

IV. Implementation

IV-A Method 1: Direct Implementation

以 STFT 為例

$$X(t, f) = \int_{-\infty}^{\infty} w(t - \tau) x(\tau) e^{-j2\pi f \tau} d\tau$$

Converting into the Discrete Form

$$t = n\Delta_t, \quad f = m\Delta_f, \quad \tau = p\Delta_t$$

$$X(n\Delta_t, m\Delta_f) = \sum_{p=-\infty}^{\infty} w((n-p)\Delta_t) x(p\Delta_t) e^{-j2\pi pm\Delta_t \Delta_f} \Delta_t$$

Suppose that $w(t) \approx 0$ for $|t| > B$, $B/\Delta_t = Q$

$$X(n\Delta_t, m\Delta_f) = \sum_{p=-Q}^{n+Q} w((n-p)\Delta_t) x(p\Delta_t) e^{-j2\pi pm\Delta_t \Delta_f} \Delta_t$$

Problem : 對 scaled Gabor transform 而言 , $Q = ?$

$$w(t) = e^{-\pi \sigma t^2}, \quad B = \frac{1.9143}{\sqrt{\sigma}}, \quad Q = \frac{1.9143}{\sqrt{\sigma} \Delta t}$$

$w((n-p)\Delta_t) \approx 0$
 for $|n-p| > \frac{B}{\Delta t}$
 $|p-h| > Q$
 Therefore, only when
 $-Q \leq p-h \leq Q$
 $w((n-p)\Delta_t)$ is
 nonzero

- **Constraint for Δ_t** (The only constraint for the direct implementation method)

To avoid the aliasing effect,

$$\underline{\Delta_t < 1/2\Omega},$$

Ω is the bandwidth of ?

$$\Omega = \Omega_x + \Omega_w$$

$$x(t) w(t-\tau)$$

$$w(\tau) \xrightarrow{f} w(f)$$

$$w(-\tau) \rightarrow w(-f)$$

$$w(t-\tau) \rightarrow e^{-j2\pi f\tau} w(-f)$$

suppose that

bandwidth of $x : \Omega_x$

bandwidth of $w : \Omega_w$

There is no constraint for Δ_f when using the direct implementation method.

Four Implementation Methods

(1) Direct implementation

Complexity: $T F(2Q+1) \asymp 2TFQ \quad \mathcal{O}(TF^2)$

假設 t -axis 有 T 個 sampling points, f -axis 有 F 個 sampling points

(2) FFT-based method

Complexity: $\mathcal{O}(TN \log N)$ unbalanced form
 $\mathcal{O}(\frac{T}{S}N \log N)$

(3) FFT-based method with recursive formula

Complexity: $\mathcal{O}(TF)$

(4) Chirp-Z transform method

Complexity: $\mathcal{O}(TN \log N)$

4 (A) Direct Implementation

Advantage : simple, flexible

$$\text{the only constraint: } \Delta t < \frac{1}{2(\Omega_x + \Omega_w)}$$

Disadvantage : higher complexity

2 (B) DFT-Based Method

Advantage : lower complexity

Disadvantage : with some constraints

$$\frac{1}{\Delta t \Delta f} = N \text{ must be an integer}$$

$$N \geq 2Q+1 = 2 \frac{B}{\Delta t} + 1$$

$$\Delta t < \frac{1}{2(\Omega_x + \Omega_w)}$$

1 (C) Recursive Method

Advantage : least complexity

Disadvantage (i) only suitable for the rectangular window
(ii) accumulated error

3 (D) Chirp Z Transform

Advantage : no constraint (except for $\Delta t < \frac{1}{2(\Omega_x + \Omega_w)}$)

Disadvantage : complexity is higher than DFT-based method

IV-B Method 2: FFT-Based Method

Constraints (i) $\Delta_t \Delta_f = 1/N$,

(ii) $N = 1/(\Delta_t \Delta_f) \geq 2Q + 1$: ($\Delta_t \Delta_f$ is 整數的倒數)

$$X(n\Delta_t, m\Delta_f) = \sum_{p=-Q}^{n+Q} w((n-p)\Delta_t) x(p\Delta_t) e^{-j\frac{2\pi pm}{N}} \Delta_t$$

$w((n-Q-n+Q)\Delta_t) \times ((Q+n-Q)\Delta_t)$

Note that the input of the FFT has less than N points (others are set to zero).

Standard form of the DFT $Y[m] = \sum_{n=0}^{N-1} y[n] e^{-j\frac{2\pi mn}{N}}$

$$X(n\Delta_t, m\Delta_f) = \Delta_t e^{j\frac{2\pi(Q-n)m}{N}} \sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}}, \quad q = p - (n-Q) \rightarrow p = (n-Q) + q$$

window $\times x(n) \quad (n-Q \leq n \leq n+Q)$
 $w((Q-Q)\Delta_t) \quad N-1 \geq 2Q$

where $x_1(q) = w((Q-q)\Delta_t) x((n-Q+q)\Delta_t)$

$$x_1(q) = 0 \quad -Q \leq Q-Q \leq Q$$

$$0 \leq q \leq 2Q \quad n-Q \leq n-Q+q \leq n+Q$$

for $0 \leq q \leq 2Q$,
 for $2Q < q < N$.

$$n-Q \leq n-Q+q \leq n+Q$$

$$Q \geq Q - q \geq -Q$$

$$\boxed{X(n\Delta_t, m\Delta_f) = \Delta_t e^{j\frac{2\pi(Q-n)m}{N}} \sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}}}$$

where $x_1(q) = w(k\Delta_t) x((n+k)\Delta_t)$ for $0 \leq q \leq 2Q$, $k = q - Q$ $-Q \leq k \leq Q$

$x_1(q) = 0$ for $2Q < q < N$.

(Suppose that $w(t) = w(-t)$)

注意：

(1) 可以使用 Matlab 的 FFT 指令來計算 $\sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}} = X_1(m)$

If $m < 0$, we can apply

$$X_1(m+N) = X_1(m)$$

$$X_1(m+N) = \sum_{q=0}^{N-1} X_1(q) e^{-j\frac{2\pi qm}{N}} e^{-j\frac{2\pi qN}{N}} = X_1(m)$$

$\Theta(N \log N)$

$$\begin{aligned} q &= 0, 1, \dots, N-1 \\ m &= 0, 1, \dots, N-1 \end{aligned}$$

(2) 對每一個固定的 n ，都要計算一次下方的式子

$$\begin{aligned} X(n\Delta_t, m\Delta_f) &= \Delta_t e^{j\frac{2\pi(Q-n)m}{N}} \sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}} \\ (\text{fixed } n) \quad T &\quad F + N \log N \quad \begin{aligned} T(F + N \log N) \\ \cong TN \log N \end{aligned} \end{aligned}$$

假設 $t = n_0\Delta_t, (n_0+1)\Delta_t, (n_0+2)\Delta_t, \dots, (n_0+T-1)\Delta_t$

$$f = m_0\Delta_f, (m_0+1)\Delta_f, (m_0+2)\Delta_f, \dots, (m_0+F-1)\Delta_f$$

Step 1: Calculate n_0, m_0, T, F, N, Q

$$\begin{aligned} N &= \frac{1}{\Delta t \Delta f} \\ \theta &= \frac{B}{\Delta t} \end{aligned}$$

Step 2: $n = n_0$

Step 3: Determine $x_1(q)$

Step 4: $X_1(m) = \underline{\text{FFT}}[x_1(q)]$

Step 5: Convert $X_1(m)$ into $X(n\Delta_t, m\Delta_f)$

} page 115

$$X(n\Delta_t, m\Delta_f) = X_1(?) \times ?$$

$$X_1(\underbrace{\text{mod}(m, N)}_{m \neq N}) e^{j \frac{2\pi}{N} (Q-n)m} \Delta t$$

$$\text{mod}(m, N) = m + N$$

$$\text{if } -N \leq m < 0$$

$$X_1[m] = \sum_{q=0}^{N-1} x_1(q) e^{-j \frac{2\pi q m}{N}}$$

Step 6: Set $n = n+1$ and return to Step 3 until $n = n_0+T-1$.

$$m = f / \Delta_f$$

$$m_1 = \text{mod}(m, N) + 1$$

Matlab
語法

$$\text{ex: if } N=400 \\ X, [-100]=X, [300]$$

$$X_1[m] = X_1[m + N]$$

IV-C Method 3: Recursive Method

- A very fast way for implementing the **rec-STFT**

(n 和 $n-1$ 有 recursive 的關係)

限制

$$\Delta t + \Delta f = \frac{1}{N}, \quad W(t) \mid_{\text{for } |t| \leq B}$$

$$X(n\Delta_t, m\Delta_f) = \sum_{p=n-Q}^{n+Q} x(p\Delta_t) e^{-j\frac{2\pi pm}{N}} \Delta_t$$

$$X((n-1)\Delta_t, m\Delta_f) = \sum_{p=n-1-Q}^{n-1+Q} x(p\Delta_t) e^{-j\frac{2\pi pm}{N}} \Delta_t$$

- Calculate $X(\min(n)\Delta_t, m\Delta_f)$ by the N -point FFT

$$X(n_0\Delta_t, m\Delta_f) = \Delta_t e^{j\frac{2\pi(Q-n_0)m}{N}} \sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}}, \quad n_0 = \min(n),$$

$$x_1(q) = x((n-Q+q)\Delta_t) \quad \text{for } q \leq 2Q, \quad x_1(q) = 0 \quad \text{for } q > 2Q$$

$\mathcal{O}(TF)$

- Applying the recursive formula to calculate $X(n\Delta_t, m\Delta_f)$,

$$F + N \log N + (T-1)(2F) \approx (T-1)(2F) \underset{p=n-1-Q}{\approx} 2TF$$

$$n = n_0 + 1 \sim \max(n)$$

$$X(n\Delta_t, m\Delta_f) = X((n-1)\Delta_t, m\Delta_f) - x((n-Q-1)\Delta_t) e^{-j2\pi(n-Q-1)m/N} \Delta_t$$

T 點 F 點

$$+ x((n+Q)\Delta_t) e^{-j2\pi(n+Q)m/N} \Delta_t$$

$\nwarrow p = n+Q$

IV-D Method 4: Chirp Z Transform

$$\exp(-j2\pi pm\Delta_t\Delta_f) = \exp(-j\pi p^2\Delta_t\Delta_f) \exp(j\pi(p-m)^2\Delta_t\Delta_f) \exp(-j\pi m^2\Delta_t\Delta_f)$$

$p^2 \geq pm + m^2$

For the STFT

$$X(n\Delta_t, m\Delta_f) = \sum_{p=n-Q}^{n+Q} w((n-p)\Delta_t) x(p\Delta_t) e^{-j2\pi pm\Delta_t\Delta_f}$$

$$Y[n] = X[n] * h[n]$$

$$Y[n] = \sum_m X[m] h[n-m]$$

$$Y(f) = X(f) H(f)$$

$$Y[n] = \text{IFT}(\text{FT}(x(n)) \text{FT}(h[n]))$$

$$X(n\Delta_t, m\Delta_f) = \Delta_t e^{-j\pi m^2\Delta_t\Delta_f} \sum_{p=n-Q}^{n+Q} w((n-p)\Delta_t) x(p\Delta_t) e^{-j\pi p^2\Delta_t\Delta_f} e^{j\pi(p-m)^2\Delta_t\Delta_f}$$

Step 1 multiplication

Step 2 convolution

Step 3 multiplication

$$T \cdot 3N \log N \quad O(TN \log N)$$

Step 1 $x_1[p] = w((n-p)\Delta_t) x(p\Delta_t) e^{-j\pi p^2 \Delta_t \Delta_f}$ $n-Q \leq p \leq n+Q$

Step 2 $\underline{X_2[n,m]} = \sum_{p=n-Q}^{n+Q} \underline{x_1[p]} \underline{c[m-p]} \quad c[m] = e^{j\pi m^2 \Delta_t \Delta_f}$

Step 3 $X(n\Delta_t, m\Delta_f) = \Delta_t e^{-j\pi m^2 \Delta_t \Delta_f} X_2[n, m]$

Step 2 在計算上，需要用到 linear convolution 的技巧

Question: Step 2 要用多少點的 DFT?

