

XIV. Orthogonal Transform and Multiplexing

© 14-A Orthogonal and Dual Orthogonal

Any $M \times N$ discrete linear transform can be expressed as the matrix form:

$$\begin{array}{c}
 \left[\begin{array}{c} y[0] \\ y[1] \\ y[2] \\ \vdots \\ y[M-1] \end{array} \right] = \left[\begin{array}{ccccc} \phi_0^*[0] & \phi_0^*[1] & \phi_0^*[2] & \cdots & \phi_0^*[N-1] \\ \phi_1^*[0] & \phi_1^*[1] & \phi_1^*[2] & \cdots & \phi_1^*[N-1] \\ \phi_2^*[0] & \phi_2^*[1] & \phi_2^*[2] & \cdots & \phi_2^*[N-1] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \phi_{M-1}^*[0] & \phi_{M-1}^*[1] & \phi_{M-1}^*[2] & \cdots & \phi_{M-1}^*[N-1] \end{array} \right] \left[\begin{array}{c} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[N-1] \end{array} \right] \\
 \mathbf{Y} = \qquad \qquad \qquad \mathbf{A} \qquad \qquad \qquad \mathbf{X}
 \end{array}$$

$$y[m] = \langle x[n], \phi_m[n] \rangle = \sum_{n=0}^{N-1} x[n] \phi_m^*[n]$$

\uparrow
 inner product

Orthogonal: $\langle \phi_k[n], \phi_h[n] \rangle = \sum_{n=0}^{N-1} \phi_k[n] \phi_h^*[n] = 0$ **when $k \neq h$**

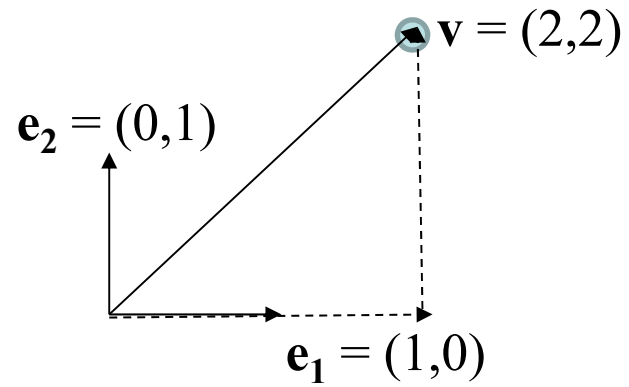
orthogonal transforms 的例子：

- discrete Fourier transform
- discrete cosine, sine, Hartley transforms
- Walsh Transform, Haar Transform
- discrete Legendre transform
- discrete orthogonal polynomial transforms
Hahn, Meixner, Krawtchouk, Charlier

為什麼在信號處理上，我們經常用 orthogonal transform?

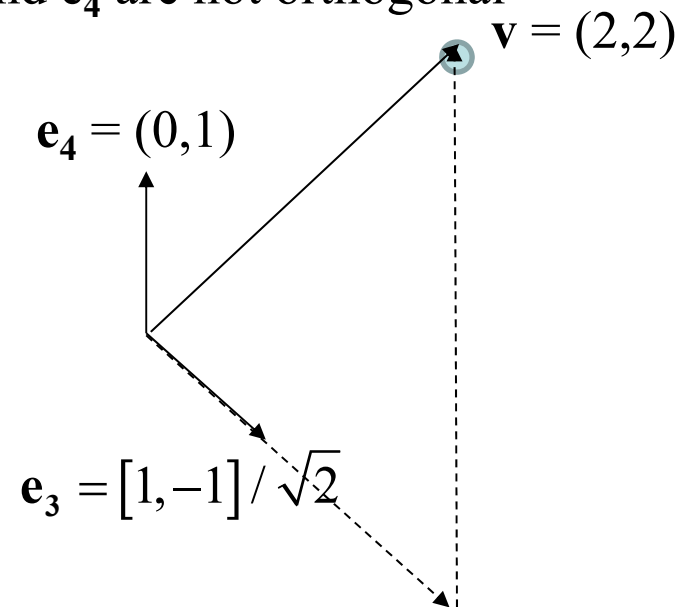
Orthogonal transform 最大的好處何在？

\mathbf{e}_1 and \mathbf{e}_2 are orthogonal



$$\mathbf{v} = 2\mathbf{e}_1 + 2\mathbf{e}_2$$

\mathbf{e}_3 and \mathbf{e}_4 are not orthogonal



$$\mathbf{v} = 2\sqrt{2}\mathbf{e}_3 + 4\mathbf{e}_4$$

- If partial terms are used for reconstruction

for orthogonal case,

perfect reconstruction: $x[n] = \sum_{m=0}^{N-1} C_m^{-1} y[m] \phi_m[n]$

partial reconstruction: $x_K[n] = \sum_{m=0}^{K-1} C_m^{-1} y[m] \phi_m[n] \quad K < N$

