

III. Gabor Transform

III-A Definition

Standard Definition:

$$G_x(t, f) = \int_{-\infty}^{\infty} e^{-\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau$$

Alternative Definitions:

$$G_{x,1}(t, f) = \int_{-\infty}^{\infty} e^{-\pi(\tau-t)^2} e^{-j2\pi f(\tau-\frac{t}{2})} x(\tau) d\tau$$

$$G_{x,2}(t, f) = \sqrt[4]{2} \int_{-\infty}^{\infty} e^{-\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau \leftarrow \text{normalization}$$

$$G_{x,3}(t, \omega) = \int_{-\infty}^{\infty} e^{-(\tau-t)^2/2} e^{-j\omega\tau} x(\tau) d\tau$$

$$G_{x,4}(t, \omega) = \sqrt{\frac{1}{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(\tau-t)^2}{2}} e^{-j\omega(\tau-\frac{t}{2})} x(\tau) d\tau$$

Main Reference

- S. Qian and D. Chen, [Sections 3-2 ~ 3-6](#) in *Joint Time-Frequency Analysis: Methods and Applications*, Prentice-Hall, 1996.

Other References

- D. Gabor, “Theory of communication”, *J. Inst. Elec. Eng.*, vol. 93, pp. 429-457, Nov. 1946. (最早提出 Gabor transform)
- M. J. Bastiaans, “Gabor’s expansion of a signal into Gaussian elementary signals,” *Proc. IEEE*, vol. 68, pp. 594-598, 1980.
- R. L. Allen and D. W. Mills, *Signal Analysis: Time, Frequency, Scale, and Structure*, Wiley- Interscience.
- S. C. Pei and J. J. Ding, “Relations between Gabor transforms and fractional Fourier transforms and their applications for signal processing,” *IEEE Trans. Signal Processing*, vol. 55, no. 10, pp. 4839-4850, Oct. 2007.

註：

許多文獻把 Gabor transform 直接就稱作 short-time Fourier transform (STFT)，實際上，Gabor transform 是 STFT 當中的一個 special case.

III-B Approximation of the Gabor Transform

Although the range of integration is from $-\infty$ to ∞ , due to the fact that

$$e^{-\pi a^2} < 0.00001 \quad \text{when } |a| > 1.9143$$

$$e^{-a^2/2} < 0.00001 \quad \text{when } |a| > 4.7985$$

the Gabor transform can be simplified as:

$$G_x(t, f) \approx \int_{t-1.9143}^{t+1.9143} e^{-\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau$$

$$G_{x,3}(t, \omega) = \sqrt{\frac{1}{2\pi}} \int_{t-4.7985}^{t+4.7985} e^{-\frac{(\tau-t)^2}{2}} e^{-j\omega(\tau-\frac{t}{2})} x(\tau) d\tau$$

III-C Why Do We Choose the Gaussian Function as a Mask

(1) Among all functions, the Gaussian function has the advantage that the **area** in time-frequency distribution is **minimal**.

(和其他的 STFT 相比，比較能夠同時讓 time-domain 和 frequency domain 擁有較好的清晰度)

$w(t)$ 太寬 \rightarrow time domain 的解析度較差

$w(t)$ 太窄 $\rightarrow W(f) = FT[w(t)]$ 太寬 \rightarrow frequency domain 的解析度較差

(2) 由於 Gaussian function 是 FT 的 eigenfunction，因此 Gabor transform 在 time domain 和 frequency domain 的性質將互相對稱

$$\int_{-\infty}^{\infty} e^{-\pi t^2} e^{-j2\pi f t} dt = e^{-\pi f^2}$$

$$\int_{-\infty}^{\infty} e^{-t^2/2} e^{-j\omega t} dt = e^{-f^2/2}$$

Uncertainty Principle (Heisenberg, 1927)

For a signal $x(t)$, if $\sqrt{t} x(t) = 0$ when $|t| \longrightarrow \infty$, then

$$\sigma_t \sigma_f \geq 1/4\pi$$

where $\sigma_t^2 = \int (t - \mu_t)^2 P_x(t) dt$ $\sigma_f^2 = \int (f - \mu_f)^2 P_X(f) df$,

$$\mu_t = \int t P_x(t) dt,$$

$$\mu_f = \int f P_X(f) df$$

$$P_x(t) = \frac{|x(t)|^2}{\int |x(t)|^2 dt},$$

$$P_X(f) = \frac{|X(f)|^2}{\int |X(f)|^2 df},$$

(Proof of Henseinberg's uncertainty principle):

