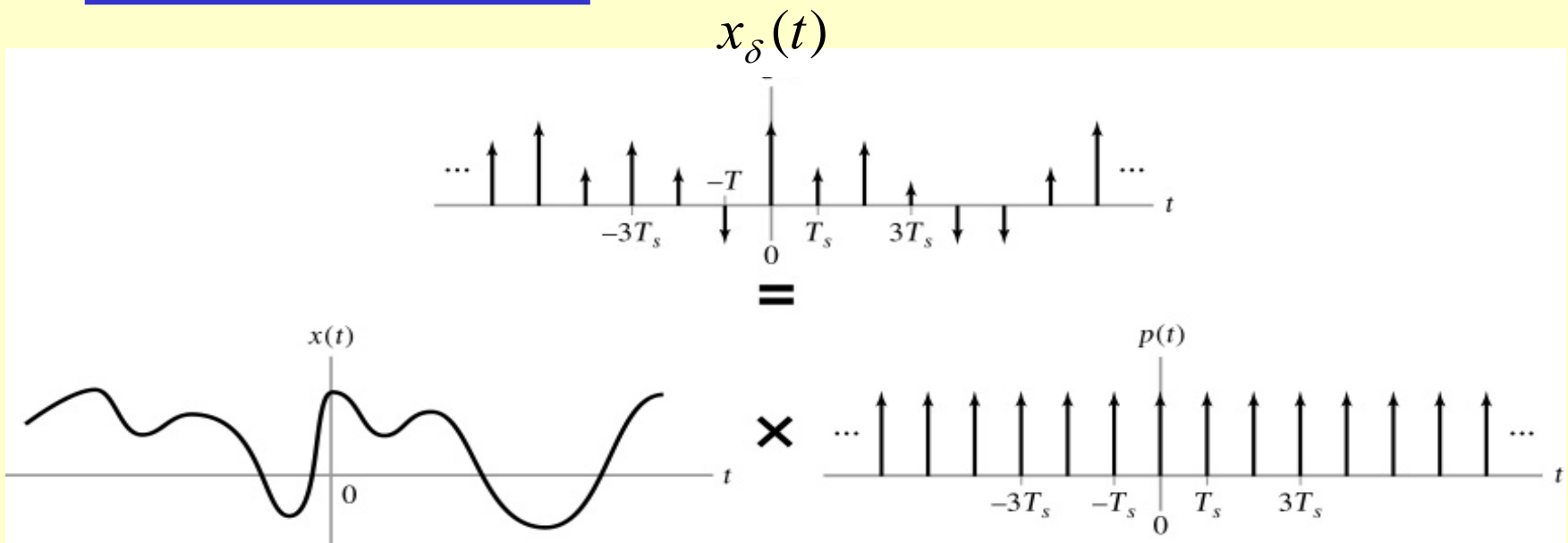
- 樣本訊號的 FT 是原始訊號的 FT ,
經過不同頻率平移後的無窮和。
- 平移後頻譜(spectrum)的位置距離原來位置是 ω_s 的整數倍。
- 若 ω_s 不夠大 , 平移後的頻譜可能會重疊。
- 原始訊號和其頻率位移的頻譜重疊情形稱為頻疊(aliasing)。



Mathematical Representation of Sampling

- Product of a Given Time Signal and an Impulse Train
- $x[n]$ 的連續時間表示法 $x_d(t)$ or $x_\delta(t)$ 可由取樣程序獲得:

$$x_\delta(t) = x(t) \cdot p(t)$$





Aliasing ?

The FT of a sampled signal for different sampling frequencies.

(a) Spectrum of continuous-time signal.

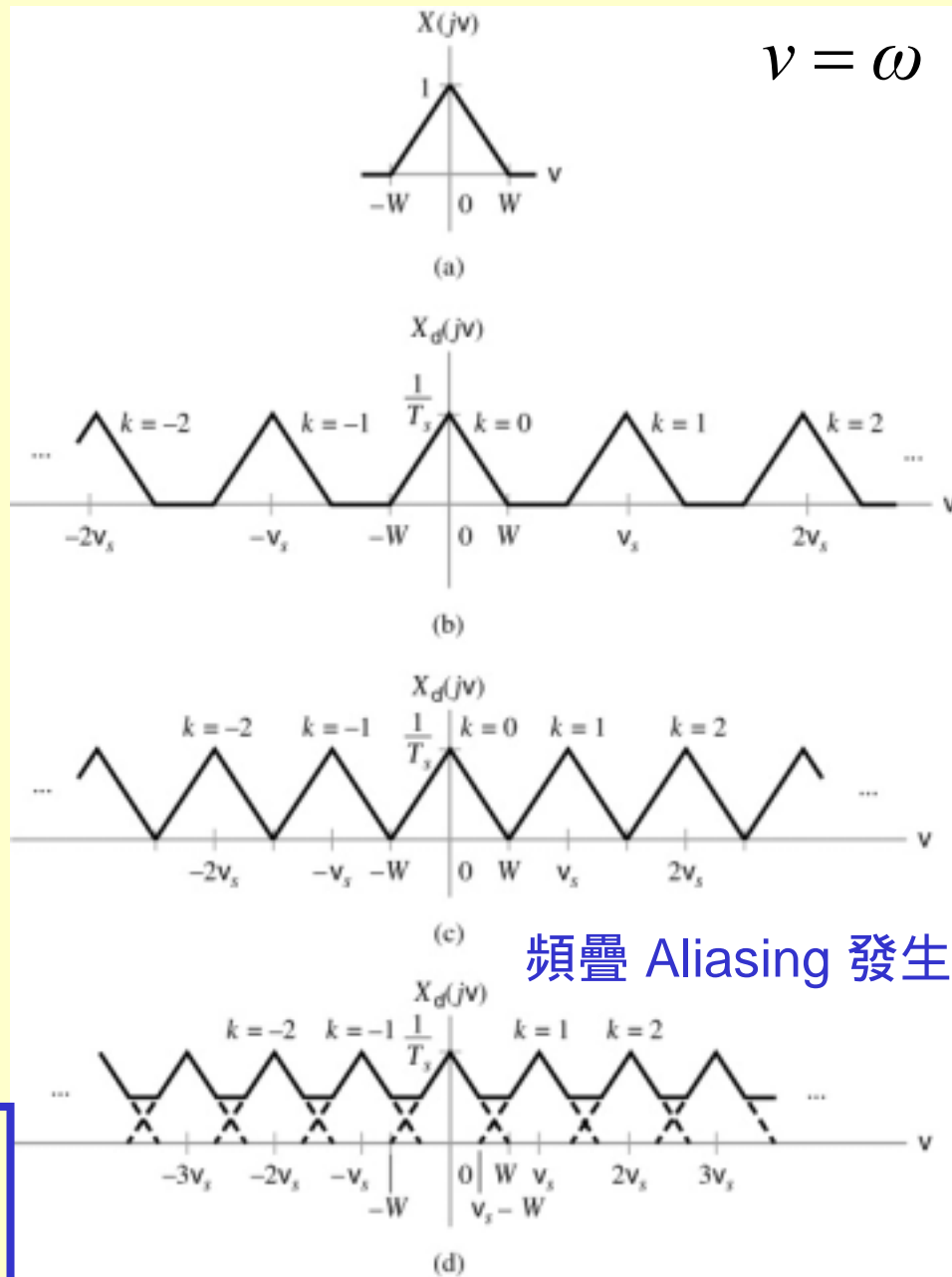
(b) Spectrum of sampled signal when $\omega_s = 3W$.

(c) Spectrum of sampled signal when $\omega_s = 2W$.

(d) Spectrum of sampled signal when $\omega_s = 1.5W$.

$$X_\delta(j\omega) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

$$\nu = \omega$$



頻疊 Aliasing 發生

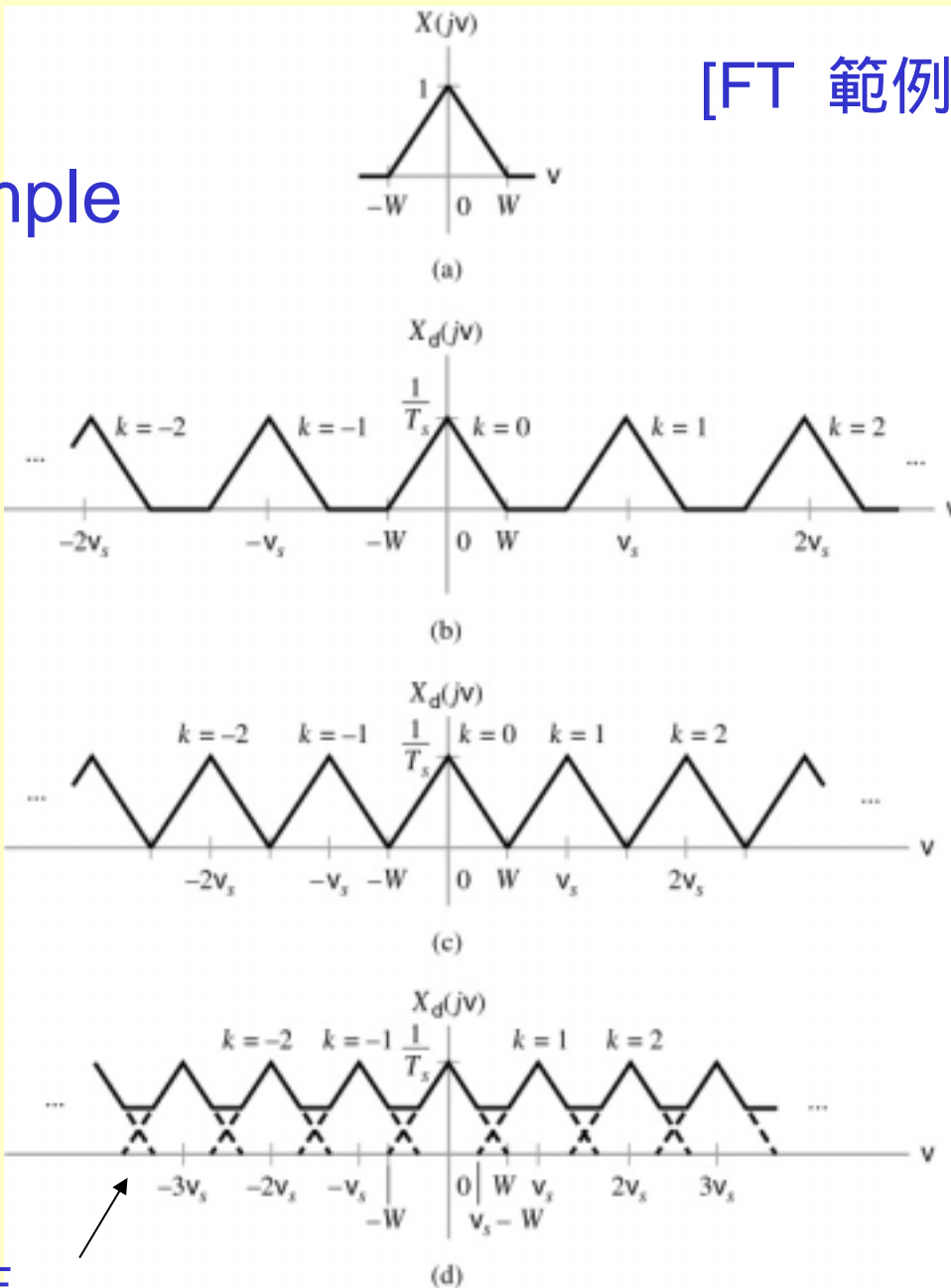


Aliasing in FT Example

[FT 範例]

$$\nu = \omega$$

$$\nu_s = \omega_s$$

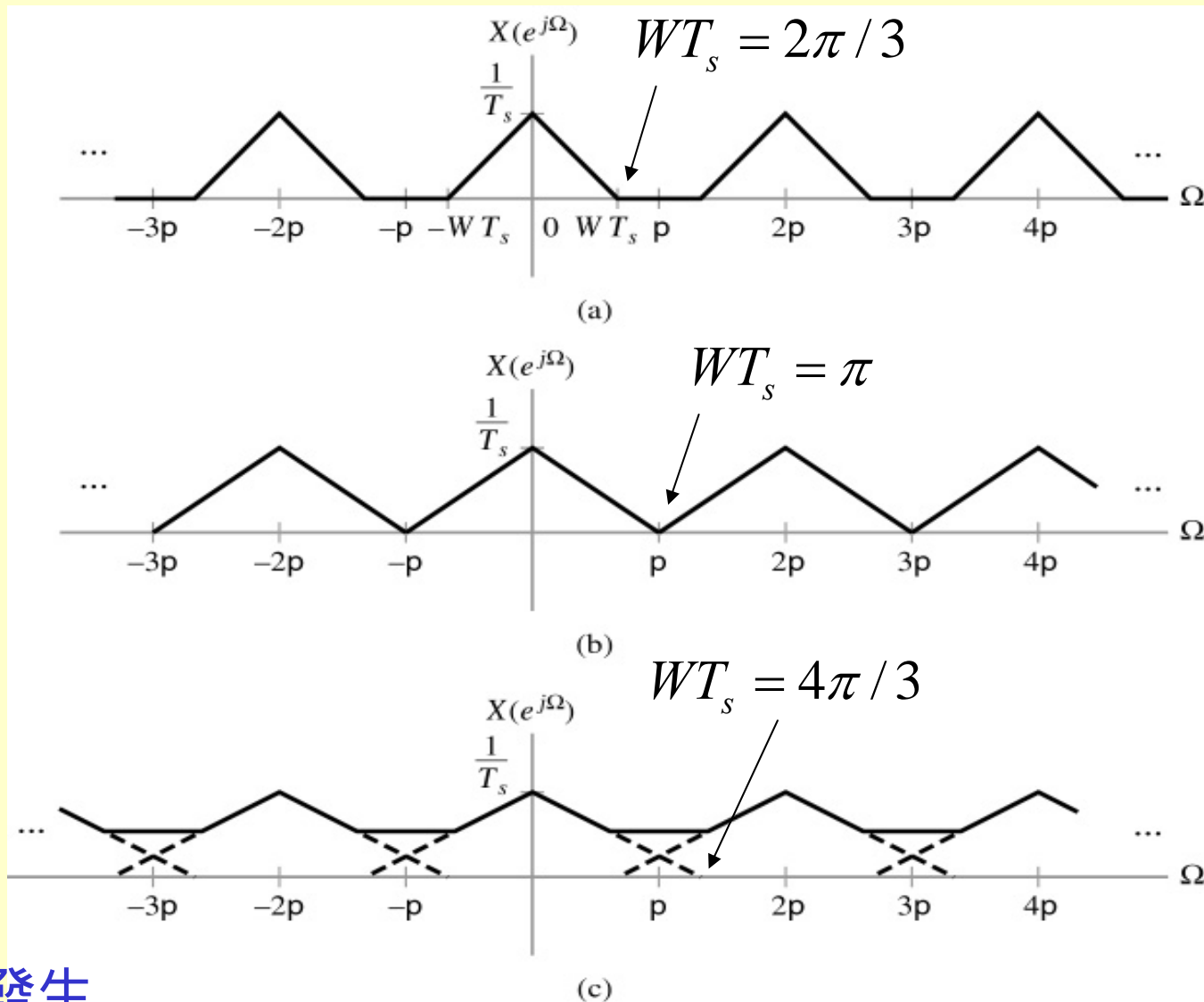


頻疊 Aliasing 發生



Aliasing in DTFT Example

$$\begin{aligned} \because \Omega &= \omega T_s \\ \therefore \Omega_s &= \omega_s T_s \\ &= \frac{2\pi}{T_s} T_s \\ &= 2\pi \end{aligned}$$



頻疊 Aliasing 發生

The DTFTs corresponding to the FTs, [DTFT 範例]
 (a) $\omega_s = 3W$. (b) $\omega_s = 2W$. (c) $\omega_s = 1.5W$.



The DTFTs corresponding to the FTs,
 (a) $\omega_s = 3W$. (b) $\omega_s = 2W$. (c) $\omega_s = 1.5W$.

$$\because \Omega = \omega T_s, \quad \therefore \Omega_s = \omega_s T_s = \frac{2\pi}{T_s} T_s = 2\pi$$

$$\because \text{FT spectrum bandwidth, } \omega_m = W$$

$$\therefore \text{DTFT spectrum bandwidth, } \Omega_m = \omega_m T_s = WT_s$$

$$(a) \quad \omega_s = 2\pi / T_s = 3W, \quad \therefore T_s = 2\pi / 3W$$

$$\therefore \underline{WT_s = 2\pi / 3}$$

$$(b) \quad \omega_s = 2\pi / T_s = 2W, \quad \therefore T_s = \pi / W$$

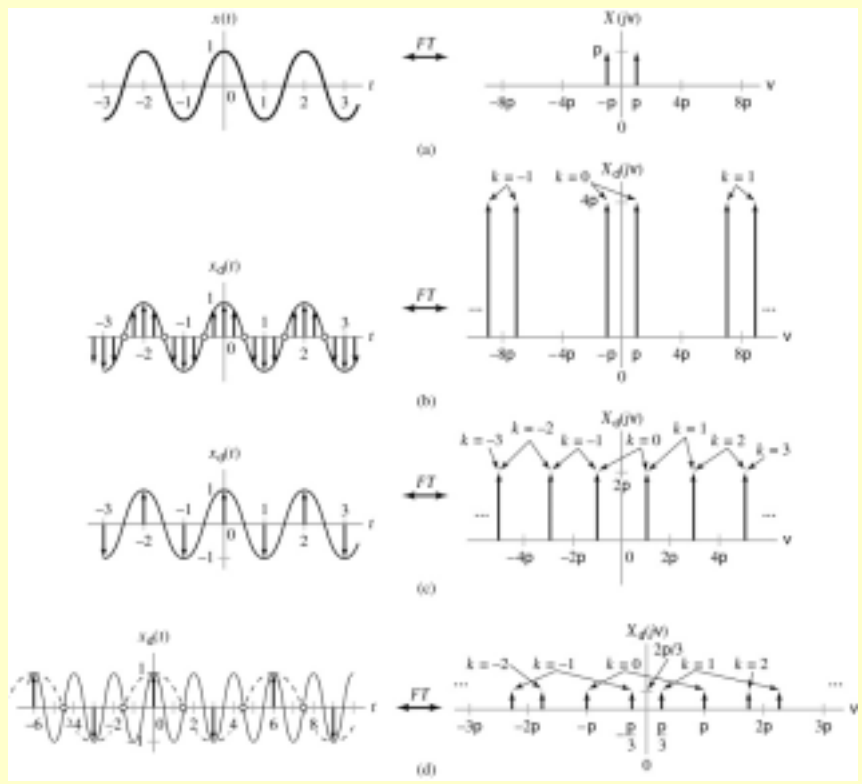
$$\therefore \underline{WT_s = \pi}$$

$$(c) \quad \omega_s = 2\pi / T_s = 3W / 2, \quad \therefore T_s = 4\pi / 3W$$

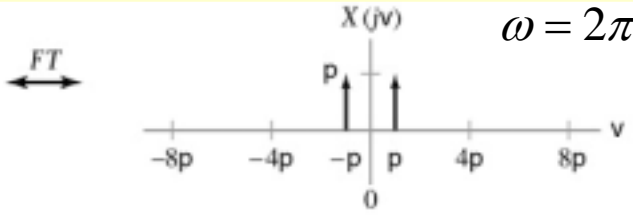
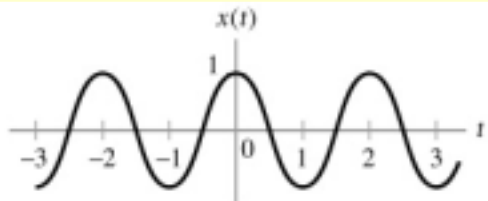
$$\therefore \underline{WT_s = 4\pi / 3}$$



Sampling a Sinusoid at Different Rates

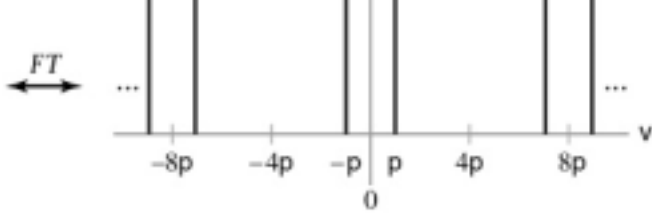
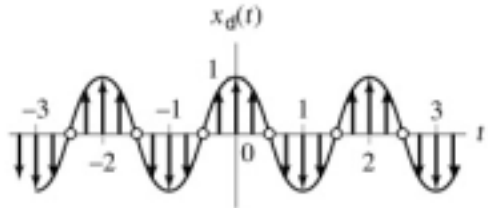


- The effect of sampling a sinusoid at different rates (Example 4.9). 對弦波取樣範例
- (a) Original signal and FT.
 - (b) Original signal, impulse sampled representation and FT for $T_s = 1/4$.
 - (c) Original signal, impulse sampled representation and FT for $T_s = 1$.
 - (d) Original signal, impulse sampled representation and FT for $T_s = 3/2$.
- A cosine of frequency $\pi/3$ is shown as the dashed line.



