

# Advanced Information Engineering

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Kenjiro T. Miura

# Discrete Sinusoidal Wave Signal

- Discretize  $g_a(x,y,t)$  by sampling.

$$\begin{aligned} g(n_1, n_2, n) &= g_a(x, y, t)|_{x=n_1T_{s_1}, y=n_2T_{s_2}, t=nT_s} \\ &= A \cos(\Omega_1 n_1 T_{s_1} + \Omega_2 n_2 T_{s_2} + \Omega n T_s) \\ &= A \cos(\omega_1 n_1 + \omega_2 n_2 + \omega n) \end{aligned}$$

where

$$\omega_1 = 2\pi f_1 = \Omega_1 T_{s_1} = 2\pi F_1 / F_{s_1}$$

$$\omega_2 = 2\pi f_2 = \Omega_2 T_{s_2} = 2\pi F_2 / F_{s_2}$$

$$\omega = 2\pi f = \Omega T_s = 2\pi F / F_s$$

$f$ :normalized frequency,  $\omega$ :normalized angular frequency

$F$ :non-normalized frequency,  $\Omega$ :non-normalized angular frequency

# Periodicity of Frequency Spectrum

When

$$F'_1 = F_1 + kF_{s_1}$$

$$F'_2 = F_2 + iF_{s_2}$$

For

$$g_a(x, y) = A \cos(2\pi(F_1x + F_2y))$$

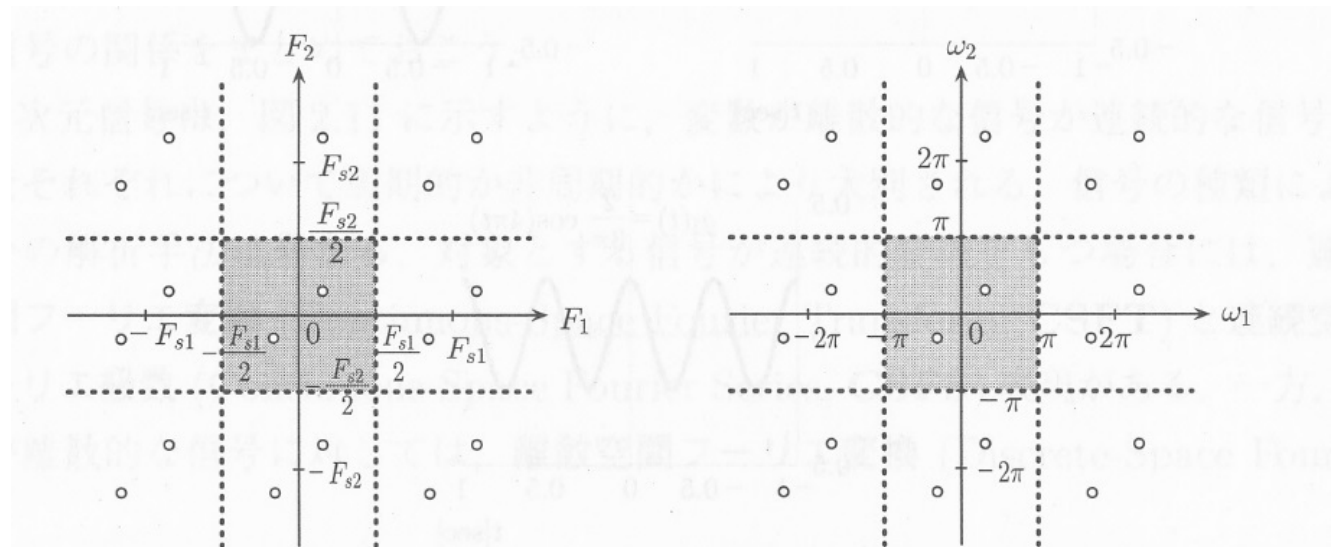
$$g'_a(x, y) = A \cos(2\pi(F'_1x + F'_2y))$$

their sample values are identical. (*proof skipped*)

Sinusoidal waves with frequencies different by the sampling frequency multiplied by integer cannot be distinguished.

# Periodicity of Frequency Spectrum

- Sinusoidal waves with frequencies different by the sampling frequency multiplied by integer cannot be distinguished.
- What kind of signals in the figure?



(a) 非正規化

(b) 正規化

図 2.9 スペクトルの周期性

Spectrum of discrete signals

# Exercise Example

- Let's consider  $g(n_1, n_2) = \cos(\pi n_1 + \pi n_2/2)$  .
- Calculate normalized frequencies  $f_1$  and  $f_2$ .
- When the sampling frequencies are  $F_{s1}$  and  $F_{s2}$ , Calculate non-normalized frequencies  $F_1$  and  $F_2$ .

# Answers

- Since  $\omega_1, \omega_2$  are  $\pi, \pi/2$ , respectively,  
 $f_1 = 1/2, f_2 = 1/4$
- $F_1 = F_{S1}/2, F_2 = F_{S2}/4$

# Fourier Analysis of Signal

- A non-sinusoidal wave is generated by adding two sinusoidal waves with different frequencies.
- There are cases where a non-sinusoidal wave is decomposed into plural sinusoidal waves.
- Since the law of superposition is satisfied for linear systems, the process for a non-sinusoidal wave comes down to those for plural sinusoidal waves by Fourier analysis.

