

# Advanced Information Engineering

#8 November 30 (Mon), 2020

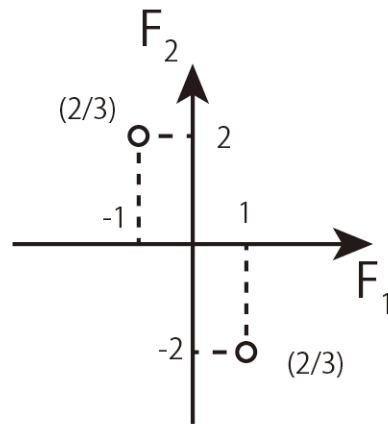
Kenjiro T. Miura

# Assignment #2 Ex.1 Answer

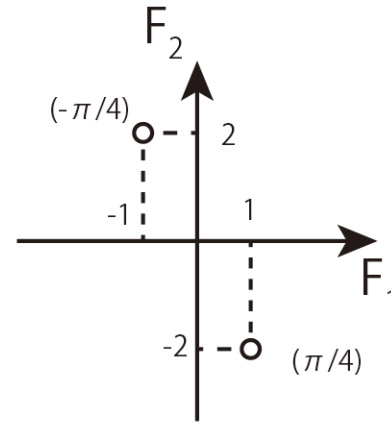
Consider 2-dimensional sinusoidal wave signal  
 $g_a(x,y)=3 \cos (2 \pi x - 4 \pi y + \pi/4)$

Draw spectrum.

$$g_a(x,y)=\frac{3}{2}\left(e^{j(2\pi(x-2y)+\frac{\pi}{4})} + e^{j(-2\pi(x-2y)-\frac{\pi}{4})}\right)$$



Amplitude

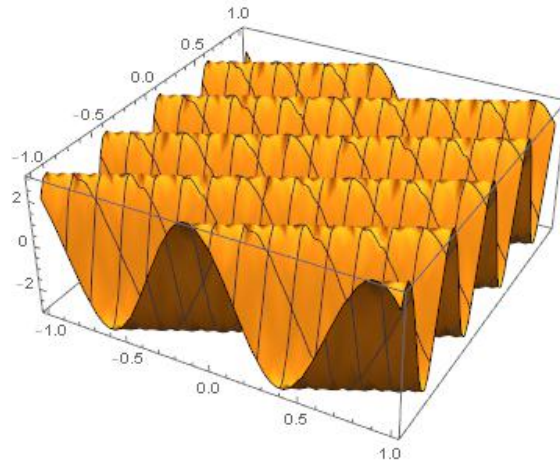


Phase

# Assignment #2 Ex.1 Answer

Consider 2-dimensional sinusoidal wave signal  
 $g_a(x,y)=3 \cos (2 \pi x - 4 \pi y + \pi/4)$

Illustrate this wave assuming that its variables are  $x, y$ .



# Assignment #2 Ex.2 Answer

Calculate a continuous signal which has spectrums given by Fig. 1.

$$(a) \ g_a(x,y) = 2\cos(2\pi y) + \cos(4\pi y)$$

$$(b) \ g_a(x,y) = \cos\left(4\pi x + 2\pi y - \frac{\pi}{4}\right)$$

# Assignment #2 Ex.3 Answer

Perform discrete spatial Fourier transform of each of the signals shown in Fig. 2.

$$\begin{aligned} \text{(a) } G(\varpi_1, \varpi_2) &= 1 + e^{-j\varpi_1} + e^{-j\varpi_2} + e^{-j(\varpi_1 + \varpi_2)} \\ &= 4 \cos\left(\frac{\varpi_1}{2}\right) \cos\left(\frac{\varpi_2}{2}\right) e^{-\frac{j(\varpi_1 + \varpi_2)}{2}} \end{aligned}$$

$$\begin{aligned} \text{(b) } G(\varpi_1, \varpi_2) &= 1 + e^{-j\varpi_2} + e^{-2j\varpi_2} + e^{-3j\varpi_2} \\ &= 2 \left( \cos\left(\frac{\varpi_2}{2}\right) \cos\left(\frac{3\varpi_2}{2}\right) \right) e^{-\frac{j3\varpi_2}{2}} \end{aligned}$$

$$\text{(c) } G(\varpi_1, \varpi_2) = e^{j\varpi_1} + 1 + e^{-j\varpi_1} = 1 + 2\cos(\varpi_1)$$

# Assignment #2 Ex.4 Answer

When a real-valued 2-dimensional discrete signal  $g(n_1, n_2)$  satisfies that  $g(n_1, n_2) = g(-n_1, -n_2)$ , show that its discrete spatial Fourier transform (DSFT)  $G(\omega_1, \omega_2)$  is real-valued.

$$\begin{aligned} G(\omega_1, \omega_2) &= \sum_{n_1=-\infty}^{\infty} \sum_{n_2=-\infty}^{\infty} g(n_1, n_2) e^{-j\omega_1 n_1} e^{-j\omega_2 n_2} \end{aligned}$$

# Assignment #2 Ex.4 Answer

$$\begin{aligned} G(\omega_1, \omega_2) &= \sum_{n_1=-\infty}^{\infty} \sum_{n_2=-\infty}^{\infty} g(n_1, n_2) e^{-j\omega_1 n_1} e^{-j\omega_2 n_2} \\ &= g(0, 0) + \sum_{n_1=1}^{\infty} g(n_1, 0) (e^{j\omega_1 n_1} + e^{-j\omega_1 n_1}) + \sum_{n_2=1}^{\infty} g(0, n_2) (e^{j\omega_2 n_2} + e^{-j\omega_2 n_2}) \\ &\quad + \sum_{n_1=1}^{\infty} \sum_{n_2=1}^{\infty} (g(n_1, n_2) (e^{j(\omega_1 n_1 + \omega_2 n_2)} + e^{-j(\omega_1 n_1 + \omega_2 n_2)}) \\ &\quad + g(-n_1, n_2) (e^{j(\omega_1 n_1 + \omega_2 n_2)} + e^{-j(\omega_1 n_1 + \omega_2 n_2)})) \\ &= g(0, 0) + \sum_{n_1=1}^{\infty} 2g(n_1, 0) \cos(\omega_1 n_1) + \sum_{n_2=1}^{\infty} 2g(0, n_2) \cos(\omega_2 n_2) \\ &\quad + \sum_{n_1=1}^{\infty} \sum_{n_2=1}^{\infty} 2(g(n_1, n_2) + g(-n_1, n_2)) \cos(\omega_1 n_1 + \omega_2 n_2) \end{aligned}$$