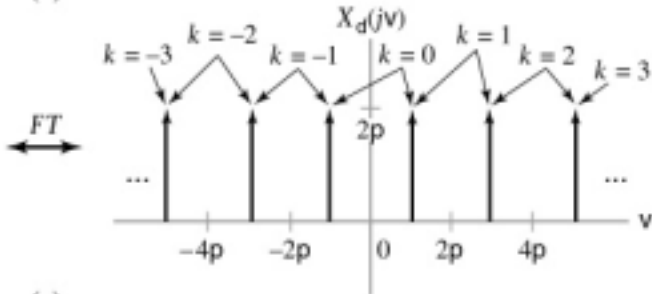
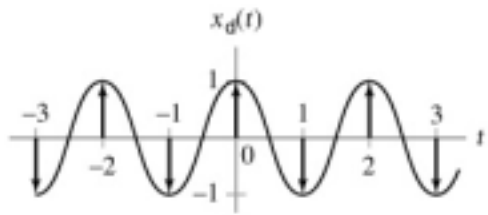
$$\omega = 2\pi / T = 2\pi / 2 = \pi$$

(a)



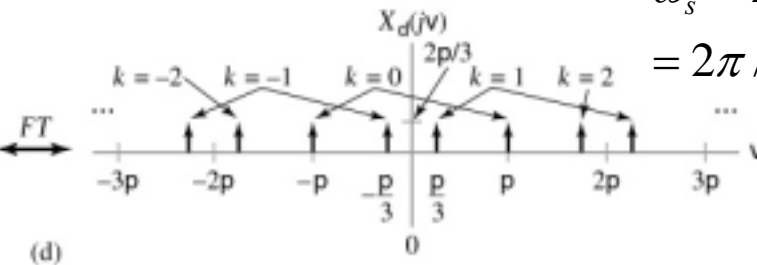
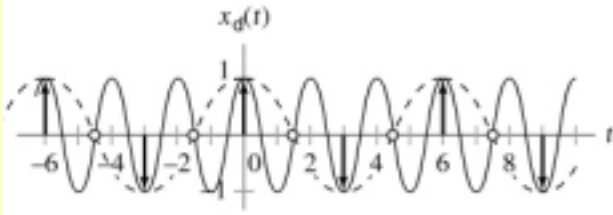
$$\omega_s = 2\pi / T_s = 2\pi / (1/4) = 8\pi$$

(b)



$$\omega_s = 2\pi / T_s = 2\pi / 1 = 2\pi$$

(c)



$$\omega_s = 2\pi / T_s = 2\pi / (3/2) = 4\pi / 3$$

(d)

$T_s = 1/4$

$T_s = 1$

$T_s = 3/2$

頻疊 Aliasing 發生



Aliasing in a Movie

- (a) Wheel rotating at ω (radians per second) and moving from right to left at v (meters per second).
- (b) Sequence of movie frames, assuming that the wheel rotates less than one-half turn between frames.
- (c) Sequences of movie frames, assuming that the wheel rotates between one-half and one turn between frames.
- (d) Sequence of movie frames, assuming that the wheel rotates one turn between frames.

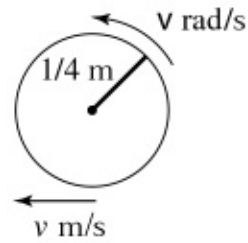


電影拍攝速度：30 frames/sec., $T_s = 1/30$ 秒

- (a) 輪子轉動 ω radians/sec. ($r = 1/4\text{m}$) ,
由右向左移動 v meters/sec. = $\omega r = \omega/4$ (m/sec.)
- (b) 電影拍攝取樣所拍到角度變化: $\omega T_s < \text{半圈}$ (i.e. $< \pi$).
 → $\omega < 30\pi$ (輪子轉動速率 $\omega < \text{電影拍速率的一半}$)
 → 正常
- (c) 電影拍攝取樣所拍到角度變化: $\text{半圈} < \omega T_s < \text{一圈}$ (i.e. $\pi < \omega T_s < 2\pi$).
 → $30\pi < \omega < 60\pi \rightarrow 30\pi/4 < \omega/4 < 60\pi/4$
 → $23.5 \text{ m/sec.} < v < 47.12 \text{ m/sec.}$
 → 輪子有反轉錯覺(頻疊產生)
- (d) 電影拍攝取樣所拍到角度變化: $\omega T_s = \text{一圈}$ (i.e. $\omega T_s = 2\pi$).
 → $\omega = 60\pi \rightarrow v = 47.12 \text{ m/sec.}$
 → 輪子有靜止錯覺(頻疊產生)



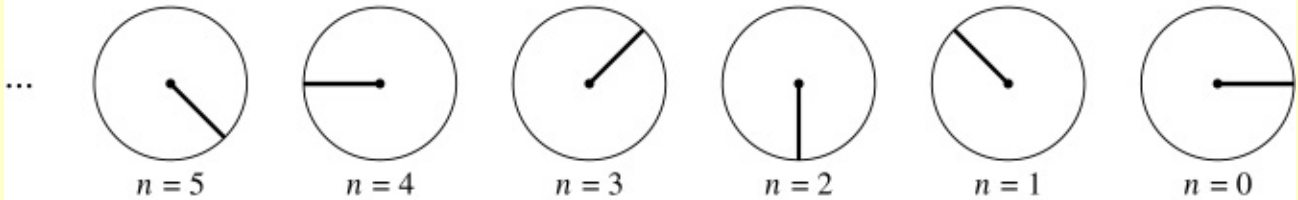
(a)



(a)

← direction of travel

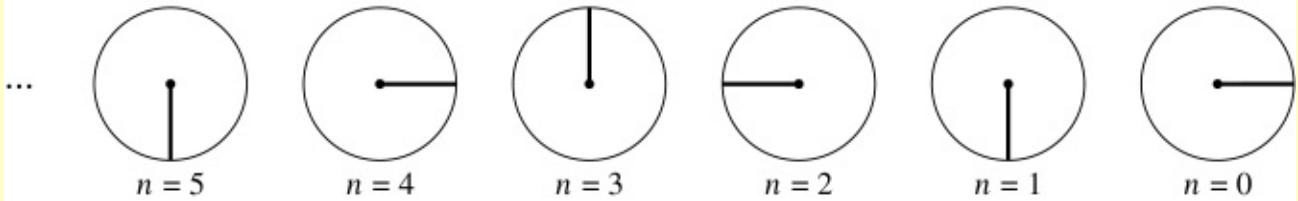
(b)



(b)

← direction of travel

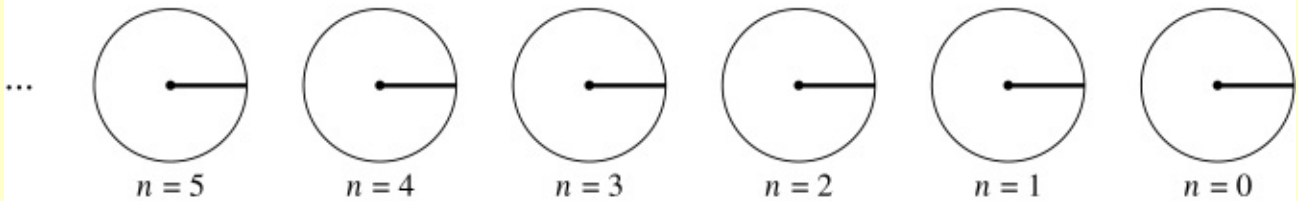
(c)



(c)

← direction of travel

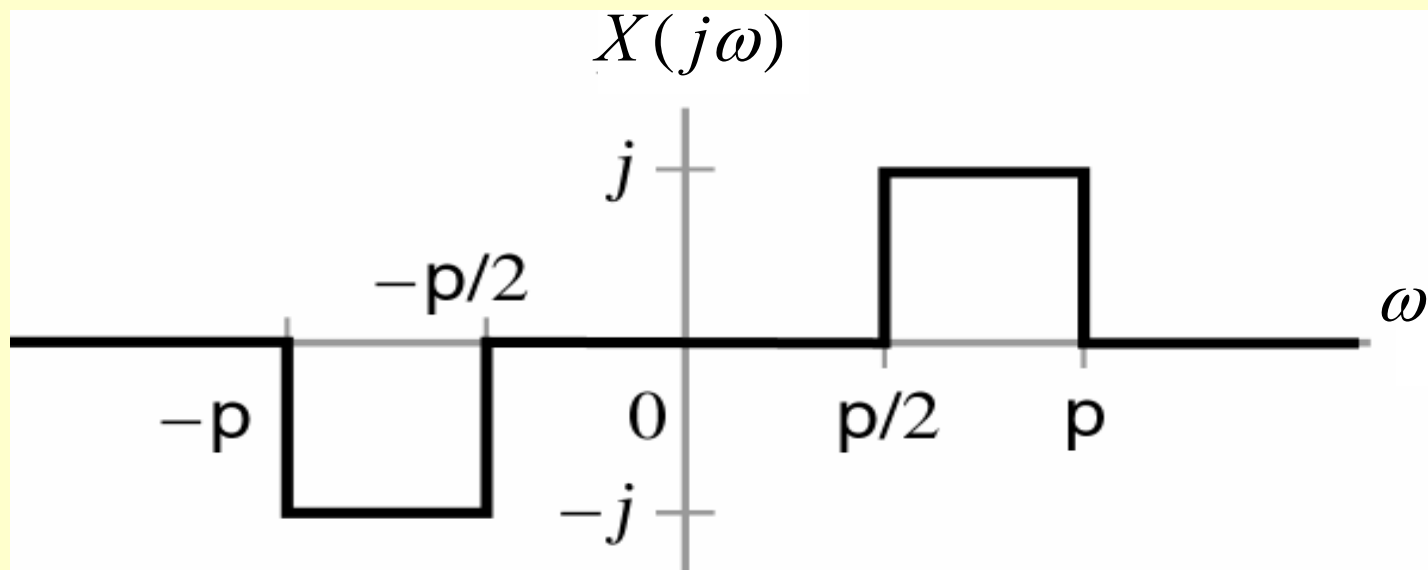
(d)



(d)



Problem 4.10 下圖為連續時間訊號的 FT 頻譜，試分別對
 (a) $T_s = 1/2$ (b) $T_s = 2$ 求取樣後的 FT 並繪出。



Sampled Spectra:
$$X_\delta(j\omega) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

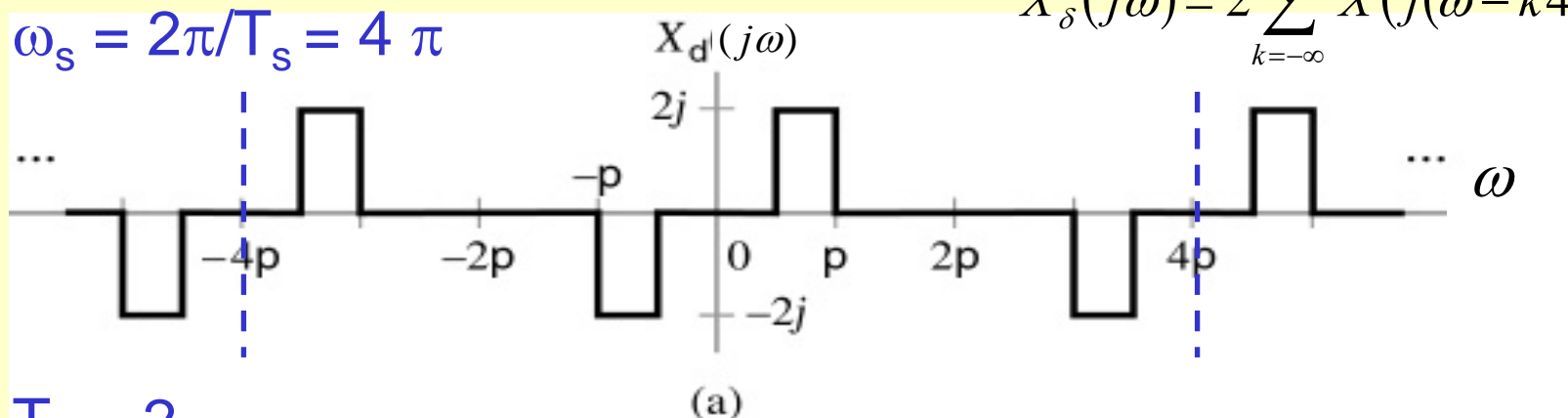


Solution to Problem 4.10

$$T_s = 1/2$$

$$\omega_s = 2\pi/T_s = 4\pi$$

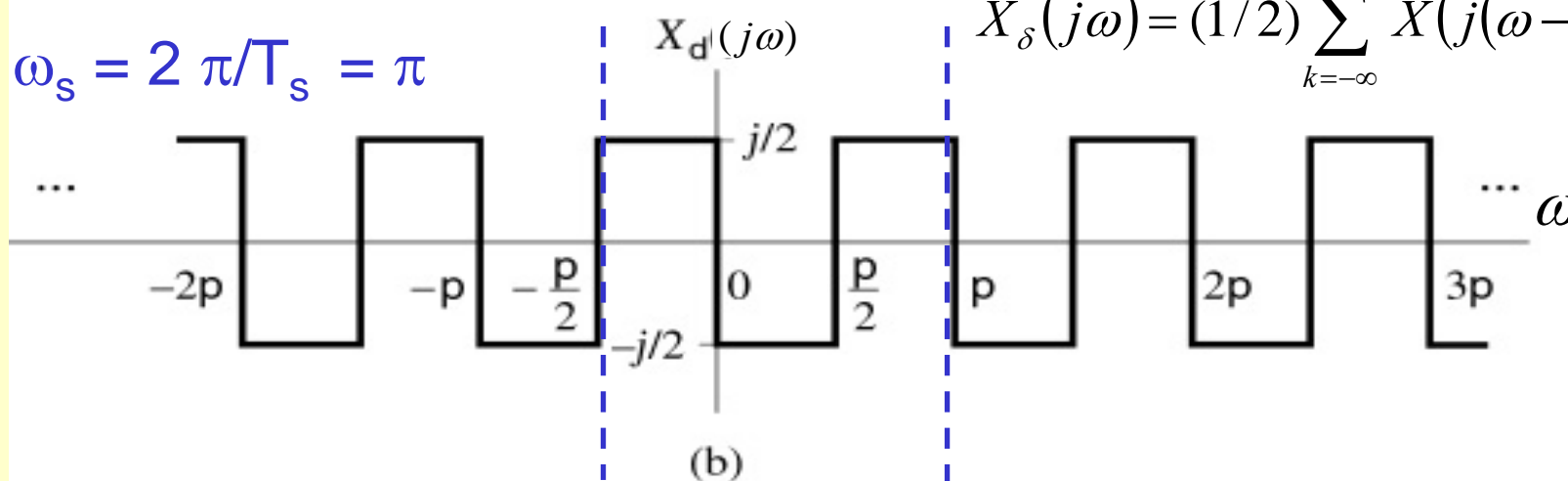
$$X_\delta(j\omega) = 2 \sum_{k=-\infty}^{\infty} X(j(\omega - k4\pi))$$



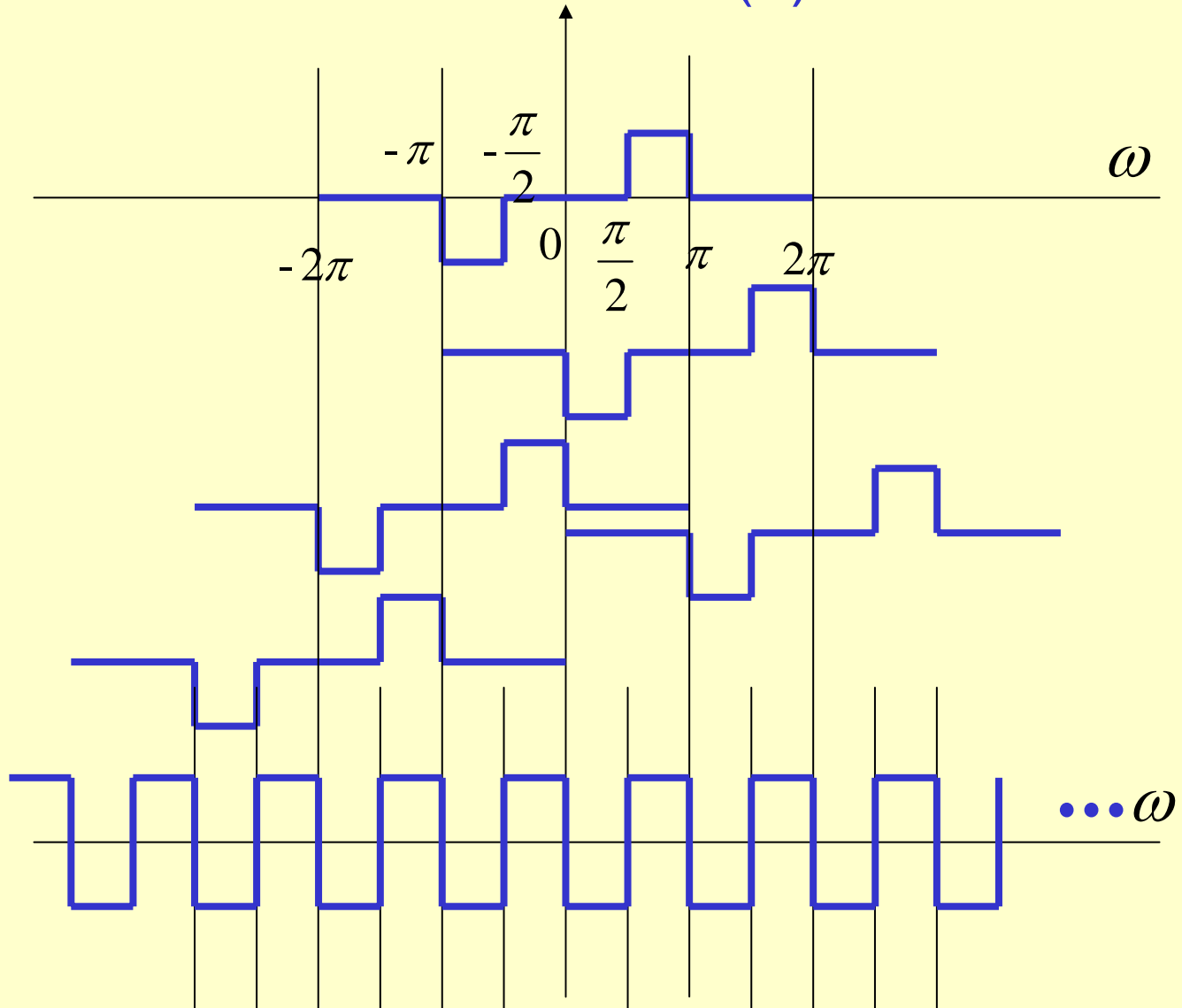
$$T_s = 2$$

$$\omega_s = 2\pi/T_s = \pi$$

$$X_\delta(j\omega) = (1/2) \sum_{k=-\infty}^{\infty} X(j(\omega - k\pi))$$



Problem 4.10(b)





Sub-Sampling: Sampling Discrete-Time Signals

次取樣：離散時間訊號的取樣

- $y[n] = x[q n]$ 為經過次取樣的樣本， q 需為正整數。
- $x[n]$ 原表示為連續性時間訊號 $x(t)$ 的離散樣本。
- $y[n]$ 也可表示為連續性時間訊號 $x(t)$ 的離散樣本，取樣間距為原取樣的 q 倍，仍需注意 aliasing 的發生與否。

$$x[n] = x(nT_s)$$

$$y[n] = x[nq] = x(nqT_s)$$



Problem 4.11

若 $q = 2$, or $q = 5$. 繪出經過次取樣後訊號 $y[n] = x[qn] = ?$

$$x[n] = 2 \cos\left(\frac{\pi}{3}n\right) = e^{j\frac{\pi}{3}n} + e^{-j\frac{\pi}{3}n}$$

$$\Leftrightarrow X(e^{j\Omega}) = 2\pi\delta\left(\Omega - \frac{\pi}{3}\right) + 2\pi\delta\left(\Omega + \frac{\pi}{3}\right)$$

< q = 2 case >

$$y[n] = x[2n] = 2 \cos\left(\frac{\pi}{3}2n\right) = e^{j\frac{\pi}{3}2n} + e^{-j\frac{\pi}{3}2n}$$

$$\Leftrightarrow Y(e^{j\Omega}) = \pi \delta\left(\Omega - \frac{2\pi}{3}\right) + \pi \delta\left(\Omega + \frac{2\pi}{3}\right), \quad -\pi < \Omega < +\pi$$