# Fourier Transform of Discrete Signals

- The Fourier transform of 1-dimensional non-periodic discrete signal is given

$$g(n) = \frac{1}{2\pi} \int_{-\infty}^{\infty} G(e^{j\omega}) e^{j\omega n} d\omega$$

$$G(\omega) = \sum_{n=-\infty}^{\infty} g(n) e^{-j\omega n}$$



# Fourier Transform of Discrete Signals

- The Fourier transform of 2-dimensional non-periodic discrete signal is given by

$$g(n_1, n_2) = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G(e^{j\omega_1}, e^{j\omega_2}) e^{j\omega_1 n_1} e^{j\omega_2 n_2} d\omega_1 d\omega_2$$

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

# Amplitude and Phase Spectrums

Even  $g(n_1, n_2)$  is a real valued function,  $G(\omega_1, \omega_2)$  is generally a complex valued function.

As  $G(\omega_1, \omega_2) = A(\omega_1, \omega_2)e^{j\theta(\omega_1, \omega_2)}$ , complex values are represented by polar coordinate system and draw spectrum by calculating amplitude and phase spectrums.

# Exercise Example

- Perform Fourier transform the following 2-dimensional non-periodic discrete signal.
- Furthermore, as  $G(\omega_1, \omega_2) = A(\omega_1, \omega_2)e^{j\theta(\omega_1, \omega_2)}$  by representing by polar coordinate and calculate amplitude and phase spectrums.
- Draw the amplitude spectrum if possible.

$$g(n_1, n_2) = \begin{cases} 1 & (0 \leq n_1 \leq L_1 - 1, \text{ and } 0 \leq n_2 \leq L_2 - 1) \\ 0 & (\text{otherwise}) \end{cases}$$

# Answer

$$G(\omega_1, \omega_2) = \sum_{n_1=0}^{L_1-1} \sum_{n_2=0}^{L_2-1} e^{-j(\omega_1 n_1 + \omega_2 n_2)}$$

$$= \sum_{n_1=0}^{L_1-1} e^{-j\omega_1 n_1} \cdot \sum_{n_2=0}^{L_2-1} e^{-j\omega_2 n_2}$$

$$= \frac{1 - e^{-j\omega_1 L_1}}{1 - e^{-j\omega_1}} \cdot \frac{1 - e^{-j\omega_2 L_2}}{1 - e^{-j\omega_2}}$$

# Answer

$$= \frac{1 - e^{-j\omega_1 L_1}}{1 - e^{-j\omega_1}} \cdot \frac{1 - e^{-j\omega_2 L_2}}{1 - e^{-j\omega_2}}$$

$$= \frac{e^{\frac{-j\omega_1 L_1}{2}} \left( e^{\frac{j\omega_1 L_1}{2}} - e^{\frac{-j\omega_1 L_1}{2}} \right)}{e^{\frac{-j\omega_1}{2}} \left( e^{\frac{j\omega_1}{2}} - e^{\frac{-j\omega_1}{2}} \right)} \cdot \frac{e^{\frac{-j\omega_2 L_2}{2}} \left( e^{\frac{j\omega_2 L_2}{2}} - e^{\frac{-j\omega_2 L_2}{2}} \right)}{e^{\frac{-j\omega_2}{2}} \left( e^{\frac{j\omega_2}{2}} - e^{\frac{-j\omega_2}{2}} \right)}$$

$$= e^{\frac{-j\omega_1(L_1-1)}{2}} \cdot \frac{\sin \frac{\omega_1 L_1}{2}}{\sin \frac{\omega_1}{2}} \cdot e^{\frac{-j\omega_2(L_2-1)}{2}} \cdot \frac{\sin \frac{\omega_2 L_2}{2}}{\sin \frac{\omega_2}{2}}$$

# Answer

$$= e^{\frac{-j\omega_1(L_1-1)}{2}} \cdot \frac{\sin \frac{\omega_1 L_1}{2}}{\sin \frac{\omega_1}{2}} \cdot e^{\frac{-j\omega_2(L_2-1)}{2}} \cdot \frac{\sin \frac{\omega_2 L_2}{2}}{\sin \frac{\omega_2}{2}}$$

$$= \frac{\sin \frac{\omega_1 L_1}{2}}{\sin \frac{\omega_1}{2}} \cdot \frac{\sin \frac{\omega_2 L_2}{2}}{\sin \frac{\omega_2}{2}} \cdot e^{j\left(\frac{-\omega_1(L_1-1)}{2} + \frac{-\omega_2(L_2-1)}{2}\right)}$$

$$A(\omega_1, \omega_2) = \left| \frac{\sin \frac{\omega_1 L_1}{2}}{\sin \frac{\omega_1}{2}} \cdot \frac{\sin \frac{\omega_2 L_2}{2}}{\sin \frac{\omega_2}{2}} \right| \quad \leftarrow \text{図示すると}$$

$$\theta(\omega_1, \omega_2) = \left| \frac{-\omega_1(L_1-1)}{2} + \frac{-\omega_2(L_2-1)}{2} \right|$$