- Illustration for the Question on Page 120

$$y[n] = \sum_k x[n-k]h[k]$$

- Case 1

When $\text{length}(x[n]) = N$, $\text{length}(h[n]) = K$, N and K are finite,

—————> $\text{length}(y[n]) = N+K-1$,

Using the $(N+K-1)$ -point DFTs (學信號處理的人一定要知道的常識)

- Case 2

$x[n]$ has finite length but $h[n]$ has infinite length ????

$$y[n] = \sum_k x[n-k]h[k]$$

- Case 2

$x[n]$ has finite length but $h[n]$ has infinite length

$x[n]$ 的範圍為 $n \in [n_1, n_2]$ ，範圍大小為 $N = n_2 - n_1 + 1$

$h[n]$ 無限長

$$y[n] = \sum_k x[n-k]h[k] \quad y[n] \text{ 每一點都有值 (範圍無限大)}$$

但我們只想求出 $y[n]$ 的其中一段

希望算出的 $y[n]$ 的範圍為 $n \in [m_1, m_2]$ ，範圍大小為 $M = m_2 - m_1 + 1$

$h[n]$ 的範圍？

要用多少點的 FFT？

$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[n-k]h[k]$$

改寫成 $y[n] = x[n] * h[n] = \sum_{s=n_1}^{n_2} x[s]h[n-s]$

$$\begin{aligned} y[n] &= x[n_1]h[n-n_1] + x[n_1+1]h[n-n_1-1] + x[n_1+2]h[n-n_1-2] \\ &\quad + \dots + x[n_2]h[n-n_2] \end{aligned}$$

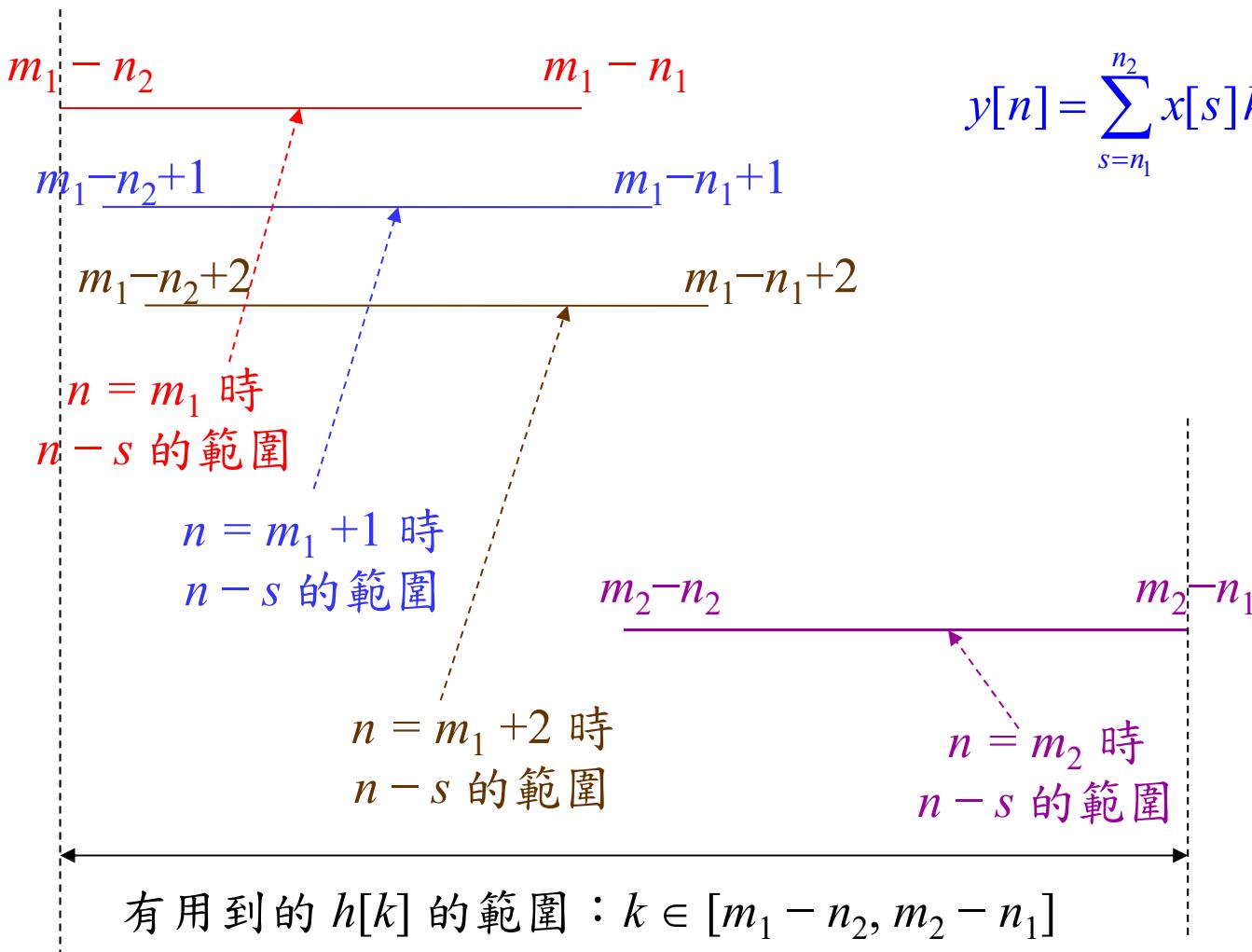
當 $n = m_1$

$$\begin{aligned} y[m_1] &= x[n_1]h[m_1-n_1] + x[n_1+1]h[m_1-n_1-1] + x[n_1+2]h[m_1-n_1-2] \\ &\quad + \dots + x[n_2]h[m_1-n_2] \end{aligned}$$

當 $n = m_2$

$$\begin{aligned} y[m_2] &= x[n_1]h[m_2-n_1] + x[n_1+1]h[m_2-n_1-1] + x[n_1+2]h[m_2-n_1-2] \\ &\quad + \dots + x[n_2]h[m_2-n_2] \end{aligned}$$

$$y[n] = \sum_{s=n_1}^{n_2} x[s]h[n-s]$$



所以，有用到的 $h[k]$ 的範圍是 $k \in [m_1 - n_2, m_2 - n_1]$

範圍大小為 $m_2 - n_1 - m_1 + n_2 + 1 = N + M - 1$

FFT implementation for Case 2

$$x_1[n] = x[n + n_1] \quad \text{for } n = 0, 1, 2, \dots, N-1$$

$$x_1[n] = 0 \quad \text{for } n = N, N+1, N+2, \dots, L-1 \quad L = N + M - 1$$

$$h_1[n] = h[n + m_1 - n_2] \quad \text{for } n = 0, 1, 2, \dots, L-1$$

$$y_1[n] = IFFT_L \left(FFT_L \{x_1[n]\} FFT_L \{h_1[n]\} \right)$$

$$y[n] = y_1[n - m_1 + N - 1] \quad \text{for } n = m_1, m_1+1, m_1+2, \dots, m_2$$

IV-E Unbalanced Sampling for STFT and WDF

將 pages 111 and 115 的方法作修正

$$\begin{aligned} X(t, f) &= \int_{-\infty}^{\infty} w(t - \tau) x(\tau) e^{-j2\pi f \tau} d\tau \\ &\quad \downarrow \\ X(n\Delta_t, m\Delta_f) &= \sum_{p=nS-Q}^{nS+Q} w((nS - p)\Delta_\tau) x(p\Delta_\tau) e^{-j2\pi pm\Delta_\tau \Delta_f} \Delta_\tau \end{aligned}$$

$w(t - \tau)$
 $= w(n\Delta_t - p\Delta_\tau)$
 $= w(nS\Delta_\tau - p\Delta_\tau)$

where $t = n\Delta_t$, $f = m\Delta_f$, $\tau = p\Delta_\tau$, $B = Q\Delta_\tau$ (假設 $w(t) \approx 0$ for $|t| > B$),

$$S = \Delta_t / \Delta_\tau$$

$$\Delta_t \neq \Delta_\tau$$

Ex: $\Delta_\tau = \frac{1}{44100}$
 $\Delta_t = \frac{1}{100}$, $S = 441$

註： Δ_τ (sampling interval for the input signal)

Δ_t (sampling interval for the output t-axis) can be different.

However, it is better that $S = \Delta_t / \Delta_\tau$ is an integer.

When (1) $\Delta_\tau \Delta_f = 1/N$, (2) $N = 1/(\Delta_\tau \Delta_f) > 2Q + 1$: ($\Delta_\tau \Delta_f$ 只要是整數的倒數即可)

(3) $\Delta_\tau < 1/2\Omega$, Ω is the bandwidth of $w(\tau - t)x(\tau)$

i.e., $|FT\{w(\tau - t)x(\tau)\}| = |X(t, f)| \approx 0$ when $|f| > \Omega$

$$X(n\Delta_t, m\Delta_f) = \sum_{p=nS-Q}^{nS+Q} w((nS-p)\Delta_\tau) x(p\Delta_\tau) e^{-j\frac{2\pi pm}{N}} \Delta_\tau$$

令 $q = p - (nS - Q) \rightarrow p = (nS - Q) + q$

$$X(n\Delta_t, m\Delta_f) = \Delta_\tau e^{j\frac{2\pi(Q-nS)m}{N}} \sum_{q=0}^{N-1} x_1(q) e^{-j\frac{2\pi qm}{N}}$$

不同一： $n \Rightarrow nS$
 不同二： $\Delta_\tau \Delta_f : \frac{1}{N}$
 (not Δ_t)

$$x_1(q) = w((Q-q)\Delta_\tau) x((nS-Q+q)\Delta_\tau) \quad \text{for } 0 \leq q \leq 2Q,$$

$$x_1(q) = 0 \quad \text{for } 2Q < q < N.$$

If $w(t) = w(-t)$

$$x_1(q) = w(k\Delta_\tau) x((nS+k)\Delta_\tau) \quad \text{for } 0 \leq q \leq 2Q, \quad k = q - Q$$

$$x_1(q) = 0 \quad \text{for } 2Q < q < N.$$

假設 $t = c_0\Delta_t, (c_0+1)\Delta_t, (c_0+2)\Delta_t, \dots, (c_0+C-1)\Delta_t$

$$= c_0S\Delta_\tau, (c_0S+S)\Delta_\tau, (c_0S+2S)\Delta_\tau, \dots, [c_0S+(C-1)S]\Delta_\tau$$

$$f = m_0\Delta_f, (m_0+1)\Delta_f, (m_0+2)\Delta_f, \dots, (m_0+F-1)\Delta_f$$

$$\tau = n_0\Delta_\tau, (n_0+1)\Delta_\tau, (n_0+2)\Delta_\tau, \dots, (n_0+T-1)\Delta_\tau$$

$$S = \Delta_t / \Delta_\tau$$

Step 1: Calculate $c_0, m_0, n_0, C, F, T, N, Q$

Step 2: $n = c_0$

Step 3: Determine $x_1(q)$

*compared with the original method,
 $n \Rightarrow nS$*

Step 4: $X_1(m) = \text{FFT}[x_1(q)]$

Step 5: Convert $X_1(m)$ into $X(n\Delta_t, m\Delta_f)$

Step 6: Set $n = n+1$ and return to Step 3 until $n = c_0+C-1$.

Complexity = ?