# Assignment #2 Ex.5 Answer

Consider a 2-dimensional sinusoidal wave  $g_a(x,y)=\cos(2\pi x + 4\pi y)$ . In order to satisfy the sampling theorem, we would like to sample as follows:

$$g(n_1,n_2)=g_a(x,y) | x=n_1 T_{s1}, y=n_2 T_{s2}.$$

Indicate the conditions for sampling intervals  $T_{s1}$  and  $T_{s2}$  to be satisfied.

$$F_{s1} = \frac{1}{T_{s1}} > 2 \text{ and } F_{s2} = \frac{1}{T_{s2}} > 4$$



# Fast Fourier Transform (FFT)

- FFT is a fast calculation version of discrete Fourier transform (DFT) .

# Discrete Fourier Transform (DFT)

- DSFT with frequency discretization
- In case where  $g(n_1, n_2)$  is defined in  $N_1 \times N_2$ , a finite domain, i.d. 2-dimensional image.

$$g(n_1, n_2) = \frac{1}{N_1 N_2} \sum_{k_1=0}^{N_1-1} \sum_{k_2=0}^{N_2-1} G(k_1, k_2) W_{N_1}^{-k_1 n_1} W_{N_2}^{-k_2 n_2}$$
$$G(k_1, k_2) = \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} g(n_1, n_2) W_{N_1}^{k_1 n_1} W_{N_2}^{k_2 n_2}$$

Where

$$W_{N_1} = e^{-j2\pi/N_1}, W_{N_2} = e^{-j2\pi/N_2}$$

Thus the values of DFT are sampled ones of DSFT obtained by the intervals uniform intervals of spectrum period/ $N_1$  and  $N_2$ .

# Periodicity of DFT

$$W_N^{nk} = W_N^{n(k+N)} = W_N^{(n+N)k}$$

に注意すると,

$$G(k_1, k_2) = G(k_1 + N_1, k_2)$$

$$= G(k_1, k_2 + N_2)$$

$$g(n_1, n_2) = g(n_1 + N_1, n_2)$$

$$= g(n_1, n_2 + N_2)$$

- The number of independent points in both of the spatial and frequency domains is  $N_1 \times N_2$  and we assume their periodicity and perform calculations.

# Exercise Example

- 1 dimensional discrete signal of  $N$  points

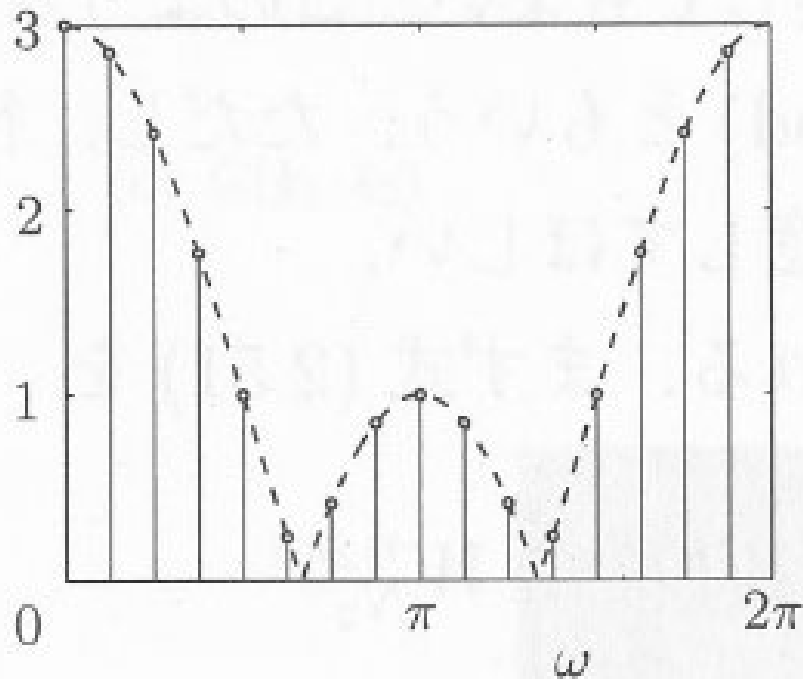
$$g(n) = \begin{cases} 1, & n=0 \cdots, L-1 \\ 0, & n=L, L+1, \cdots, N-1 \end{cases}$$