# Answer

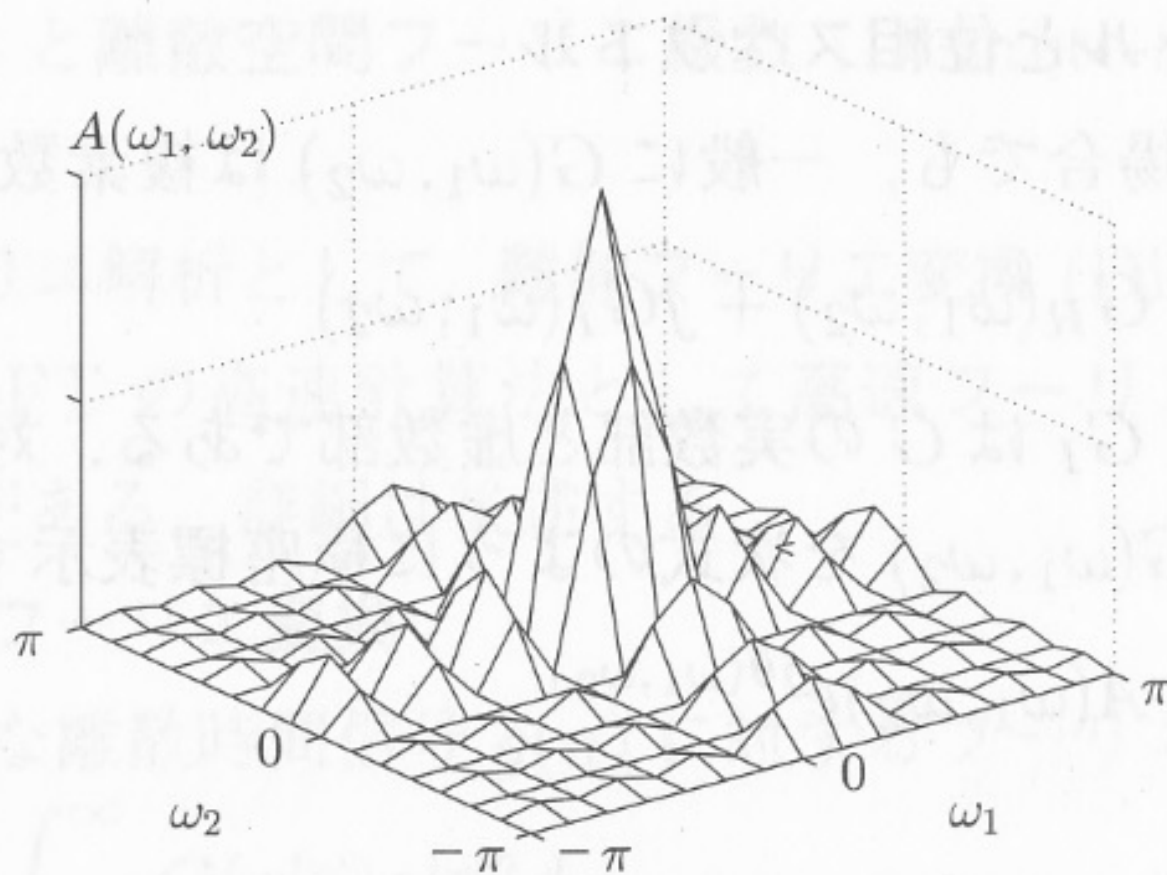
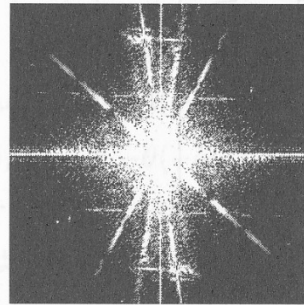


図 2.12 振幅スペクトル例 ( $L_1 = L_2 = 8$ )

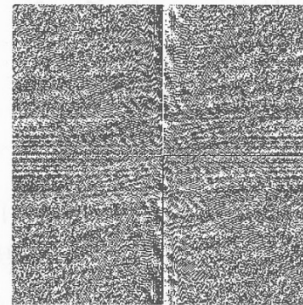
# Amplitude-only and Phase-only Images



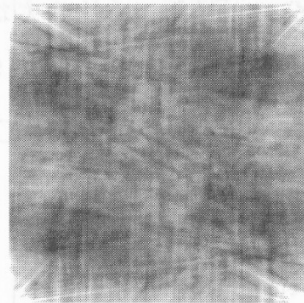
(a) 原画像



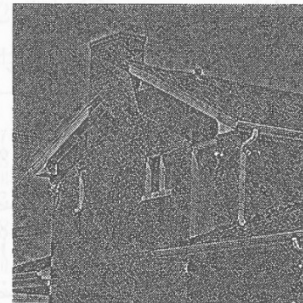
(b) 振幅スペクトル



(c) 位相スペクトル



(d) 振幅限定画像



(e) 位相限定画像

図 2.13 振幅限定画像と位相限定画像



# Symmetry of Spectrum

- In case where  $g(n_1, n_2)$  is a real valued function, its discrete Fourier transform given by