P 4.11 (cont.) $y[n] = x[5n]$

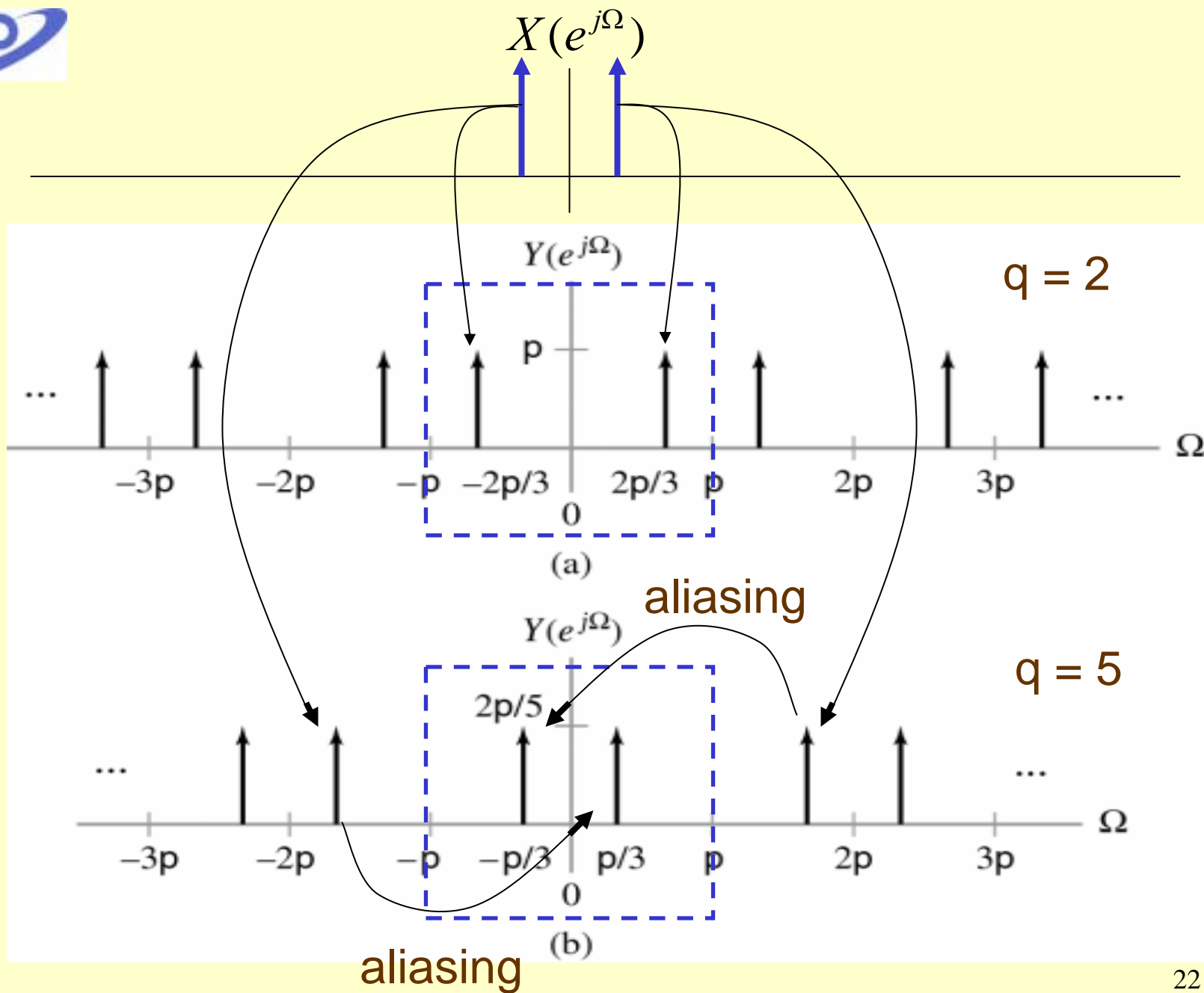
$$x[n] = 2 \cos\left(\frac{\pi}{3}n\right) = e^{j\frac{\pi}{3}n} + e^{-j\frac{\pi}{3}n}$$

$$\Leftrightarrow X(e^{j\Omega}) = 2\pi\delta\left(\Omega - \frac{\pi}{3}\right) + 2\pi\delta\left(\Omega + \frac{\pi}{3}\right)$$

$\langle q = 5 \text{ case} \rangle$

$$y[n] = x[5n] = 2 \cos\left(\frac{\pi}{3}5n\right) = e^{j\frac{\pi}{3}5n} + e^{-j\frac{\pi}{3}5n}$$

$$\begin{aligned} \Leftrightarrow Y(e^{j\Omega}) &= \frac{2\pi}{5} \delta\left(\Omega - \frac{5\pi}{3} \pm 2\pi\right) + \frac{2\pi}{5} \delta\left(\Omega + \frac{5\pi}{3} \pm 2\pi\right) \\ &= \frac{2\pi}{5} \delta\left(\Omega + \frac{\pi}{3}\right) + \frac{2\pi}{5} \delta\left(\Omega - \frac{\pi}{3}\right), \quad -\pi < \Omega < \pi \end{aligned}$$



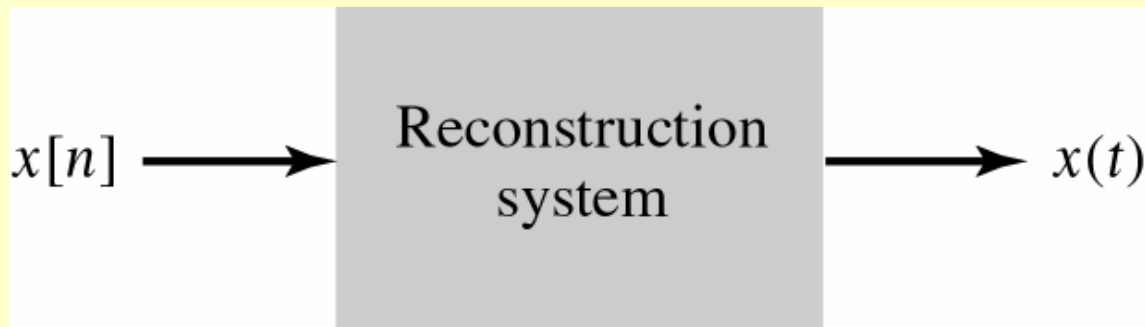


從離散樣本重建連續時間性訊號

取樣定理 (Sampling Theorem)

理想的訊號重建 (Ideal Reconstruction)

重建基本方法 - 零階保持器 (Zero-Order Hold)





Sampling Theorem

取樣定理：

令 $x(t) \xleftrightarrow{FT} X(j\omega)$ 表一頻寬受限的訊號，頻寬 ω_m

其頻譜 $X(j\omega) = 0, \quad \forall |\omega| > \omega_m$

取樣頻率 $\omega_s = 2\pi/T_s \geq 2 \omega_m$

(or $f_s \geq 2 f_m$)

奈奎斯特頻率(Nyquist frequency)： (取樣頻率的底線)

$\omega_n = 2\pi/T_s = 2 \omega_m$ (or $f_n = 2 f_m$)



Not One to One Mapping

離散樣本 $x[n]$ (火柴棒形狀) 有可能是來自於下述兩組連續時間訊號的任一信號。

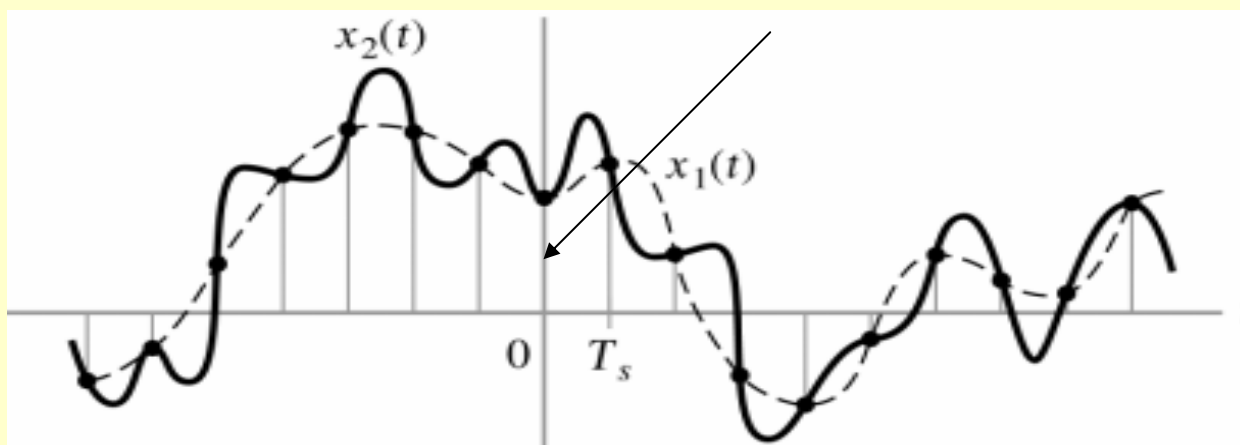
$$x[n] = x_1(nT_s) = x_2(nT_s)$$

連續時間與離散時間訊號無法一對一對應，
頻疊 (aliasing) 因而產生。

$x_1(t)$ -----

$x_2(t)$ _____

$x[n]$





Example 4.12

假設 $x(t) = \frac{\sin(\pi t)}{\pi t}$ ，試求取樣間距應滿足的條件 $T_s = ?$ ，使得 $x[n] = x(nT_s)$ 能唯一的表示。

Solution:

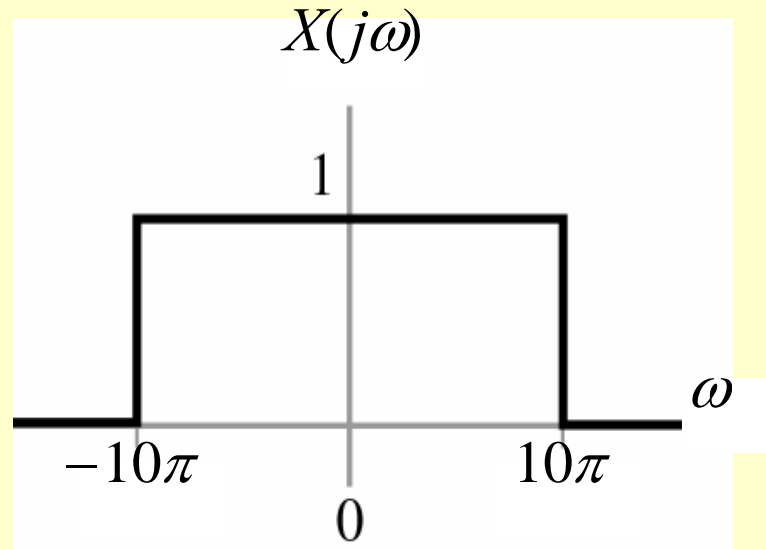
$$\therefore x(t) = \frac{\sin(\pi t)}{\pi t} \stackrel{FT}{\leftrightarrow} X(j\omega) = \begin{cases} 1, & |\omega| \leq 10\pi \\ 0, & |\omega| > 10\pi \end{cases}$$

$$\therefore \omega_m = 10\pi$$

$$\therefore \frac{2\pi}{T_s} \geq 2\omega_m = 20\pi,$$

$$\therefore \frac{1}{T_s} \geq 10,$$

$$\therefore T_s < 0.1$$





Problem 4.12

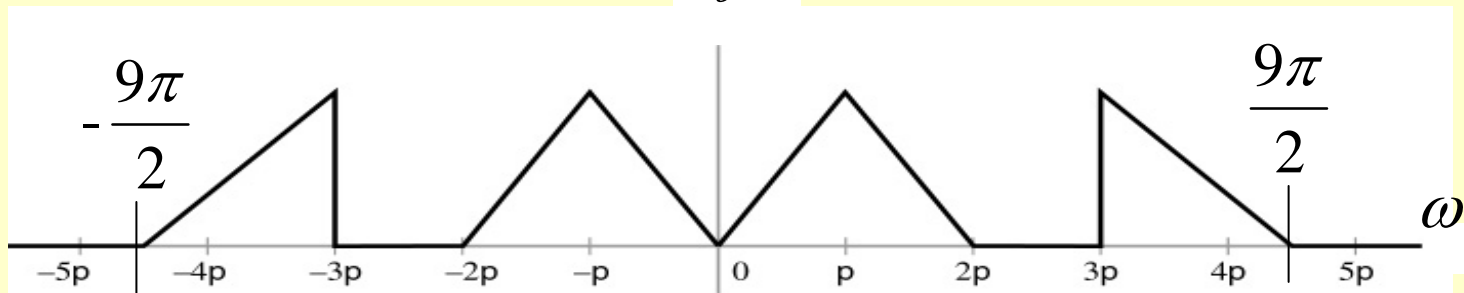
試求取樣間距應滿足的條件 $T_s = ?$, 使得 $x[n] = x(nT_s)$ 能唯一的表示 : 亦即無頻疊發生 (下圖為 $x(t)$ 的 FT)

Solution:

$$\therefore \omega_m = \frac{9\pi}{2}, \quad \frac{2\pi}{T_s} \geq 2\omega_m = 9\pi,$$

$$\therefore \frac{2}{T_s} \geq 9, \quad \therefore T_s < \frac{2}{9}.$$

$X(j\omega)$





理想的訊號重建 (Ideal Reconstruction)

取樣定理指出我們須以多快的速率取樣，才可以使樣本唯一代表連續時間訊號。

時域取樣過程： (原始連續訊號與脈衝串列的乘積)

$$\begin{aligned}x_{\delta}(t) &= x(t) \cdot p(t) = x(t) \cdot \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \\ &= \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s) \\ &= \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s)\end{aligned}$$



理想的訊號重建 (cont.)

頻域取樣過程：

樣本訊號的 FT 是原來訊號的 FT 經過不同頻率平移之後的無窮和。(若 ω_s 不夠大，頻率平移版本有可能會重疊)

$$\begin{aligned} X_\delta(j\omega) &= \frac{1}{2\pi} X(j\omega) * P(j\omega) \\ &= \frac{1}{2\pi} X(j\omega) * \frac{2\pi}{T_s} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) \\ &= \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s)) \end{aligned}$$



重建的目標是對 $X_\delta(j\omega)$ 做運算以便轉換成 $X(j\omega)$ ，這樣運算用來刪除下式中的 $k\omega_s$ 項。

$$X_\delta(j\omega) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

欲轉換成 $X(j\omega)$ ，需做下述乘積：Low Pass Filtering

$$X(j\omega) = X_\delta(j\omega) \cdot H_r(j\omega)$$

其中

$$H_r(j\omega) = \begin{cases} T_s, & |\omega| \leq \omega_s / 2 \\ 0, & |\omega| > \omega_s / 2 \end{cases}$$



Time Signal Reconstruction

時域重建運算：

$$\begin{aligned}
 x(t) &= x_{\delta}(t) * h_r(t) = h_r(t) * x_{\delta}(t) \\
 &= h_r(t) * \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s) = \sum_{n=-\infty}^{\infty} x[n] h_r(t - nT_s)
 \end{aligned}$$

$$= \sum_{n=-\infty}^{\infty} x[n] \operatorname{sinc}(\omega_s (t - nT_s) / (2\pi))$$

其中

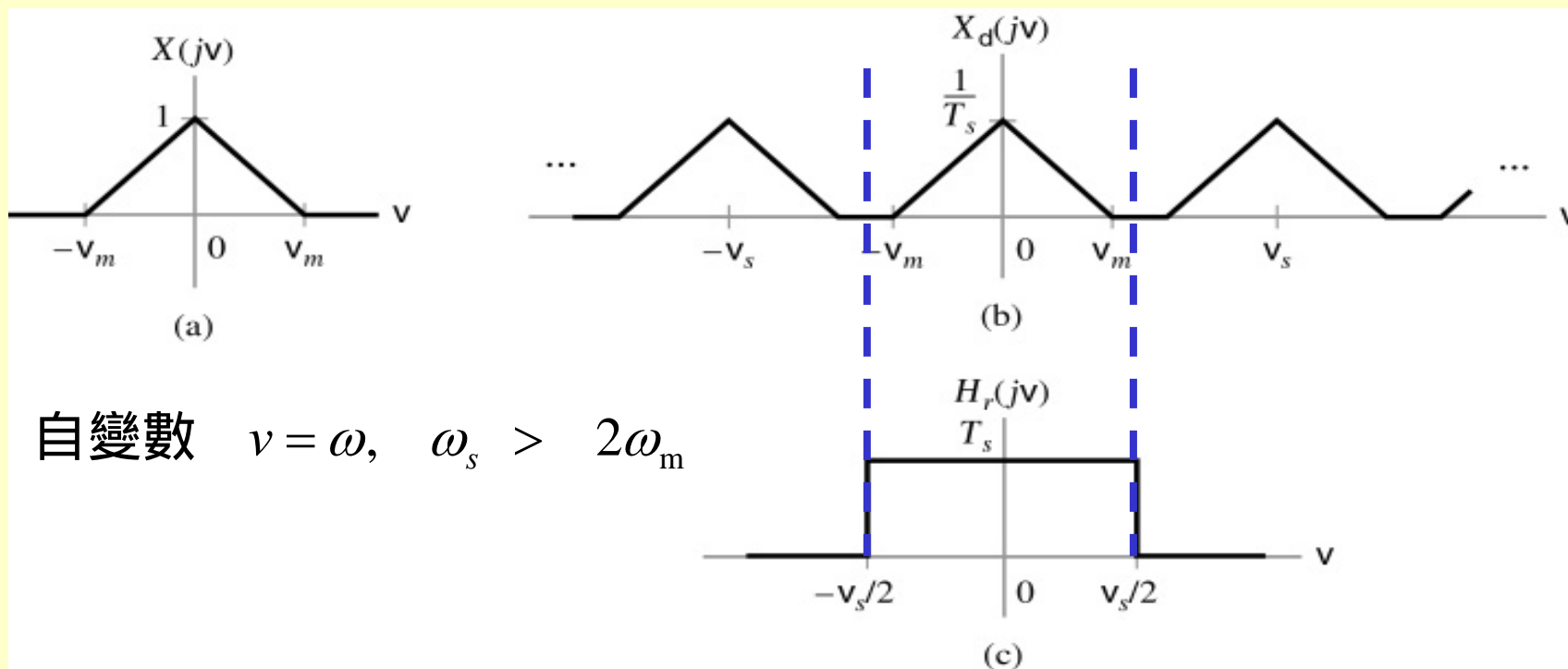
$$h_r(t) = \frac{T_s \sin\left(\frac{\omega_s}{2} t\right)}{\pi t}$$

$$\operatorname{sinc}(x) = \frac{\sin(\pi x)}{\pi x}$$



Ideal Reconstruction

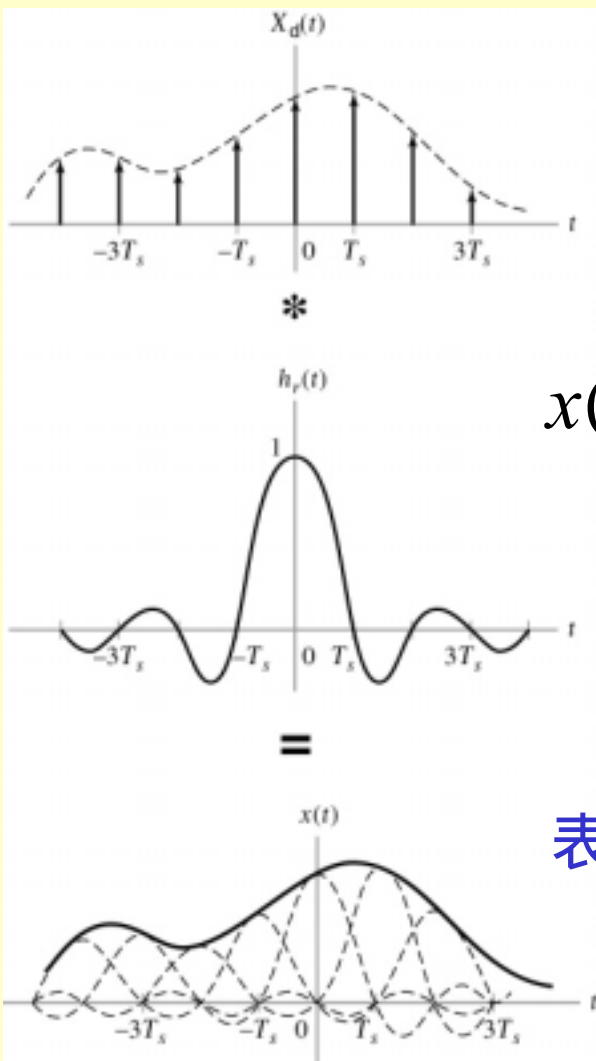
原訊號理想重建



- (a) Spectrum of original signal.
- (b) Spectrum of sampled signal.
- (c) Frequency response of reconstruction filter.