を考える．以下の問いに答えよ．

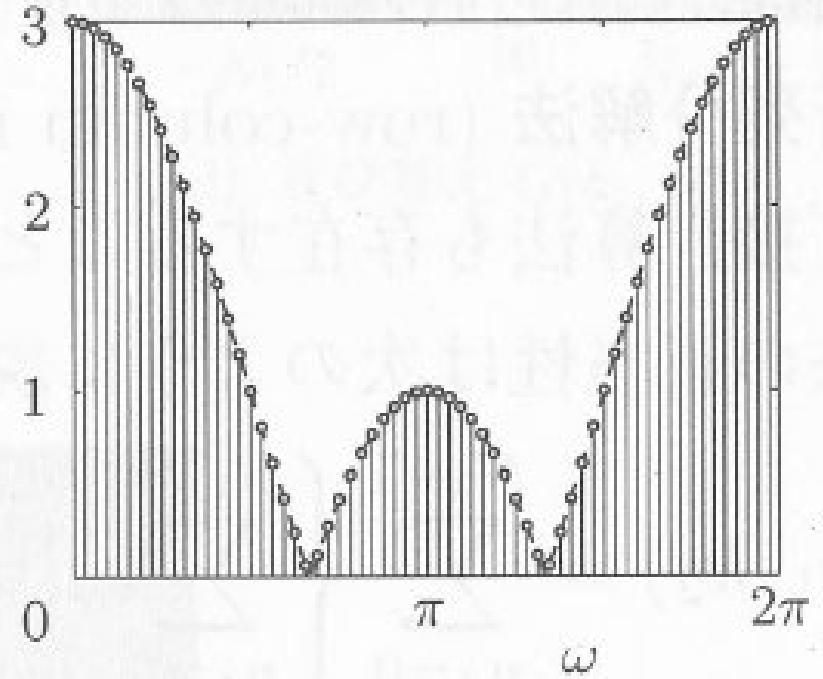
- (a)  $L = 3, N = 16$  として  $N$  点 DFT を求めよ．
- (b)  $L = 3, N = 64$  として  $N$  点 DFT を求めよ．

You can use a calculator to calculate amplitude.  
Since (b) takes long time , please do (a).

# Answer



(a)  $L = 3, N = 16$



(b)  $L = 3, N = 64$

- The larger  $N$ , the more sufficient sampling density
- For Fourier image analysis, how to make sure to get sufficient sampling density?

# Fast Fourier Transform (FFT)

- FFT is to make DFT (a lot of computational cost) faster.
- Without approximation error, it can perform DFT strictly.
- The method which takes advantage of Matrix decomposition method (decomposability of DFT) .

# Matrix Decomposition (decomposability of DFT)

- Direct 2D DFT repeats  $N_1 \times N_2$  DFT by  $N_1 \times N_2$  times.
- In case of matrix decomposition, for horizontal row data perform  $N_1$  1D DFT by  $N_2$  times, then for column data perform  $N_2$  1D DFT by  $N_1$  times.

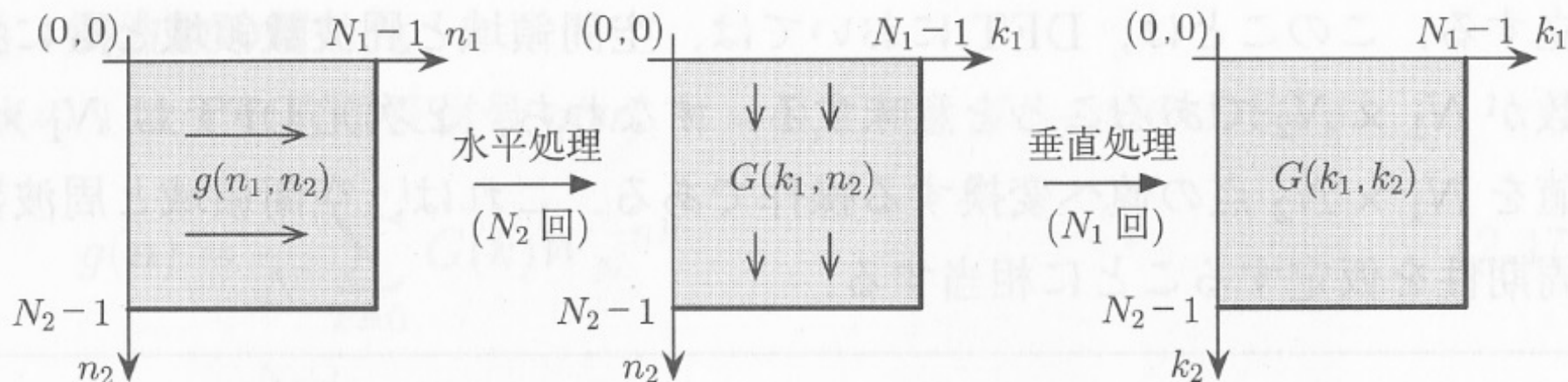


图 2.16 行列分解法

# Example

- (a): 2D image signal
- (b): for (a), perform DFT horizontally
- (c): for (b), perform DFT vertically.
- (d): By using the periodicity of DFT, put DC component at the center.

