

# Lecture 10 – Transform and Image Compression

This lecture will cover:

– Other Transform

- Discrete Cosine Transform (余弦变换)
- Walsh Transform (沃尔什变换)
- Discrete Wavelet Transform (小波变换)

– Image compression

# Cosine Transform (余弦变换)

**Discrete Cosine Transform (DCT):**

$$F(0) = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} f(x) \quad F(u) = \sqrt{\frac{2}{N}} \sum_{x=0}^{N-1} f(x) \cos \frac{(2x+1)u\pi}{2N}$$

**Inverse Discrete Cosine Transform (IDCT):**

$$f(x) = \frac{1}{\sqrt{N}} F(0) + \sqrt{\frac{2}{N}} \sum_{u=1}^{N-1} F(u) \cos \frac{(2x+1)u\pi}{2N}$$

**Analytic form:** 
$$\begin{cases} F(0) = 0.500f(0) + 0.500f(1) + 0.500f(2) + 0.500f(3) \\ F(1) = 0.653f(0) + 0.271f(1) - 0.271f(2) - 0.653f(3) \\ F(2) = 0.500f(0) - 0.500f(1) - 0.500f(2) + 0.500f(3) \\ F(3) = 0.271f(0) - 0.653f(1) + 0.653f(2) - 0.271f(3) \end{cases}$$

**Matrix Form:** 
$$\begin{bmatrix} F(0) \\ F(1) \\ F(2) \\ F(3) \end{bmatrix} = \begin{bmatrix} 0.500 & 0.500 & 0.500 & 0.500 \\ 0.653 & 0.271 & -0.271 & -0.653 \\ 0.500 & -0.500 & -0.500 & 0.500 \\ 0.271 & -0.653 & 0.653 & -0.271 \end{bmatrix} \begin{bmatrix} f(0) \\ f(1) \\ f(2) \\ f(3) \end{bmatrix}$$

Forward Matrix form:

$$[F(u)] = [A][f(x)]$$

Inverse Matrix form:

$$[f(x)] = [A]^T [F(u)]$$



# 2D DCT and IDCT

Forward Transform:

$$F(0,0) = \frac{1}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x,y)$$

$$F(u,0) = \frac{\sqrt{2}}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x,y) \cos \frac{(2x+1)u\pi}{2N}$$

$$F(0,v) = \frac{\sqrt{2}}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x,y) \cos \frac{(2y+1)v\pi}{2N}$$

$$F(u,v) = \frac{2}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x,y) \cos \frac{(2x+1)u\pi}{2N} \cos \frac{(2y+1)v\pi}{2N}$$

DCT

Inverse Transform:

$$f(x,y) = \frac{1}{N} F(0,0)$$

$$+ \frac{\sqrt{2}}{N} \sum_{u=1}^{N-1} F(u,0) \cos \frac{(2x+1)u\pi}{2N}$$

$$+ \frac{\sqrt{2}}{N} \sum_{v=1}^{N-1} F(0,v) \cos \frac{(2y+1)v\pi}{2N}$$

$$+ \frac{2}{N} \sum_{x=1}^{N-1} \sum_{y=1}^{N-1} F(u,v) \cos \frac{(2x+1)u\pi}{2N} \cos \frac{(2y+1)v\pi}{2N}$$

Forward Matrix form:  $[F(u,v)] = [A][f(x,y)][A]^T$

Inverse Matrix form:  $[f(x,y)] = [A]^T[F(u,v)][A]$



# Calculate DCT(IDCT) by DFT(IDFT)

**DCT:**

$$\begin{aligned}
 F(0) &= \frac{1}{\sqrt{N}} \sum_{x=0}^{2N-1} f_e(x) \\
 F(u) &= \frac{2}{N} \sum_{x=0}^{N-1} f(x) \cos \frac{(2x+1)u\pi}{2N} \\
 &= \sqrt{\frac{2}{N}} \sum_{x=0}^{2N-1} f_e(x) \cos \frac{2(x+1)u\pi}{2N} \\
 &= \sqrt{\frac{2}{N}} \operatorname{Re} \left\{ \sum_{x=0}^{2N-1} f_e(x) e^{-j \frac{(2x+1)u\pi}{2N}} \right\} \\
 &= \sqrt{\frac{2}{N}} \operatorname{Re} \left\{ e^{-j \frac{u\pi}{2N}} \sum_{x=0}^{2N-1} f_e(x) e^{-j \frac{2\pi ux}{2N}} \right\} \\
 &= \sqrt{\frac{2}{N}} \operatorname{Re} \left\{ e^{-j \frac{u\pi}{2N}} \operatorname{DFT}[f_e(x)] \right\}
 \end{aligned}$$