$G(\omega_1, \omega_2) = A(\omega_1, \omega_2)e^{j\theta(\omega_1, \omega_2)}$  satisfies

$$A(\omega_1, \omega_2) = A(-\omega_1, -\omega_2)$$

$$\theta(\omega_1, \omega_2) = -\theta(-\omega_1, -\omega_2)$$

(proof skipped)

Amplitude spectrum : even symmetry

Phase spectrum : odd symmetry

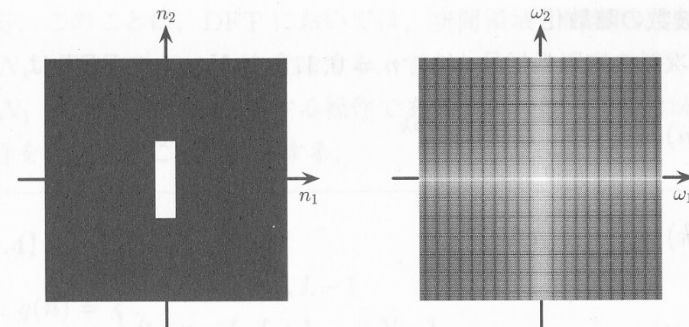
# Signal Shift

- Signal  $g(n_1, n_2)$  and its discrete Fourier transform  $G(\omega_1, \omega_2)$
- For integers  $k_1, k_2$ , and signal  $g(n_1 - k_1, n_2 - k_2)$ , its discrete Fourier transform is given by

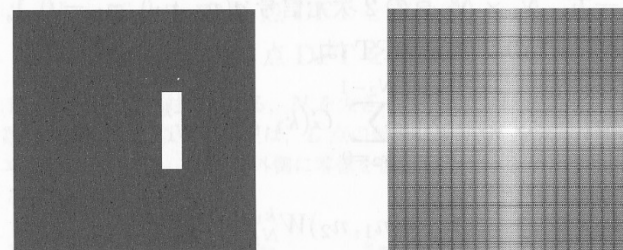
$$G(\omega_1, \omega_2) e^{-i(\omega_1 k_1 + \omega_2 k_2)}.$$

(proof skipped)

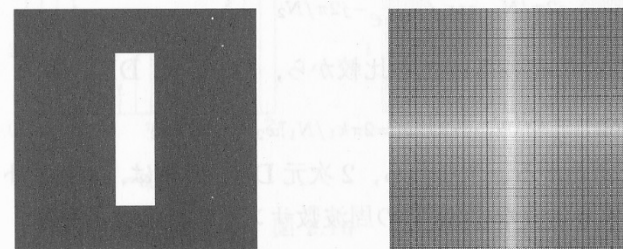
- No effect on amplitude spectrum
- Why ?



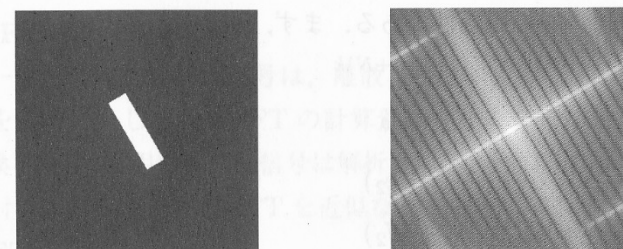
(a) 原画像



(b) シフト



(c) 拡大



(d) 回転

# Discrete Spatial Fourier Transform (DSFT)

$$g(n_1, n_2) = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G(e^{j\omega_1}, e^{j\omega_2}) e^{j\omega_1 n_1} e^{j\omega_2 n_2} d\omega_1 d\omega_2$$

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

The Fourier analysis performed by computer is done by FFT.

# Discrete Spatial Fourier Transform (DSFT)

Consider the case where  $g(n_1, n_2)$  is defined within finite domain  $N_1 \times N_2$ , i.e. 2-dimensional image signal.

$$G(\omega_1, \omega_2) = \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} g(n_1, n_2) e^{-j\omega_1 n_1} e^{-j\omega_2 n_2}$$

Can the computer perform DSFT ?