From simplification, we consider the case where $\mu_t = \mu_f = 0$

Then, use Parseval's theorem

$$\sigma_t^2 \sigma_f^2 = \frac{1}{4\pi^2} \frac{\int t^2 |x(t)|^2 dt \int |x'(t)|^2 dt}{\int |x(t)|^2 dt \int |x(t)|^2 dt}$$

$$\int |x(t)|^2 dt = \int |X(f)|^2 df \quad \text{if } X(f) = FT[x(t)]$$

From Schwarz's inequality $\langle x(t), x(t) \rangle \langle y(t), y(t) \rangle \geq |\langle x(t), y(t) \rangle|^2$

$$\begin{aligned}
 \int t^2 |x(t)|^2 dt \int |x'(t)|^2 dt &\geq \left(\left| \int tx^*(t) \frac{d}{dt} x(t) dt \right|^2 + \left| \int tx(t) \frac{d}{dt} x^*(t) dt \right|^2 \right) / 2 \\
 &\geq \left| \int \left(tx^*(t) \frac{d}{dt} x(t) + tx(t) \frac{d}{dt} x^*(t) \right) dt \right|^2 / 4 \quad (\text{using } |a+b|^2 + |a-b|^2 \geq 2|a|^2) \\
 &= \left| \int t \frac{d}{dt} [x(t)x^*(t)] dt \right|^2 / 4 = \left| tx(t)x^*(t) \Big|_{-\infty}^{\infty} - \int x^*(t)x(t) dt \right|^2 / 4 \\
 &= \left| [tx(t)x^*(t) \Big|_{t \rightarrow \infty} - tx(t)x^*(t) \Big|_{t \rightarrow -\infty}] - \int x^*(t)x(t) dt \right|^2 / 4 \\
 &= \left| \int |x(t)|^2 dt \right|^2 / 4
 \end{aligned}$$

$$\sigma_t^2 \sigma_f^2 \geq \frac{1}{16\pi^2} \implies \sigma_t \sigma_f \geq \frac{1}{4\pi}$$

For Gaussian function

$$x(t) = e^{-\pi t^2} \quad X(f) = e^{-\pi f^2}$$

$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} e^{-2\pi t^2} dt = ?$$

$$\text{use } \int_{-\infty}^{\infty} e^{-(at^2+bt)} dt = \sqrt{\pi/a} \cdot e^{b^2/4a}$$

$$\int_{-\infty}^{\infty} t^2 |x(t)|^2 dt = \int_{-\infty}^{\infty} t^2 e^{-2\pi t^2} dt = 2 \int_0^{\infty} t^2 e^{-2\pi t^2} dt = ?$$

$$\text{use } \int_0^{\infty} t^m e^{-at^2} dt = \frac{\Gamma[(m+1)/2]}{2a^{(m+1)/2}}$$

$$\Gamma(1/2) = \sqrt{\pi}$$

$$\Gamma(n+1) = n\Gamma(n)$$

[工具書] M. R. Spiegel, *Mathematical Handbook of Formulas and Tables*, McGraw-Hill, 3rd Ed., 2009.

$$\sigma_t^2 = \frac{\int t^2 |x(t)|^2 dt}{\int |x(t)|^2 dt} = \frac{1}{4\pi},$$

$$\sigma_t = \sqrt{\frac{1}{4\pi}}$$

同理, $\sigma_f = \sqrt{\frac{1}{4\pi}}$

所以對 Gaussian function 而言，

$$\sigma_t \sigma_f = \frac{1}{4\pi}$$

滿足下限

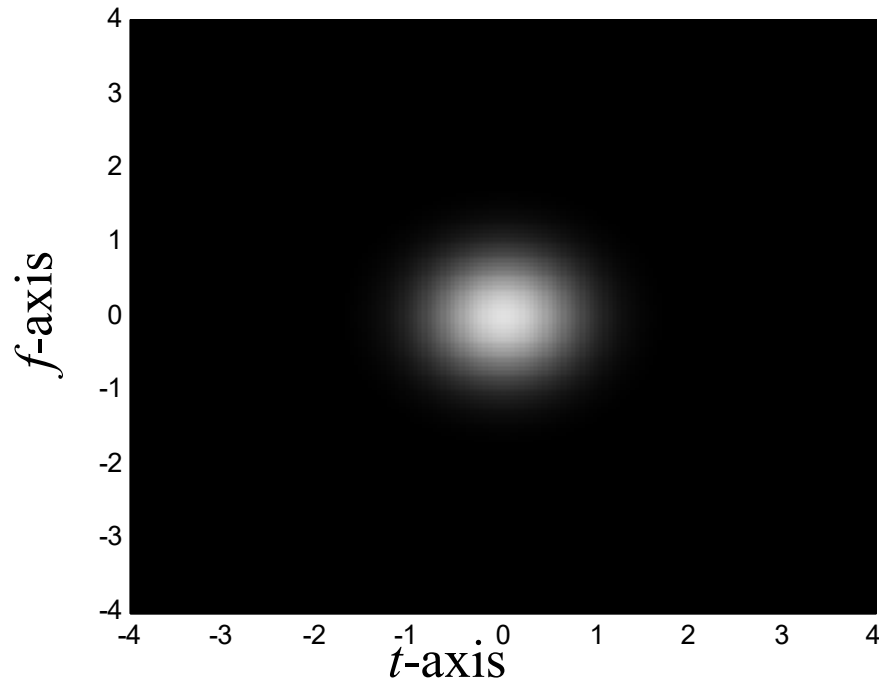
Special relation between the Gaussian function and the rectangular function