reconstruction error of partial reconstruction

$$\begin{aligned} \|x[n] - x_K[n]\|^2 &= \sum_{n=0}^{N-1} \left\| \sum_{m=K}^{N-1} C_m^{-1} y[m] \phi_m[n] \right\|^2 \\ &= \sum_{n=0}^{N-1} \sum_{m=K}^{N-1} C_m^{-1} y[m] \phi_m[n] \sum_{m_1=K}^{N-1} C_{m_1}^{-1} y^*[m_1] \phi_{m_1}^*[n] \\ &= \sum_{m=K}^{N-1} \sum_{m_1=K}^{N-1} C_m^{-1} y[m] C_{m_1}^{-1} y^*[m_1] \sum_{n=0}^{N-1} \phi_m[n] \phi_{m_1}^*[n] \\ &= \sum_{m=K}^{N-1} \sum_{m_1=K}^{N-1} C_m^{-1} y[m] C_{m_1}^{-1} y^*[m_1] C_m \delta[m - m_1] = \sum_{m=K}^{N-1} C_m^{-1} |y[m]|^2 \end{aligned}$$

由於 $C_m^{-1} |y[m]|^2$ 一定是正的，可以保證 K 越大, reconstruction error 越小

For non-orthogonal case,

perfect reconstruction: $x[n] = \sum_{m=0}^{N-1} B[n, m] y[m] \quad \mathbf{B} = \mathbf{A}^{-1}$

partial reconstruction: $x_K[n] = \sum_{m=0}^{K-1} B[n, m] y[m] \quad K < N$

reconstruction error of partial reconstruction

$$\begin{aligned} \|x[n] - x_K[n]\|^2 &= \sum_{n=0}^{N-1} \left\| \sum_{m=K}^{N-1} B[n, m] y[m] \right\|^2 \\ &= \sum_{n=0}^{N-1} \sum_{m=K}^{N-1} B[n, m] y[m] \sum_{m_1=K}^{N-1} B^*[n, m_1] y^*[m_1] \\ &= \sum_{m=K}^{N-1} \sum_{m_1=K}^{N-1} y[m] y^*[m_1] \sum_{n=0}^{N-1} B[n, m] B^*[n, m_1] \end{aligned}$$

由於 $y[m] y^*[m_1] \sum_{n=0}^{N-1} B[n, m] B^*[n, m_1]$ 不一定是正的，
無法保證 K 越大, reconstruction error 越小

◎ 14-B Frequency and Time Division Multiplexing

傳統 Digital Modulation and Multiplexing : 使用 Fourier transform

- **Frequency-Division Multiplexing (FDM)**

$$z(t) = \sum_{n=0}^{N-1} X_n \exp(j2\pi f_n t) \quad X_n = 0 \text{ or } 1$$

X_n can also be set to be -1 or 1

When (1) $t \in [0, T]$ (2) $f_n = n/T$

$$z(t) = \sum_{n=0}^{N-1} X_n \exp\left(j \frac{2\pi n t}{T}\right)$$

it becomes the **orthogonal frequency-division multiplexing (OFDM)** in the continuous case.

Furthermore, if the time-axis is also sampled

$$t = mT/N, \quad m = 0, 1, 2, \dots, N-1$$

$$z\left(m\frac{T}{N}\right) = \sum_{n=0}^{N-1} X_n \exp\left(j\frac{2\pi nm}{N}\right)$$

$t \in [0, T]$
sampling for t-axis

(OFDM in the discrete case)

then the OFDM is equivalent to the transform matrix of the **inverse discrete Fourier transform (IDFT)**, which is one of the discrete orthogonal transform.