IV-F Non-Uniform Δ_t

(A) 先用較大的 Δ_t

(B) 如果發現 $|X(n\Delta_t, m\Delta_f)|$ 和 $|X((n+1)\Delta_t, m\Delta_f)|$ 之間有很大的差異

則在 $n\Delta_t, (n+1)\Delta_t$ 之間選用較小的 sampling interval Δ_{t1}

$(\Delta_\tau < \Delta_{t1} < \Delta_t, \Delta_t/\Delta_{t1} \text{ 和 } \Delta_{t1}/\Delta_\tau \text{ 皆為整數})$

再用 page 127 的方法算出

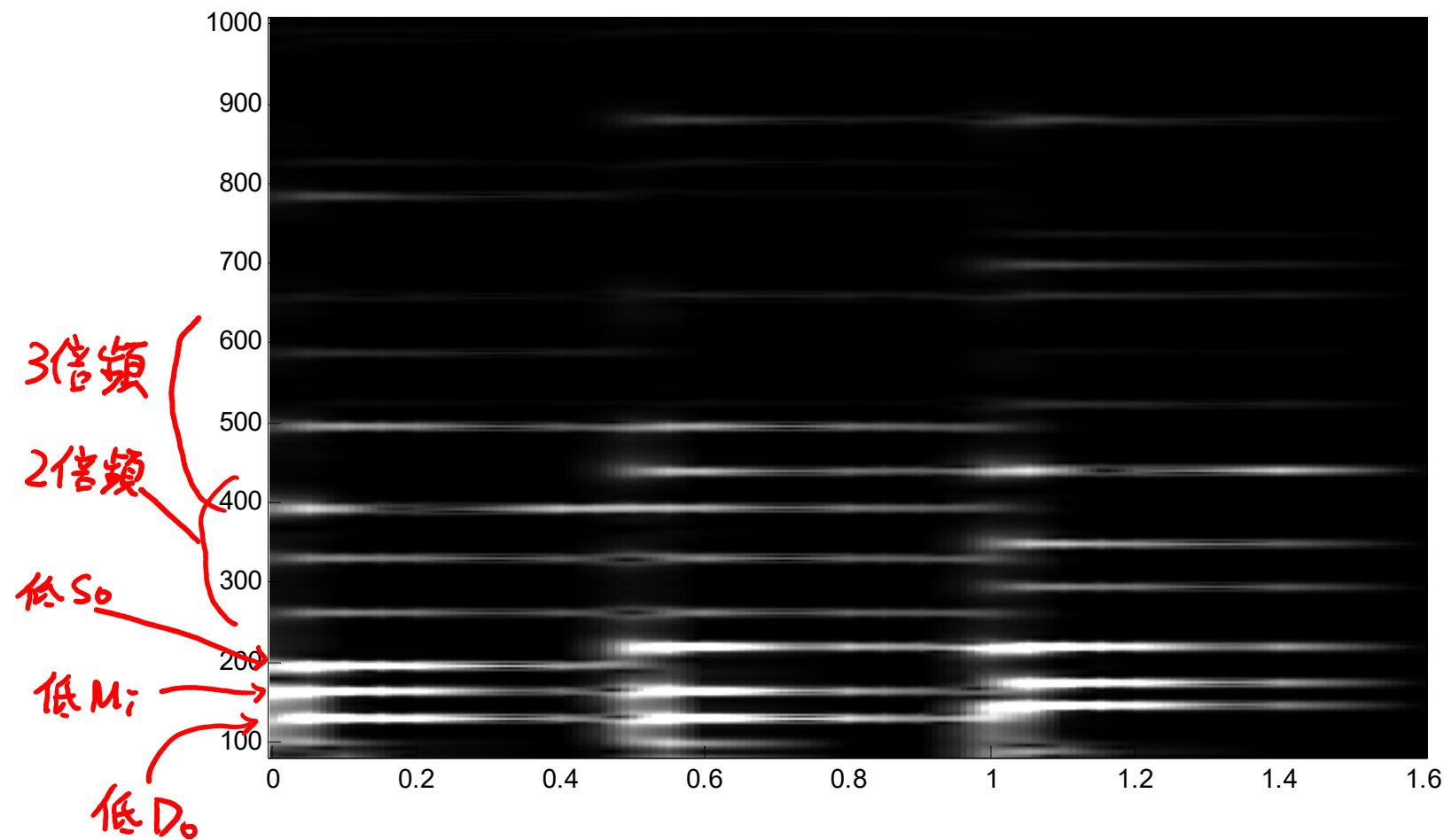
$$X(n\Delta_t + \Delta_{t1}, m\Delta_f), X(n\Delta_t + 2\Delta_{t1}, m\Delta_f), \dots, X((n+1)\Delta_t - \Delta_{t1}, m\Delta_f)$$

(C) 以此類推，如果 $|X(n\Delta_t + k\Delta_{t1}, m\Delta_f)|, |X(n\Delta_t + (k+1)\Delta_{t1}, m\Delta_f)|$

的差距還是太大，則再選用更小的 sampling interval Δ_{t2}

$(\Delta\tau < \Delta_{t2} < \Delta_{t1}, \Delta_{t1}/\Delta_{t2} \text{ 和 } \Delta_{t2}/\Delta\tau \text{ 皆為整數})$

Gabor transform of a music signal



$$\Delta_\tau = 1/44100 \text{ (總共有 } 44100 \times 1.6077 \text{ sec} + 1 = 70902 \text{ 點)}$$

(A) Choose $\Delta_t = \Delta_\tau$

running time = out of memory

(B) Choose $\Delta_t = 0.01 = 441\Delta_\tau$ ($1.6/0.01 + 1 = 161$ points)

running time = 1.0940 sec (2008年)

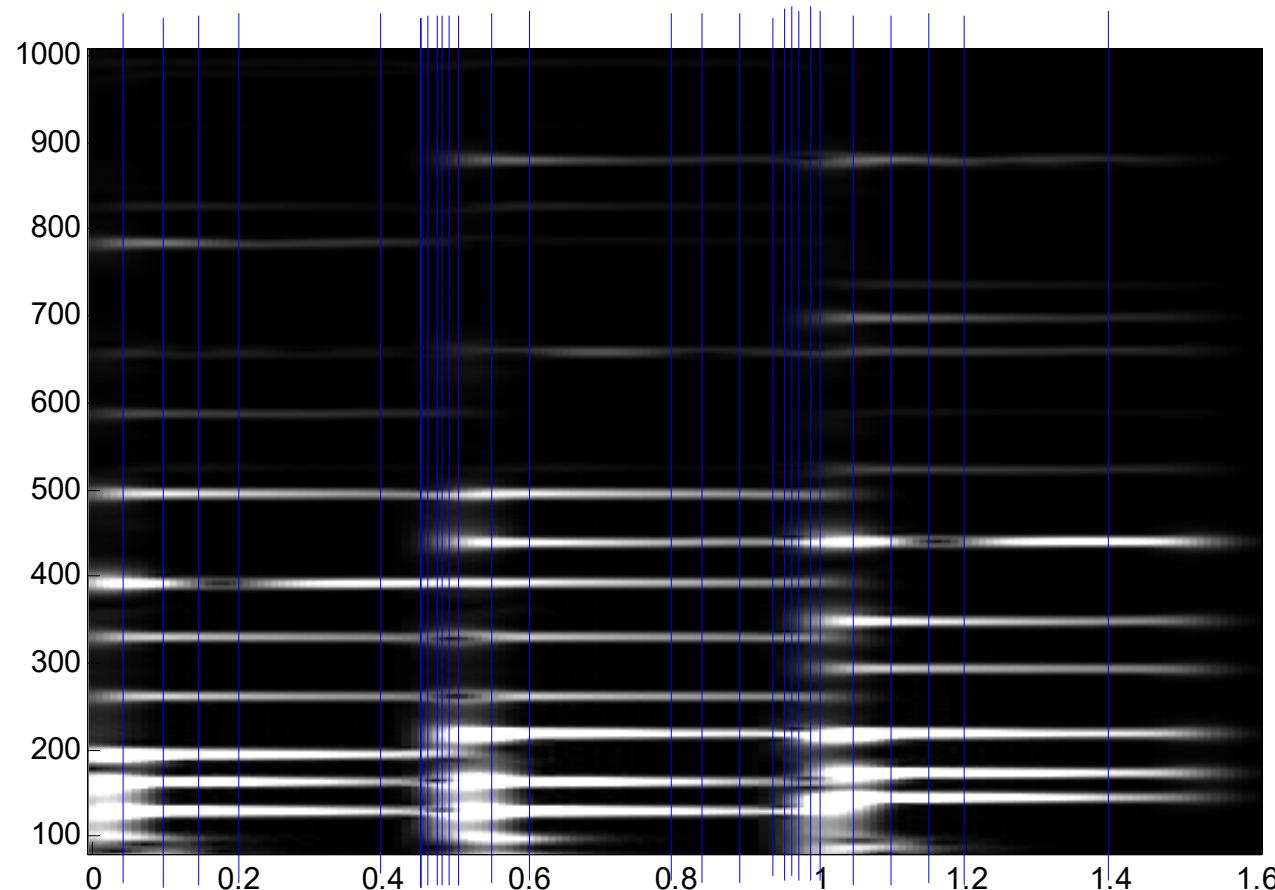
(C) Choose the sampling points on the t -axis as

$$t = \underline{0}, \underline{0.05}, \underline{0.1}, \underline{0.15}, \underline{0.2}, \underline{0.4}, \underline{0.45}, 0.46, 0.47, 0.48, 0.49, \underline{0.5}, \underline{0.55}, \underline{0.6}, \underline{0.8}, \\ \underline{0.85}, \underline{0.9}, \underline{0.95}, 0.96, 0.97, 0.98, 0.99, \underline{1}, \underline{1.05}, \underline{1.1}, \underline{1.15}, \underline{1.2}, \underline{1.4}, \underline{1.6}$$

(29 points)

running time = 0.2970 sec $\Delta t = 0.2 \Rightarrow \Delta t = 0.05 \Rightarrow \Delta t = 0.01$

with adaptive output sampling intervals



附錄七 和 Dirac Delta Function 相關的常用公式

$$(1) \quad \int_{-\infty}^{\infty} e^{-j2\pi t f} dt = \delta(f)$$

$$\int \delta(t) dt = 1$$

$$(2) \quad \delta(t) = |a| \delta(at) \quad (\text{scaling property})$$

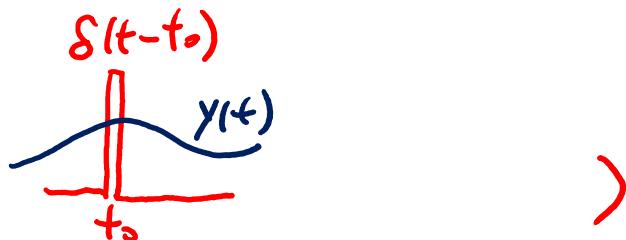
$$\delta(at)$$

$$(3) \quad \int_{-\infty}^{\infty} e^{-j2\pi t g(f)} dt = \delta(g(f)) = \sum_n |g'(f_n)|^{-1} \delta(f - f_n)$$

where f_n are the zeros of $g(f)$

$$(4) \quad \int_{-\infty}^{\infty} \delta(t - t_0) y(t, \dots) dt = y(t_0, \dots) \quad (\text{sifting property I})$$

$$(5) \quad \delta(t - t_0) y(t, \dots) = \delta(t - t_0) y(t_0, \dots) \quad (\text{sifting property II})$$



V. Wigner Distribution Function

V-A Wigner Distribution Function (WDF)

Definition 1:
$$W_x(t, f) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} d\tau$$

Definition 2:
$$W_x(t, \omega) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j\omega\tau} d\tau$$

Another way for computation from the frequency domain

(output is the same)

Definition 1: $W_x(t, f) = \int_{-\infty}^{\infty} X(f + \eta/2) \cdot X^*(f - \eta/2) e^{j2\pi\eta t} d\eta$
 where $X(f)$ is the Fourier transform of $x(t)$

Definition 2: $W_x(t, \omega) = \int_{-\infty}^{\infty} X(\omega + \eta/2) \cdot X^*(\omega - \eta/2) e^{j\eta t} d\eta$

Main Reference

[Ref] S. Qian and D. Chen, *Joint Time-Frequency Analysis: Methods and Applications*, Chap. 5, Prentice Hall, N.J., 1996.