Gaussian function is also an eigenmode in optics, radar system, and other electromagnetic wave systems.

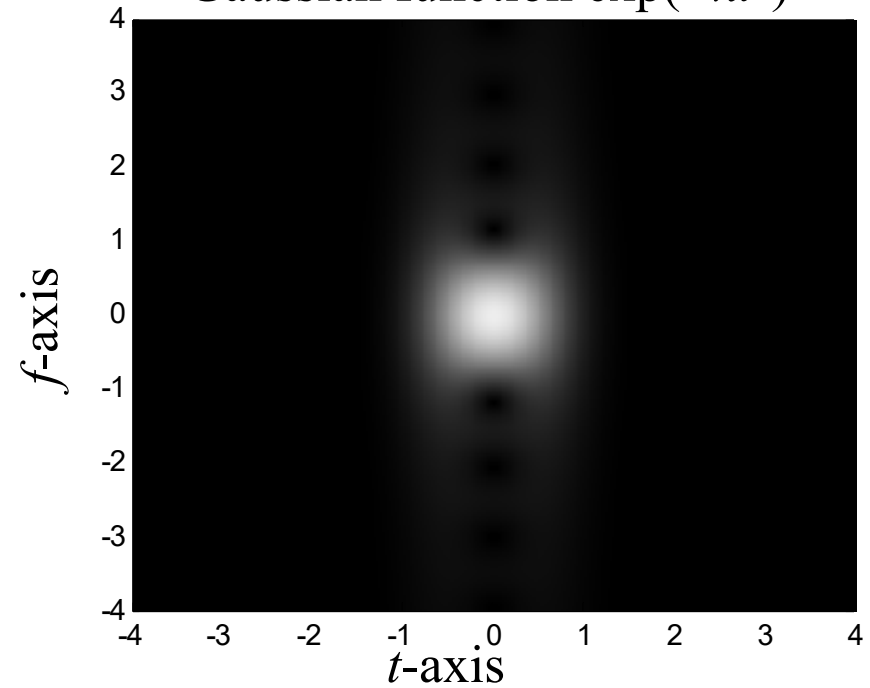
(will be illustrated in the 8th week)