**Where**

$$f_e(x) = \begin{cases} f(x), & x = 0, 1, 2, \dots, N-1 \\ 0, & x = N, N+1, N+2, \dots, 2N-1 \end{cases}$$

**IDCT:**

$$\begin{aligned}
 f(x) &= \frac{1}{\sqrt{N}} F(0) + \sqrt{\frac{2}{N}} \sum_{u=1}^{N-1} F(u) \cos \frac{(2x+1)u\pi}{2N} \\
 &= \frac{1}{\sqrt{N}} F_e(0) + \sqrt{\frac{2}{N}} \sum_{u=1}^{2N-1} F_e(u) \cos \frac{(2x+1)u\pi}{2N} \\
 &= \frac{1}{\sqrt{N}} F_e(0) + \sqrt{\frac{2}{N}} \sum_{u=1}^{2N-1} \operatorname{Re} \left\{ F_e(u) e^{j \frac{(2x+1)u\pi}{2N}} \right\} \\
 &= \frac{1}{\sqrt{N}} F_e(0) + \sqrt{\frac{2}{N}} \sum_{u=1}^{2N-1} \operatorname{Re} \left\{ F_e(u) e^{j \frac{u\pi}{2N}} e^{j \frac{2\pi ux}{2N}} \right\} \\
 &= \left( \frac{1}{\sqrt{N}} - \sqrt{\frac{2}{N}} \right) F_e(0) + \sqrt{\frac{2}{N}} \operatorname{Re} \left\{ \sum_{u=0}^{2N-1} \left\{ F_e(u) e^{j \frac{u\pi}{2N}} \right\} e^{j \frac{2\pi ux}{2N}} \right\} \\
 &= \left( \frac{1}{\sqrt{N}} - \sqrt{\frac{2}{N}} \right) F_e(0) + \sqrt{\frac{2}{N}} \operatorname{Re} \left\{ \operatorname{IDFT} \left[ F_e(u) e^{j \frac{u\pi}{2N}} \right] \right\}
 \end{aligned}$$

$$F_e(u) = \begin{cases} F(u), & u = 0, 1, 2, \dots, N-1 \\ 0, & u = N, N+1, N+2, \dots, 2N-1 \end{cases}$$

# Walsh Transform

➤ Consist of  $\pm 1$  arranged in a checkerboard pattern

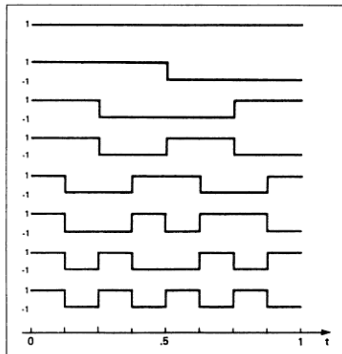
➤ Transform:

$$W(i) = \frac{1}{N} \sum_{t=0}^{N-1} f(t) \cdot \text{Wal}(i, t)$$

$$f(t) = \sum_{i=0}^{N-1} W(i) \cdot \text{Wal}(i, t)$$

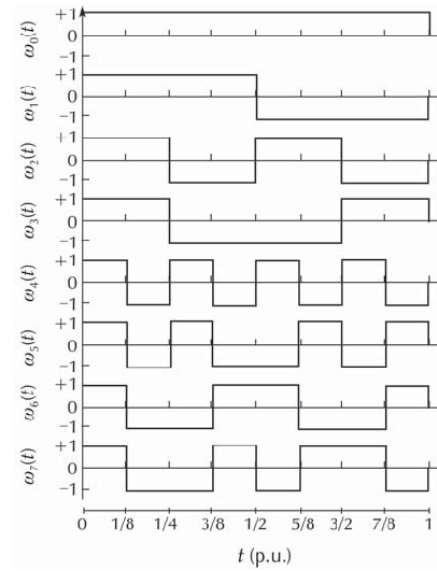
➤ Types of  $\text{Wal}(i, t)$

- Walsh Ordering (沃尔什定序)