Modulation: $Y_m = z\left(m\frac{T}{N}\right) = \sum_{n=0}^{N-1} A[m, n]X_n$

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & e^{j\frac{2\pi}{N}} & e^{j\frac{4\pi}{N}} & \dots & e^{j\frac{2(N-1)\pi}{N}} \\ 1 & e^{j\frac{4\pi}{N}} & e^{j\frac{8\pi}{N}} & \dots & e^{j\frac{4(N-1)\pi}{N}} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & e^{j\frac{2(N-1)\pi}{N}} & e^{j\frac{4(N-1)\pi}{N}} & \dots & e^{j\frac{2(N-1)(N-1)\pi}{N}} \end{bmatrix}$$

Modulation:
$$Y_m = \sum_{n=0}^{N-1} A[m, n] X_n$$

Demodulation:
$$X_n = \frac{1}{N} \sum_{m=0}^{N-1} A^*[m, n] Y_m$$

Example: $N = 8$

$$X_n = [1, 0, 1, 1, 0, 0, 1, 1] \quad (n = 0 \sim 7)$$

- **Time-Division Multiplexing (TDM)**

$$z(0) = X_0, \quad z\left(\frac{T}{N}\right) = X_1, \quad z\left(2\frac{T}{N}\right) = X_2, \quad \dots, \quad z\left((N-1)\frac{T}{N}\right) = X_{N-1}$$

$$y(m) = z\left(m\frac{T}{N}\right) = \sum_{n=0}^{N-1} A[m,n]X_n$$

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

(also a discrete orthogonal transform)

思考：

既然 time-division multiplexing 那麼簡單

那為什麼要使用 frequency-division multiplexing
和 orthogonal frequency-division multiplexing (OFDM)?

◎ 14-C Code Division Multiple Access (CDMA)

除了 **frequency**-division multiplexing 和 **time**-division multiplexing，是否還有其他 multiplexing 的方式？

使用其他的 orthogonal transforms
即 code division multiple access (CDMA)

CDMA is an important topic in **spread spectrum** communication

參考資料

[1] M. A. Abu-Rgheff, *Introduction to CDMA Wireless Communications*, Academic, London, 2007

[2] 邱國書, 陳立民譯, “CDMA 展頻通訊原理”, 五南, 台北, 2002.

CDMA 最常使用的 orthogonal transform 為 Walsh transform

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \\ y_7 \\ y_8 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix}$$

channel 1
 channel 2
 channel 3
 channel 4
 channel 5
 channel 6
 channel 7
 channel 8

channel 1

當有兩組人在同一個房間裡交談 (A 和B交談) , (C 和D交談) ,
如何才能夠彼此不互相干擾?

(1) Different Time

(2) Different Tone

(3) Different Language

CDMA 分為：

- (1) Orthogonal Type (2) Pseudorandom Sequence Type

Orthogonal Type 的例子： 兩組資料 $[1, 0, 1]$ $[1, 1, 0]$

(1) 將 0 變為 -1 $[1, -1, 1]$ $[1, 1, -1]$

(2) $1, -1, 1$ modulated by $[1, 1, 1, 1, 1, 1, 1, 1]$ (channel 1)

→ $[1, 1, 1, 1, 1, 1, 1, 1, -1, -1, -1, -1, -1, -1, -1, -1, 1, 1, 1, 1, 1, 1, 1, 1]$

$1, 1, -1$ modulated by $[1, 1, 1, 1, -1, -1, -1, -1]$ (channel 2)

→ $[1, 1, 1, 1, -1, -1, -1, -1, 1, 1, 1, 1, -1, -1, -1, -1, -1, -1, -1, -1, 1, 1, 1, 1]$

(3) 相合

$[2, 2, 2, 2, 0, 0, 0, 0, 0, 0, 0, 0, -2, -2, -2, -2, 0, 0, 0, 0, 2, 2, 2, 2]$

demodulation

[2, 2, 2, 2, 0, 0, 0, 0, 0, 0, 0, 0, -2, -2, -2, -2, 0, 0, 0, 0, 2, 2, 2, 2]

[1, 1, 1, 1, 1, 1, 1, 1]

[1, 1, 1, 1, 1, 1, 1, 1]

[1, 1, 1, 1, 1, 1, 1, 1]

↓
内積 = 8

注意：

- (1) 使用 N -point Walsh transform 時，總共可以有 N 個 channels
- (2) 除了 Walsh transform 以外，其他的 orthogonal transform 也可以使用
- (3) 使用 Walsh transform 的好處