Ideal Reconstruction in the Time-Domain



$$x(t) = x_{\delta}(t) * h_r(t) = \sum_{n=-\infty}^{\infty} x[n] h_r(t - nT_s)$$

$$= \sum_{n=-\infty}^{\infty} x[n] \text{sinc}(\omega_s(t - nT_s)/(2\pi))$$

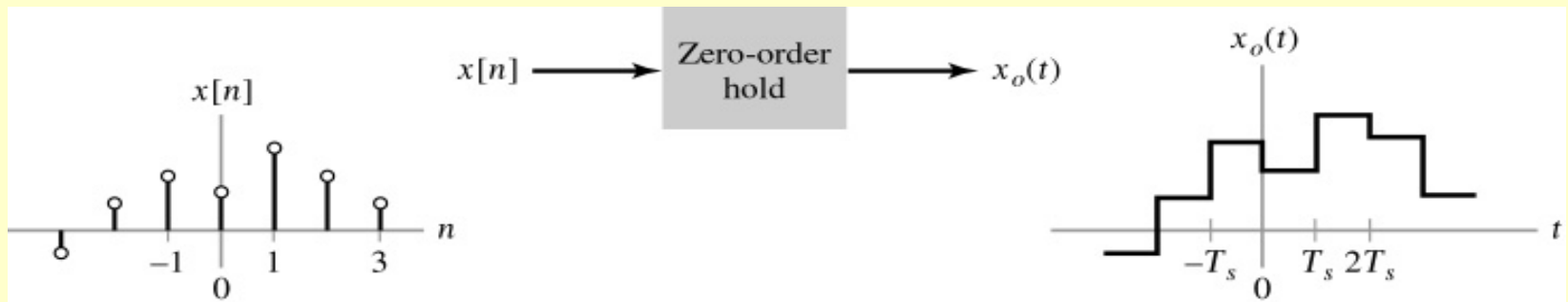
表示為許多 sinc 函數的時間平移加權疊加



Zero- Order Holder

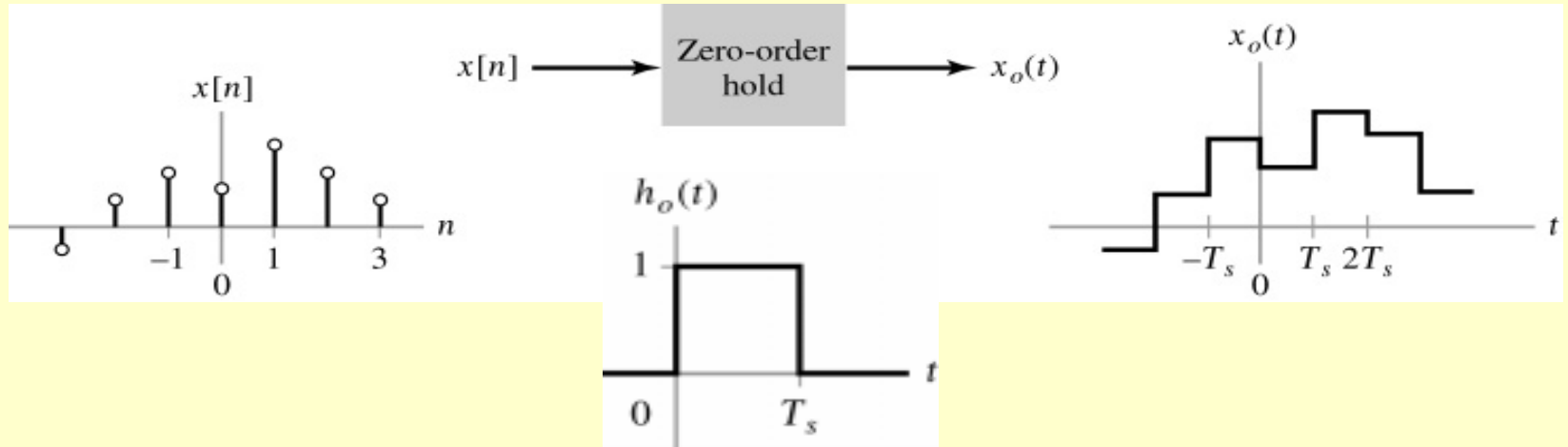
重建基本方法-零階保持器 (Zero-Order Hold)

零階保持器將數值 $x[n]$ 維持或保留 T_s 秒的時間，產生一個階梯式近似連續訊號





Reconstruction via a Zero-Order Holder



重建的訊號： $h_0(t)$ 脈波的時間平移與加權疊加

$$x_o(t) = x_\delta(t) * h_0(t) = \sum_{n=-\infty}^{\infty} x[n] h_0(t - nT_s)$$



Reconstruction in Frequency

頻域考量：

$$\begin{aligned} X_0(j\omega) &= H_0(j\omega) \cdot X_\delta(j\omega) \\ &= \left[2e^{-j\omega T_s/2} \frac{\sin(\omega T_s/2)}{\omega} \right] \cdot X_\delta(j\omega) \end{aligned}$$

$H(j\omega)$ 頻率響應考量：

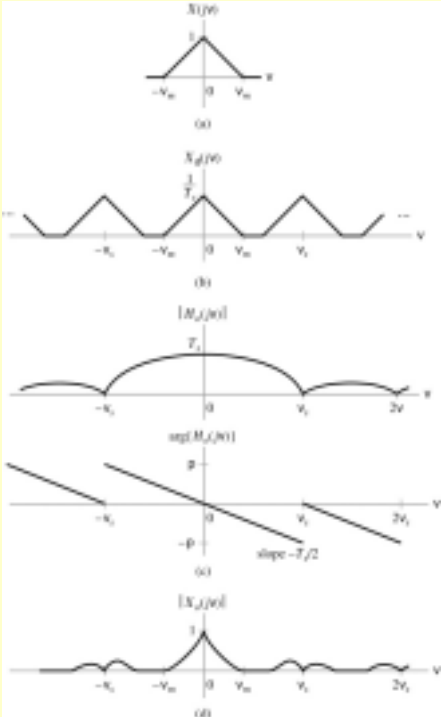
振幅響應主瓣(main lobe) 曲率造成非線性失真

$H(j\omega)$ 相位響應考量：

線性相位位移相當於 $T_s/2$ 秒時間延遲



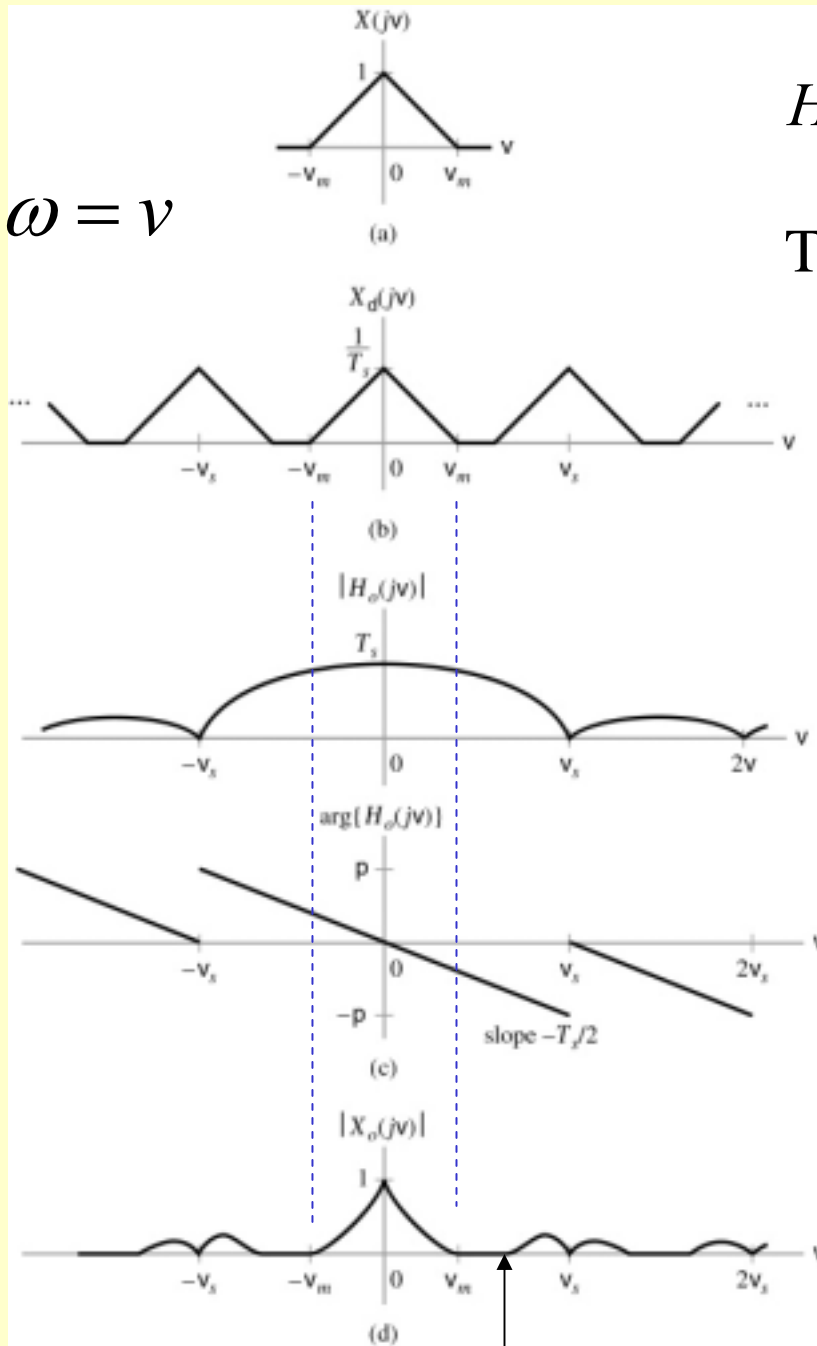
Effect of the Zero-Order Hold in the Frequency Domain



- (a) Spectrum of original continuous-time signal.
- (b) FT of sampled signal.
- (c) Magnitude and phase of $H_0(j\omega)$.
- (d) Magnitude spectrum of signal reconstructed using zero-order hold.



$\omega = \nu$



$$H_o(j\omega) = \left[2e^{-j\omega T_s/2} \frac{\sin(\omega T_s / 2)}{\omega} \right]$$

The 1st zero crossing: $\omega = \pm \omega_s$

$$\omega_s - \omega_m$$



Zero-Order Holder Disadvantages

零階保持器重建連續時間訊號引入三項被修改處：

1. 線性相位位移相當於 $T_s/2$ 秒時間延遲
2. 在 $-\omega_m$ 和 ω_m 之間有一段 $X_\delta(j\omega)$ 的失真 (因main lobe 曲率所造成)
3. 其它部份 ω_s 倍數處有失真與衰減

上述問題可經由通過一個補償濾波器來改善

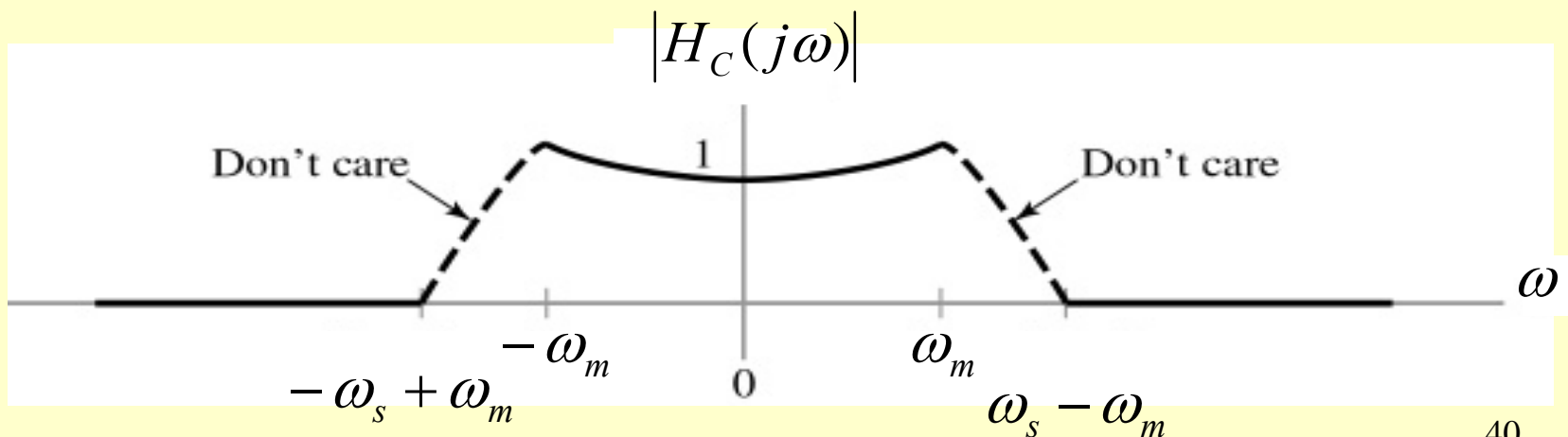
Compensation Filter (or Anti-imaging Filter)



The Compensation Filter (anti-imaging filter)

The Compensation Filter (anti-imaging filter) used to eliminate some of the distortion introduced by the zero-order hold.

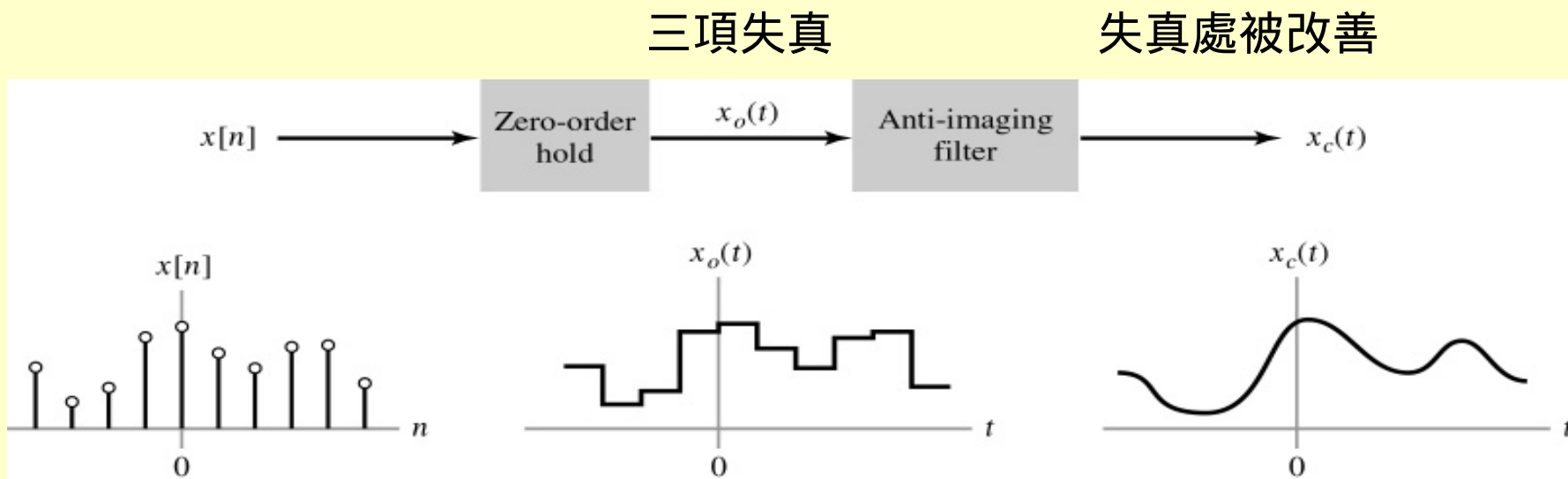
$$H_c(j\omega) = \begin{cases} \frac{\omega T_s}{2 \sin(\omega T_s / 2)}, & |\omega| < \omega_m \\ 0, & |\omega| > \omega_s - \omega_m \end{cases}$$





Block Diagram of a Practical Reconstruction System (重建系統)

- 零階保持器 重建連續訊號時，產生三項失真
- 抗映像濾波器 去除高頻映像頻譜並改善非線性失真





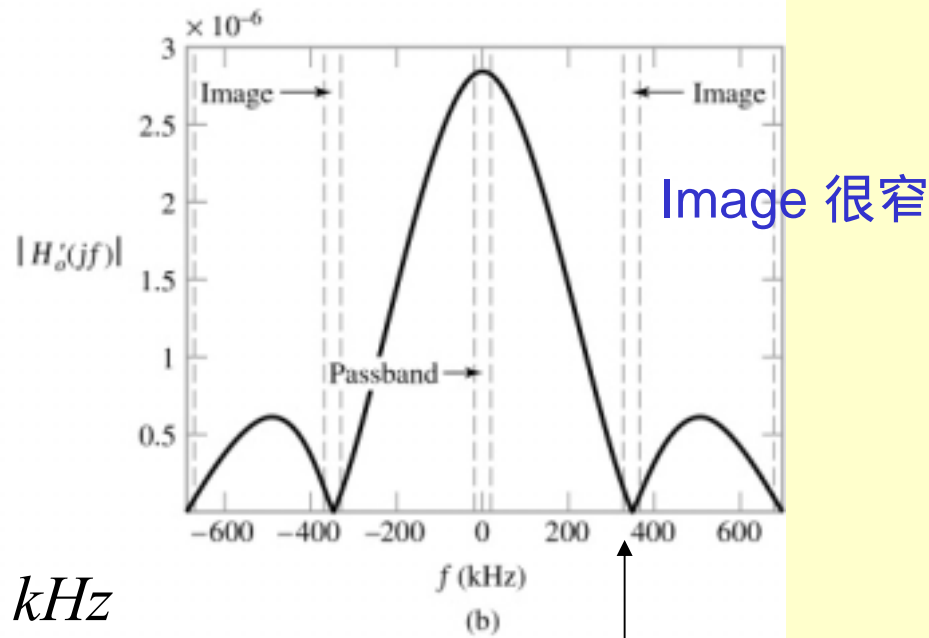
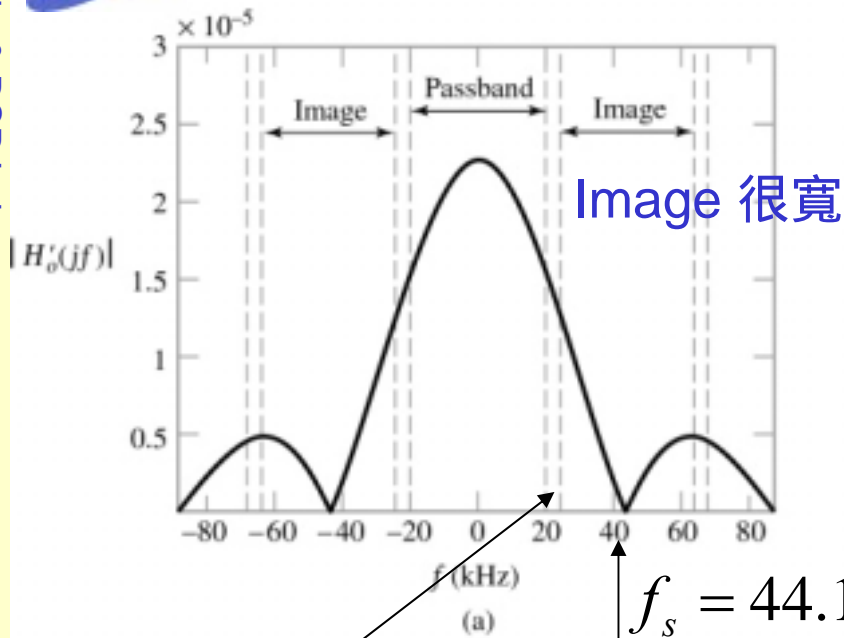
Anti-Imaging Filter Design With and Without Over-Sampling

Anti-imaging filter design with and without over-sampling.
找出 抗映像濾波器限制

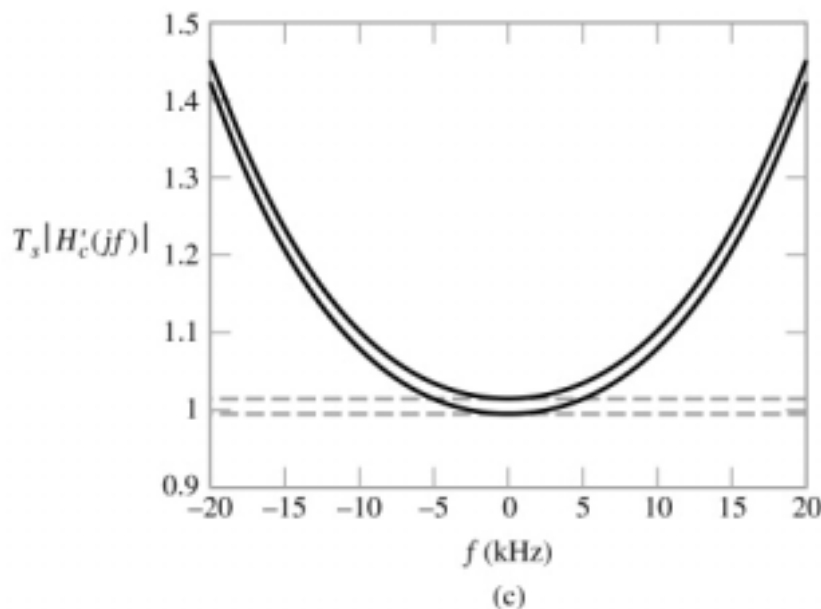
(a) Magnitude of $H_o(jf)$ for 44.1-kHz sampling rate. Dashed lines denote signal pass-band and images.