# 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

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

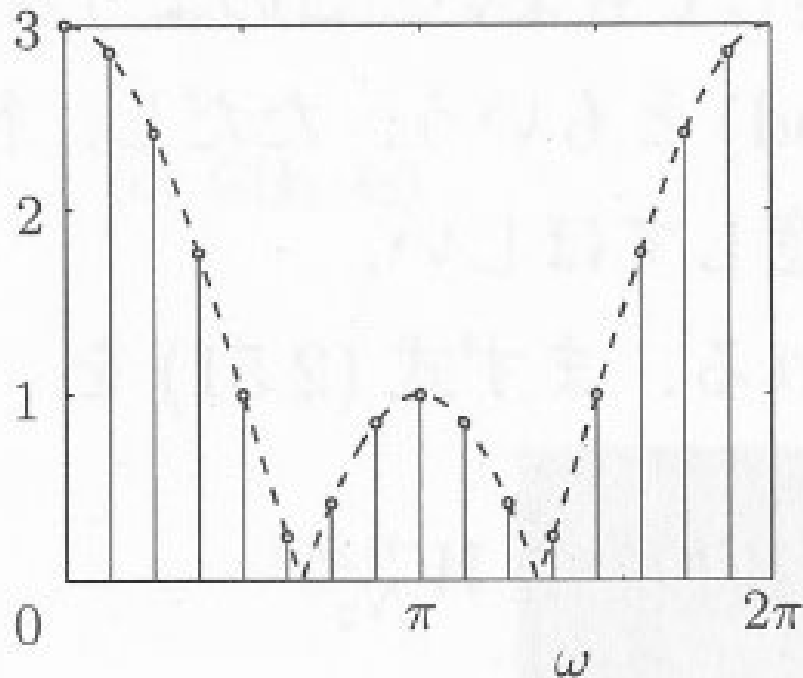
を考える．以下の問いに答えよ．

- (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．

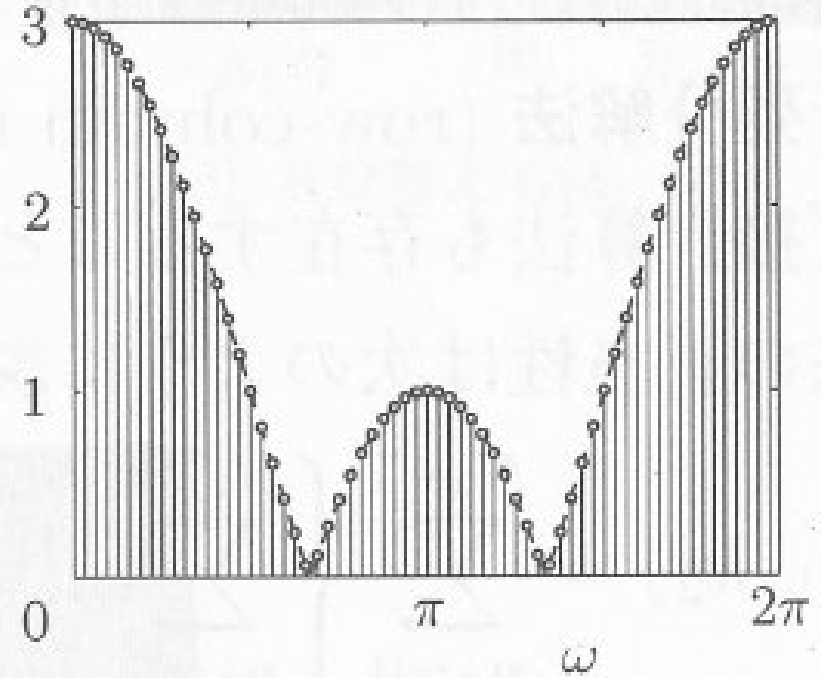
You can use a calculator to calculate amplitude.  
Since (b) takes long time , please do (a).



# Answer



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



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

- More larger  $N$ , 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.