III-D Simulations

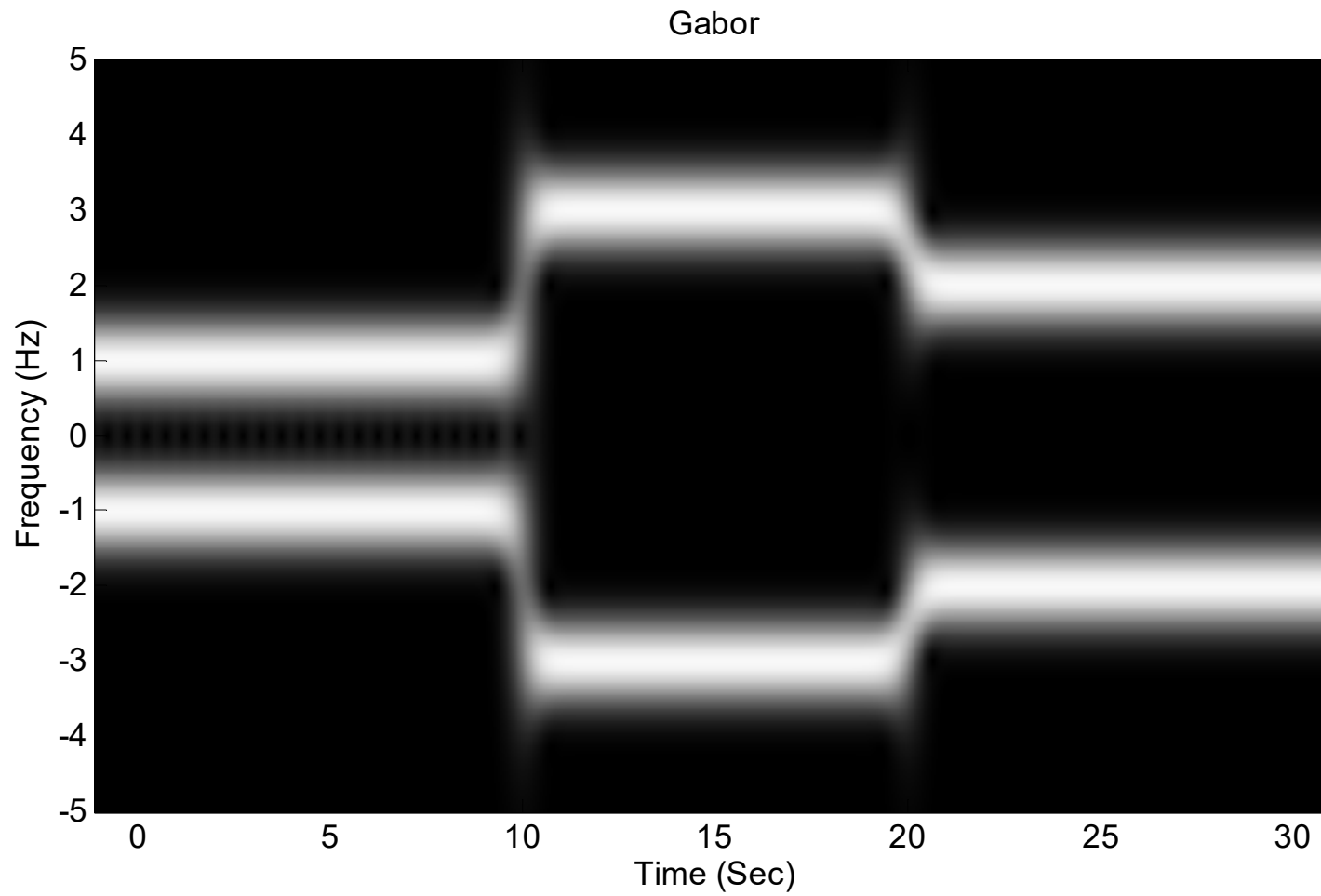
Gabor transform for
Gaussian function $\exp(-\pi t^2)$