(b) Magnitude of $H_o(jf)$ for eight-times over-sampling (352.8-kHz sampling rate). Dashed lines denote signal pass-band and images.

(c) Normalized constraints on pass-band response of anti-imaging filter. Solid lines assume a 44.1-kHz sampling rate; dashed lines assume eight-times over-sampling. The normalized filter response must lie between each pair of lines.



$$\begin{aligned}
 f_s - f_m &= 44.1 - 20 \\
 &= 22.1 \text{ kHz}
 \end{aligned}$$



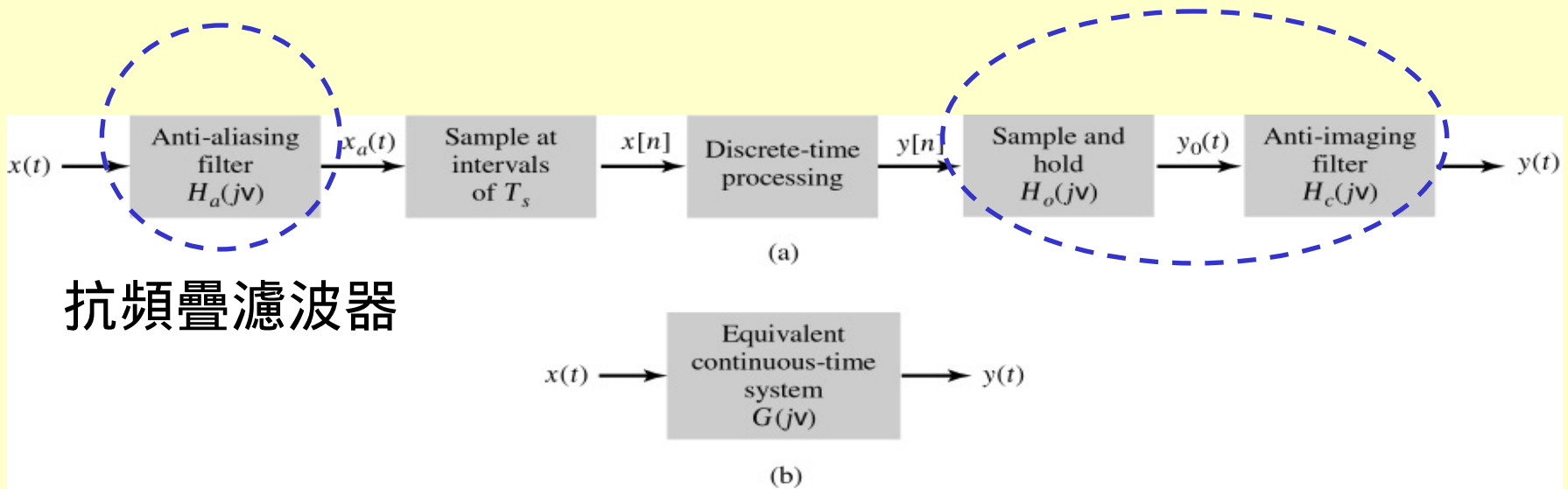
考慮：
Over sampling



Block Diagram for Discrete-Time Processing of Continuous-Time Signals

(a) A basic system.

(b) Equivalent continuous-time system.





Effect of Over-Sampling on Anti-Aliasing Filter Specifications (抗頻疊濾波器)

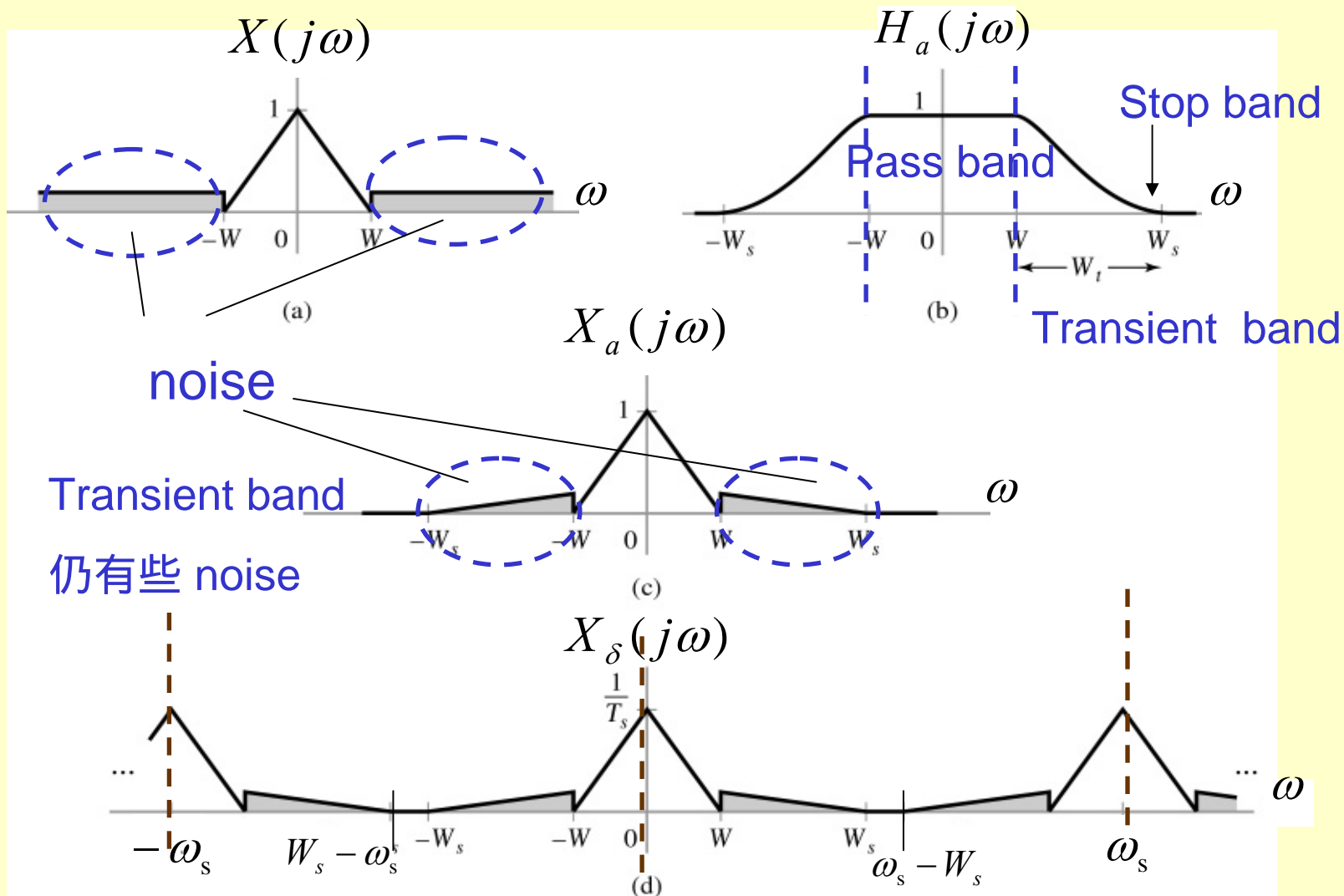
Effect of over-sampling on Anti-aliasing filter specifications:
(抗頻疊濾波器)

- (a) Spectrum of original signal.
- (b) Anti-aliasing filter frequency response magnitude.
- (c) Spectrum of signal at the anti-aliasing filter output.
- (d) Spectrum of the anti-aliasing filter output after sampling. The graph depicts the case of $\omega_s > 2W_s$.

抗頻疊濾波器作用在於使非限頻訊號成為限頻訊號



Anti-aliasing Filter可限制頻寬





抗頻疊濾波器(anti-aliasing filter) 在訊號取樣前，限制訊號頻寬，才可使用取樣理論。

為避免雜訊和訊號本身產生頻疊，要求： $\omega_s - W_s > W_s$

避免雜訊和訊號 pass band 產生頻疊，更嚴格要求：

$$\omega_s - W_s > W$$

$$\because W_s = W + W_t$$

$$\therefore \omega_s - (W + W_t) > W$$

$$\omega_s - W_t > 2W \quad \text{or} \quad W_t < \omega_s - 2W$$

- 因而形成過渡頻帶 W_t 的限制條件 (限制嚴格則成本高)
- 提高 ω_s 與降低 W 可降低過渡頻帶的限制，進而節省成本
- Decimation & Interpolation 則可改變 W 頻寬



Decimation (十分法)

若以 取樣間距 T_{s1} 取樣 $x(t)$ 訊號 獲得 $x_1[n]$

若以 取樣間距 T_{s2} 取樣 $x(t)$ 訊號 獲得 $x_2[n]$

其中 取樣間距 $T_{s1} = q T_{s2}$, q 是整數

效果如下：

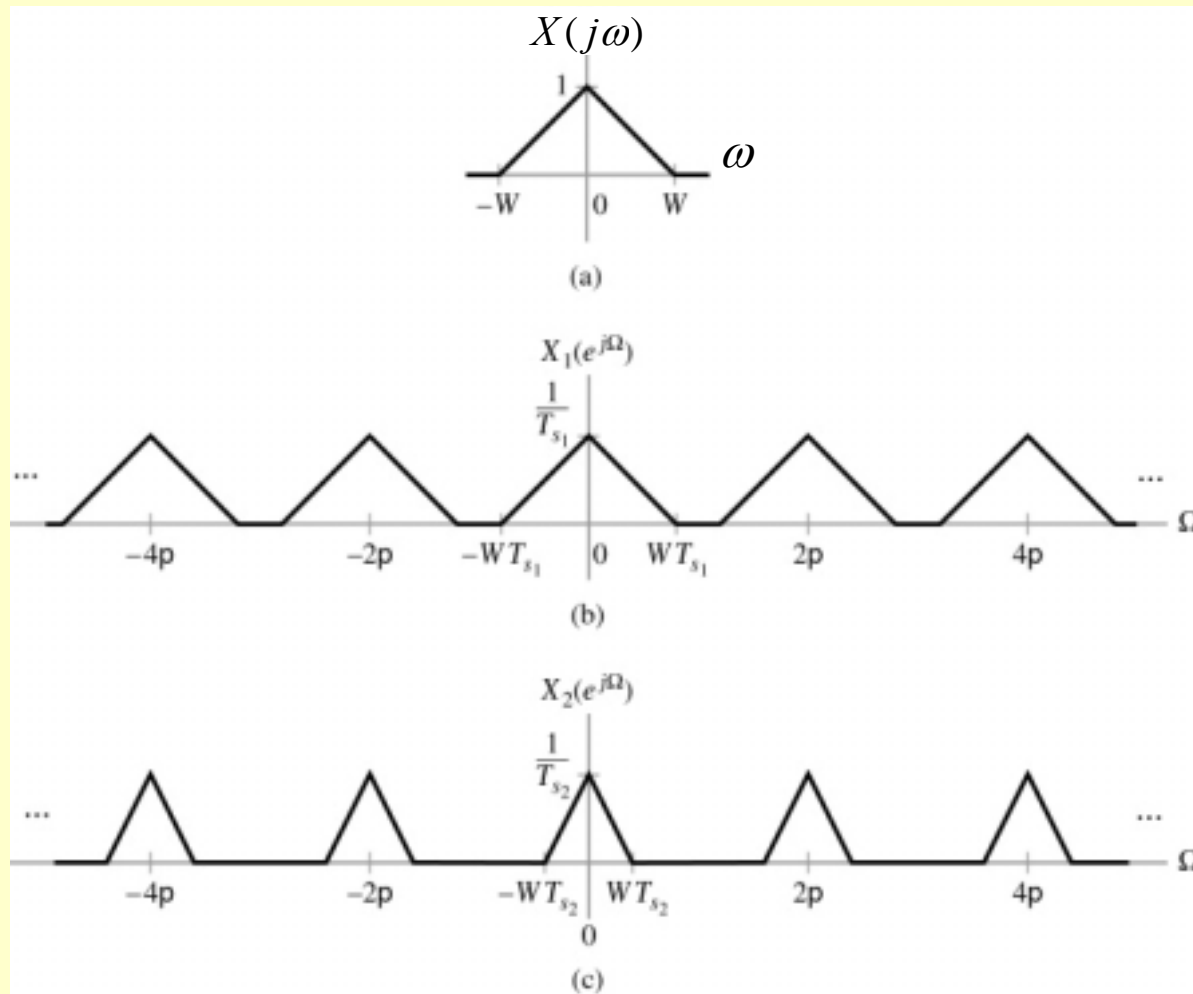
Decimation 將取樣間距 T_{s2} 增加為 T_{s1} (間距擴大)

Decimation 減少取樣比率 (WT_s 上升)

Decimation 取樣將頻譜 $X_2(e^{j\Omega})$ 改成 $X_1(e^{j\Omega})$

Decimation 取樣將訊號頻譜的頻寬 wT_{s2} 增加為 wT_{s1}

Decimation 取樣將原始訊號頻率增加 q 倍



Effect of changing the sampling rate.

(a) Underlying continuous-time signal FT.

(b) DTFT of sampled data at sampling interval T_{s1} .

(c) DTFT of sampled data at sampling interval T_{s2} .

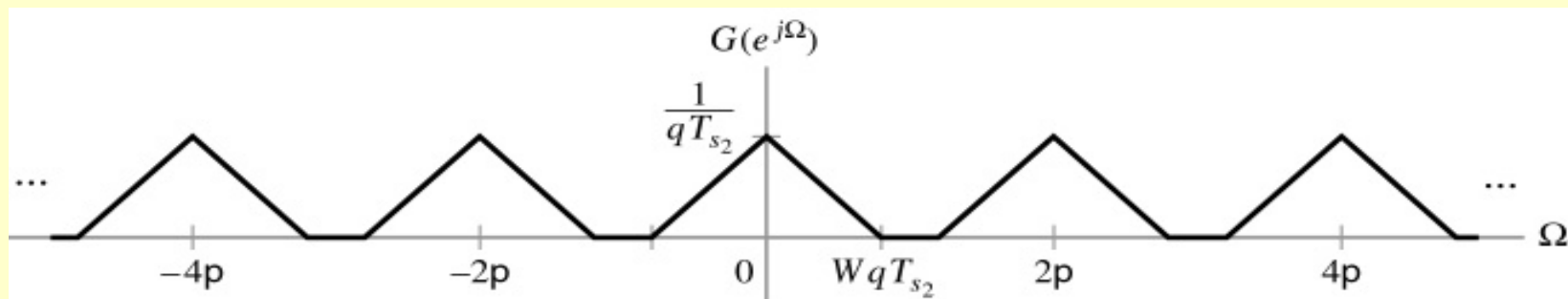


The Spectrum Results from Sub-Sampling the DTFT $X_2(e^{j\Omega})$ by a Factor of q

$g[n] = x(qT_s n)$ 訊號可由取樣 $x_2[n] = x(nT_s)$ 獲得

$$x_2[n] = x(nT_s), \quad g[n] = x_2[qn]$$

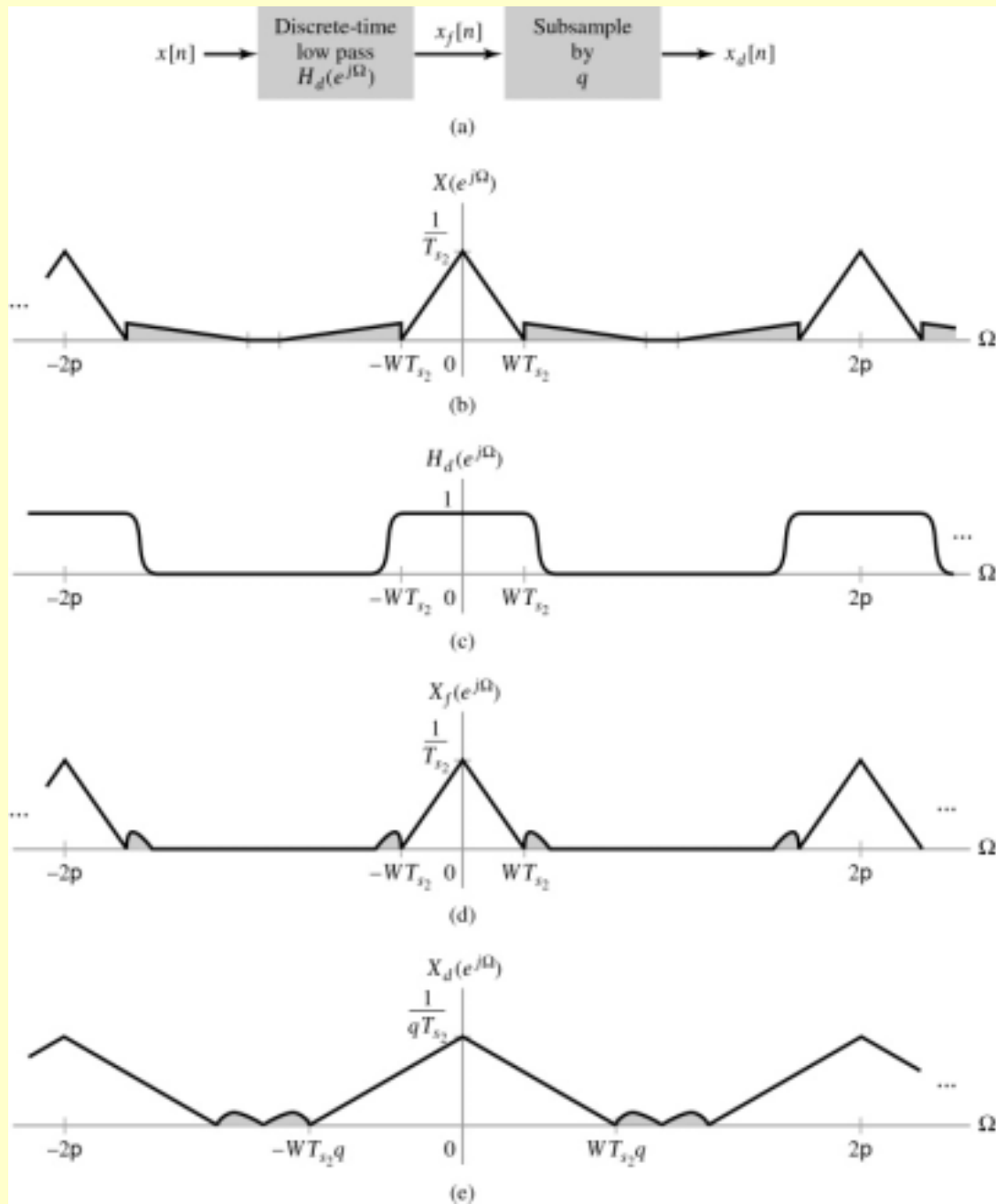
$$G(e^{j\Omega}) = \frac{1}{q} \sum_{m=0}^{q-1} X_2\left(e^{j((\Omega - m2\pi)/q)}\right)$$





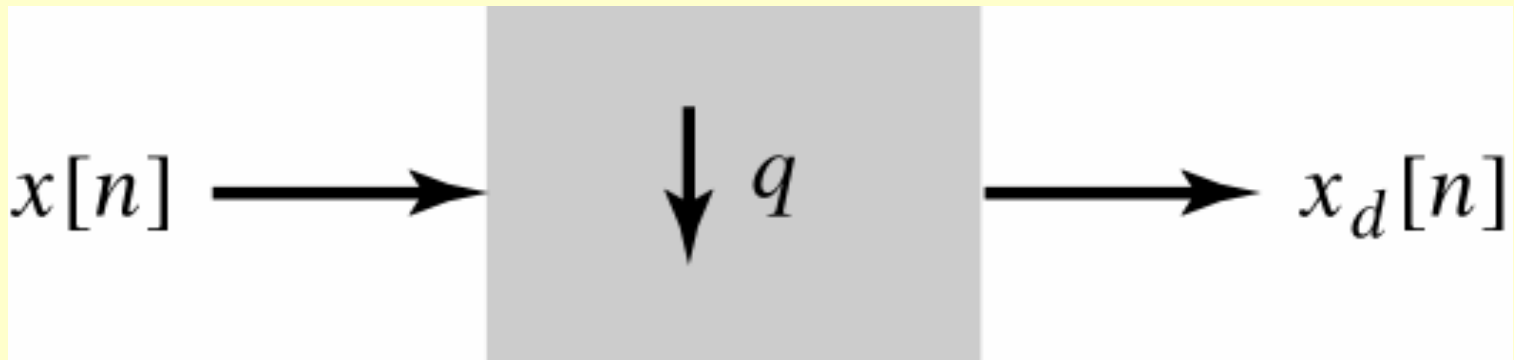
Frequency-domain interpretation of decimation

- (a) Block diagram of decimation system.
- (b) Spectrum of over-sampled input signal. Noise is depicted as the shaded portions of the spectrum.
- (c) Filter frequency response.
- (d) Spectrum of filter output.
- (e) Spectrum after sub-sampling.