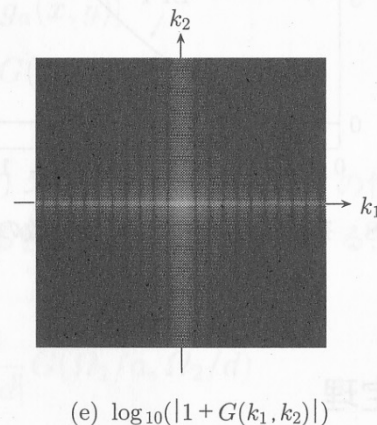
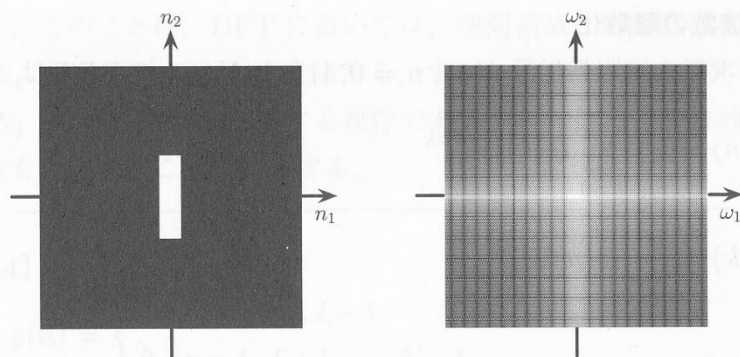
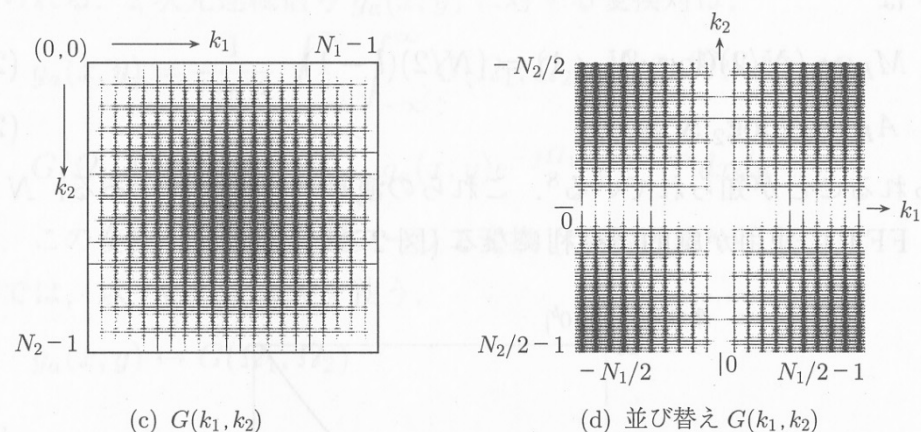
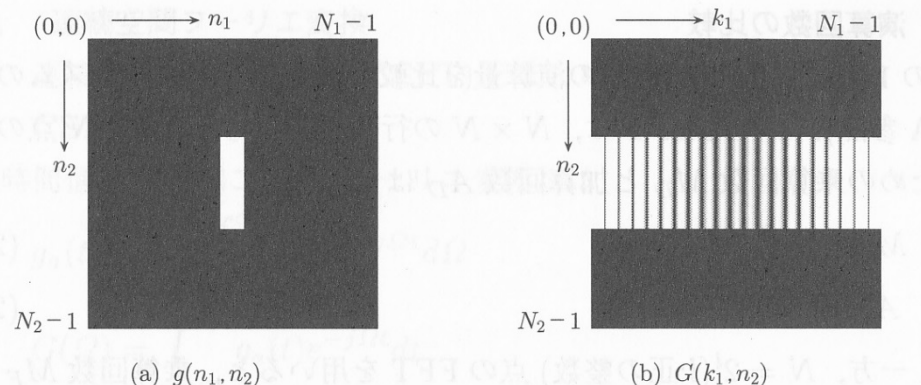
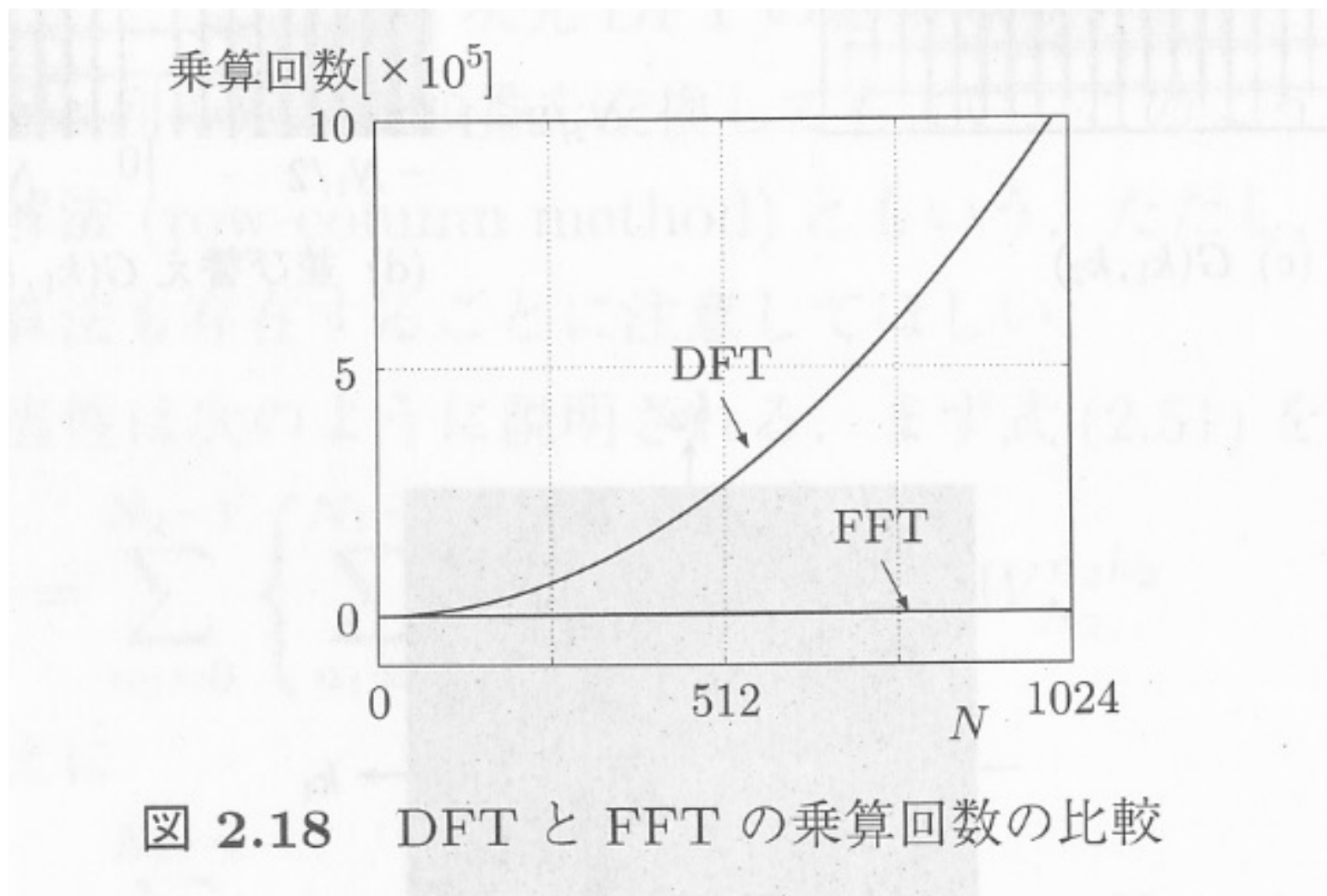