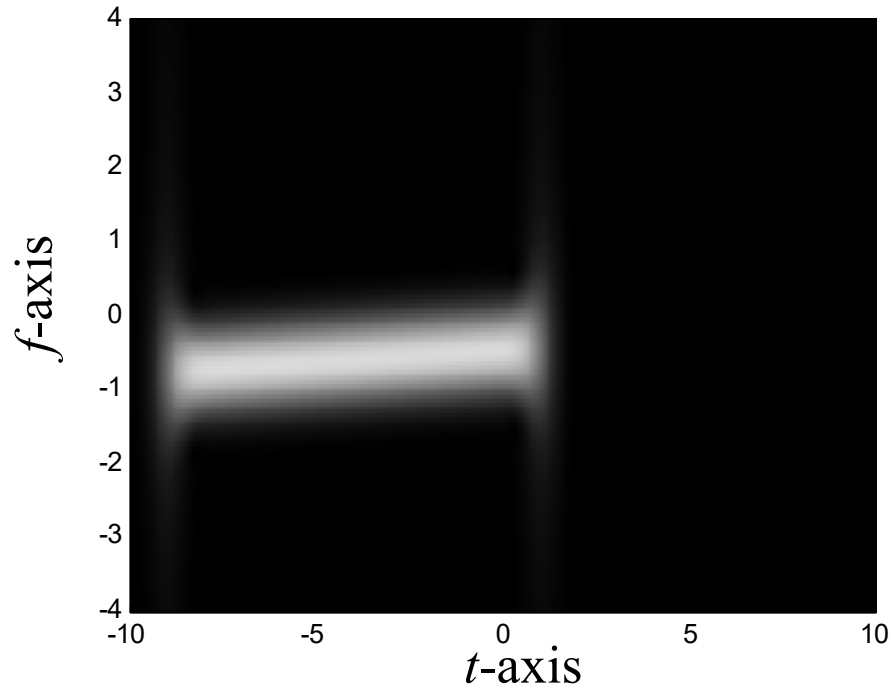


rec-STFT, $B = 0.5$ for
Gaussian function $\exp(-\pi t^2)$



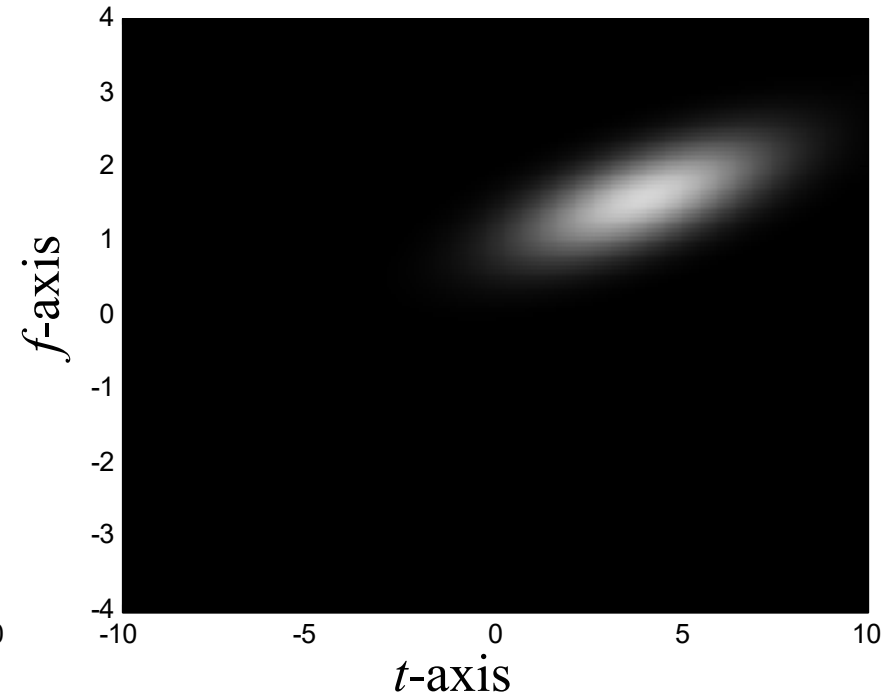
$$x(t) = \cos(2\pi t) \text{ when } t < 10,$$
$$x(t) = \cos(6\pi t) \text{ when } 10 \leq t < 20,$$
$$x(t) = \cos(4\pi t) \text{ when } t \geq 20$$



Gabor transform of $s(t)$ 

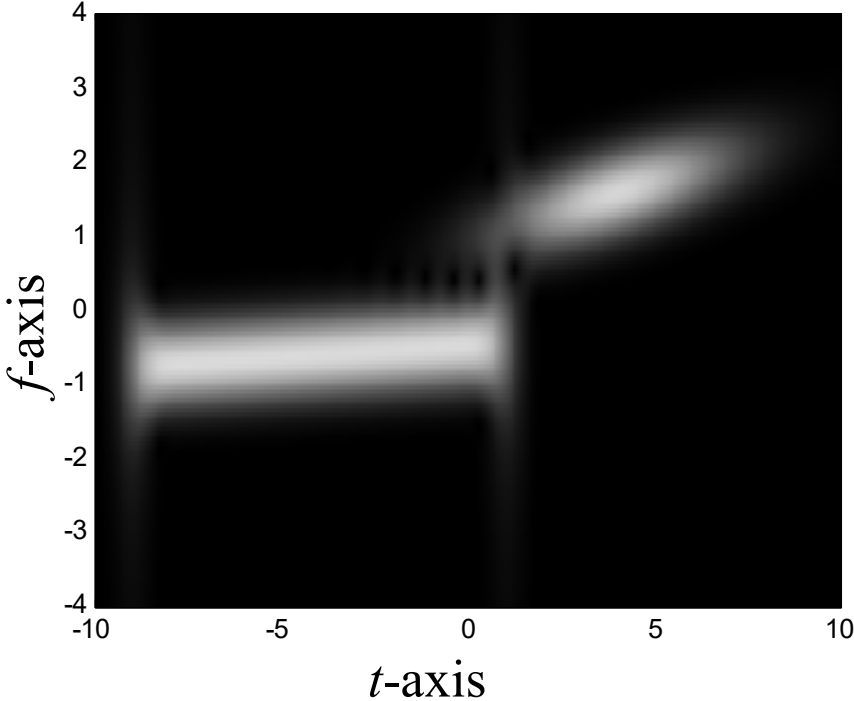
$$s(t) = \exp(jt^2 / 10 - j3t) \text{ for } -9 \leq t \leq 1,$$

$$s(t) = 0 \text{ otherwise,}$$

Gabor transform of $r(t)$ 

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

Gabor transform for $s(t) + r(t)$



III-E Properties of Gabor Transforms

$$G_x(t, f) = \int_{-\infty}^{\infty} e^{-j2\pi f\tau} e^{-\pi(\tau-t)^2} x(\tau) d\tau$$

(1) Integration property

When $k \neq 0$,
$$\int_{-\infty}^{\infty} G_x(t, f) e^{j2\pi k t f} df = e^{-\pi(k-1)^2 t^2} x(kt)$$

When $k = 0$,
$$\int_{-\infty}^{\infty} G_x(t, f) df = e^{-\pi t^2} x(0)$$

When $k = 1$,
$$\int_{-\infty}^{\infty} G_x(t, f) e^{j2\pi t f} df = x(t) \quad (\text{recovery property})$$