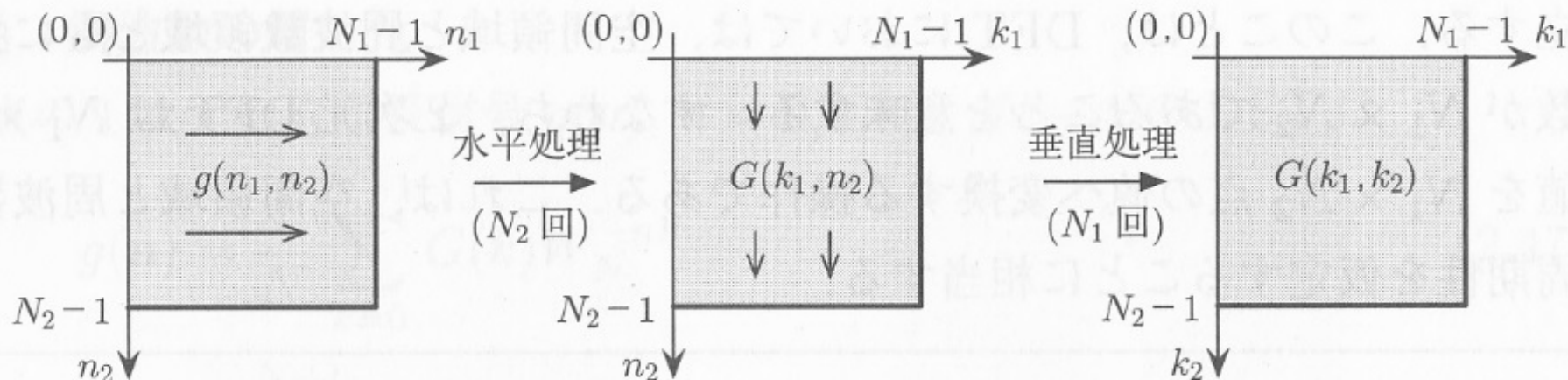


图 2.16 行列分解法

# 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.

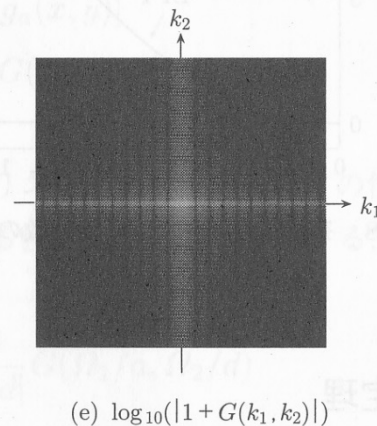
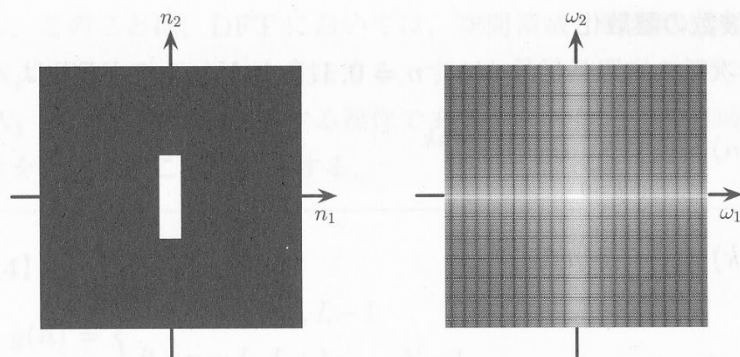
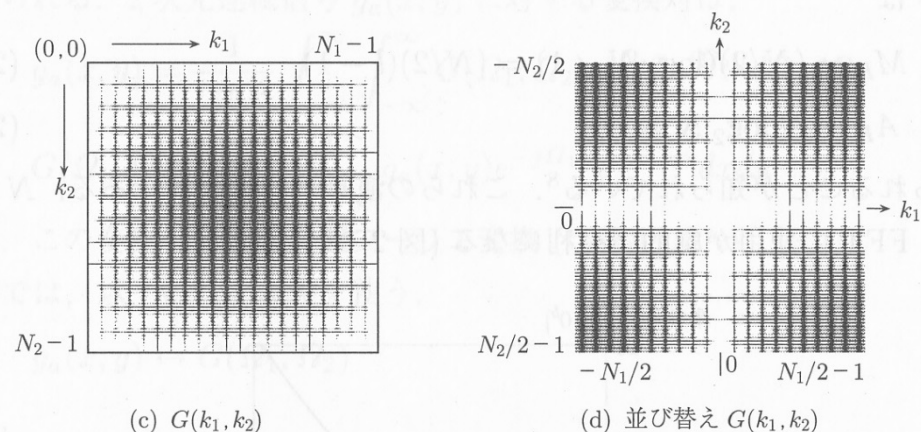
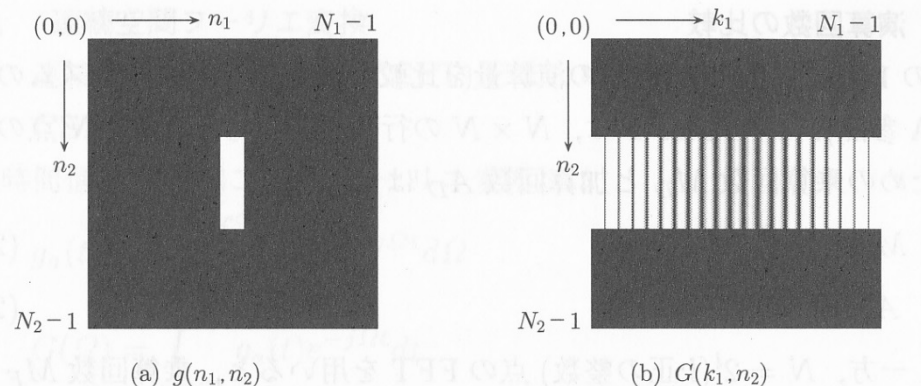
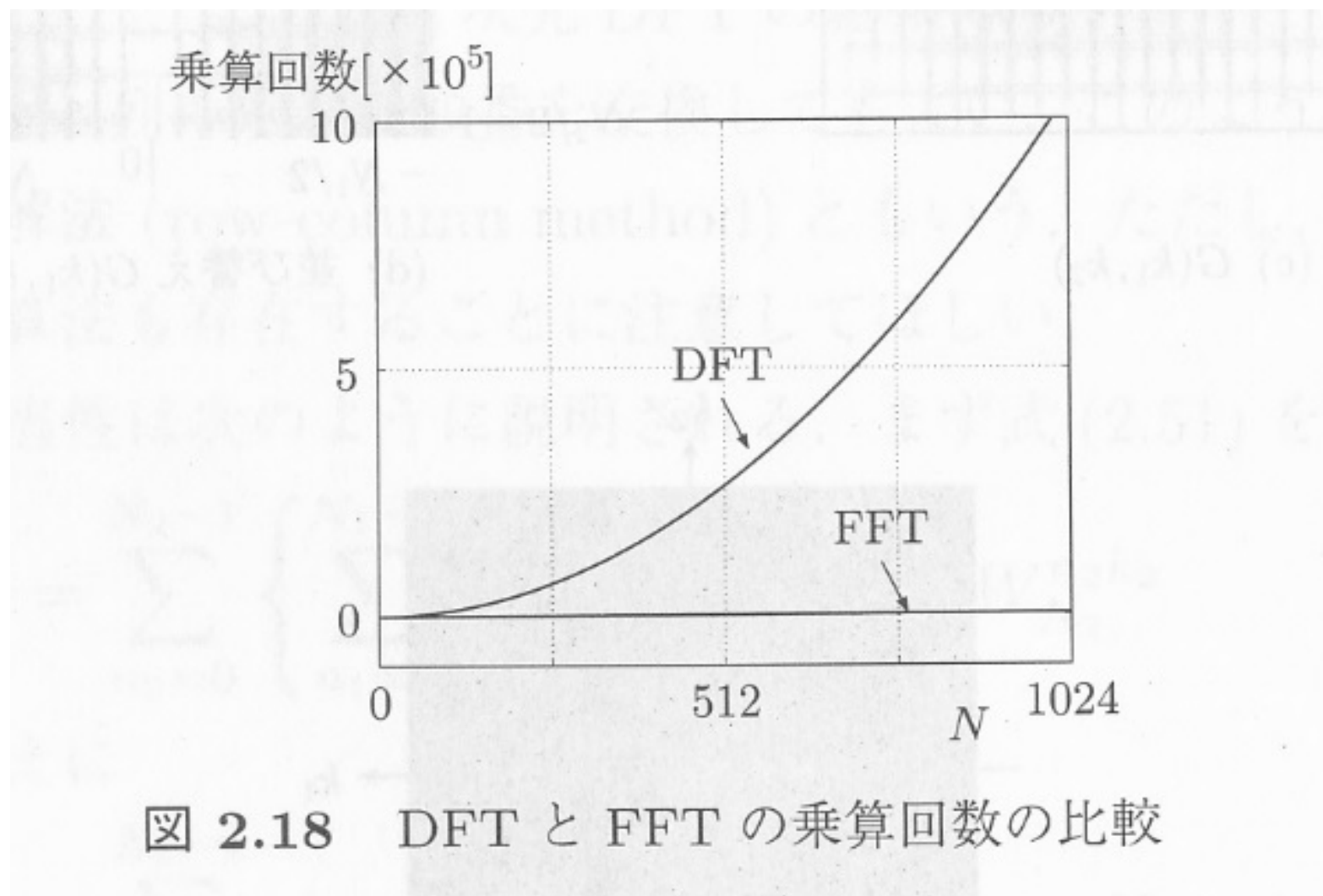