図 2.17 行列分解法による FFT 計算例



# Comparison of Operation Number



# Sampling Effect

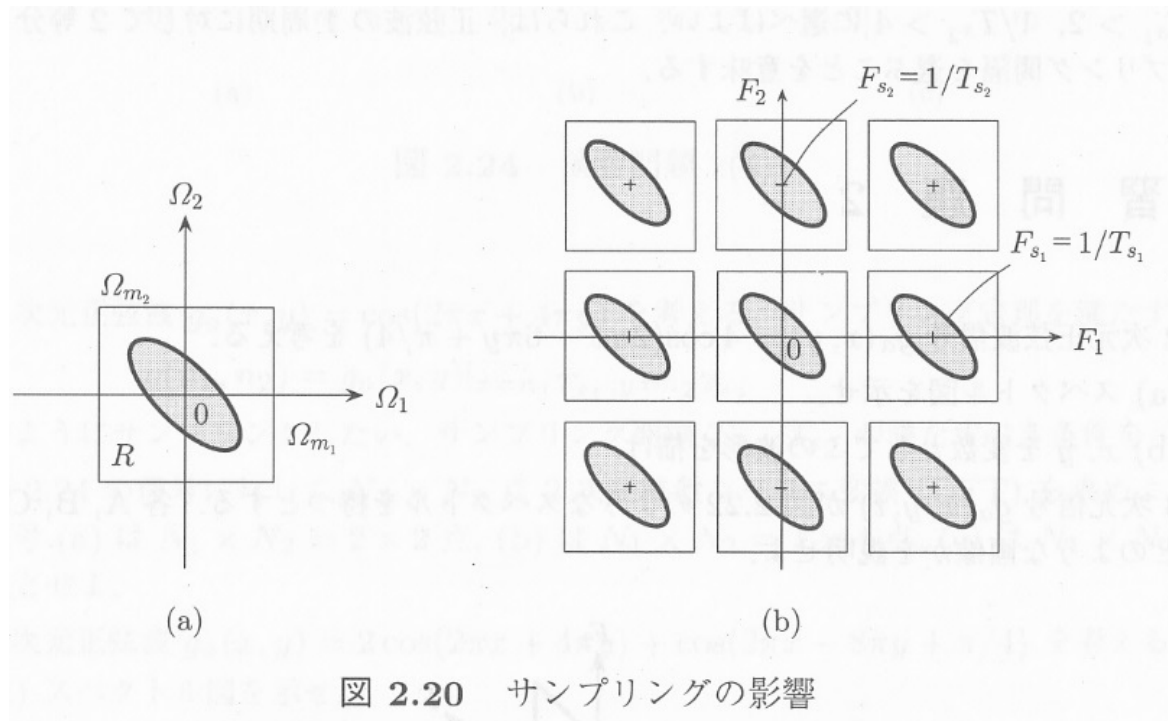
- Sampling generally gives signals distortions (aliasing).

# Sampling Theorem

- Theorem that give some condition to avoid the effects of sampling

# Sampling Theorem

- Fig.(a): 2D continuous signal's bandwidth is limited by angular frequency  $\Omega_{m1}$  and  $\Omega_{m2}$ . (No signal exists outside of the limited bandwidth.)
- Fig.(b): Assume rectangular sampling, 2D discrete signal has rectangular periodic spectrum.



# Sampling Theorem

$$F_{s_1} = 1/T_{s_1} > 2F_{m_1}, \text{ かつ } F_{s_2} = 1/T_{s_2} > 2F_{m_2}$$

No overlap exists for spectrum.

- Theoretically it is possible to reconstruct perfectly the original signal from sample values by filtering.

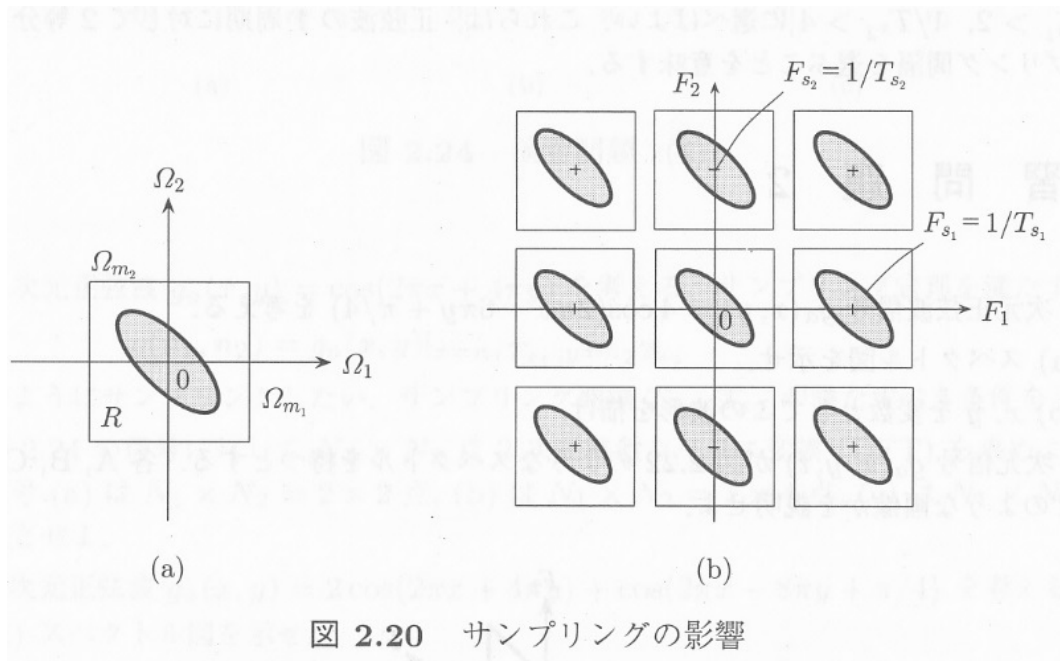
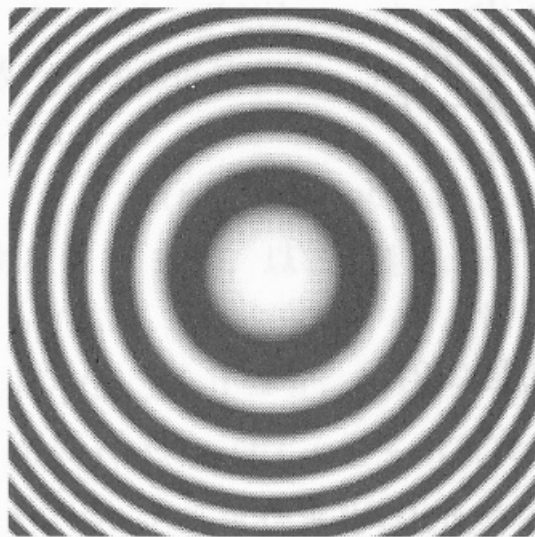