(2) Shifting property

If $y(t) = x(t - t_0)$, then
$$G_y(t, f) = G_x(t - t_0, f) e^{-j2\pi f t_0}.$$

(3) Modulation property

If $y(t) = x(t) \exp(j2\pi f_0 t)$, then
$$G_y(t, f) = G_x(t, f - f_0)$$

(4) Special inputs:

(a) When $x(\tau) = \delta(\tau)$, $G_x(t, f) = e^{-\pi t^2}$

(b) When $x(\tau) = 1$, $G_x(t, f) = e^{-j2\pi ft} e^{-\pi f^2}$

(symmetric for the time and frequency domains)

(5) Linearity property

If $z(\tau) = \alpha x(\tau) + \beta y(\tau)$ and $G_z(t, f)$, $G_x(t, f)$ and $G_y(t, f)$ are their Gabor transforms, then

$$G_z(t, f) = \alpha G_x(t, f) + \beta G_y(t, f)$$

(6) Power integration property:

$$\int_{-\infty}^{\infty} |G_x(t, f)|^2 df = \int_{-\infty}^{\infty} e^{-2\pi(\tau-t)^2} |x(\tau)|^2 d\tau \approx \int_{u-1.9143}^{u+1.9143} e^{-2\pi(\tau-u)^2} |x(\tau)|^2 d\tau$$

(7) Power decayed property

- If $x(t) = 0$ for $t > t_0$, then

$$\int_{-\infty}^{\infty} |G_x(t, f)|^2 df < e^{-2\pi(t-t_0)^2} \int_{-\infty}^{\infty} |G_x(t_0, f)|^2 df$$

i.e., $\underset{\text{(fix } t, \text{ vary } f)}{\text{average of } |G_x(t, f)|^2} < e^{-2\pi(t-t_0)^2} \times \underset{\text{(fix } t_0, \text{ vary } f)}{\text{average of } |G_x(t_0, f)|^2}$ for $t > t_0$.

(Proof):

$$G_x(t, f) = \int_{-\infty}^{t_0} e^{-\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau \quad G_x(t_0, f) = \int_{-\infty}^{t_0} e^{-\pi(\tau-t_0)^2} e^{-j2\pi f\tau} x(\tau) d\tau$$

Since $(\tau-t)^2 > (\tau-t_0)^2 + (t_0-t)^2$ $e^{-\pi(\tau-t)^2} < e^{-\pi(\tau-t_0)^2} e^{-\pi(t_0-t)^2}$

$$G_x(t, f) < e^{-\pi(t-t_0)^2} G_x(t_0, f)$$

- If $X(f) = FT[x(t)] = 0$ for $f > f_0$, then

average of $|G_x(t, f)|^2$ $\underset{\text{(fix } f, \text{ vary } t)}{<} e^{-2\pi(f-f_0)^2} \times \underset{\text{(fix } f_0, \text{ vary } t)}{\text{average of } |G_x(t, f_0)|^2}$ for $f > f_0$.

(8) Energy sum property

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G_x(t, f) G_y^*(t, f) df dt = \int_{-\infty}^{\infty} x(\tau) y^*(\tau) d\tau$$

where $G_x(t, f)$ and $G_y(t, f)$ are the Gabor transforms of $x(\tau)$ and $y(\tau)$, respectively.

III-F Scaled Gabor Transforms

$$G_x(t, f) = \sqrt[4]{\sigma} \int_{-\infty}^{\infty} e^{-\sigma\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau$$



(finite interval form)

larger σ : higher resolution in the time domain

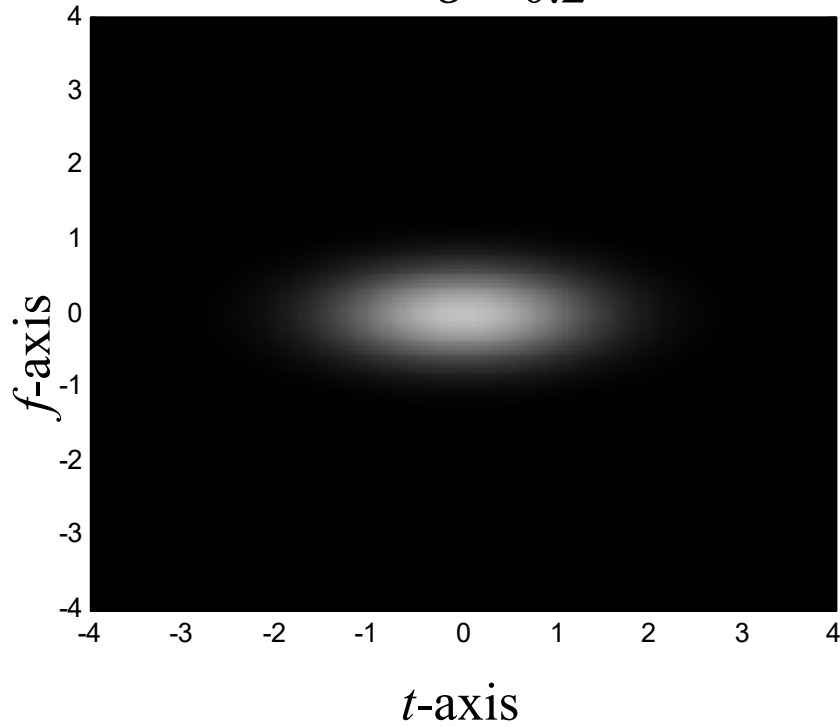
lower resolution in the frequency domain