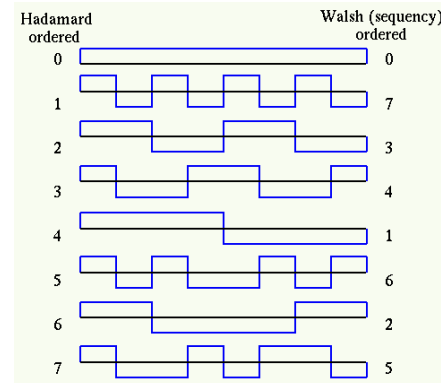
$$W_8 = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \end{pmatrix}$$

- Paley Ordering (佩利定序)



$$W_8 = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \end{pmatrix}$$

- Hadamard Matrix Ordering (哈达玛矩阵定序)



$$V_8 = \begin{pmatrix} W_4 & W_4 \\ W_4 & -W_4 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \end{pmatrix}$$

# Relationship Between Ordering

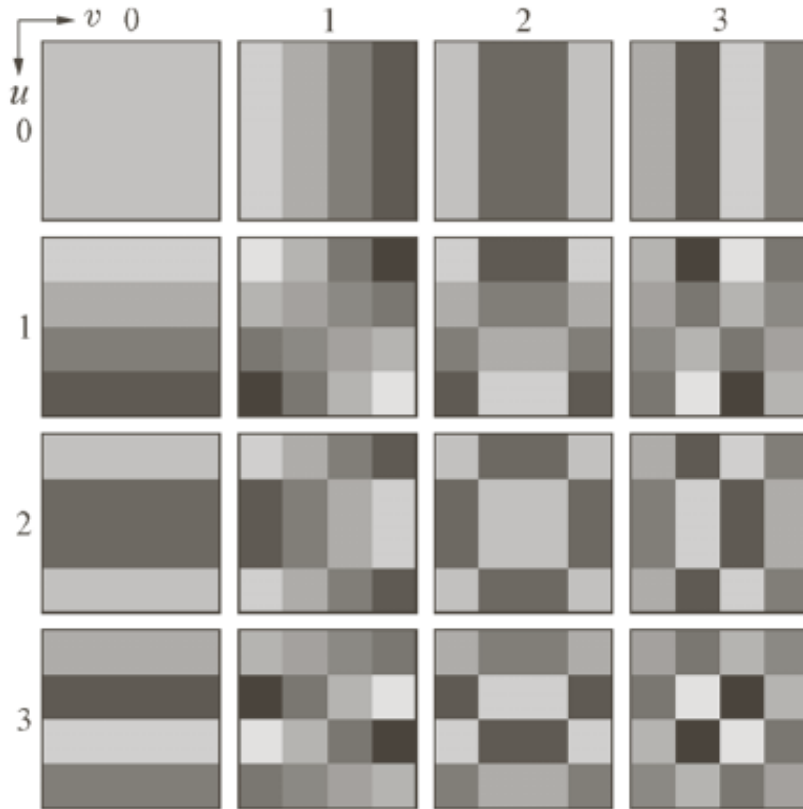
Walsh ordering ( <i>Signal processing</i> )	Paley ordering ( <i>Control Engineering</i> )	Hadamard ordering ( <i>Mathematics</i> )	$W(m,n)$
$Wal_w(0,t)$	$Wal_p(0,t)$	$Wal_H(0,t)$	[1 1 1 1 1 1 1 1]
$Wal_w(1,t)$	$Wal_p(1,t)$	$Wal_H(4,t)$	[1 1 1 1 -1 -1 -1 -1]
$Wal_w(2,t)$	$Wal_p(3,t)$	$Wal_H(6,t)$	[1 1 -1 -1 -1 -1 1 1]
$Wal_w(3,t)$	$Wal_p(2,t)$	$Wal_H(2,t)$	[1 1 -1 -1 1 1 -1 -1]
$Wal_w(4,t)$	$Wal_p(6,t)$	$Wal_H(3,t)$	[1 -1 -1 1 1 -1 -1 1]
$Wal_w(5,t)$	$Wal_p(7,t)$	$Wal_H(7,t)$	[1 -1 -1 1 -1 1 1 -1]
$Wal_w(6,t)$	$Wal_p(5,t)$	$Wal_H(5,t)$	[1 -1 1 -1 -1 1 -1 1]
$Wal_w(7,t)$	$Wal_p(4,t)$	$Wal_H(1,t)$	[1 -1 1 -1 1 -1 1 -1]