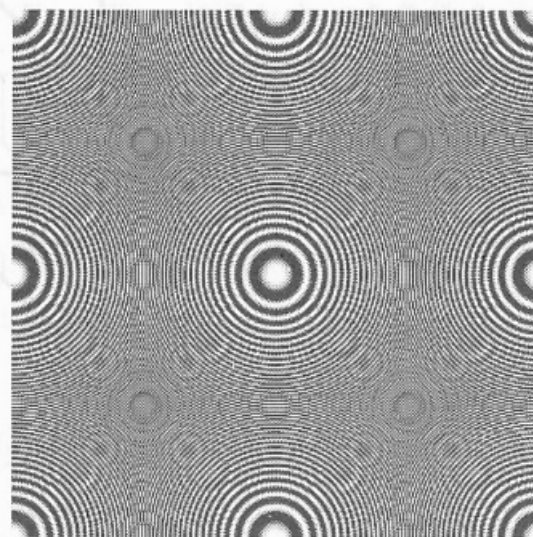
図 2.20 サンプリングの影響

# 折り返し歪み (Aliasing)

- By sampling without keeping sampling theorem, the spectrums overlap and distort the continuous signal. This distortion (overlap of spectrums) is called aliasing.



(a) エイリアジングなし



(b) エイリアジングあり

図 2.21 エイリアジングの発生例

# Example

$$g_a(x, y) = \cos(2\pi x + 4\pi y)$$

Calculate the maximum sampling intervals  $T_{s1}$  and  $T_{s2}$  to keep the sampling theorem.

# Answer

- Since the spatial frequency is 1, 2 respectively in the x and y directions, the minimum sampling frequencies are 2, and 4 and their corresponding sampling intervals  $T_{s1}$  and  $T_{s2}$  are  $\frac{1}{2}$  and  $\frac{1}{4}$ , respectively.



# Basics of Multi-dimensional Filter

- Most of image processing perform filtering to remove or enhance specific frequency components.
- Today we will study about filtering in the spatial domain and frequency domain.

# Typical Signals

- 2 D sinusoidal wave signal

$$g(n_1, n_2) = A \cos(\omega_1 n_1 + \omega_2 n_2 + \theta)$$

- 2 D complex sinusoidal wave signal

$$g(n_1, n_2) = A e^{j(\omega_1 n_1 + \omega_2 n_2)}$$

- 2 D unit sample signal (2D unit impulse)

$$\delta(n_1, n_2) = \begin{cases} 1, & n_1 = n_2 = 0 \\ 0, & \text{その他} \end{cases}$$

- 2 D unit step signal

$$u(n_1, n_2) = \begin{cases} 1, & n_1 \geq 0 \text{ かつ } n_2 \geq 0 \\ 0, & \text{その他} \end{cases}$$

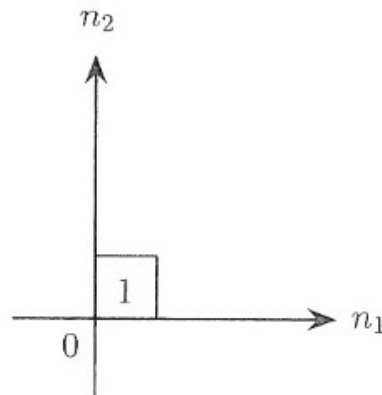
# Typical Signals

- 2D unit sample signal (2D impulse signal)

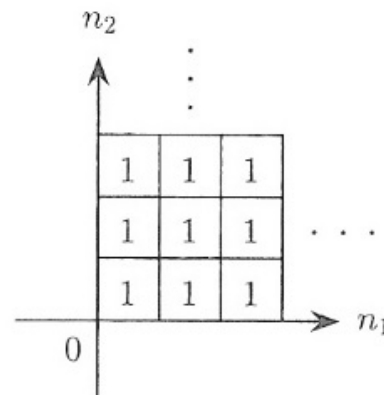
$$\delta(n_1, n_2) = \begin{cases} 1, & n_1 = n_2 = 0 \\ 0, & \text{その他} \end{cases}$$

- 2 D unit step signal

$$u(n_1, n_2) = \begin{cases} 1, & n_1 \geq 0 \text{ かつ } n_2 \geq 0 \\ 0, & \text{その他} \end{cases}$$



(a)  $\delta(n_1, n_2)$



(b)  $u(n_1, n_2)$

# Continuous Delta Function $\delta(t)$

$$\int_{-\infty}^{\infty} \delta(t) dt = 1 \text{ であり, } \delta(t) = \begin{cases} \infty, & t = 0 \\ 0, & t \neq 0 \end{cases}$$