smaller σ : higher resolution in the frequency domain

lower resolution in the time domain

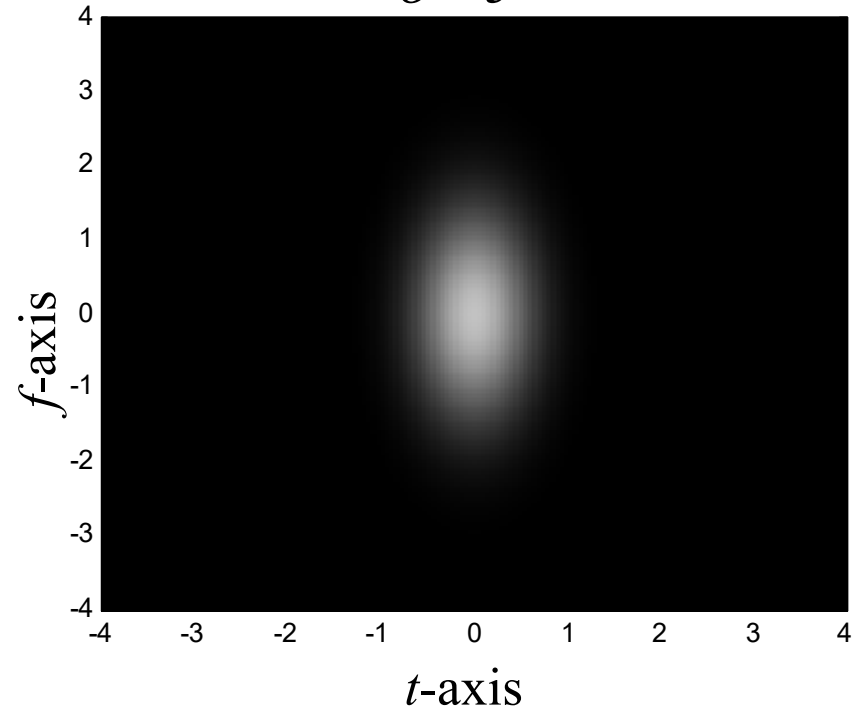
Gabor transform for
Gaussian function $\exp(-\pi t^2)$

$\sigma = 0.2$



Gabor transform for
Gaussian function $\exp(-\pi t^2)$

$\sigma = 5$



處理對 time resolution 相對上比 frequency resolution 敏感的信號

(1) Using the generalized Gabor transform with larger σ

(2) Using other time unit instead of second

例如，原本 t (單位：sec) f (單位：Hz)

對聲音信號可以改成

t (單位：0.1 sec) f (單位：10 Hz)

III-G Gabor Transforms with Adaptive Window Width

For a signal,

when the instantaneous frequency varies fast \rightarrow larger σ

when instantaneous frequency varies slowly \rightarrow smaller σ

$$G_x(t, f) = \sqrt[4]{\sigma(t)} \int_{-\infty}^{\infty} e^{-\sigma(t)\pi(\tau-t)^2} e^{-j2\pi f\tau} x(\tau) d\tau$$

$\sigma(t)$ is a function of t

S. C. Pei and S. G. Huang, "STFT with adaptive window width based on the chirp rate," *IEEE Trans. Signal Processing*, vol. 60, issue 8, pp. 4065-4080, 2012.