Symbol for Decimation by a Factor of q





內插法取樣 (Interpolation)

以取樣間距 T_{s1} 取樣 $x(t)$ 訊號 獲得 $x_1[n]$

以取樣間距 T_{s2} 取樣 $x(t)$ 訊號 獲得 $x_2[n]$

取樣間距 $T_{s1} = q T_{s2}$, q 是整數

內插 取樣將取樣間距 T_{s1} 減少為 T_{s2}

內插 取樣增加取樣比率

內插 取樣將頻譜 $X_1(e^{j\Omega})$ 改成 $X_2(e^{j\Omega})$

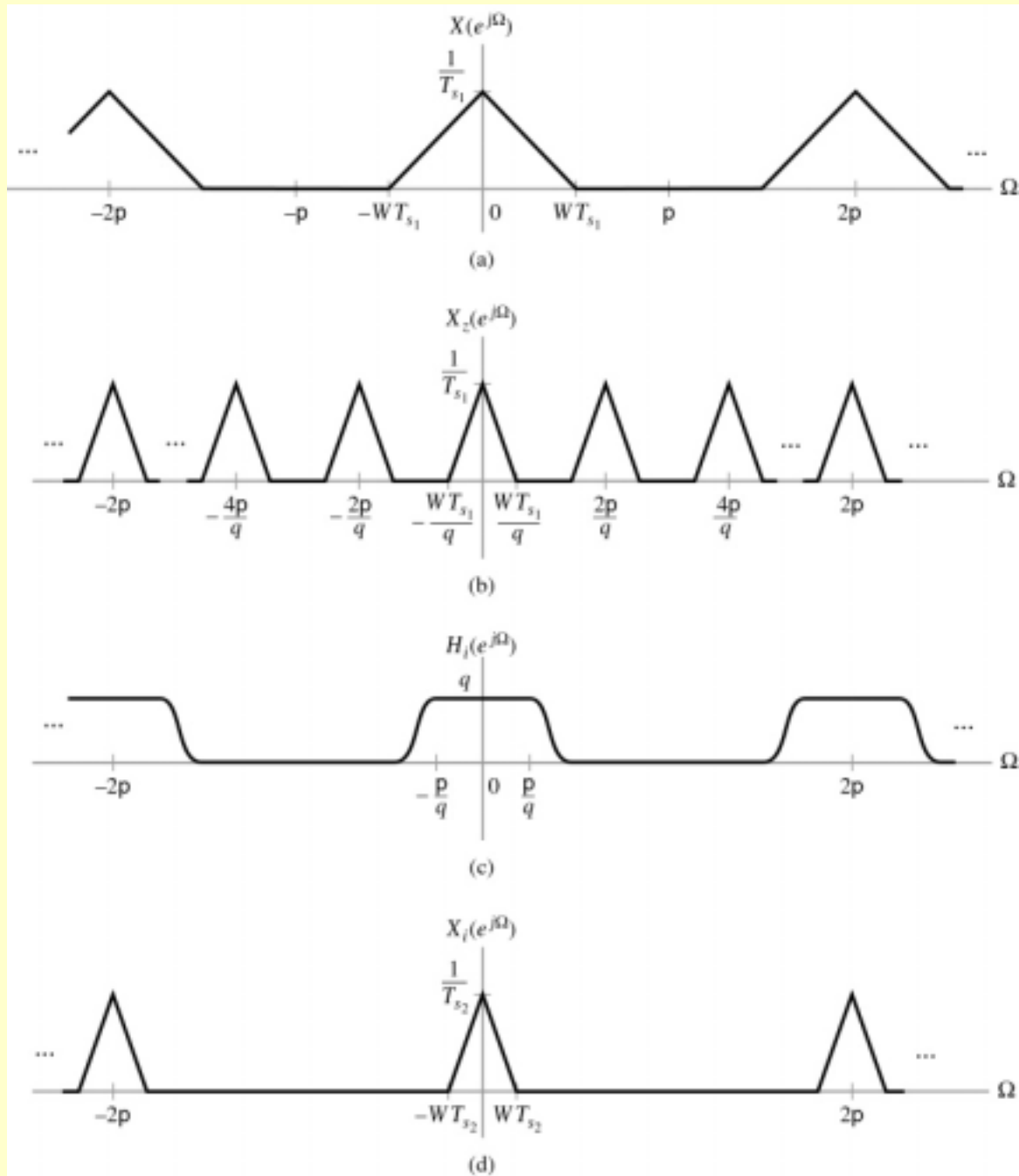
內插 取樣將頻譜頻寬 wT_{s1} 減少為 $wT_{s2} = wT_{s1}/q$

內插 取樣將原始訊號頻率減少 q 倍



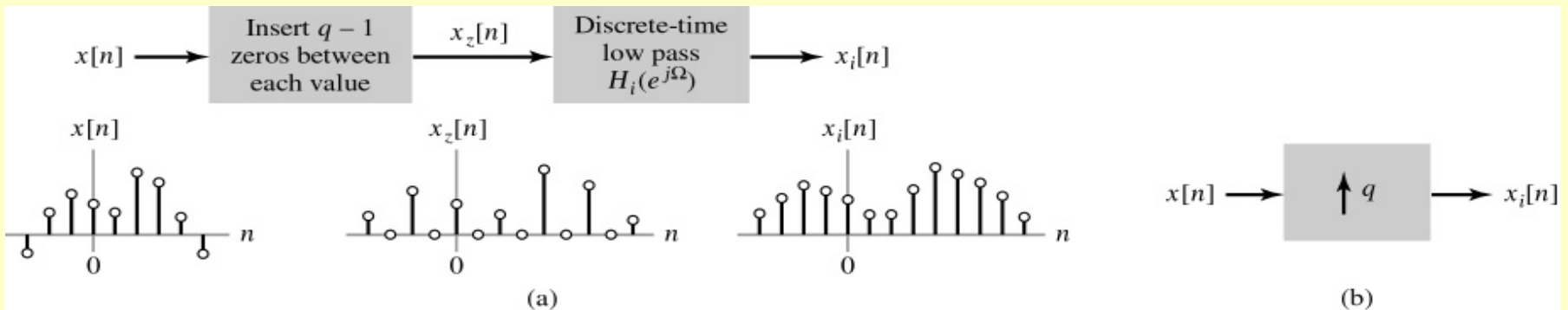
Frequency-Domain Interpretation of Interpolation

- (a) Spectrum of original sequence.
- (b) Spectrum after inserting $q - 1$ zeros in between every value of the original sequence.
- (c) Frequency response of a filter for removing undesired replicates located at $\pm 2\pi/q, \pm 4\pi/q, \dots, \pm (q - 1)2\pi/q$.
- (d) Spectrum of interpolated sequence.





Interpolation System Diagram

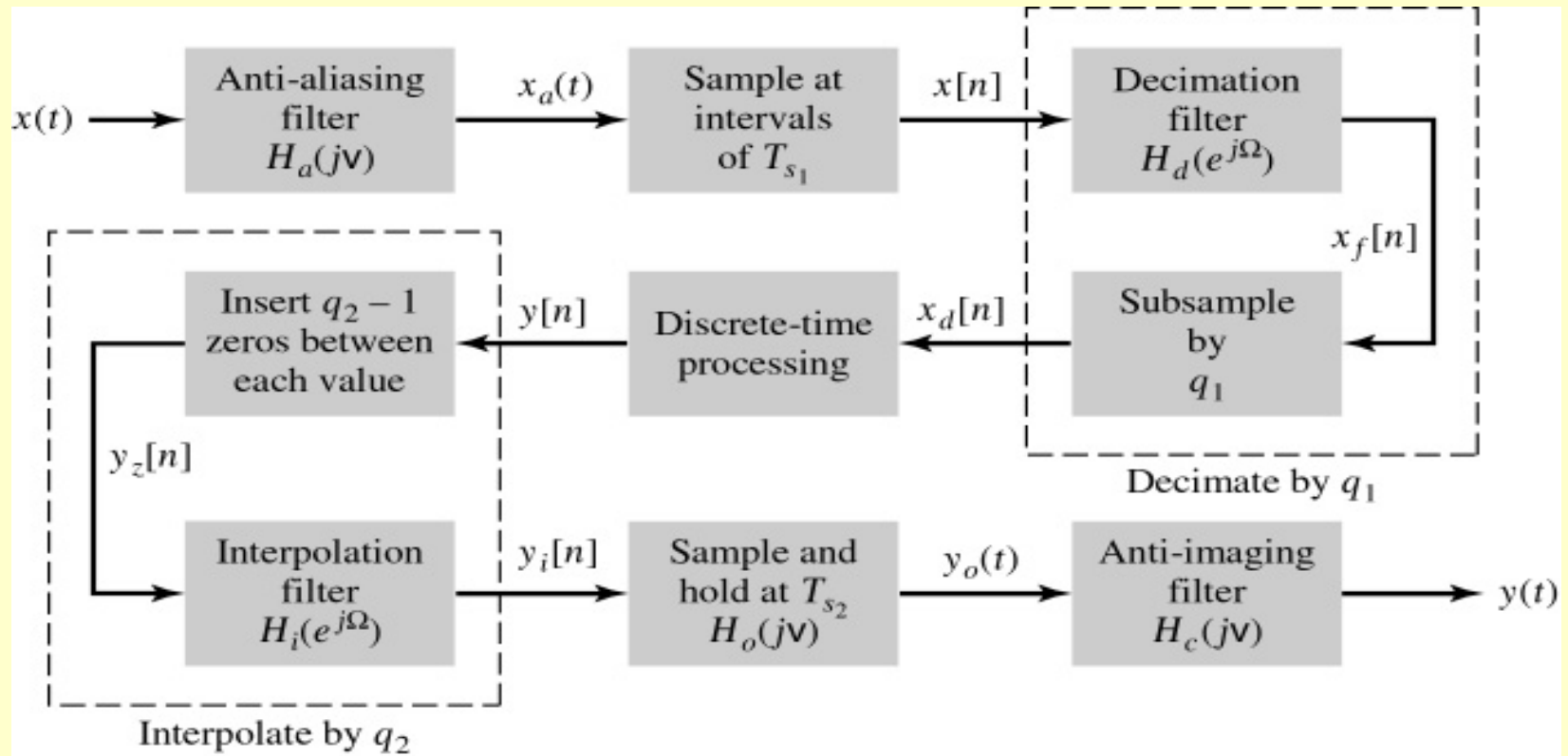


(a) Block diagram of an interpolation system.

(b) Symbol denoting interpolation by a factor of q .



Block Diagram of a System for Discrete-Time Processing of Continuous-Time Signals Including Decimation and Interpolation.





FS for Finite-Duration Non-periodic Signals

Relating the DTFS to the DTFT:

$x[n]$ 為一個長度為 M 的有限時間訊號：

$$x[n] = 0, \quad n < 0 \quad \text{or} \quad n \geq M$$

$x[n]$ 的 DTFT :

$$X(e^{j\Omega}) = \sum_{n=0}^{M-1} x[n] e^{-j\Omega n}$$



擴充 $x[n]$ 為一個週期為 $N \geq M$ 的週期訊號： $\tilde{x}[n]$

$$\tilde{x}[n] = \sum_{m=-\infty}^{\infty} x[n + mN]$$

$\tilde{x}[n]$ 的 DTFS：

$$\tilde{X}[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jk\Omega_0 n} = \frac{1}{N} \sum_{n=0}^{M-1} x[n] e^{-jk\Omega_0 n}$$



比較 $X(e^{j\Omega})$ 和 $\tilde{X}[k]$:

$$X(e^{j\Omega}) = \sum_{n=0}^{M-1} x[n] e^{-j\Omega n} \quad \text{DTFT 頻譜}$$

$$\tilde{X}[k] = \frac{1}{N} \sum_{n=0}^{M-1} x[n] e^{-jk\Omega_0 n} \quad \text{DTFS 頻譜}$$

可得:

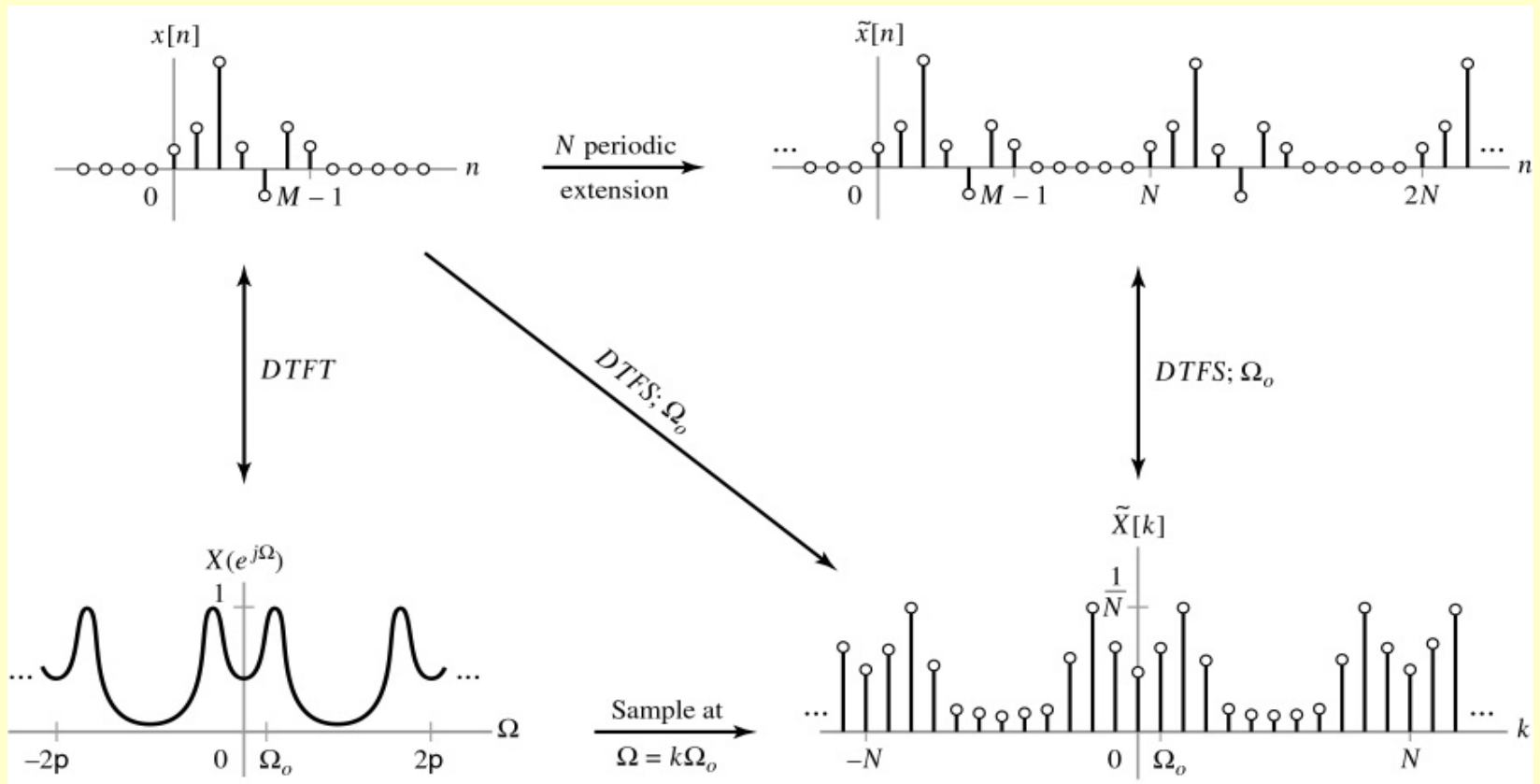
$$\tilde{X}[k] = \frac{1}{N} X(e^{j\Omega}) \Big|_{\Omega = k\Omega_0}$$



The DTFS of a Finite-Duration Non-Periodic Signal

Duration = M

Expanded to Duration N

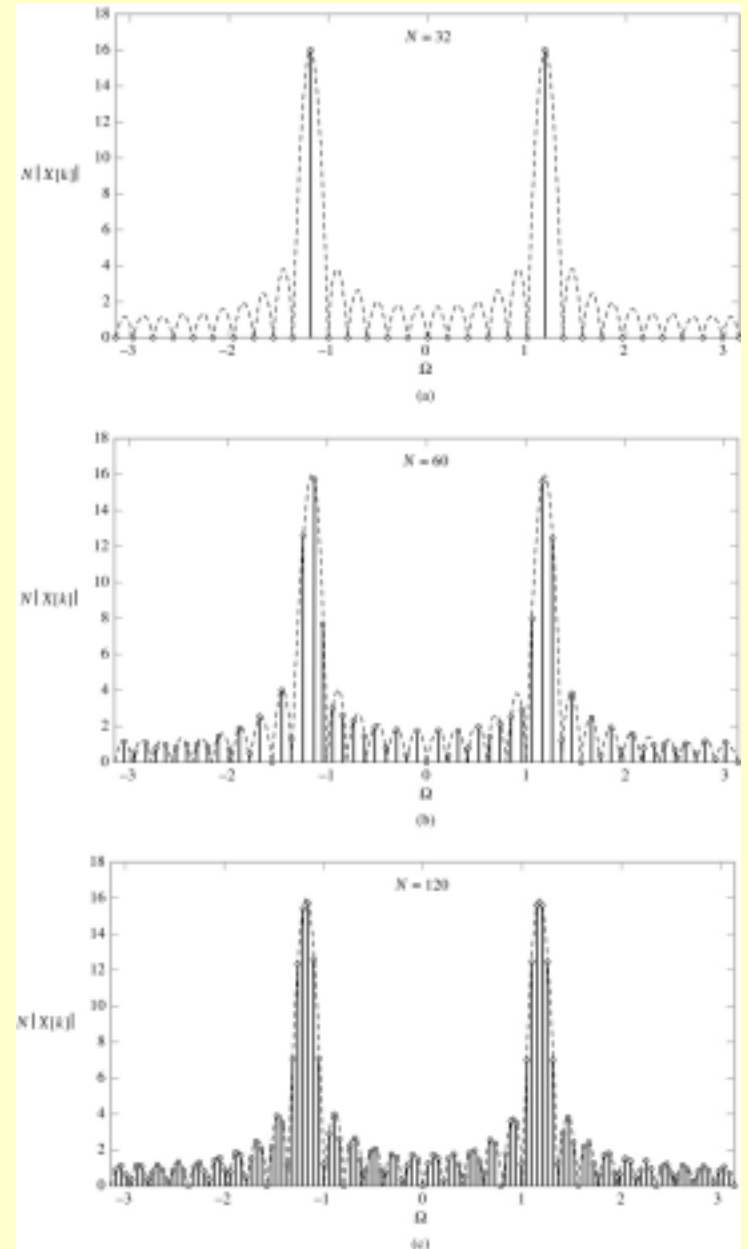




The DTFT and length- N DTFS of a 32-point cosine.

The dashed line denotes $|X(e^{j\Omega})|$, while the stems represent $N|X[k]|$.

- (a) $N = 32$,
- (b) $N = 60$,
- (c) $N = 120$.





Fourier Series Representations of Finite-Duration Non-periodic Signals

Relating the FS to the FT:

$x(t)$ 為一個長度為 T_0 的有限時間訊號：

$$x(t) = 0, \quad t < 0 \quad \text{or} \quad t \geq T_0$$

$x(t)$ 的 FT：

$$X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$



Relating the FS to the FT (cont.)