# Example

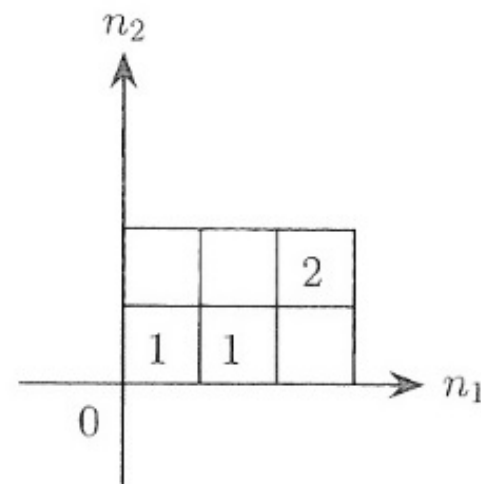
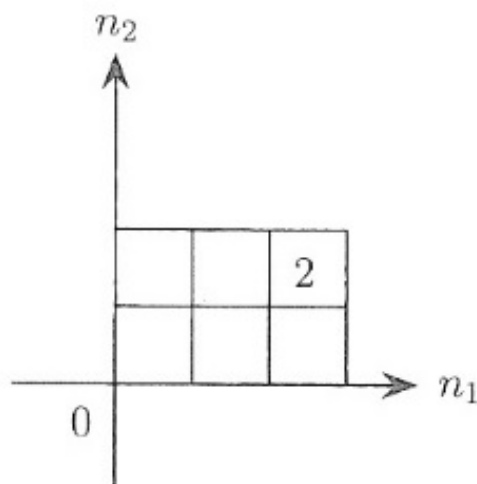
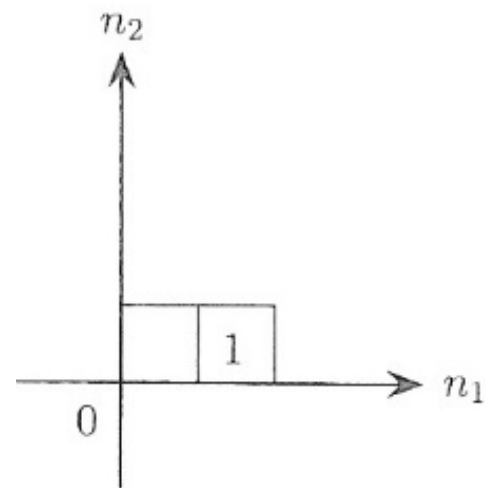
Represent impulse  $\delta(n_1, n_2)$  by 2D unit step signal  $u(n_1, n_2)$ .

# Answer

$$\delta(n_1, n_2) = u(n_1, n_2) - u(n_1 - 1, n_2) - u(n_1, n_2 - 1) + u(n_1 - 1, n_2 - 1)$$

# Example : Impulse Signal

- 2D image is a set of 2D impulse signals.
- Represent the following 2D images by using impulse  $\delta(n_1, n_2)$ .



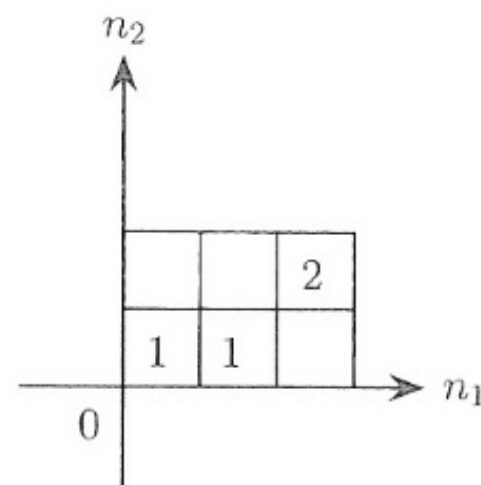
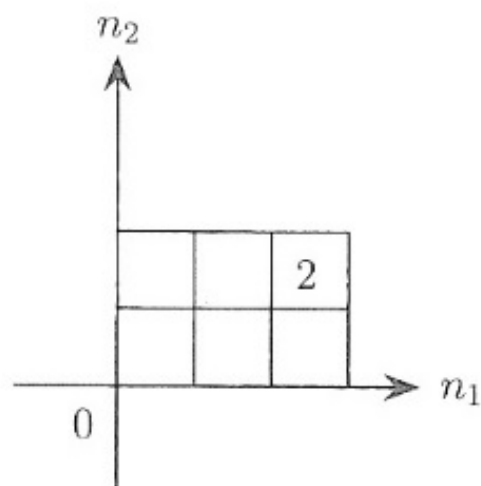
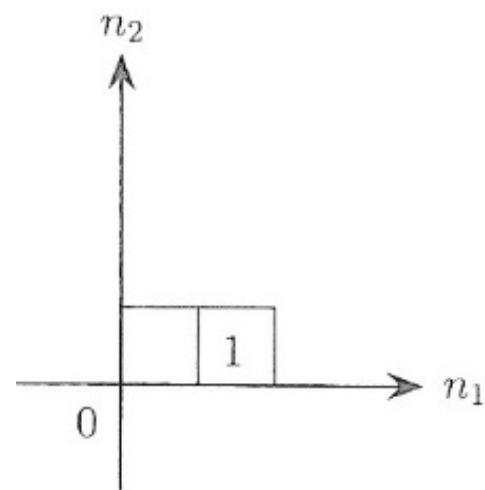
インパルス  $\delta(n_1, n_2)$

# Answers

$$g_1(n_1, n_2) = \delta(n_1 - 1, n_2)$$

$$g_2(n_1, n_2) = 2\delta(n_1 - 2, n_2 - 1)$$

$$g(n_1, n_2) = \delta(n_1, n_2) + \delta(n_1 - 1, n_2) + 2\delta(n_1 - 2, n_2 - 1)$$





# Representation of 2D Image by Impulse

- By generalization,

$$g(n_1, n_2) = \sum_{k_1=-\infty}^{\infty} \sum_{k_2=-\infty}^{\infty} g(k_1, k_2) \delta(n_1 - k_1, n_2 - k_2)$$

# Example Exercise

【例題 3.2】 図 3.3 の信号をインパルスを用いて表せ.