Other References

- [Ref] E. P. Wigner, “On the quantum correlation for thermodynamic equilibrium,” *Phys. Rev.*, vol. 40, pp. 749-759, 1932.
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- [Ref] F. Hlawatsch and G. F. Boudreux–Bartels, “Linear and quadratic time-frequency signal representation,” *IEEE Signal Processing Magazine*, pp. 21-67, Apr. 1992.
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The operators that are related to the WDF:

(a) Signal auto-correlation function:

$$C_x(t, \tau) = x(t + \tau/2) \cdot x^*(t - \tau/2)$$

*tt 較: in random process
auto-correlation =*

$$R(t, \tau) = E(x(t)x^*(t + \tau))$$

(b) Spectrum auto-correlation function:

$$S_x(\eta, f) = X(f + \eta/2) \cdot X^*(f - \eta/2)$$

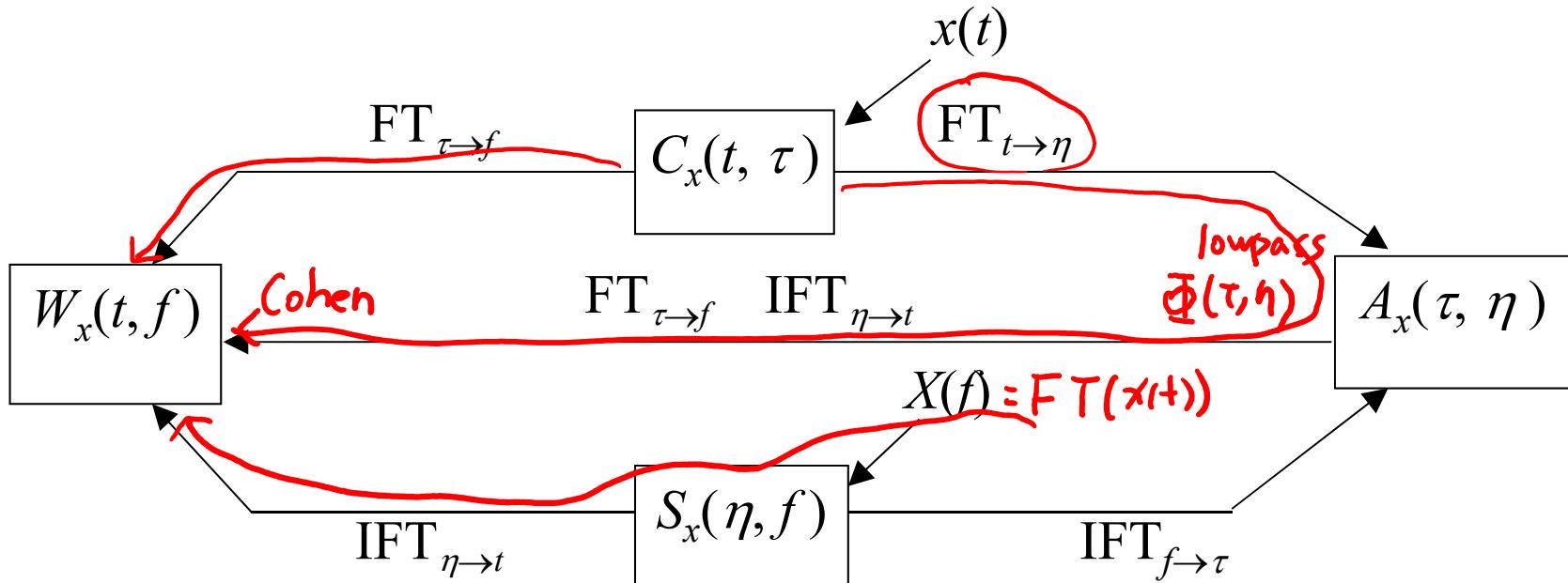
$$\mathcal{F}_{\tau \rightarrow f}(R(t, \tau)) = S(t, f)$$

*power spectral
density (PSD)*

(c) Ambiguity function (AF):

$$A_x(\tau, \eta) = \int_{-\infty}^{\infty} x(t + \tau/2) x^*(t - \tau/2) e^{-j2\pi t\eta} dt$$

$\eta: / \text{rad}/$



V-B Why the WDF Has Higher Clarity?

Due to signal auto-correlation function

$$(1) \text{ If } x(t) = 1 \quad h=0$$

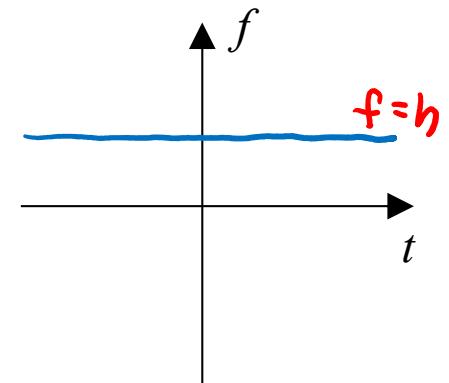
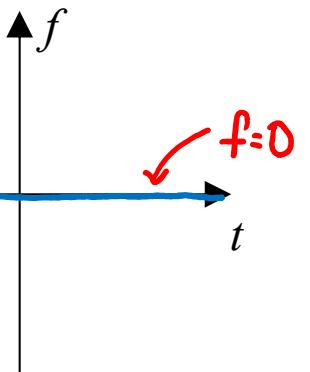
$$(2) \text{ If } x(t) = \exp(j2\pi h t) \quad x(t+\frac{\tau}{2}) x^*(t-\frac{\tau}{2}) \quad \frac{\phi'(t)}{2\pi} = h \text{ for all } t$$

$$W_x(t, f) = \int_{-\infty}^{\infty} e^{j2\pi h(t+\tau/2)} e^{-j2\pi h(t-\tau/2)} \cdot e^{-j2\pi \tau f} d\tau$$

$$= \int_{-\infty}^{\infty} e^{j2\pi h\tau} \cdot e^{-j2\pi \tau f} d\tau$$

$$= \int_{-\infty}^{\infty} e^{-j2\pi \tau(f-h)} d\tau \quad g(f) = f - h \quad \text{page 133 (3)}$$

$$= \underline{\delta(f-h)}$$



Comparing: for the case of the STFT

$$\frac{\partial'(t)}{2\pi} = 2kt$$

(3) If $x(t) = \exp(j2\pi k t^2)$

$$\begin{aligned} W_x(t, f) &= \int x(t + \frac{\tau}{2}) x^*(t - \frac{\tau}{2}) e^{-j2\pi f \tau} d\tau \\ &= \int e^{j2\pi k(t + \frac{\tau}{2})^2} e^{-j2\pi k(t - \frac{\tau}{2})^2} e^{-j2\pi f \tau} d\tau \\ &= \int e^{j2\pi k(2t\tau)} e^{-j2\pi f \tau} d\tau \\ &= \int e^{-j2\pi \tau(f - 2tk)} d\tau \\ &= \delta(f - 2kt) \quad (\text{page 133 (3)}) \end{aligned}$$

(4) If $x(t) = \delta(t)$

$$\begin{aligned} W_x(t, f) &= \int_{-\infty}^{\infty} \delta(t + \tau/2) \cdot \delta(t - \tau/2) e^{-j2\pi \tau f} d\tau \\ &\stackrel{t \rightarrow \tau, t_0 \rightarrow -2t}{=} 4 \int_{-\infty}^{\infty} \delta(2t + \tau) \underbrace{\delta(2t - \tau) e^{-j2\pi \tau f} d\tau}_{y(\tau)} \\ &= 4 \delta(4t) e^{j4\pi t f} = \delta(t) e^{j4\pi t f} = \delta(t) \end{aligned}$$

Page 133
公式(2)

Page 133

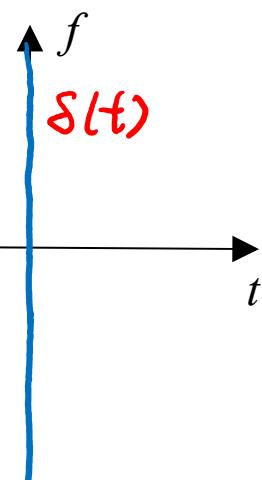
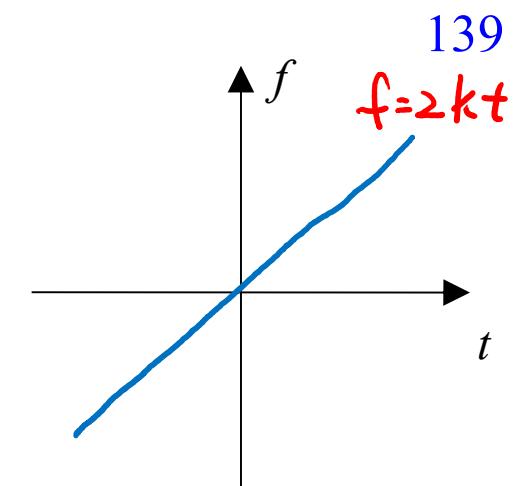
公式(4)

Page 133

公式(2)

Page 133

公式(5), $t_0 = 0$



V-C The WDF is not a Linear Distribution

$$W_x(t, f) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau$$

If $h(t) = \alpha g(t) + \beta s(t)$

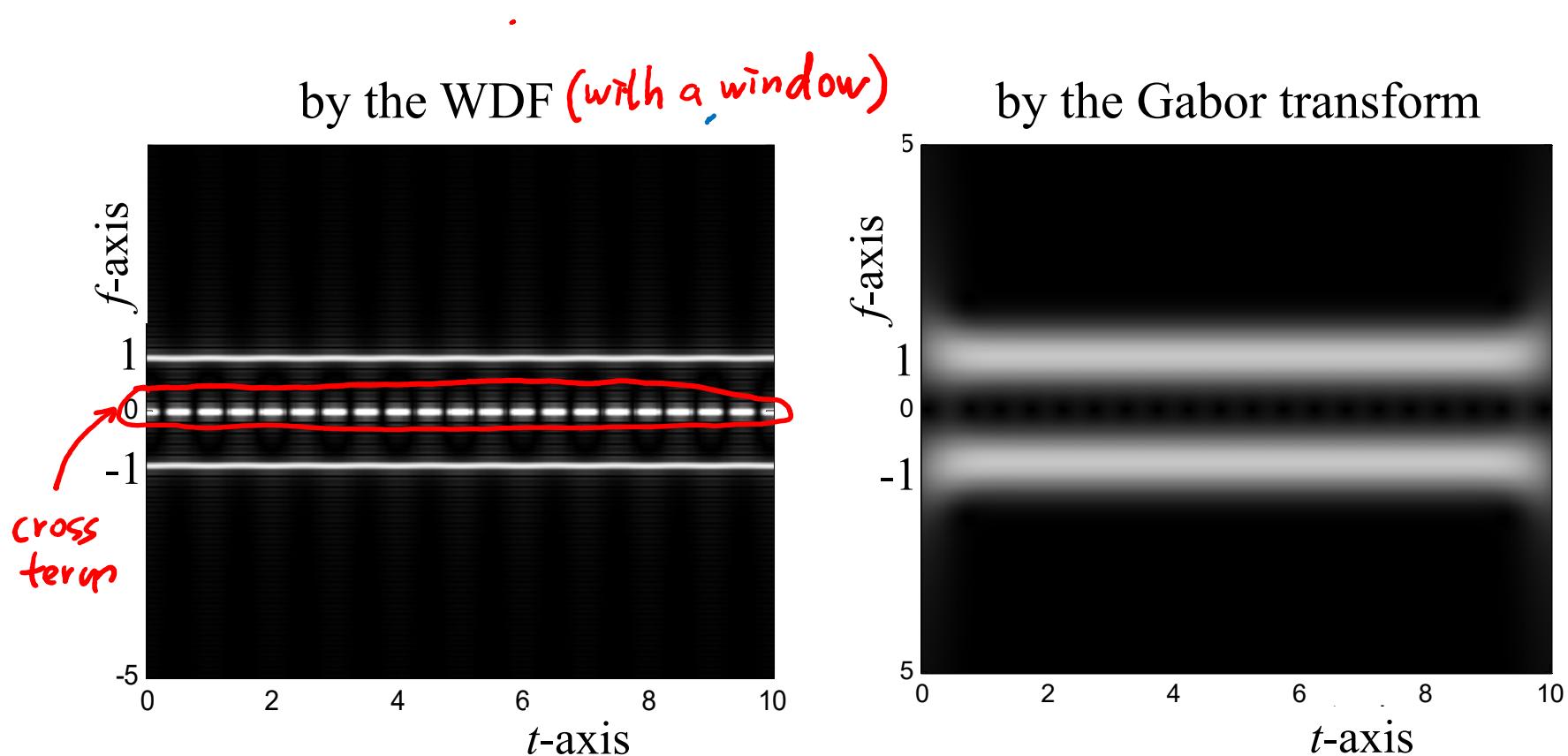
$$\begin{aligned}
 W_h(t, f) &= \int_{-\infty}^{\infty} h(t + \tau/2) \cdot h^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau \\
 &= \int_{-\infty}^{\infty} [\alpha g(t + \tau/2) + \beta s(t + \tau/2)] [\alpha^* g^*(t - \tau/2) + \beta^* s^*(t - \tau/2)] e^{-j2\pi\tau f} d\tau \\
 &= \int_{-\infty}^{\infty} [|\alpha|^2 g(t + \tau/2)g^*(t - \tau/2) + |\beta|^2 s(t + \tau/2)s^*(t - \tau/2) \\
 &\quad + \alpha\beta^* g(t + \tau/2)s^*(t - \tau/2) + \alpha^*\beta g^*(t - \tau/2)s(t + \tau/2)] e^{-j2\pi\tau f} d\tau \\
 &= \underbrace{|\alpha|^2 W_g(t, f)}_{\text{auto terms}} + \underbrace{|\beta|^2 W_s(t, f)}_{\text{auto terms}} \\
 &\quad + \int_{-\infty}^{\infty} [\alpha\beta^* g(t + \tau/2)s^*(t - \tau/2) + \alpha^*\beta g^*(t - \tau/2)s(t + \tau/2)] e^{-j2\pi\tau f} d\tau
 \end{aligned}$$