# Basic Function for DCT and WHT

$$T(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) r(x, y, u, v)$$

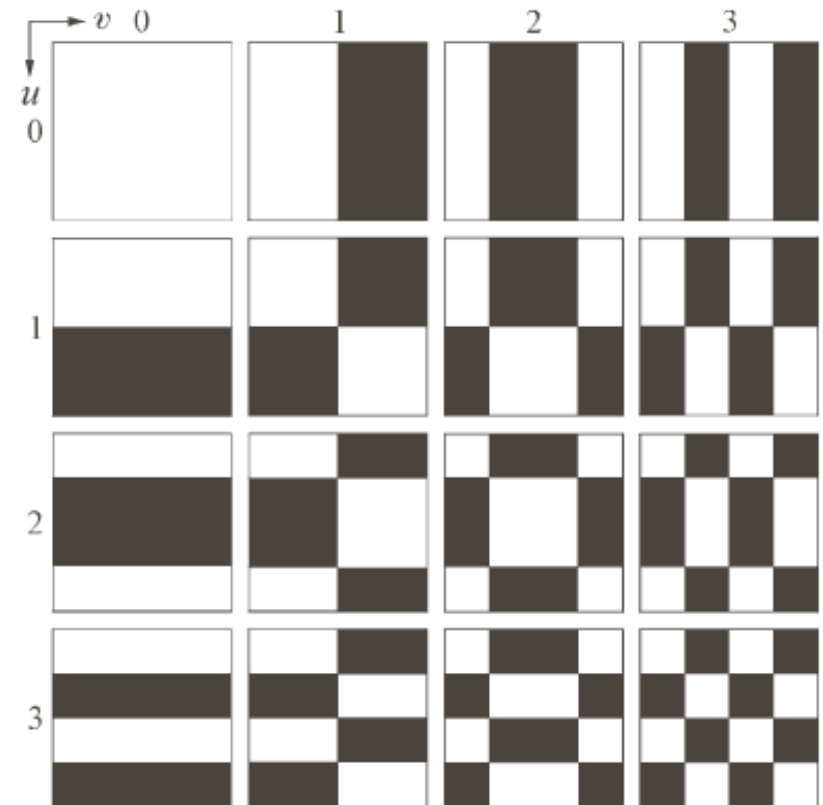
**FIGURE 8.23**

Discrete-cosine basis functions for  $n = 4$ . The origin of each block is at its top left.



**FIGURE 8.22**

Walsh-Hadamard basis functions for  $n = 4$ . The origin of each block is at its top left.