- Orthogonal Transform 共通的問題: 需要同步 synchronization

$$\mathbf{R}_1 = [1, 1, 1, 1, 1, 1, 1, 1]$$

$$\mathbf{R}_2 = [1, 1, 1, 1, -1, -1, -1, -1]$$

$$\mathbf{R}_5 = [1, -1, -1, 1, 1, -1, -1, 1]$$

$$\mathbf{R}_8 = [1, -1, 1, -1, 1, -1, 1, -1]$$

但是某些 basis, 就算不同步也近似 orthogonal

$$\langle \mathbf{R}_1[n], \mathbf{R}_1[n] \rangle = 8, \quad \langle \mathbf{R}_1[n], \mathbf{R}_k[n] \rangle = 0 \text{ if } k \neq 1$$

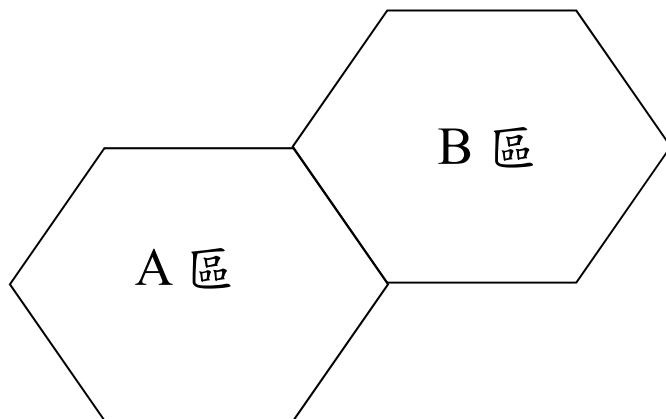
$$\langle \mathbf{R}_1[n], \mathbf{R}_k[n-1] \rangle = 2 \text{ or } 0 \quad \text{if } k \neq 1.$$

這裡的 shift 為 circular shift

CDMA 的優點：

- (1) 運算量相對於 frequency division multiplexing 減少很多
- (2) 可以減少 noise 及 interference 的影響
- (3) 可以應用在保密和安全傳輸上
- (4) 就算只接收部分的信號，也有可能把原來的信號 recover 回來
- (5) 相鄰的區域的干擾問題可以減少

相鄰的區域，使用差距最大的「語言」，則干擾最少



假設 A 區使用的 orthogonal basis 為 $\phi_k[n], k = 0, 1, 2, \dots, N-1$

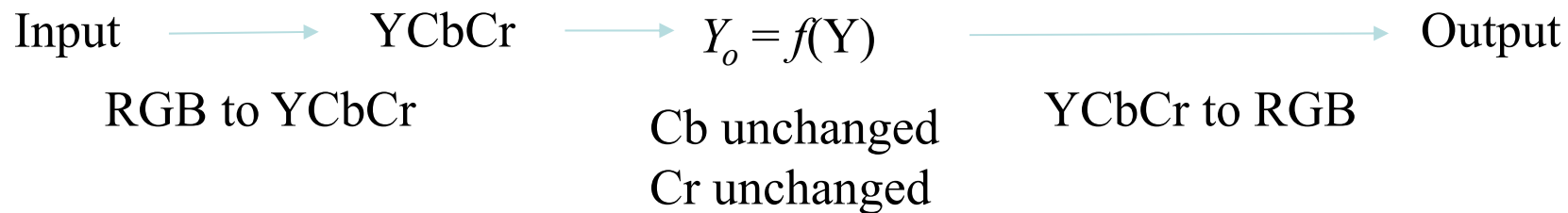
B 區使用的 orthogonal basis 為 $\mu_h[n], h = 0, 1, 2, \dots, N-1$

設法使 $\max \left(\left| \frac{\langle \phi_k[n], \mu_h[n] \rangle}{\langle \phi_k[n], \phi_h[n] \rangle} \right| \right)$ 為最小