擴充 $x(t)$ 為一個週期為 $T \geq T_0$ 的週期訊號：

$$\tilde{x}(t) = \sum_{m=-\infty}^{\infty} x(t + mT)$$

$\tilde{x}(t)$ 的 FS：

$$\begin{aligned}\tilde{X}[k] &= \frac{1}{T} \int_0^T \tilde{x}(t) e^{-jk\omega_0 t} dt \\ &= \frac{1}{T} \int_0^{T_0} x(t) e^{-jk\omega_0 t} dt\end{aligned}$$



比較 $X(j\omega)$ 和 $\tilde{X}[k]$:

$$X(j\omega) = \int_0^{T_0} x(t) e^{-j\omega t} dt \quad \text{FT 頻譜}$$

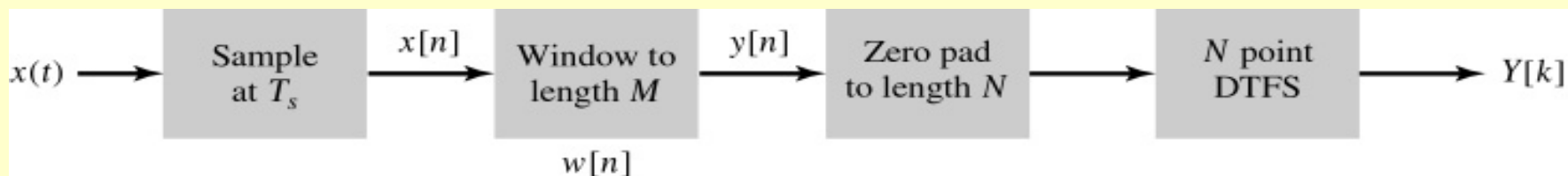
$$\tilde{X}[k] = \frac{1}{T} \int_0^{T_0} x(t) e^{-jk\omega_0 t} dt \quad \text{FS 頻譜}$$

可得:

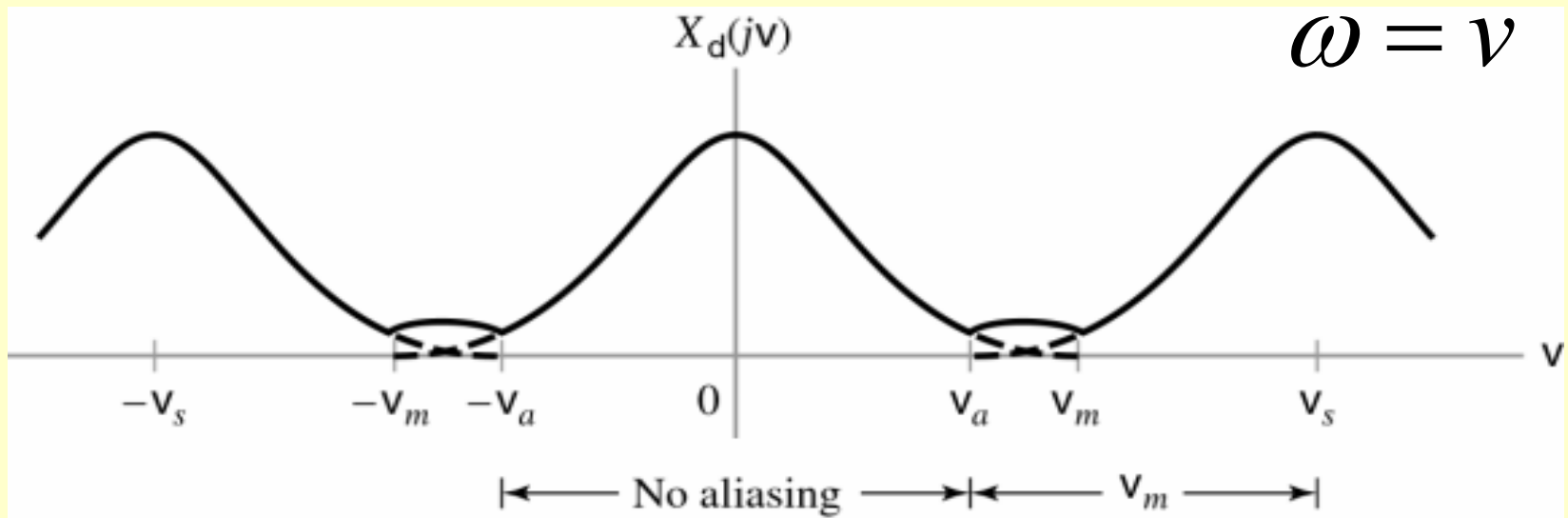
$$\tilde{X}[k] = \frac{1}{T_s} X(j\omega) \Big|_{\Omega = k\Omega_0}$$



Block Diagram Depicting the Sequence of Operations Involved in Approximating the FT with the DTFS

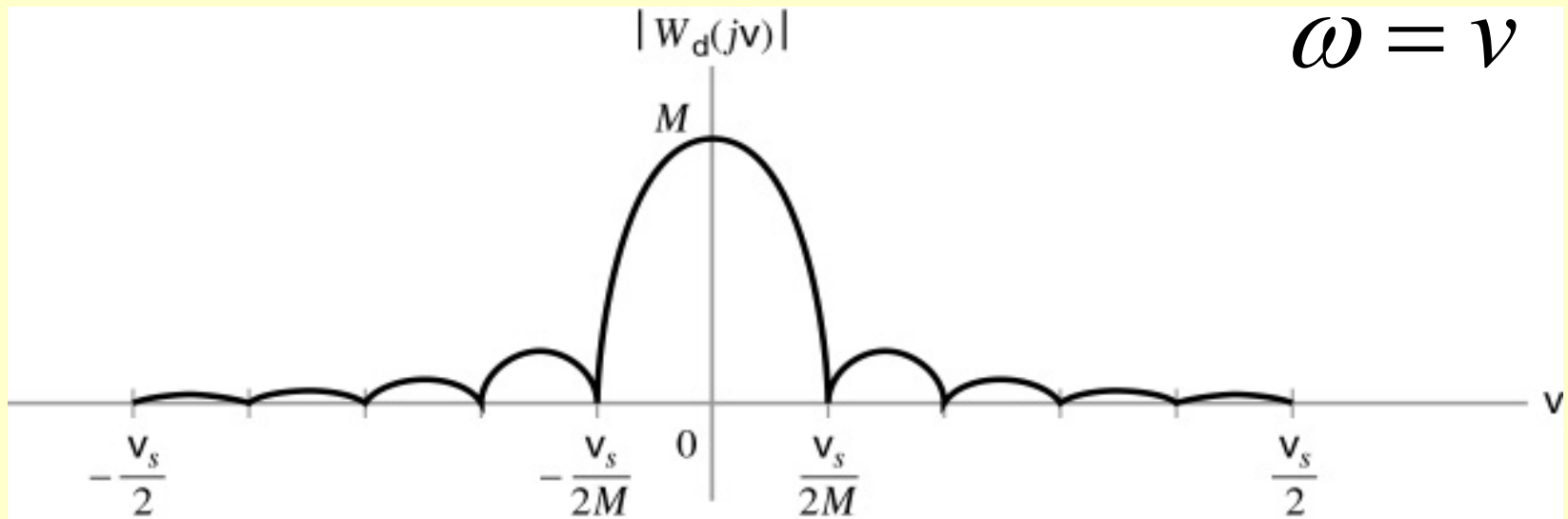


Effect of Aliasing





Magnitude response of M -point window

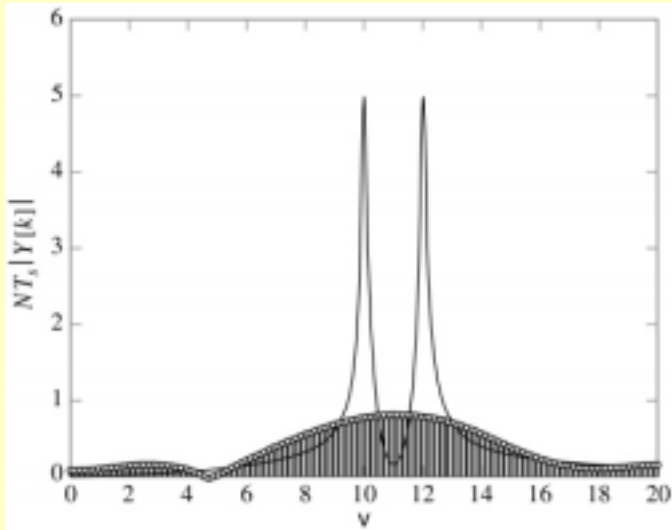




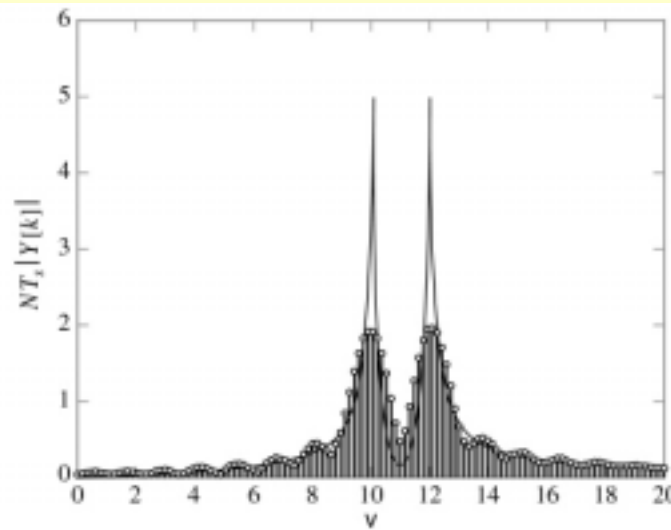
DTFT Approximation to the FT

The DTFS approximation to the FT of $x(t) = e^{-1/10} u(t)(\cos(10t) + \cos(12t))$. The solid line is the FT $|X(j\omega)|$, and the stems denote the DTFS approximation $N T_s |Y[k]|$. Both $|X(j\omega)|$ and $N T_s |Y[k]|$ have even symmetry, so only $0 < \omega < 20$ is displayed.

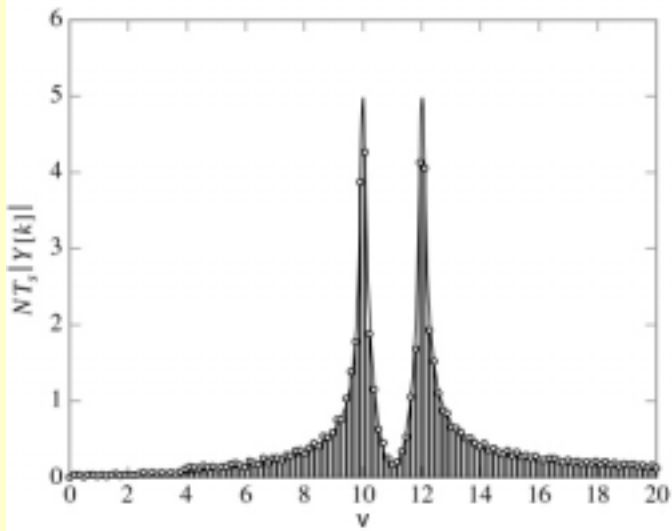
- (a) $M = 100, N = 4000$.
- (b) $M = 500, N = 4000$.
- (c) $M = 2500, N = 4000$.
- (d) $M = 2500, N = 16,0000$ for $9 < \omega < 13$.



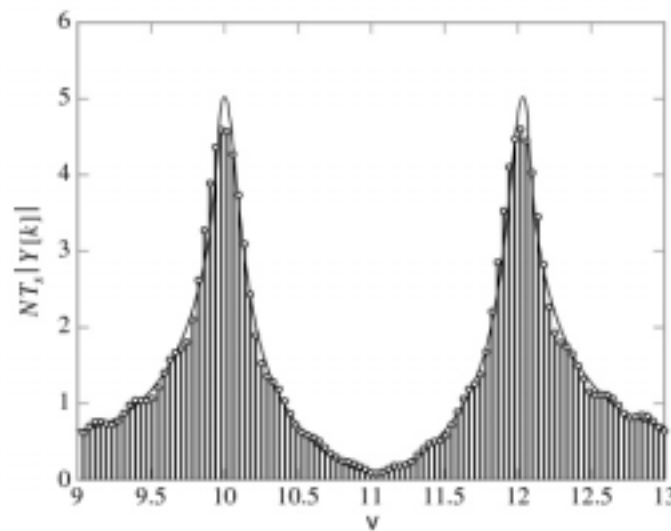
(a)



(b)



(c)



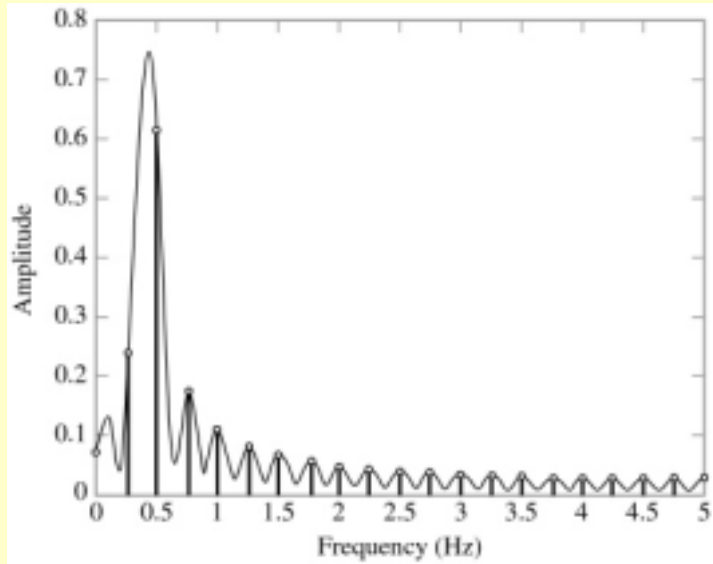
(d)



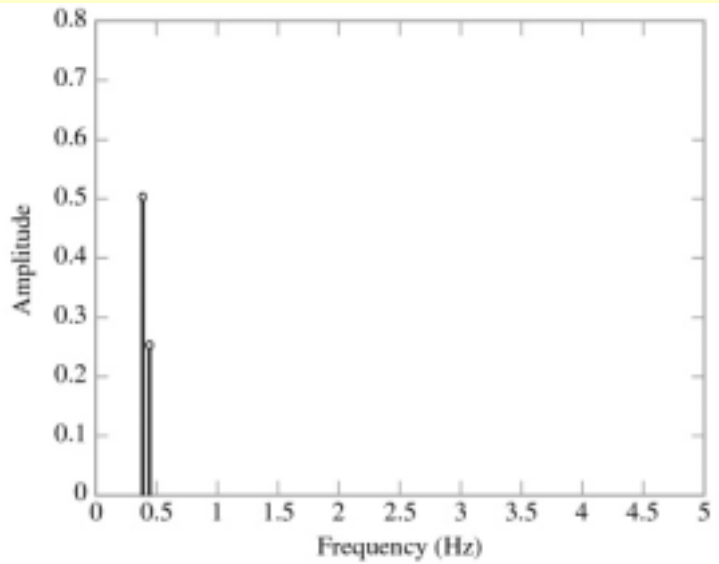
DTFS Approximation to the FT

The DTFS approximation to the FT of $x(t) = \cos(2\pi(0.4)t) + \cos(2\pi(0.45)t)$. The stems denote $|Y[k]|$, while the solid lines denote $(1/M|Y\delta(j\omega)|$. The frequency axis is displayed in units of Hz for convenience, and only positive frequencies are illustrated.

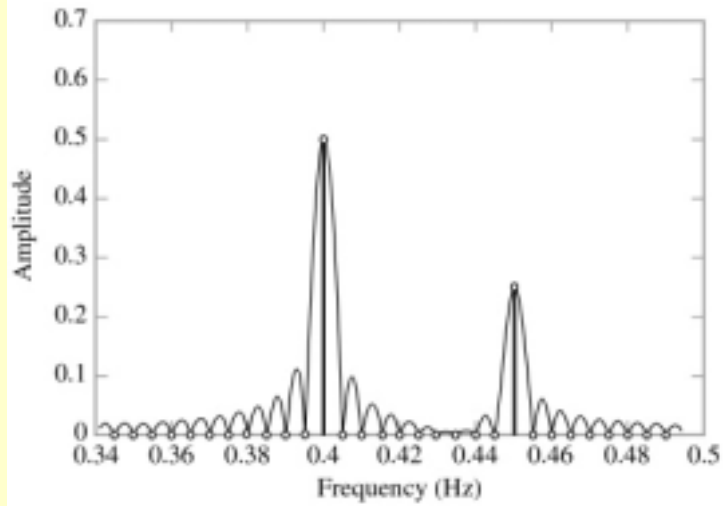
- (a) $M = 40$.
- (b) $M = 2000$. Only the stems with nonzero amplitude are depicted.
- (c) Behavior in the vicinity of the sinusoidal frequencies for $M = 2000$.
- (d) Behavior in the vicinity of the sinusoidal frequencies for $M = 2010$.



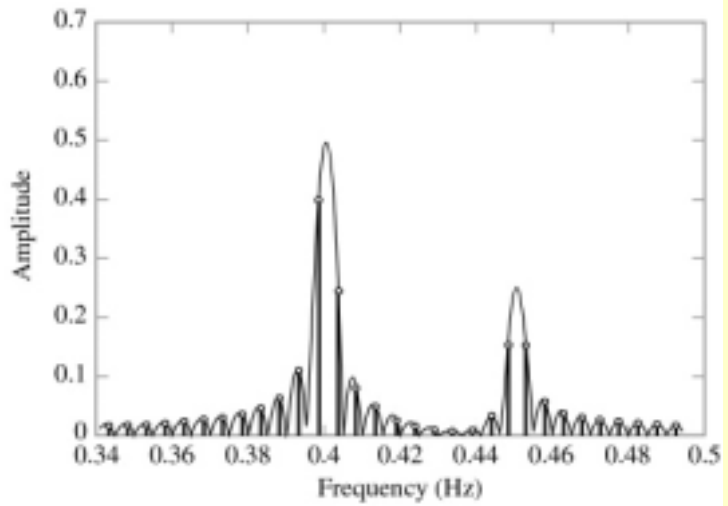
(a)



(b)



(c)



(d)



DTFS Decompositions

Block diagrams depicting the decomposition of an inverse DTFS as a combination of lower order inverse DTFS's.

- (a) Eight-point inverse DTFS represented in terms of two four-point inverse DTFS's.
- (b) four-point inverse DTFS represented in terms of two-point inverse DTFS's.
- (c) Two-point inverse DTFS.