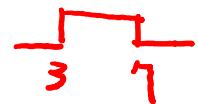
cross terms

V-D Examples of the WDF

Simulations

$$x(t) = \cos(2\pi t) = 0.5[\exp(j2\pi t) + \exp(-j2\pi t)]$$



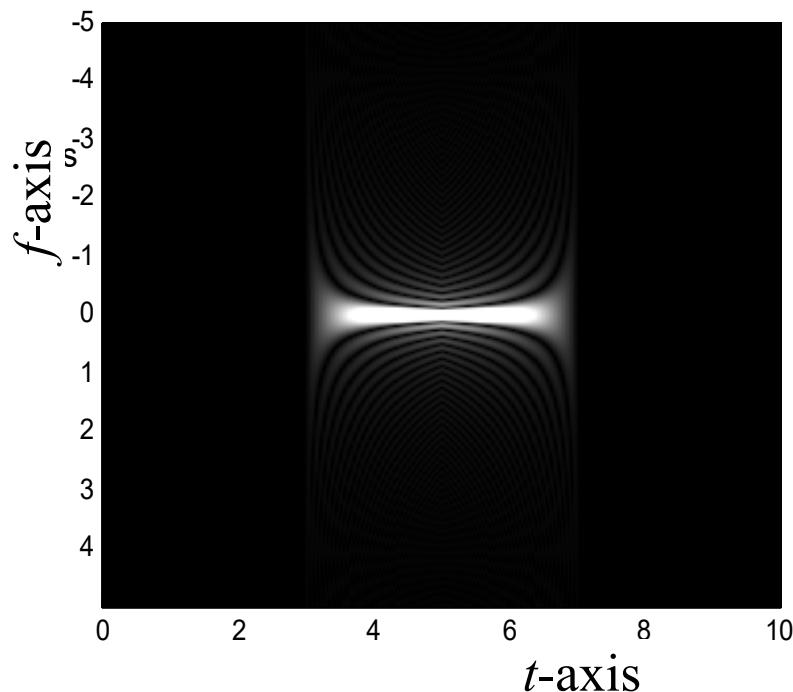


$$x(t) = \Pi((t - 5)/4)$$

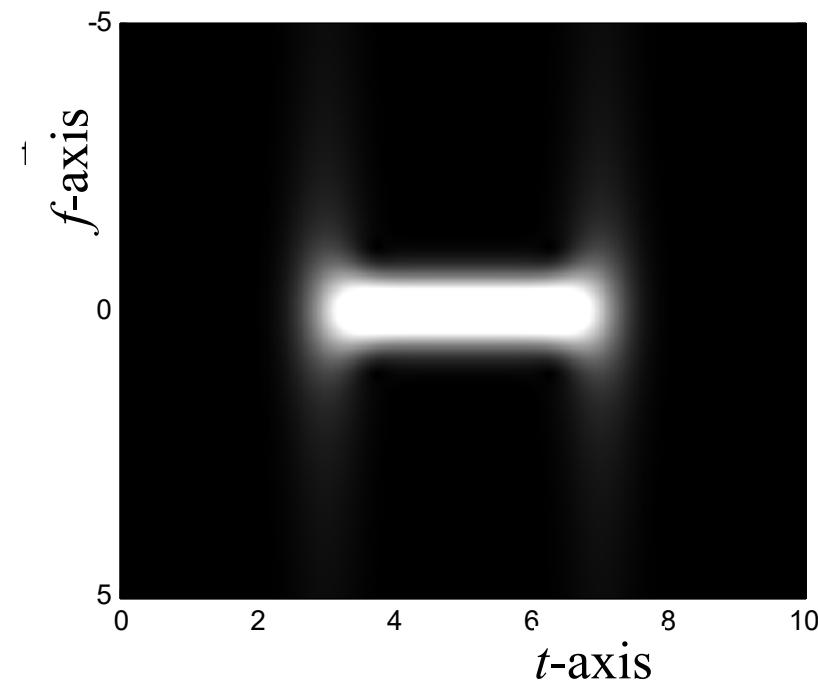
center width

Π : rectangular function

by the WDF

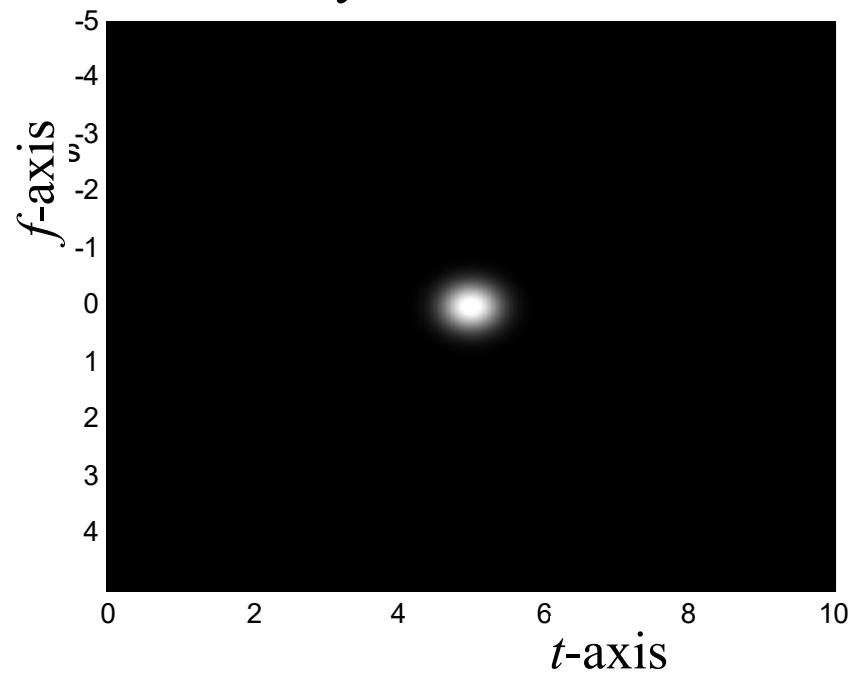


by the Gabor transform

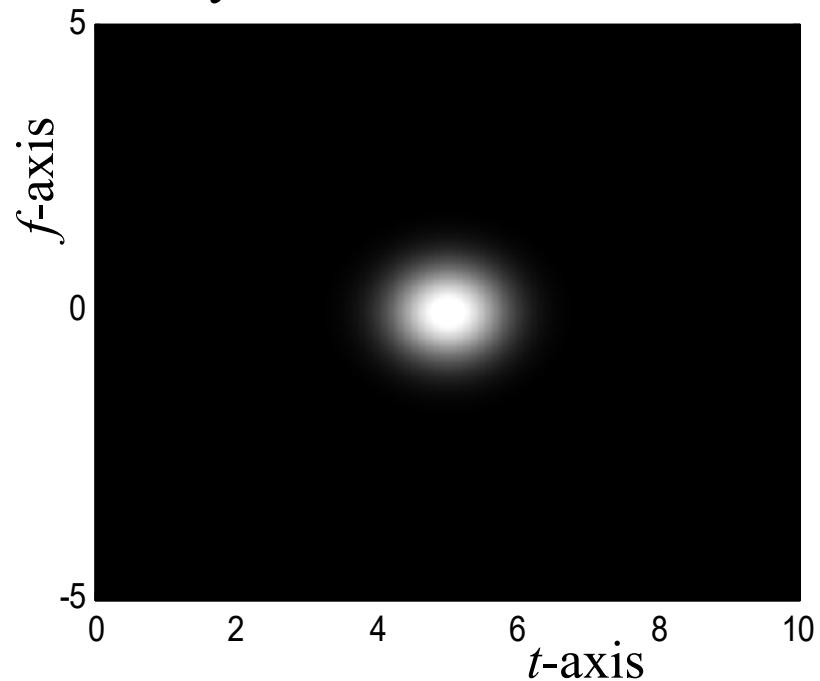


$$x(t) = \exp[-\pi(t-5)^2]$$

by the WDF



by the Gabor transform



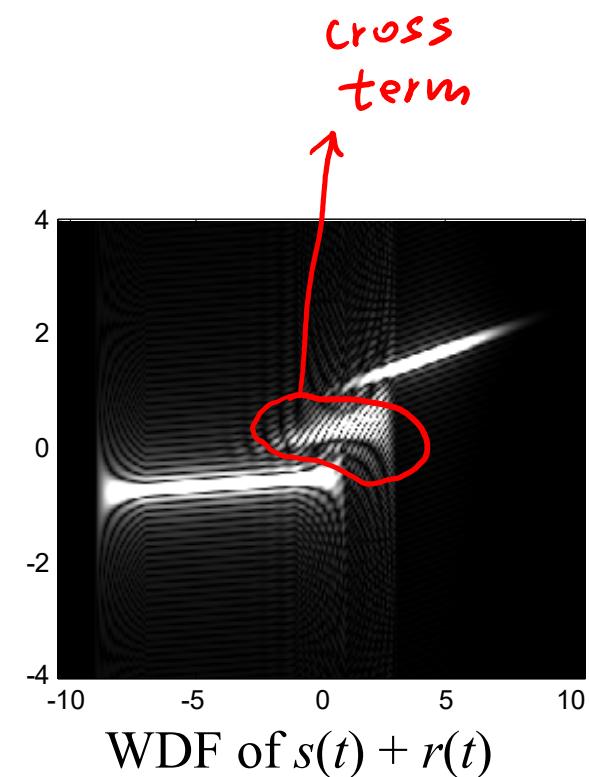
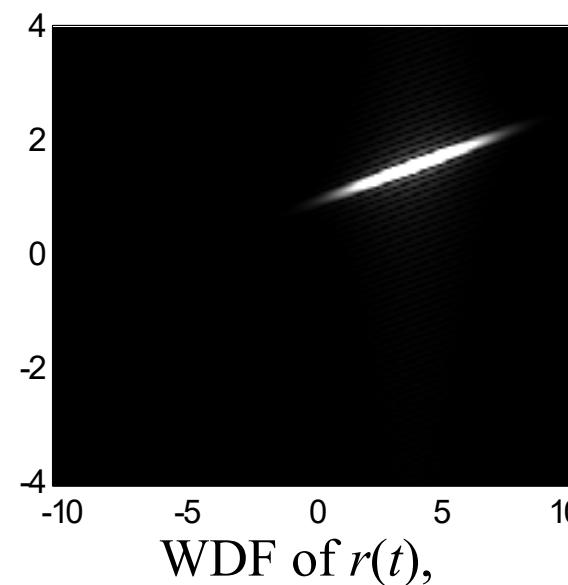
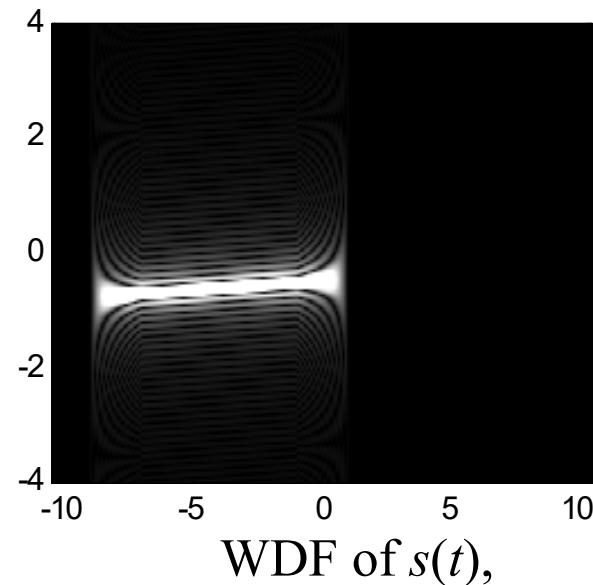
Gaussian function: $e^{-\pi t^2} \xrightarrow{FT} e^{-\pi f^2}$

Gaussian function's T-F area is minimal.