附錄四：Matlab 寫程式的原則

- (1) 迴圈能避免就儘量避免
- (2) 儘可能使用 Matrix 及 Vector operation
- (3) 能夠不在迴圈內做的運算，則移到迴圈外
- (4) 寫一部分即測試，不要全部寫完再測試(縮小範圍比較容易 debug)
- (5) 先測試簡單的例子，成功後再測試複雜的例子

註：作業 Matlab Program (or Python program) 鼓勵各位同學儘量用精簡而快速的方式寫。Program 執行速度越快，分數就越高。

一些重要的 Matlab 指令

(1) **function**: 放在第一行，可以將整個程式函式化

(2) **tic, toc**: 計算時間

tic 為開始計時，toc 為顯示時間

(3) **find**: 找尋一個 vector 當中不等於 0 的 entry 的位置

範例： $\text{find}([1\ 0\ 0\ 1]) = [1, 4]$

$\text{find}(\text{abs}([-5:5]) \leq 2) = [4, 5, 6, 7, 8]$

(因為 $\text{abs}([-5:5]) \leq 2 = [0\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 0]$)

(4) **'**: Hermitian (transpose + conjugation)，**.'**: transpose

(5) **imread**: 讀圖，**image, imshow, imagesc**: 將圖顯示出來，

(註：較老的 Matlab 版本 imread 要和 double 並用

$A = \text{double}(\text{imread}('Lena.bmp'));$

(6) **imwrite**: 製做圖檔

- (7) `xlsread`: 由 Excel 檔讀取資料
- (8) `xlswrite`: 將資料寫成 Excel 檔
- (9) `aviread`: 讀取 video 檔，限副檔名為 avi
- (10) `VideoReader`: 讀取 video 檔
- (11) `VideoWriter`: 製作 video 檔
- (12) `dlmread`: 讀取 *.txt 或其他類型檔案的資料
- (13) `dlmwrite`: 將資料寫成 *.txt 或其他類型檔案

附錄五：使用 Python 將時頻分析的圖畫出來

事前安裝模組

```
pip install numpy
```

```
pip install matplotlib
```

假設 y 為時頻分析結果(應為二維的矩陣數列)，將 y 以灰階方式畫出來

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
C = 400
```

```
y = np.abs(y) / np.max(np.abs(y)) * C
```

```
plt.imshow(y, cmap='gray', origin='lower')
```

```
# 加上 origin='lower' 避免上下相反
```

```
plt.xlabel('Time (Sec)')
```

```
plt.ylabel('Frequency (Hz)')
```

```
plt.show()
```

感謝2021年擔任助教的蔡昌廷同學

若要加上座標軸數值(在plt.show()之前加上以下程式碼)

```
x_label = ['0', '10', '20', '30'] # 橫軸座標值
y_label = ['-5', '0', '5']      # 縱軸座標值
plt.xticks(np.arange(0, x_max, step=int(x_max/(len(x_label)-1))), x_label)
plt.yticks(np.arange(0, y_max, step=int(y_max/(len(y_label)-1))), y_label)
```

Reference :

https://matplotlib.org/stable/api/as_gen/matplotlib.pyplot.xticks.html

附錄六：寫 Python 版本程式可能會用到的重要指令

建議必安裝模組

```
pip install numpy
```

```
pip install scipy
```

```
pip install scipy
```

```
pip install opencv-python
```

(1) 定義函式：使用 def

(2) 計算時間

```
import time
```

```
start_time = time.time() #獲取當前時間
```

```
end_time = time.time()
```

```
total_time = end_time - start_time #計算時間差來得到總執行時間
```

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(3) 讀取圖檔、輸出圖檔(建議使用opencv)

```
import cv2
image = cv2.imread(file_name) #預設color channel為BGR
cv2.imwrite(file_name, image) #需將color channel轉為BGR
```

(4) 尋找array中滿足特定條件的值的位址

(相當於 Matlab 的 find 指令)

```
import numpy as np
a = np.array([0, 1, 2, 3, 4, 5])
index = np.where(a > 3) # 回傳array([4, 5])
print(index)
      (array([4, 5], dtype=int64),)
index[0][0]
      4
index[0][1]
      5
```

```
A1= np.array([[1,3,6],[2,4,5]])
```

```
index = np.where(A1 > 3)
```

```
print(index)
```

```
(array([0, 1, 1], dtype=int64), array([2, 1, 2], dtype=int64))
```

(代表滿足 $A1 > 3$ 的點的位置座標為 [0, 2], [1, 1], [1, 2])

```
[index[0][0], index[1][0]]
```

```
[0, 2]
```

```
[index[0][1], index[1][1]]
```

```
[1, 1]
```

```
[index[0][2], index[1][2]]
```

```
[1, 2]
```

$$\mathbf{A}_1 = \begin{bmatrix} 1 & 3 & 6 \\ 2 & 4 & 5 \end{bmatrix}$$

(5) Hermitian 、transpose

```
import numpy as np
result = np.conj(matrix.T) # Hermitian
result = matrix.T # transpose
```

(6) 讀取 Matlab 當中的 mat 檔

```
data = scipy.io.loadmat('***.mat')
y = np.array(data['y']) # 假設 y 是 ***.mat 當中儲存的資料
```