# Block Transform



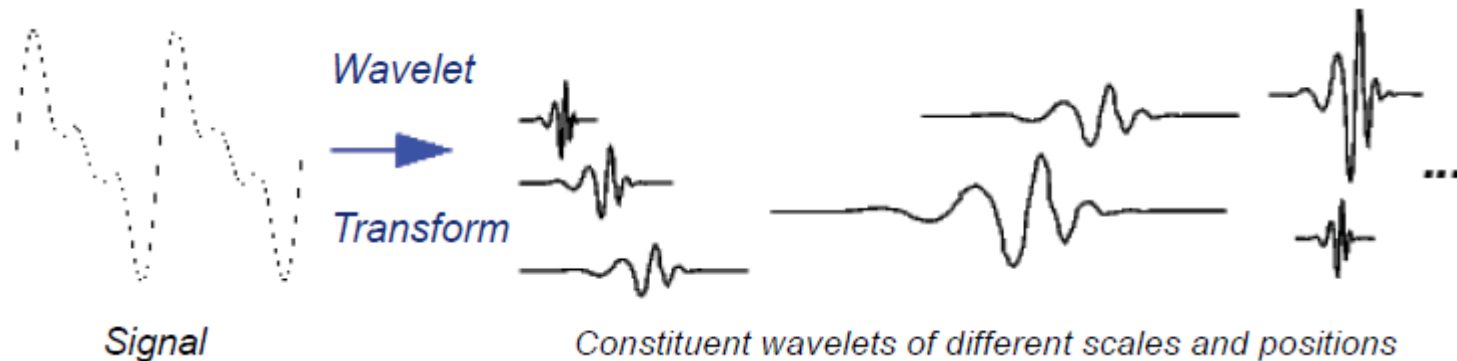
a	b	c
d	e	f

**FIGURE 8.24** Approximations of Fig. 8.9(a) using the (a) Fourier, (b) Walsh-Hadamard, and (c) cosine transforms, together with the corresponding scaled error images in (d)–(f).



# Wavelet Transform (小波变换)

- Based on small waves called Wavelets – 1) limited; 2) oscillation
- Mother wavelet (母小波) : Translation & Scaling
- Varying frequency and limited duration
- localized in both time and frequency



# Continuous Wavelet Transform (连续小波变换)

## ➤ Continuous Wavelet Transform (CWT)

$$W_{\psi}(s, \tau) = \int_{-\infty}^{\infty} f(x) \psi_{s,\tau}(x) dx$$

Where  $\psi_{s,\tau}(x) = \frac{1}{\sqrt{s}} \psi\left(\frac{x-\tau}{s}\right)$

$s$  : scale parameter (尺度参数)  $\tau$  : translation parameter (平移参数)

## ➤ Inverse Continuous Wavelet Transform (ICWT)

$$f(x) = \frac{1}{C_{\psi}} \int_0^{\infty} \int_{-\infty}^{\infty} W_{\psi}(s, \tau) \frac{\psi_{s,\tau}(x)}{s} d\tau ds$$

Where  $C_{\psi} = \int_{-\infty}^{\infty} \frac{|\Psi(\mu)|^2}{|\mu|} d\mu$ ,  $\Psi(\mu)$  is Fourier transform of  $\psi(x)$



# Discrete Wavelet Transform (离散小波变换)

## ➤ Discrete Wavelet Transform (DWT)

$$W_{\varphi}(j_0, k) = \frac{1}{\sqrt{M}} \sum_n f(n) \varphi_{j_0, k}(n)$$

$$W_{\psi}(j, k) = \frac{1}{\sqrt{M}} \sum_n f(n) \psi_{j, k}(n) \quad j \geq j_0$$

## ➤ Inverse Continuous Wavelet Transform (ICWT)

$$f(n) = \frac{1}{\sqrt{M}} \sum_k W_{\varphi}(j_0, k) \varphi_{j_0, k}(n) + \frac{1}{\sqrt{M}} \sum_{j=j_0}^{\infty} \sum_k W_{\psi}(j, k) \psi_{j, k}(n)$$