$$s(t) = \exp(jt^2/10 - j3t) \quad \text{for } -9 \leq t \leq 1, s(t) = 0 \text{ otherwise,}$$

$$r(t) = \exp(jt^2/2 + j6t) \exp[-(t-4)^2/10]$$

$$f(t) = s(t) + r(t)$$

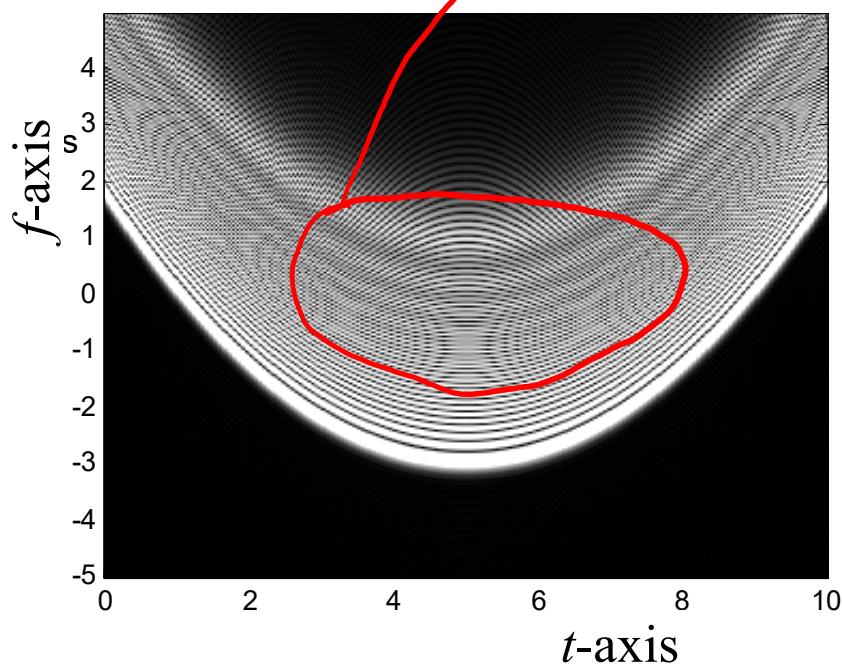


橫軸: t -axis, 縱軸: f -axis

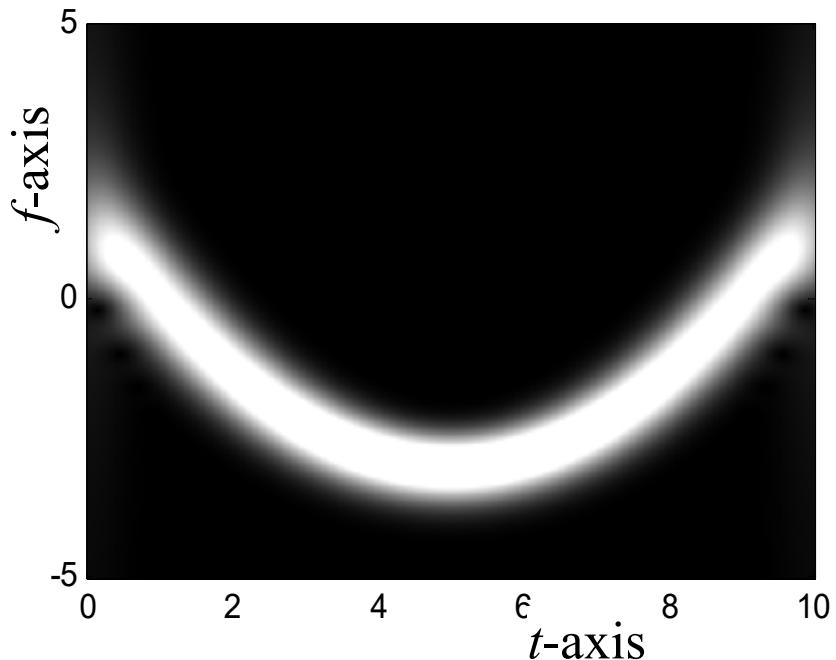
$$\frac{\phi'(t)}{2\pi} = \frac{3}{2\pi}(t-5)^2 - 3$$

$$x(t) = \exp(j(t-5)^3 - j6\pi t)$$

by the WDF
 ↗ cross term
 (with a window)



by the Gabor transform



V-E Digital Implementation of the WDF

$$W_x(t, f) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau ,$$

$$W_x(t, f) = 2 \int_{-\infty}^{\infty} x(t + \tau') \cdot x^*(t - \tau') e^{-j4\pi\tau' f} \cdot d\tau' \quad (\text{using } \underline{\tau' = \tau/2})$$

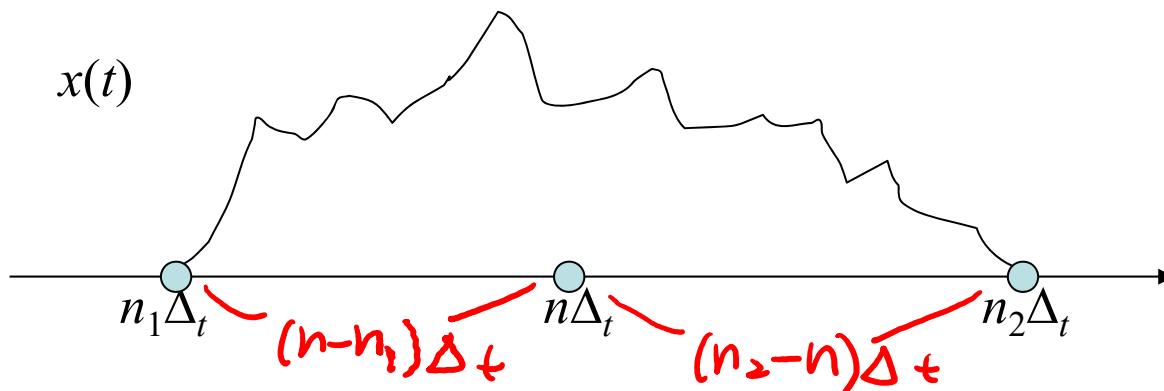
$d\tau' = d\tau/2$
 $d\tau = 2d\tau'$

Sampling: $t = n\Delta_t$, $f = m\Delta_f$, $\tau' = p\Delta_t$

$$W_x(n\Delta_t, m\Delta_f) = 2 \sum_{p=-\infty}^{\infty} x((n+p)\Delta_t) x^*((n-p)\Delta_t) \exp(-j4\pi mp\Delta_t \Delta_f) \Delta_t$$

When $x(t)$ is not a time-limited signal, it is hard to implement.

Suppose that $\underline{x(t) = 0}$ for $\underline{t < n_1 \Delta_t}$ and $\underline{t > n_2 \Delta_t}$



$$x((n+p)\Delta_t)x^*((n-p)\Delta_t) = 0 \quad \text{if } n+p \notin [n_1, n_2] \\ \text{or } n-p \notin [n_1, n_2]$$

• p 的範圍的問題 (當 n 固定時)

$$\underline{n_1 \leq n+p \leq n_2} \longrightarrow \underline{n_1 - n \leq p \leq n_2 - n}$$

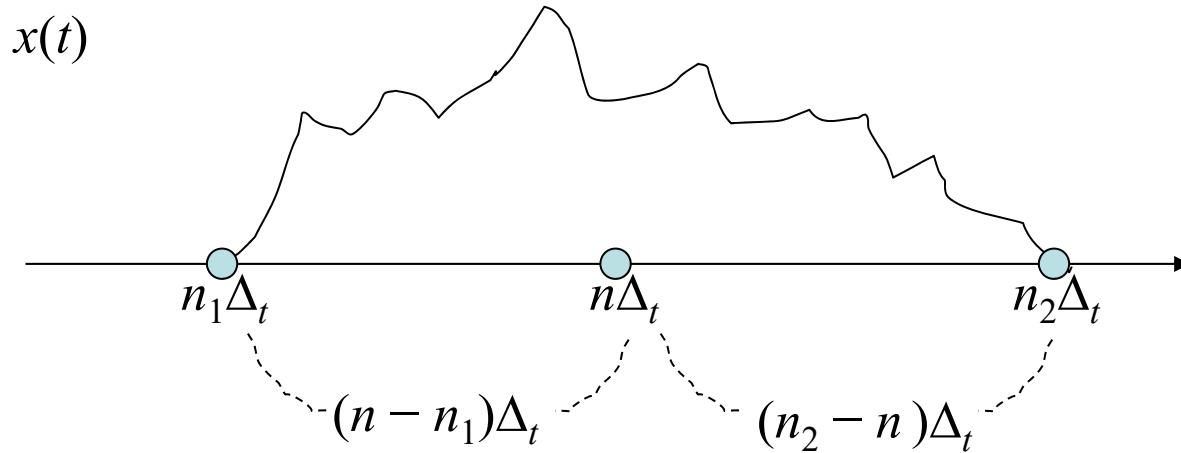
$$\underline{n_1 \leq n-p \leq n_2} \longrightarrow n_1 - n \leq -p \leq n_2 - n, \quad \underline{n - n_2 \leq p \leq n - n_1}$$

$$\max(n_1 - n, n - n_2) \leq p \leq \min(n_2 - n, n - n_1)$$

$$\max(-3, -5)$$

$$\underline{-\min(n_2 - n, n - n_1) \leq p \leq \min(n_2 - n, n - n_1)}$$

$$= -\min(3, 5)$$



$$-\min(n_2 - n, n - n_1) \leq p \leq \min(n_2 - n, n - n_1)$$

$\neg Q$ Q

$(n_2 - n) \Delta_t, (n - n_1) \Delta_t$: 離兩個邊界的距離

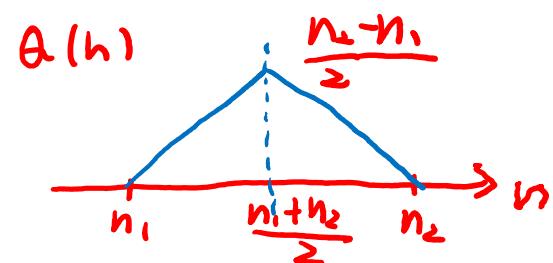
注意：當 $n > n_2$ 或 $n < n_1$ 時，

將沒有 p 能滿足上面的不等式

If $x(t) = 0$ for $t < n_1 \Delta_t$ and $t > n_2 \Delta_t$

$$W_x(n\Delta_t, m\Delta_f) = \sum_{p=-Q}^Q x((n+p)\Delta_t)x^*((n-p)\Delta_t) \exp(-j4\pi mp\Delta_t\Delta_f)\Delta_t$$

T點 F點



$$\underline{Q} = \min(n_2 - n_1, n - n_1). \quad (\text{varies with } n)$$

$$p \in [-Q, Q], \quad n \in [n_1, n_2],$$

possible for implementation

Method 1: Direct Implementation (brute force method)

唯一的限制條件？

$$\text{Since WDF} = \lim_{\tau \rightarrow f} F(x(t+\frac{\tau}{2})x^*(t-\frac{\tau}{2}))$$

If Ω_x is the bandwidth of $x(\tau)$

bandwidth of $x(t+\frac{\tau}{2})x^*(t-\frac{\tau}{2})$

$$= \frac{1}{2} \text{ bw of } x(2t+\tau)x^*(2t-\tau) = \frac{1}{2}(2\Omega_x)$$

complexity: $TF \cdot \text{mean}(Q)$

$$= TF \frac{n_2 - n_1}{4} \cong \frac{T^2 F}{4}$$

$\Theta(T^2 F)$

$$\tau' = \frac{\tau}{2}, \Delta\tau' = \frac{\Delta\tau}{2} = \Delta t$$

$$= \Omega_x \because \Delta\tau < \frac{1}{2\Omega_x}, 2\Delta t < \frac{1}{2\Omega_x}, \Delta t < \frac{1}{4\Omega_x}$$

3 大限制條件

Method 2: Using the DFT

When $\Delta_t \Delta_f = \frac{1}{2N}$ and $N \geq 2\text{Max}(Q)+1 = 2(n_2-n_1)/2+1 = n_2-n_1+1 = T$

$$W_x(n\Delta_t, m\Delta_f) = 2\Delta_t \sum_{p=-Q}^Q x((n+p)\Delta_t)x^*((n-p)\Delta_t)e^{-j\frac{2\pi mp}{N}}$$

T點 F點

$$q = p+Q \rightarrow p = q - Q$$

complexity: $\Theta(TN \log N) = \Theta(T^2 \log T)$

$$W_x(n\Delta_t, m\Delta_f) = 2\Delta_t e^{j\frac{2\pi mQ}{N}} \sum_{q=0}^{2Q} x((n+q-Q)\Delta_t)x^*((n-q+Q)\Delta_t)e^{-j\frac{2\pi mq}{N}}$$

$$W_x(n\Delta_t, m\Delta_f) = 2\Delta_t e^{j\frac{2\pi mQ}{N}} \sum_{q=0}^{N-1} c_1(q) e^{-j\frac{2\pi mq}{N}}$$

$$Q = \min(n_2-n, n-n_1).$$

$$n \in [n_1, n_2],$$

$$c_1(q) = x((n+q-Q)\Delta_t)x^*((n-q+Q)\Delta_t) \quad \text{for } 0 \leq q \leq 2Q$$

$$\text{i.e., } c_1(Q+k) = x((n+k)\Delta_t)x^*((n-k)\Delta_t) \quad \text{for } -Q \leq k \leq Q \quad (k = q-Q)$$

$$c_1(q) = 0 \quad \text{for } 2Q+1 \leq q \leq N-1$$

假設 $t = n_0\Delta_t, (n_0+1)\Delta_t, (n_0+2)\Delta_t, \dots, n_1\Delta_t$

$f = m_0\Delta_f, (m_0+1)\Delta_f, (m_0+2)\Delta_f, \dots, m_1\Delta_f$

Step 1: Calculate n_0, n_1, m_0, m_1, N

Step 2: $n = n_0$

Step 3: Determine Q

Step 4: Determine $c_1(q)$

Step 5: $C_1(m) = \text{FFT}[c_1(q)]$

Step 6: Convert $C_1(m)$ into $C(n\Delta_t, m\Delta_f)$

Step 7: Set $n = n+1$ and return to Step 3 until $n = n_1$.