図 2.17 行列分解法による FFT 計算例

# Comparison of Operation Number



# Sampling Effect

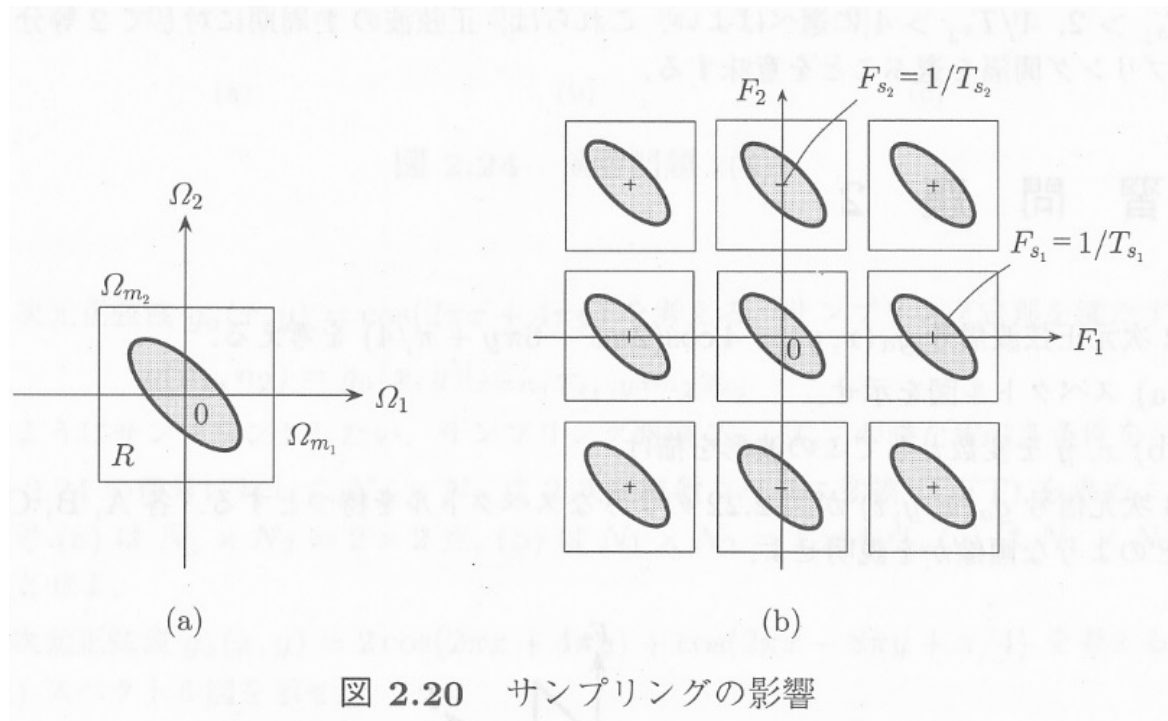
- Sampling generally gives signals distortions (aliasing).