$k = 0, 1, 2, \dots, N-1, h = 0, 1, 2, \dots, N-1$

附錄十七 常用的影像修飾方法

(1) Lightening and Darkening

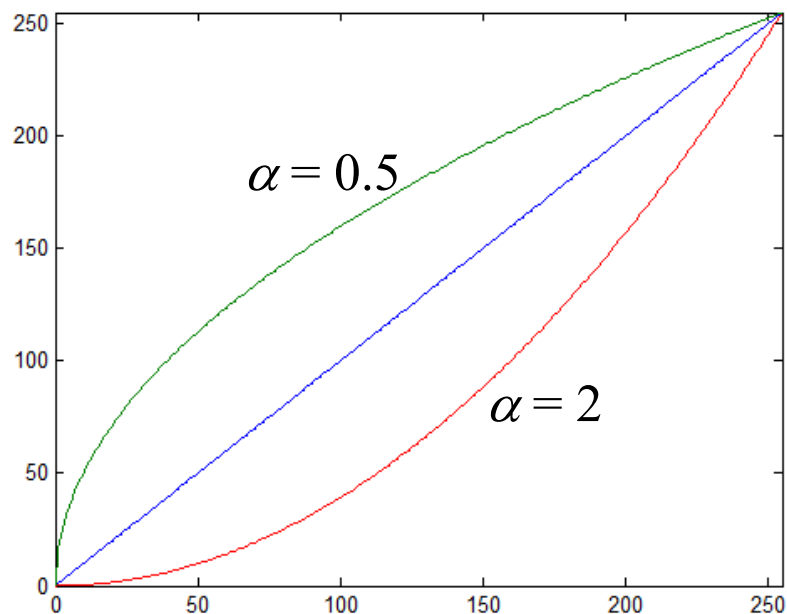


Example:

$$f(Y) = 255 \left(\frac{Y}{255} \right)^\alpha$$

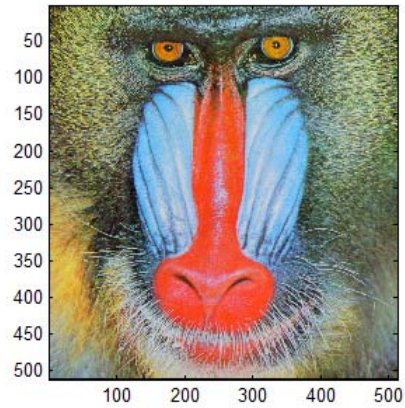
$\alpha < 1$: lightening

$\alpha > 1$: darkening

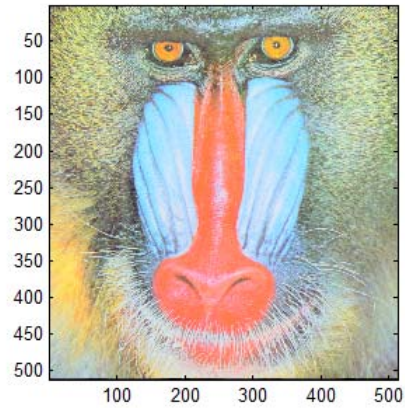


附錄十七 常用的影像修飾方法

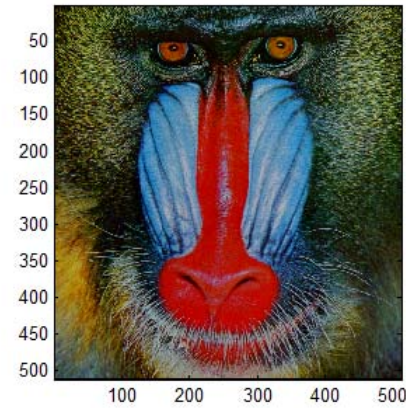
original image



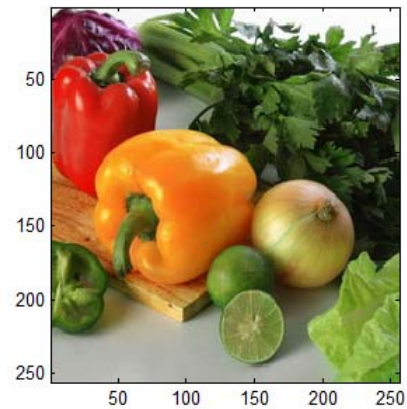
lighten



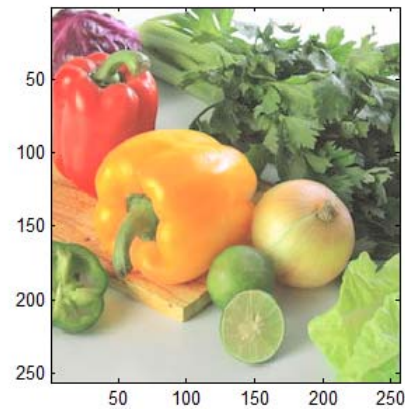
darken



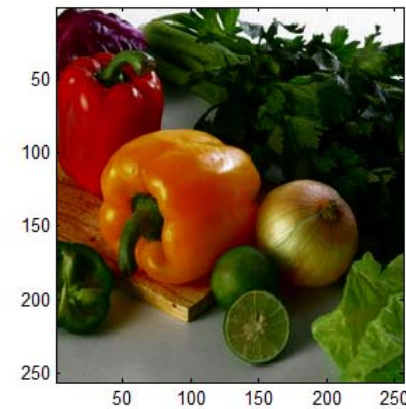
original image



lighten



darken

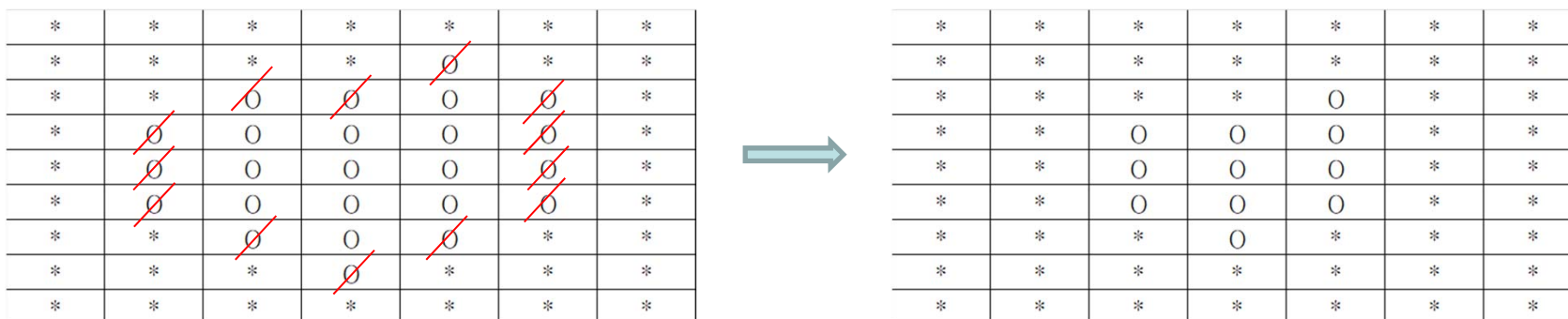


附錄十七 常用的影像修飾方法

(2) Morphology

(2-1) Erosion (去除區域外圍)

$$A[m,n] = A[m,n] \& A[m-1,n] \& A[m+1,n] \& A[m,n-1] \& A[m,n+1]$$



Erosion for a Non-binary Image


$$A[m,n] = \min \{ A[m,n], A[m-1,n], A[m+1,n], A[m,n-1], A[m,n+1] \}$$

附錄十七 常用的影像修飾方法

(2-2) Dilation (擴大區域)

$$A[m,n] = A[m,n] \vee A[m-1,n] \vee A[m+1,n] \vee A[m,n-1] \vee A[m,n+1]$$

*	*	*	*	*	*	*	*
*	*	*	*	*	*	*	*
*	*	*	*	*	*	*	*
*	*	*	*	0	*	*	*
*	*	*	0	0	0	0	*
*	*	0	0	0	0	*	*
*	*	*	*	0	0	0	*
*	*	*	*	*	*	*	*
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*	*	*	*	*	*	*	*
*	*	*	*	*	*	*	*



*	*	*	*	*	*	*	*
*	*	*	*	*	*	*	*
*	*	*	*	0	*	*	*
*	*	*	0	0	0	0	*
*	*	0	0	0	0	0	0
*	0	0	0	0	0	0	*
*	*	0	0	0	0	0	0
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Dilation for a Non-binary Image

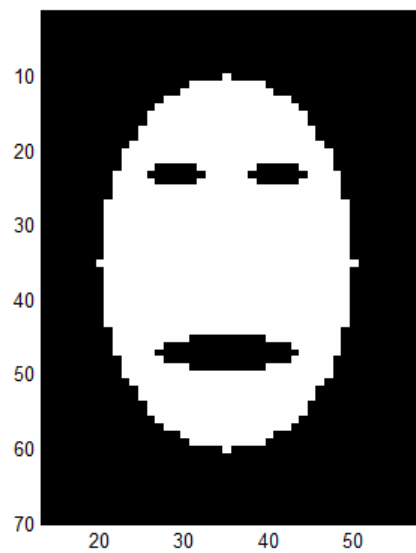
$$A[m,n] = \text{Max}\{A[m,n], A[m-1,n], A[m+1,n], A[m,n-1], A[m,n+1]\}$$