Method 3: Using the Chirp Z Transform

$$W_x(n\Delta_t, m\Delta_f) = \underset{T}{2} \sum_{p=-Q}^Q x((n+p)\Delta_t) x^*((n-p)\Delta_t) \exp(-j4\pi mp\Delta_t\Delta_f) \Delta_t$$

$$W_x(n\Delta_t, m\Delta_f) = 2\Delta_t e^{-j2\pi m^2 \Delta_t \Delta_f} \sum_{p=-Q}^Q x((n+p)\Delta_t) x^*((n-p)\Delta_t) e^{-j2\pi p^2 \Delta_t \Delta_f} e^{j2\pi(p-m)^2 \Delta_t \Delta_f}$$

Step 1 $x_1(n, p) = x((n+p)\Delta_t) x^*((n-p)\Delta_t) e^{-j2\pi p^2 \Delta_t \Delta_f}$

Step 2 $X_2[n, m] = \sum_{p=-Q}^Q x_1[n, p] c[m-p] \quad c[m] = e^{j2\pi m^2 \Delta_t \Delta_f}$

Step 3 $X(n\Delta_t, m\Delta_f) = 2\Delta_t e^{-j2\pi m^2 \Delta_t \Delta_f} X_2[n, m]$

complexity = ? $\Theta()$

思考：Method 1 的複雜度為多少

思考：Method 2 的複雜度為多少

思考：Method 3 的複雜度為多少

The computation time of the WDF is more than those of the rec-STFT and the Gabor transform.

V-F Properties of the WDF

(1) Projection property	$ x(t) ^2 = \int_{-\infty}^{\infty} W_x(t, f) df \quad X(f) ^2 = \int_{-\infty}^{\infty} W_x(t, f) dt$
(2) Energy preservation property	$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W_x(t, f) dt df = \int_{-\infty}^{\infty} x(t) ^2 dt = \int_{-\infty}^{\infty} X(f) ^2 df$
(3) Recovery property	$\int_{-\infty}^{\infty} W_x(t/2, f) e^{j2\pi f t} df = x(t) \cdot x^*(0)$ $x^*(0)$ 已知 $\int_{-\infty}^{\infty} W_x(t, f/2) e^{-j2\pi f t} dt = X(f) \cdot X^*(0)$
(4) Mean condition frequency and mean condition time	If $x(t) = x(t) \cdot e^{j2\pi \phi(t)}$, $X(f) = X(f) \cdot e^{j2\pi \Psi(f)}$ then $\phi'(t) = x(t) ^{-2} \cdot \int_{-\infty}^{\infty} f \cdot W_x(t, f) \cdot df$ $-\Psi'(f) = X(f) ^{-2} \int_{-\infty}^{\infty} t \cdot W_x(t, f) \cdot dt$
(5) Moment properties	$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} t^n W_x(t, f) dt df = \int_{-\infty}^{\infty} t^n x(t) ^2 dt$, $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f^n W_x(t, f) dt df = \int_{-\infty}^{\infty} f^n X(f) ^2 df$

(6) $W_x(t, f)$ is bound to be real	$\overline{W_x(t, f)} = W_x(t, f)$
(7) Region properties	If $x(t) = 0$ for $t > t_2$ then $W_x(t, f) = 0$ for $t > t_2$ If $x(t) = 0$ for $t < t_1$ then $W_x(t, f) = 0$ for $t < t_1$
(8) Multiplication theory	If $y(t) = x(t)h(t)$, then $W_y(t, f) = \int_{-\infty}^{\infty} W_x(t, \rho)W_h(t, f - \rho) \cdot d\rho$
(9) Convolution theory	If $y(t) = \int_{-\infty}^{\infty} x(t - \tau)h(\tau) d\tau$, then $W_y(t, f) = \int_{-\infty}^{\infty} W_x(\rho, f) \cdot W_h(t - \rho, f) \cdot d\rho$
(10) Correlation theory	If $y(t) = \int_{-\infty}^{\infty} x(t + \tau)h^*(\tau) d\tau$, then $W_y(t, f) = \int_{-\infty}^{\infty} W_x(\rho, f) \cdot W_h(-t + \rho, f) \cdot d\rho$

(11) Time-shifting property	If $y(t) = x(t - t_0)$, then $W_y(t, f) = W_x(t - t_0, f)$
(12) Modulation property	If $y(t) = \exp(j2\pi f_0 t)x(t)$, then $W_y(t, f) = W_x(t, f - f_0)$
(13) Constant multiplication property	If $y(t) = cx(t)$, then $W_y(t, f) = c ^2 W_x(t, f)$
(14) Conjugation property	If $y(t) = x^*(t)$, then $W_y(t, f) = W_x(t, -f)$
(15) Scaling property	If $y(t) = x(ct)$, then $W_y(t, f) = \frac{1}{ c } W_x(ct, \frac{1}{c}f)$

The STFT (including the rec-STFT, the Gabor transform) does not have real region, multiplication, convolution, and correlation properties.

remember: If $x(t) = x^*(-t)$
then $X(f)$ is real

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- Why the WDF is always real?

What are the advantages and disadvantages it causes?

$$W_x(t, f) = \int x(t + \frac{\tau}{2}) x^*(t - \frac{\tau}{2}) e^{-j2\pi \tau f} d\tau \quad \text{set}$$

$$\begin{aligned} W_x^*(t, f) &= \int_{-\infty}^{\infty} x^*(t + \frac{\tau}{2}) x(t - \frac{\tau}{2}) e^{j2\pi \tau f} d\tau \quad \tau_1 = -\tau \quad d\tau_1 = -d\tau \\ &= \int_{\infty}^{-\infty} x^*(t - \frac{\tau_1}{2}) x(t + \frac{\tau_1}{2}) e^{-j2\pi \tau_1 f} (-d\tau_1) \end{aligned}$$

$$= - \int_{-\infty}^{\infty} x(t + \frac{\tau_1}{2}) x^*(t - \frac{\tau_1}{2}) e^{-j2\pi \tau_1 f} d\tau_1 = W_x(t, f)$$

Note: If $y(t) = e^{j\phi} x(t)$, ϕ is a constant, then $W_y(t, f) = W_x(t, f)$

- Try to prove of the projection and recovery properties

$$W_x(t, f) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau$$

- Proof of the region properties

If $x(t) = 0$ for $t < t_0$,

$$x(t + \tau/2) = 0 \quad \text{for } \tau < (t_0 - t)/2 = -(t - t_0)/2,$$

$$x(t - \tau/2) = 0 \quad \text{for } \tau > (t - t_0)/2,$$

Therefore, if $t - t_0 < 0$, the nonzero regions of $x(t + \tau/2)$ and $x(t - \tau/2)$ does not overlap and $x(t + \tau/2) x^*(t - \tau/2) = 0$ for all τ .

The importance of region property

V-G Advantages and Disadvantages of the WDF

Advantages: clarity

many good properties

suitable for analyzing the random process

Disadvantages: cross-term problem

more time for computation, especial for the signal with long time duration

not one-to-one

not suitable for $\exp(jt^n)$, $n \neq 0, 1, 2$

V-H Windowed Wigner Distribution Function

When $x(t)$ is not time-limited, its WDF is hard for implementation

$$W_x(t, f) = \int_{-\infty}^{\infty} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau$$

↓
with mask

$$W_x(t, f) = \int_{-\infty}^{\infty} \underline{w(\tau)} x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau$$

$w(\tau)$ is real and time-limited

Advantages: (1) reduce the computation time

(2) may reduce the cross term problem

Disadvantages:

$$\tau' = \frac{\tau}{2}, \quad \tau = 2\tau'$$

$$W_x(t, f) = 2 \int_{-\infty}^{\infty} \underline{w(2\tau')} x(t + \tau') \cdot x^*(t - \tau') e^{-j4\pi\tau' f} \cdot d\tau'$$

$t = n\Delta_t, \quad \tau' = p\Delta_t \quad f = m\Delta_f$

$$W_x(n\Delta_t, m\Delta_f) = 2 \sum_{p=-\infty}^{\infty} w(2p\Delta_t) x((n+p)\Delta_t) x^*((n-p)\Delta_t) e^{-j4\pi mp\Delta_t \Delta_f} \Delta_t$$

Suppose that $w(t) = 0$ for $|t| > B$

$$w(2p\Delta_t) = 0 \quad \text{for } p < -Q \text{ and } p > Q$$

$$Q = \frac{B}{2\Delta_t}$$

$$W_x(n\Delta_t, m\Delta_f) = 2 \sum_{p=-Q}^Q w(2p\Delta_t) x((n+p)\Delta_t) x^*((n-p)\Delta_t) e^{-j4\pi mp\Delta_t \Delta_f} \Delta_t$$

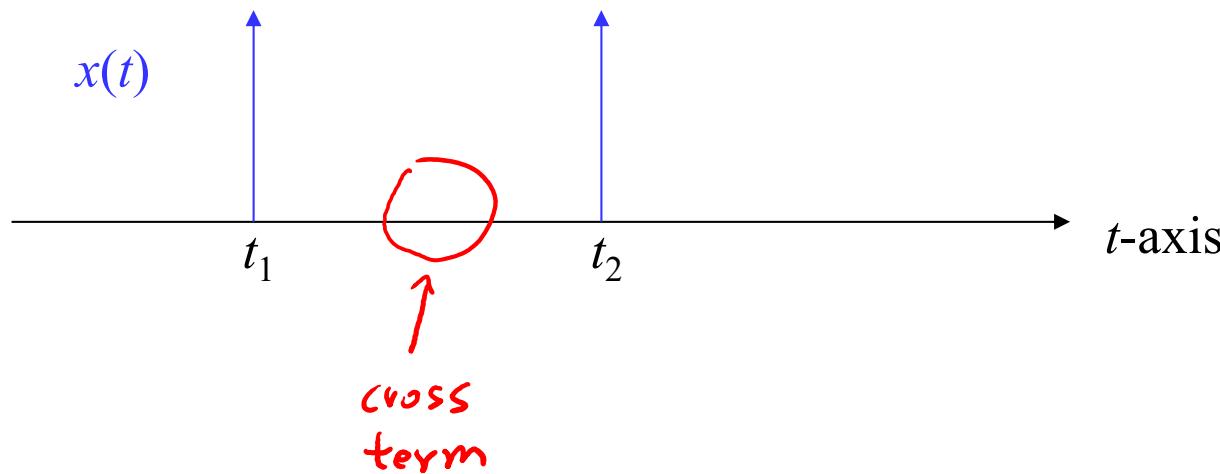
當然，乘上 mask 之後，有一些數學性質將會消失

(B) Why the cross term problem can be avoided ?

$$W_x(t, f) = \int_{-\infty}^{\infty} w(\tau) x(t + \tau/2) \cdot x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau$$