# 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.



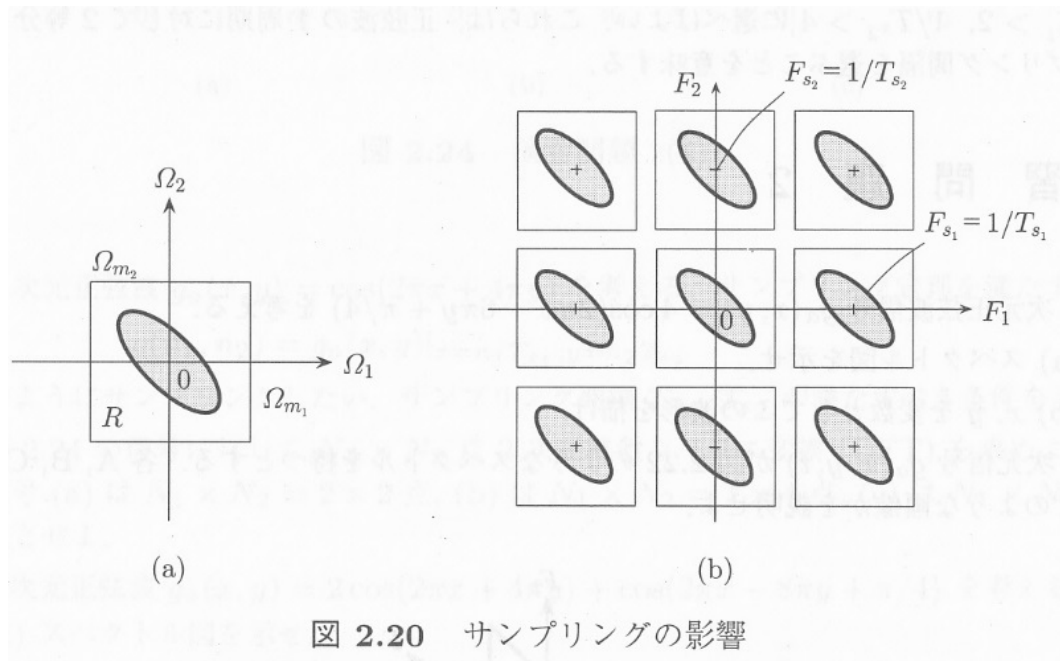


# Sampling Theorem

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

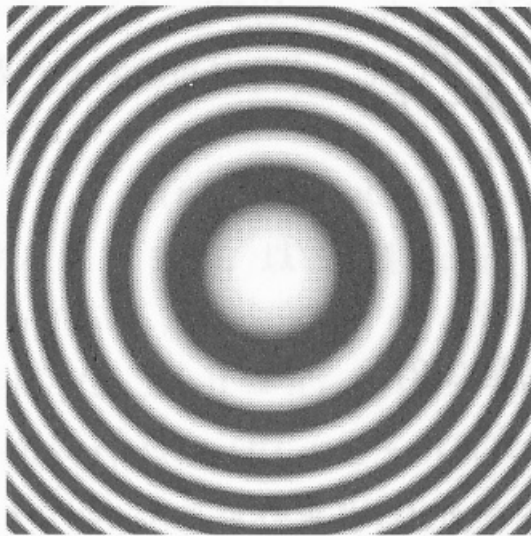
No overlap exists for spectrum.

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

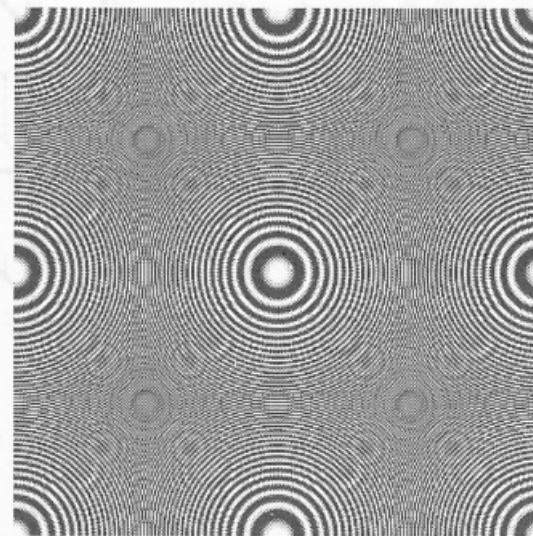


# 折り返し歪み (Aliasing)

- 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  $1/2$  and  $1/4$ , respectively.

# Assignment #2

- Hand out Assignment #2