Efficient Algorithms for Evaluating the DTFS- Fast Fourier Transform (FFT)

- FFT 是有計算 DTFS (離散訊號、頻譜) 有效率的演算法
- 將 DTFS 分割成系列較低階的 DTFS
- 應用複數弦波 $e^{jk2\pi n}$ 的對稱性與週期性

DTFS 公式：

$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jk\Omega_0 n}$$

$$x[n] = \sum_{k=0}^{N-1} X[k] e^{jk\Omega_0 n}$$

若忽略正規化因子 N 及複數指數符號，上述公式可用同一種演算法來計算



分解 DTFS 公式 (Even Number N)

$$X_{\text{even}}[k] = X[2k], \quad 0 \leq k \leq \frac{N}{2} - 1$$

$$X_{\text{odd}}[k] = X[2k + 1], \quad 0 \leq k \leq \frac{N}{2} - 1$$

DTFS 對應公式：

$$x_{\text{even}}[n] \stackrel{DTFS; \Omega_0}{\leftrightarrow} X_{\text{even}}[k]$$

$$x_{\text{odd}}[n] \stackrel{DTFS; \Omega_0}{\leftrightarrow} X_{\text{odd}}[k]$$



$$x[n] = \sum_{k=0}^{N-1} X[k] e^{jk\Omega_0 n}$$

$$= \sum_{k=0}^{\frac{N-1}{2}} X[2k] e^{j2k\Omega_0 n} + \sum_{k=0}^{\frac{N-1}{2}} X[2k+1] e^{j(2k+1)\Omega_0 n}$$

$$= \sum_{k=0}^{\frac{N-1}{2}} X[2k] e^{j2k\Omega_0 n} + e^{j\Omega_0 n} \sum_{k=0}^{\frac{N-1}{2}} X[2k+1] e^{j2k\Omega_0 n}$$

$$= \sum_{k=0}^{\frac{N-1}{2}} X[2k] e^{j2k\Omega_0 n} + e^{j\Omega_0 n} \sum_{k=0}^{\frac{N-1}{2}} X[2k+1] e^{j2k\Omega_0 n}$$

let $N' = N/2$, $\Omega_0' = 2\Omega_0 = 2\pi/N'$,

$$X_{\text{even}}[k] = X[2k], \quad X_{\text{odd}}[k] = X[2k+1]$$



分解成偶與奇項次

$$\begin{aligned}
 x[n] &= \sum_{k=0}^{N'-1} X_{even}[k] e^{jk\Omega_0'n} + e^{j\Omega_0'n} \sum_{k=0}^{N'-1} X_{odd}[k] e^{jk\Omega_0'n} \\
 &= x_{even}[n] + e^{j\Omega_0'n} \cdot x_{odd}[n]
 \end{aligned}$$

$x[n]$ 是週期性訊號，分析週期性關係：

$$x_{even}[n] = x_{even}[n + N'], \quad x_{odd}[n] = x_{odd}[n + N']$$

$$\begin{aligned}
 e^{j\Omega_0[n+N']} &= e^{j\frac{2\pi}{N}[n+N']} = e^{j\Omega_0'n} \cdot e^{j\frac{2\pi}{N} \frac{N}{2}} \\
 &= e^{j\Omega_0'n} \cdot e^{j\pi} = -e^{j\Omega_0'n}
 \end{aligned}$$



$x[n]$ 週期性訊號前半段： $x[0] \sim x[N'-1]$

$$x[n] = x_{\text{even}}[n] + e^{j\Omega_0 n} \cdot x_{\text{odd}}[n], \quad 0 \leq n \leq N'-1$$

$x[n]$ 週期性訊號後半段： $x[N'] \sim x[N-1]$

$$x[n + N'] = x_{\text{even}}[n] - e^{j\Omega_0 n} \cdot x_{\text{odd}}[n], \quad 0 \leq n \leq N'-1$$



N=2 Case

N = 2 case: 計算 DTFS $X[k] = ?$
 $N^2 = 4$ 個乘法, $N(N-1) = 2$ 個加法

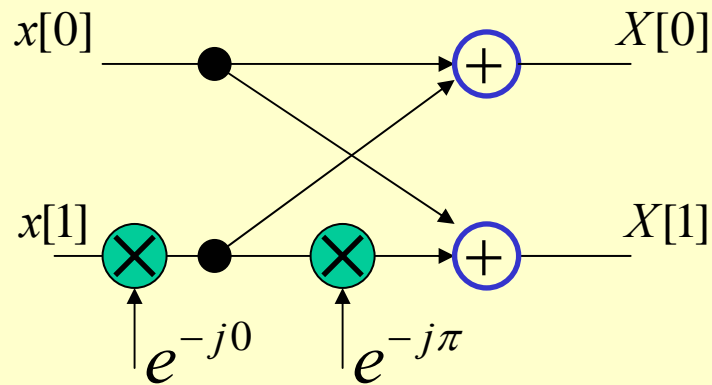
$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-jk \frac{2\pi}{N} n}$$

$N = 2$ case

$$X[k] = \sum_{n=0}^1 x[n] e^{-jk\pi n}$$

$$\Rightarrow X[0] = \sum_{n=0}^1 x[n] = x[0] \cdot e^{-j0} + x[1] \cdot e^{-j0} = x[0] + x[1]$$

$$\Rightarrow X[1] = \sum_{n=0}^1 x[n] e^{-jk\pi n} = x[0] \cdot e^{-j0} + x[1] \cdot e^{-j\pi} = x[0] - x[1]$$



For FFT: 2 個加法
 2 個乘法



N=4 Case

N = 4 case: 計算 $X[k] = ?$

$N^2 = 16$ 個乘法, $N(N-1) = 12$ 個加法

$$X[k] = \sum_{n=0}^3 x[n] e^{-jk \frac{\pi}{2} n}$$

$$X[0] = \sum_{n=0}^3 x[n] \cdot e^{-j0n} = x[0] + x[1] + x[2] + x[3]$$

$$X[2] = \sum_{n=0}^3 x[n] \cdot e^{-j\pi n} = x[0] - x[1] + x[2] - x[3]$$

$$X[1] = \sum_{n=0}^3 x[n] e^{-j \frac{\pi}{2} n} = x[0] - e^{j \frac{\pi}{2}} \cdot x[1] - x[2] + e^{j \frac{\pi}{2}} \cdot x[3]$$

$$X[3] = \sum_{n=0}^3 x[n] e^{-j \frac{3\pi}{2} n} = x[0] + e^{j \frac{\pi}{2}} \cdot x[1] - x[2] - e^{j \frac{\pi}{2}} \cdot x[3]$$

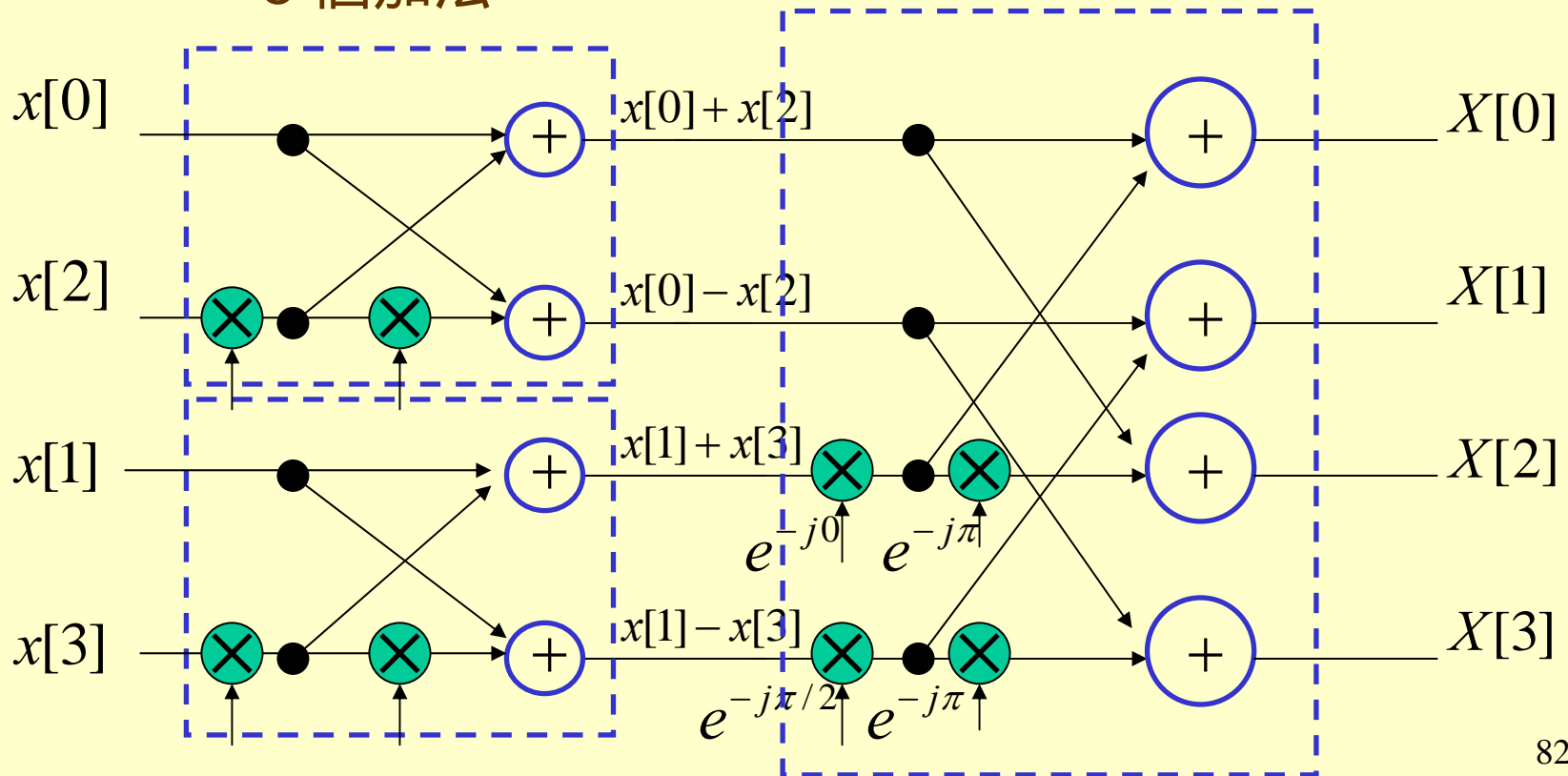


FFT for $N = 4$ case:

計算 $X[k] = ?$

$4 \log_2 4 = 8$ 個乘法

8 個加法



$$X[0] = x[0] + x[1] + x[2] + x[3]$$

$$X[2] = x[0] - x[1] + x[2] - x[3]$$

$$X[1] = x[0] - e^{j\frac{\pi}{2}} \cdot x[1] - x[2] + e^{j\frac{\pi}{2}} \cdot x[3]$$

$$X[3] = x[0] + e^{j\frac{\pi}{2}} \cdot x[1] - x[2] - e^{j\frac{\pi}{2}} \cdot x[3]$$



$N = 4$ case: 計算 DTFS $X[k] = ?$ 計算 運算量

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-jk \frac{2\pi}{N} n}, \text{ for } k = 0, \dots, N-1$$

計算 每一個 $X[k]$ 需 $N=4$ 個乘法 $N-1=3$ 個加法

若計算 N 個 $X[k]$ 需 $N^2=16$ 個乘法 $N(N-1)=12$ 個加法

使用 FFT, $N = 4$ case: 計算 $X[k] = ?$ 計算 運算量

計算 4 個 $X[k]$ 需 $4 \log_2 4 = 8$ 個乘法 8 個加法

Total : $N \log_2 N$



位元倒置 Bit Reversal

將輸入端的係數分割成索引值 **Even Index** 和 **Odd Index**

以二進位位元表示索引值

Example:

Old Sequence: $x[0], x[1], x[2], x[3], x[4], x[5], x[6], x[7]$

At $k = 0, 1, 2, 3, 4, 5, 6, 7$

→ $k = 000, 001, 010, 011, 100, 101, 110, 111$

Bit reversal

→ $k' = 000, 100, 010, 110, 001, 101, 011, 111$

At $k' = 0, 4, 2, 6, 1, 5, 3, 7$

New Sequence: $x[0], x[4], x[2], x[6], x[1], x[5], x[3], x[7]$ 84



位元倒置 Bit Reversal (cont.)

分割成 Even Index 和 Odd Index

Example 驗證:

Original: $x[0], x[1], x[2], x[3], x[4], x[5], x[6], x[7]$

Even: $x[0], x[2], x[4], x[6]$

Even: $x[0], x[4]$

Odd: $x[2], x[6]$

Odd: $x[1], x[3], x[5], x[7]$

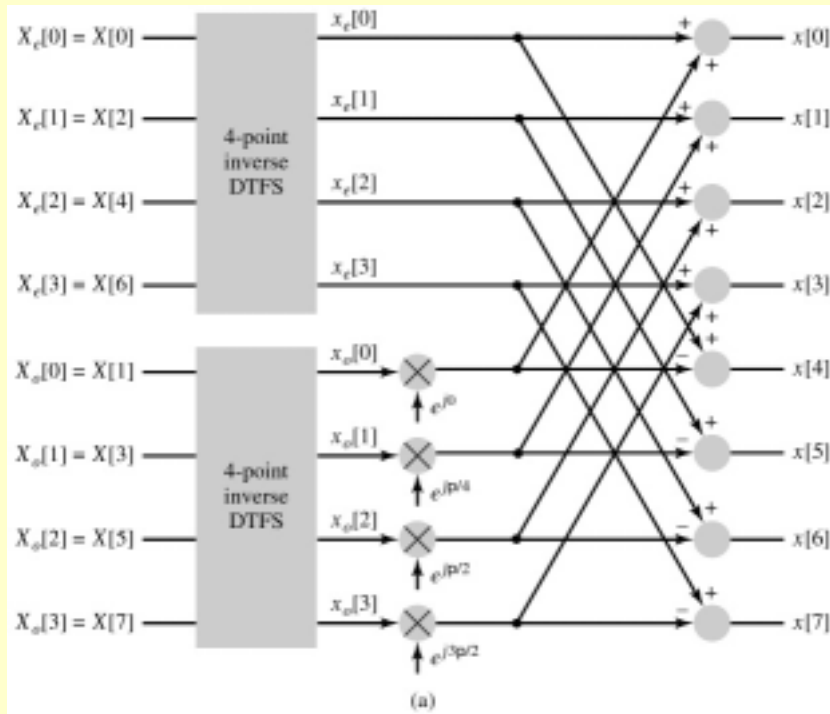
Even: $x[1], x[5]$

Odd: $x[3], x[7]$

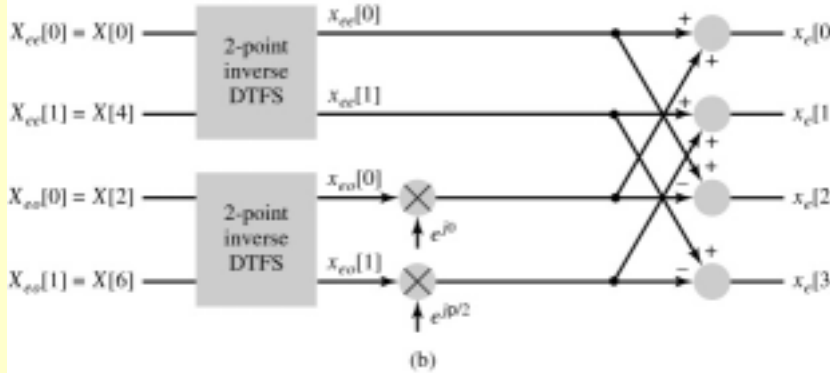
Final : $x[0], x[4], x[2], x[6], x[1], x[5], x[3], x[7]$



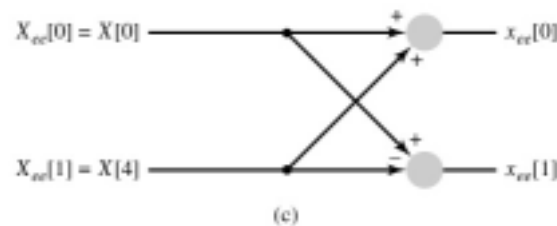
N=8



N=4



N=2





Find $x[n]$ From $X[k]$ (N=2 Case)

$$x[n] = \sum_{k=0}^{N-1} X[k] e^{jk \frac{2\pi}{N} n}$$

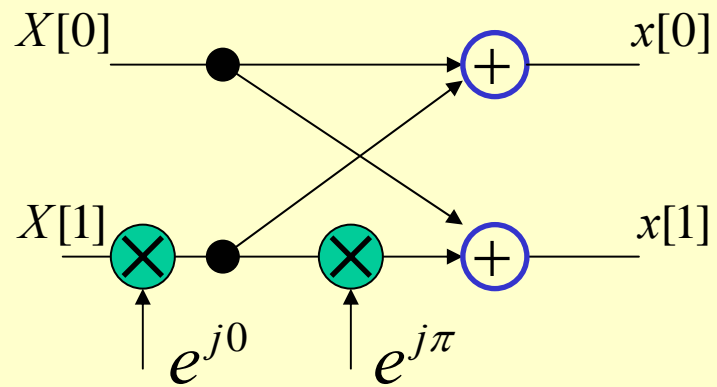
$N = 2$ case

$$x[n] = \sum_{k=0}^1 X[k] e^{jk\pi n}$$

$$\Rightarrow x[0] = \sum_{k=0}^1 X[k] e^{jk\pi n} = X[0] \cdot e^{j0} + X[1] \cdot e^{j0} = X[0] + X[1]$$

$$\Rightarrow x[1] = \sum_{k=0}^1 X[k] e^{jk\pi n} = X[0] \cdot e^{j0} + X[1] \cdot e^{j\pi} = X[0] - X[1]$$

FFT Butterfly for $N = 2$





Find $x[n]$ From $X[k]$ (N=4 Case)

$$x[n] = \sum_{k=0}^3 X[k] e^{jk\pi n/2}$$

$$\Rightarrow x[0] = \sum_{k=0}^3 X[k] e^{jk\pi n/2} = X[0] \cdot e^{j0} + X[1] \cdot e^{j0} + X[2] \cdot e^{j0} + X[3] \cdot e^{j0}$$

$$\Rightarrow x[1] = \sum_{k=0}^3 X[k] e^{jk\pi n/2} = X[0] \cdot e^{j0} + X[1] \cdot e^{j\pi/2} + X[2] \cdot e^{j\pi} + X[3] \cdot e^{j3\pi/2}$$

$$\Rightarrow x[2] = \sum_{k=0}^3 X[k] e^{jk\pi n/2} = X[0] \cdot e^{j0} + X[1] \cdot e^{j\pi} + X[2] \cdot e^{j2\pi} + X[3] \cdot e^{j3\pi}$$

$$\Rightarrow x[3] = \sum_{k=0}^3 X[k] e^{jk\pi n/2} = X[0] \cdot e^{j0} + X[1] \cdot e^{j3\pi/2} + X[2] \cdot e^{j3\pi} + X[3] \cdot e^{j9\pi/2}$$



FFT Butterfly For N = 4

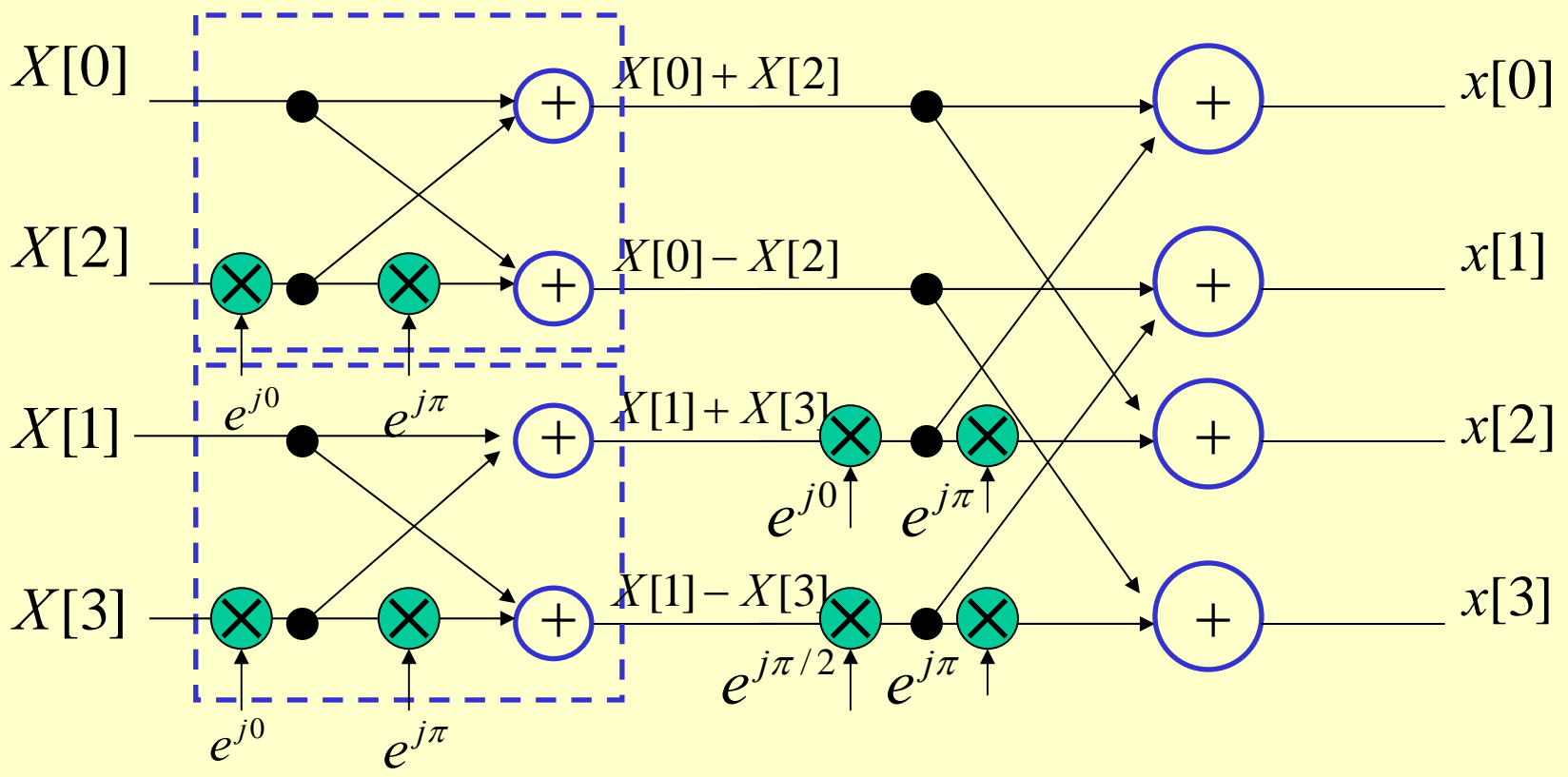




Diagram of the FFT algorithm for computing $x[n]$ from $X[k]$ for $N = 8$