$w(\tau)$ is real

Viewing the case where $x(t) = \delta(t - t_1) + \delta(t - t_2)$



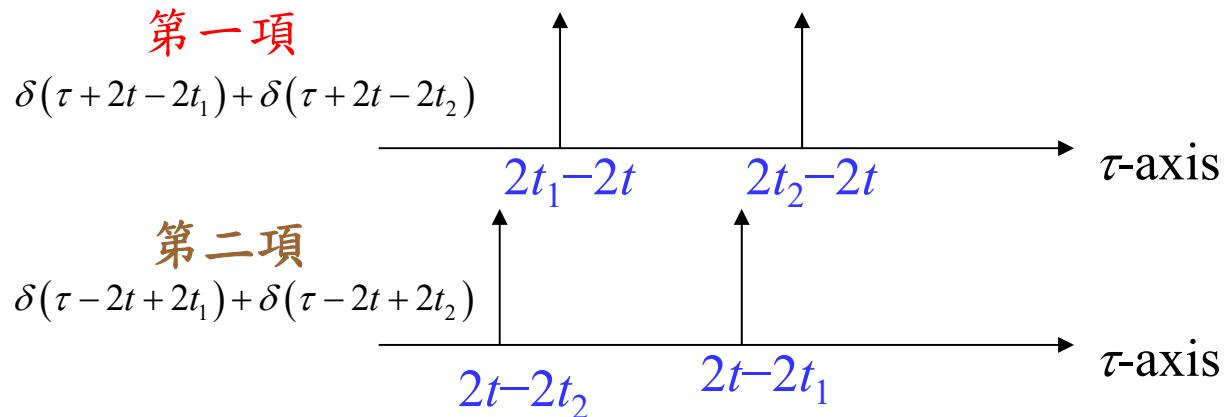
理想情形： $W_x(t, f) = 0$ for $t \neq t_1, t_2$

然而，當 mask function $w(\tau) = 1$ 時 (也就是沒有使用 mask function)

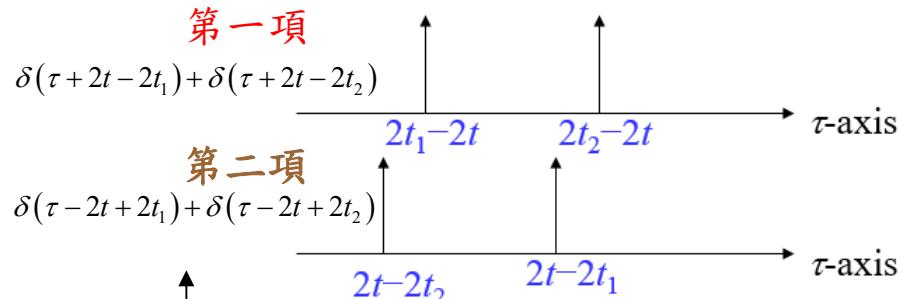
$$y(t, \tau) = x(t + \tau/2) \quad y^*(t, -\tau) = x^*(t - \tau/2)$$

$$\begin{aligned} W_x(t, f) &= \int_{-\infty}^{\infty} x(t + \tau/2)x^*(t - \tau/2)e^{-j2\pi\tau f} \cdot d\tau \\ &= \int_{-\infty}^{\infty} \left[\delta\left(t + \frac{\tau}{2} - t_1\right) + \delta\left(t + \frac{\tau}{2} - t_2\right) \right] \left[\delta\left(t - \frac{\tau}{2} - t_1\right) + \delta\left(t - \frac{\tau}{2} - t_2\right) \right] e^{-j2\pi\tau f} \cdot d\tau \\ &= 4 \int_{-\infty}^{\infty} \underbrace{[\delta(\tau + 2t - 2t_1) + \delta(\tau + 2t - 2t_2)]}_{\text{第一項}} \underbrace{[\delta(\tau - 2t + 2t_1) + \delta(\tau - 2t + 2t_2)]}_{\text{第二項}} e^{-j2\pi\tau f} \cdot d\tau \end{aligned}$$

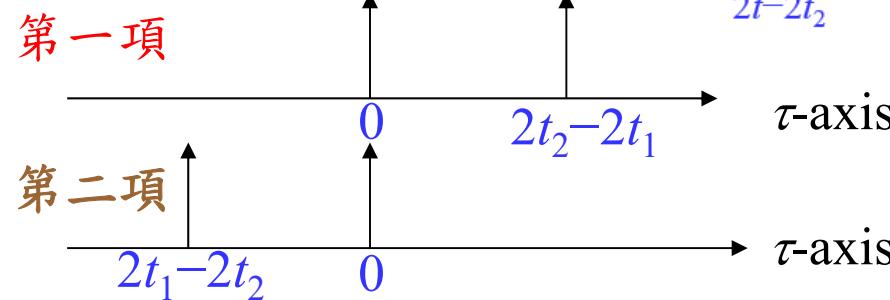
from page 133, property 2



3種情形 $W_x(t, f) \neq 0$

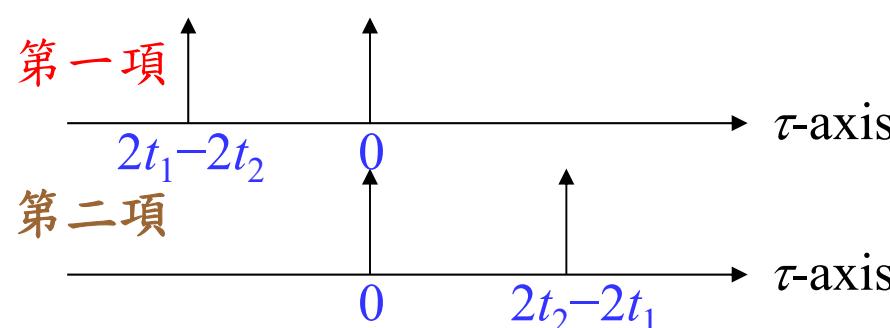


(1) If $t = t_1$
auto term



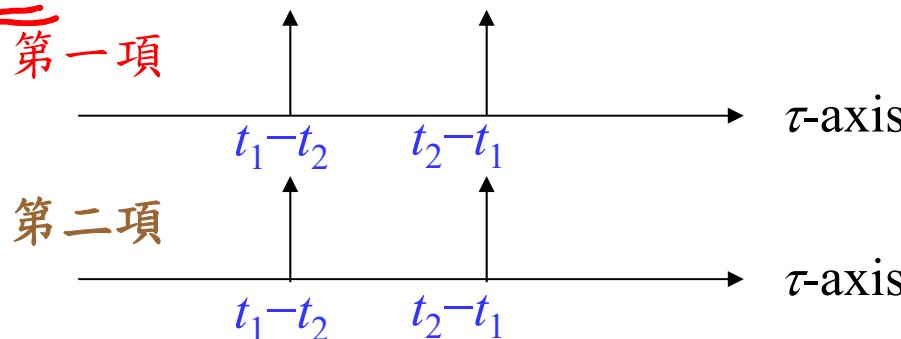
$$\begin{aligned} 2t_1 - 2t &= 2t - 2t_1 \\ 4t_1 &= 4t \\ t_1 &= t \end{aligned}$$

(2) If $t = t_2$
auto term



$$\begin{aligned} 2t_2 - 2t &= 2t - 2t_2 \\ 4t_2 &= 4t \\ t &= t_2 \end{aligned}$$

(3) If $t = (t_1 + t_2)/2$
cross term



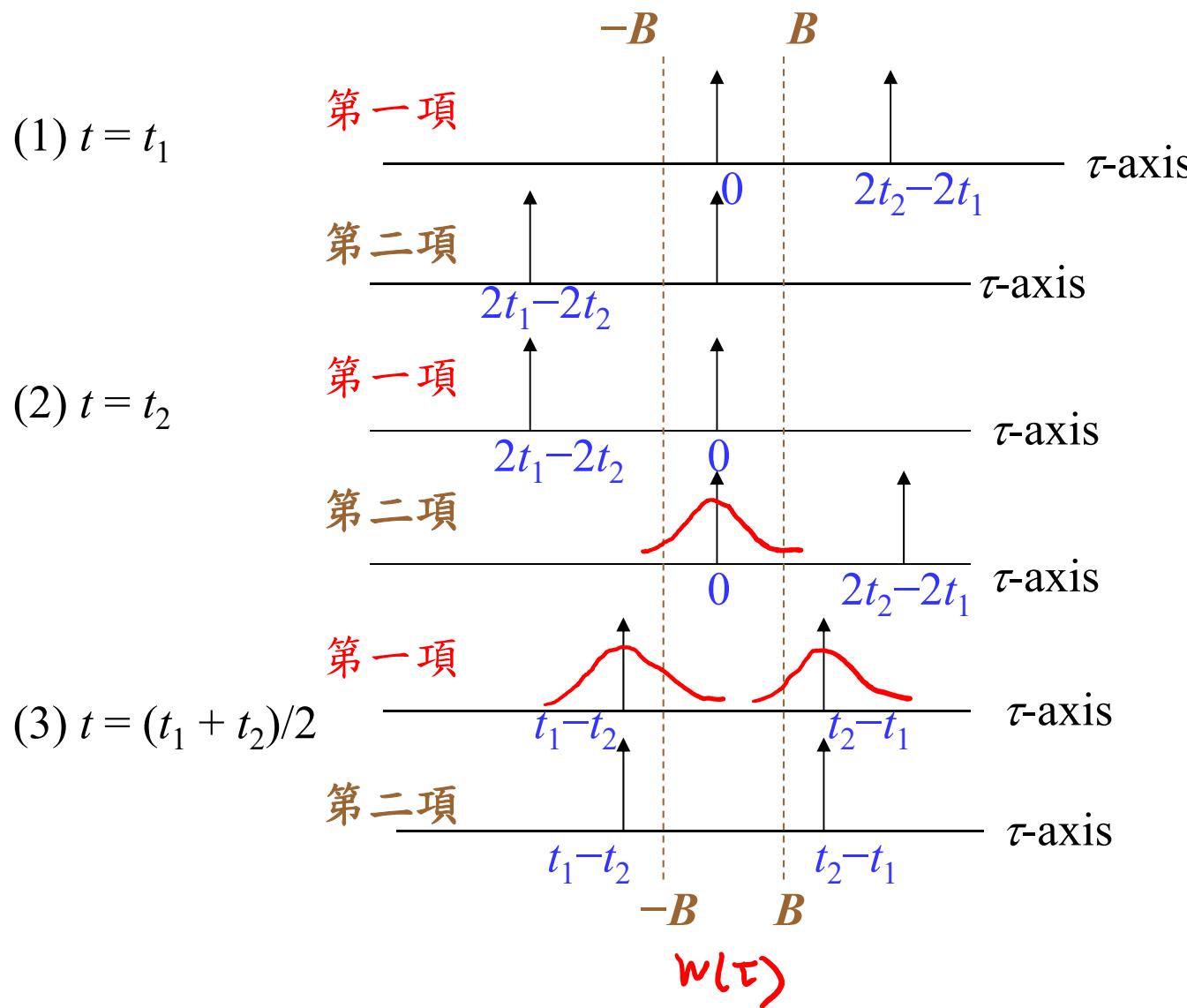
$$\begin{aligned} 2t_1 - 2t &= 2t - 2t_2 \\ 2t_1 + 2t_2 &= 4t \\ t &= \frac{t_1 + t_2}{2} \end{aligned}$$

With mask function

$$\begin{aligned}
 W_x(t, f) &= \int_{-\infty}^{\infty} w(\tau) x(t + \tau/2) x^*(t - \tau/2) e^{-j2\pi\tau f} \cdot d\tau \\
 &= \int_{-\infty}^{\infty} w(\tau) [\delta(\tau + 2t - 2t_1) + \delta(\tau + 2t - 2t_2)] \\
 &\quad \times [\delta(\tau - 2t + 2t_1) + \delta(\tau - 2t + 2t_2)] e^{-j2\pi\tau f} \cdot d\tau
 \end{aligned}$$

Suppose that $w(\tau) = 0$ for $|\tau| > B$, B is positive.

If $B < t_2 - t_1$



附錄八：研究所學習新知識把握的要點

- (1) Concepts: 這個方法的核心概念、基本精神是什麼
- (2) Comparison: 這方法和其他方法之間，有什麼相同的地方？
有什麼相異的地方
- (3) Advantages: 這方法的優點是什麼
(3-1) Why? 造成這些優點的原因是什麼
- (4) Disadvantages: 這方法的缺點是什麼
(4-1) Why? 造成這些缺點的原因是什麼
- (5) Applications: 這個方法要用來處理什麼問題，有什麼應用
- (6) Innovations: 這方法有什麼可以改進的地方
或是可以推廣到什麼地方

看過一篇論文或一個章節之後，若能夠回答(1)-(5)的問題，就代表你已經學通了這個方法

如果你的目標是發明創造出新的方法，可試著回答(3-1), (4-1), 和(6)的問題

每個領域每個月至少都有100篇以上的新論文，閱讀能力再厲害的人，也不可能都像大學讀書一樣篇篇都逐字讀過，所以，要選讀哪幾篇論文，哪些要詳讀，哪些可以只抓重點，這種擇要的能力，是研究所學生們應該練習的。