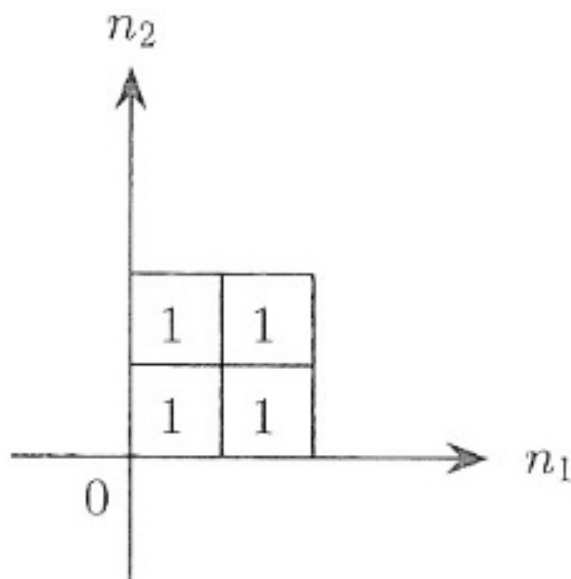


図 3.3 例題 3.2

# Answer

$$g(n_1, n_2) = \delta(n_1, n_2) + \delta(n_1 - 1, n_2) + \delta(n_1, n_2 - 1) + \delta(n_1 - 1, n_2 - 1)$$

# Exercise Example

What is the signal given by

$$g(n_1, n_2) = \sum_{k_1=0}^{\infty} \sum_{k_2=0}^{\infty} \delta(n_1 - k_1, n_2 - k_2)?$$

# Answer

2D unit step signal  $u(n_1, n_2)$

# Separability of Signal

- If a 2D signal is represented by a product of two 1D signals, it is called a separable signal.
- If not, it is called a non-separable signal.

$$g(n_1, n_2) = g_1(n_1)g_2(n_2)$$

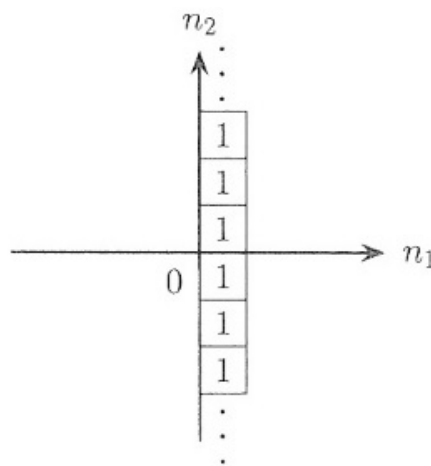
# Example : Separability of Signal

- Most general signal is non-separable.
- Are 2D unit impulse signal and 2D unit steps signal separable?

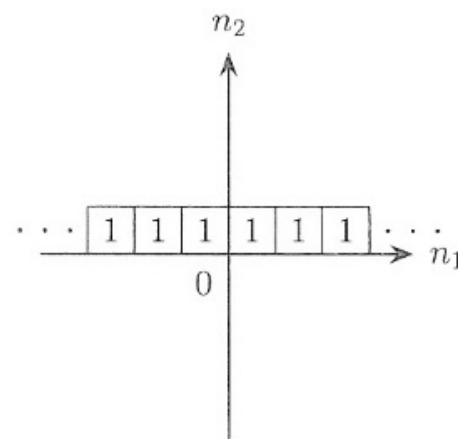
# Answer

$$g_1(n_1, n_2) = \delta(n_1)$$

$$g_2(n_1, n_2) = \delta(n_2)$$



(a)  $g_1(n_1, n_2) = \delta(n_1)$



(b)  $g_2(n_1, n_2) = \delta(n_2)$

See the left figure.

Hence,  $\delta(n_1, n_2) = \delta(n_1)\delta(n_2)$

2D unit step signal is also separable.



# Example

- Is the signal in Fig. 3.5 separable or non-separable?

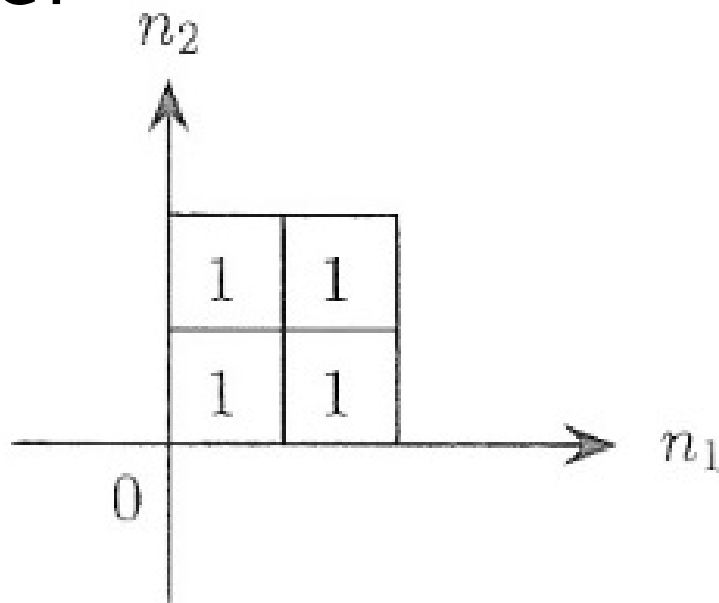


图 3.5 例题 3.4

# Answer

$$g_1(n_1) = \begin{cases} 1, & n_1 = 0, 1 \\ 0, & \text{その他} \end{cases} \quad g_2(n_2) = \begin{cases} 1, & n_2 = 0, 1 \\ 0, & \text{その他} \end{cases}$$

とすると,  $g(n_1, n_2) = g_1(n_1)g_2(n_2)$  と表現される.