附錄十七 常用的影像修飾方法

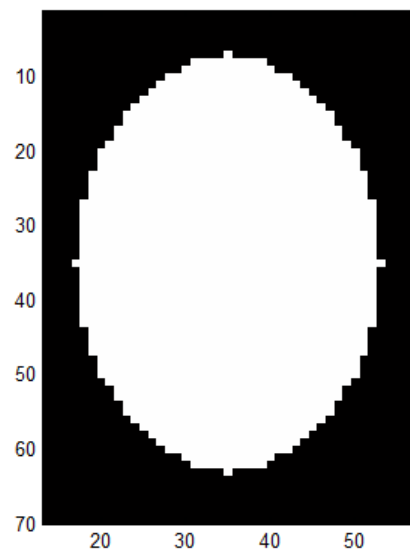
(2-3) Closing (Hole Filling)

closing = dilation k times + erosion k times

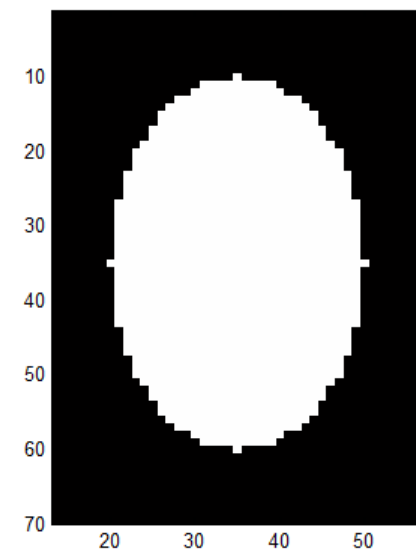
input



dilation 3 times



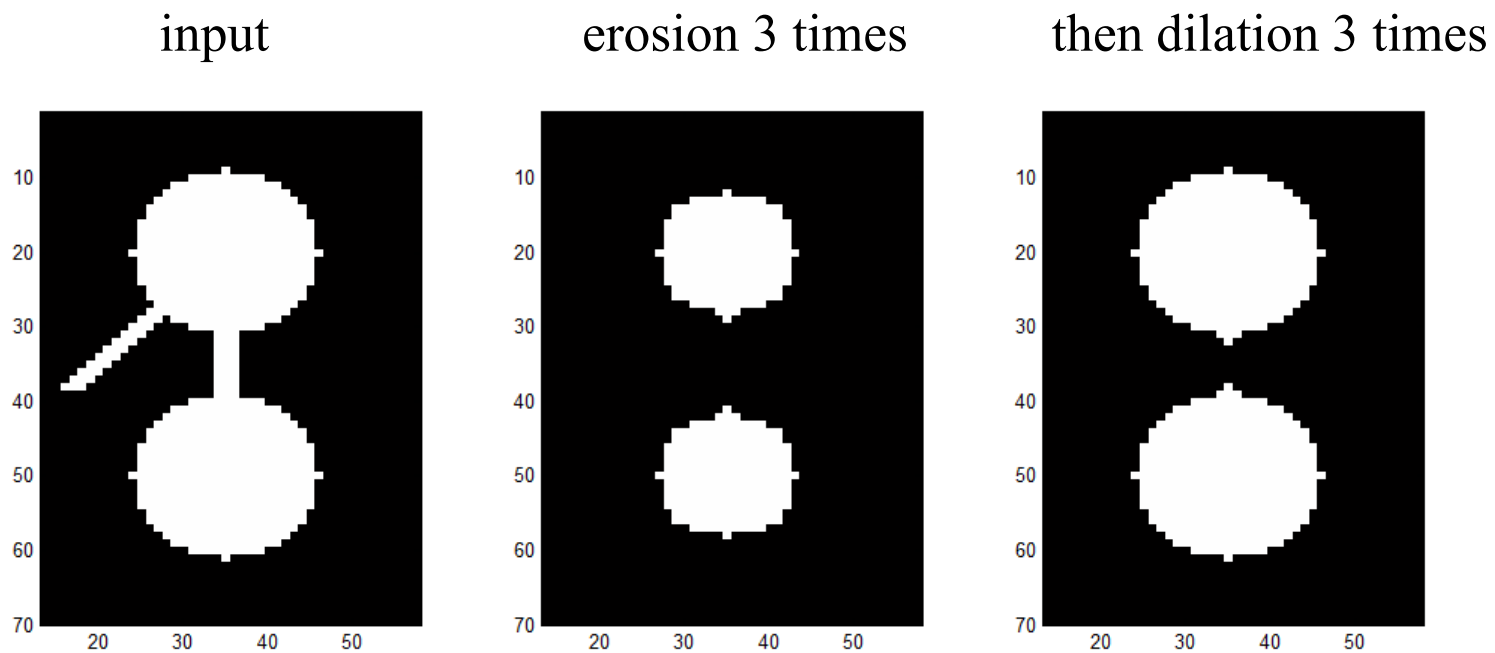
then erosion 3 times



附錄十七 常用的影像修飾方法

(2-4) Opening

opening = erosion k times + dilation k times



附錄十七 常用的影像修飾方法

(3) Edge enhancement

$$\text{input image} + \alpha |\text{edge detection output}|$$

Original image



With edge enhancement



附錄十七 常用的影像修飾方法

(4) Dehaze (除霧)



He, Kaiming, Jian Sun, and Xiaoou Tang. "Single image haze removal using dark channel prior." *IEEE Trans. Pattern Anal. Mach. Intell.*, vol. 33, pp. 2341-2353, 2011.

附錄十七 常用的影像修飾方法

Haze Model $\mathbf{I}(\mathbf{x}) = \mathbf{J}(\mathbf{x})t(\mathbf{x}) + \mathbf{A}(1 - t(\mathbf{x}))$

$\mathbf{J}(\mathbf{x})$: scene, $\mathbf{I}(\mathbf{x})$: observed image

$t(\mathbf{x})$: transmission, \mathbf{A} : intensity for the whole-haze case

$\mathbf{A}(1 - t(\mathbf{x}))$: airlight

定義 dark channel $\mathbf{J}^{dark}(\mathbf{x})$

$$J^{dark}(\mathbf{x}) = \min_{c \in \{r, g, b\}} \left(\min_{\mathbf{y} \in \Omega(\mathbf{x})} (J^c(\mathbf{y})) \right),$$

$\Omega(\mathbf{x})$: some patch (a small region)

Dark channel 為一個影像在一個小範圍區域當中，RGB 的最小值

一個正常影像的 dark channel 大多近於 0

一個受 haze 影響的影像，dark channel 常常不為 0

附錄十七 常用的影像修飾方法

Dehaze 的方法與流程

$$\mathbf{I}(\mathbf{x}) = \mathbf{J}(\mathbf{x})t(\mathbf{x}) + \mathbf{A}(1 - t(\mathbf{x}))$$

$$J^{dark}(\mathbf{x}) = \min_c \left(\min_{\mathbf{y} \in \Omega(\mathbf{x})} (J^c(\mathbf{y})) \right) = 0. \quad \min_c \left(\min_{\mathbf{y} \in \Omega(\mathbf{x})} \left(\frac{J^c(\mathbf{y})}{A^c} \right) \right) = 0$$

$$\frac{\mathbf{I}(\mathbf{x})}{A^c} = \frac{\mathbf{J}(\mathbf{x})}{A^c} t(\mathbf{x}) + 1 - t(\mathbf{x})$$

$$\tilde{t}(\mathbf{x}) = 1 - \min_c \left(\min_{\mathbf{y} \in \Omega(\mathbf{x})} \left(\frac{I^c(\mathbf{y})}{A^c} \right) \right)$$

find the transmission $t(\mathbf{x})$

\mathbf{A} : the 95% largest intensity of $\mathbf{I}(\mathbf{x})$

$$\mathbf{J}(\mathbf{x}) = \frac{\mathbf{I}(\mathbf{x})}{t(\mathbf{x})} + \mathbf{A} \left(1 - \frac{1}{t(\mathbf{x})} \right)$$

recover the original image

He, Kaiming, Jian Sun, and Xiaoou Tang. "Single image haze removal using dark channel prior." *IEEE Trans. Pattern Anal. Mach. Intell.*, vol. 33, pp. 2341-2353, 2011.

期末的勉勵

- 人生難免會有挫折，最重要的是，我們面對挫折的態度是什麼
- 長遠的願景要美麗，短期的目標要務實

祝各位同學暑假愉快！

各位同學在研究上或工作上，有任何和 digital signal processing 或 time frequency analysis 方面的問題，歡迎找我來一起討論。