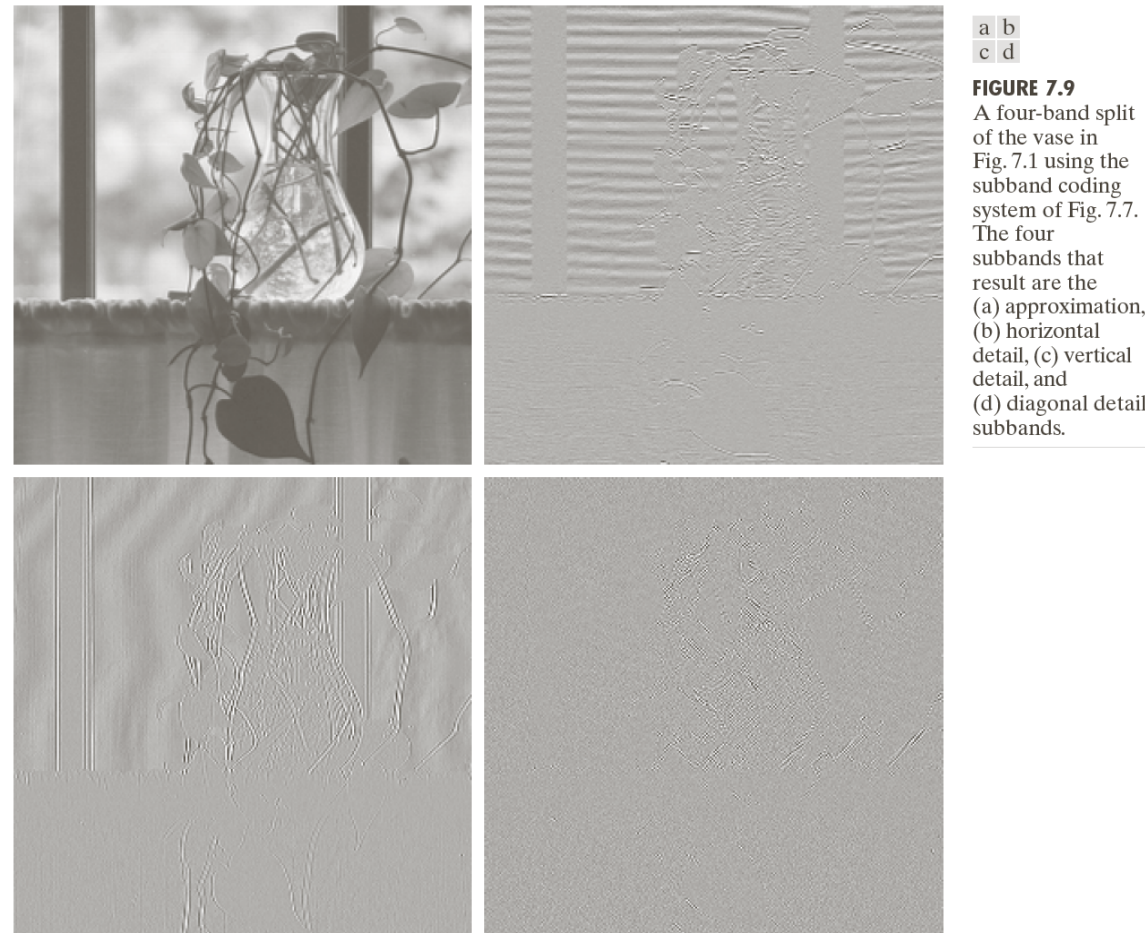
Where

$\varphi_{j_0, k}(n)$  : scaling function (尺度函数)

$\psi_{j, k}(n)$  : Wavelet (小波)

$W_{\varphi}(j_0, k)$  : Approximation coefficients (近似系数)

$W_{\psi}(j, k)$  : detail coefficients (细节系数)



# 2D DWT

**Define** 2D scale function (二维尺度函数) :

$$\varphi(x, y) = \varphi(x)\varphi(y)$$

“Directionally sensitive” wavelet (“方向敏感” 小波)

$$\psi^H(x, y) = \psi(x)\varphi(y) \quad \psi^V(x, y) = \varphi(x)\psi(y) \quad \psi^D(x, y) = \psi(x)\psi(y)$$

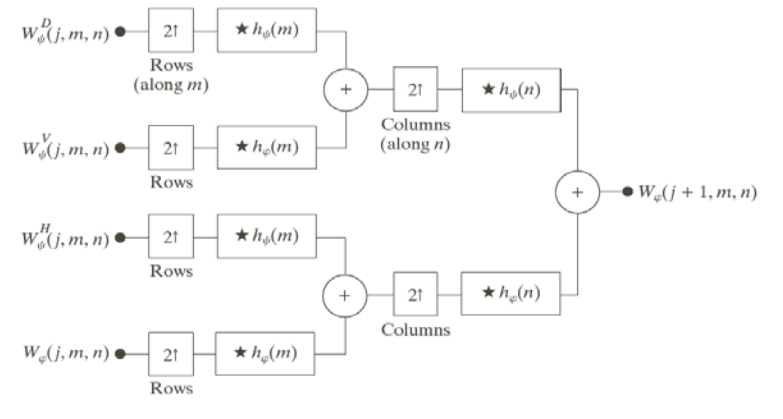
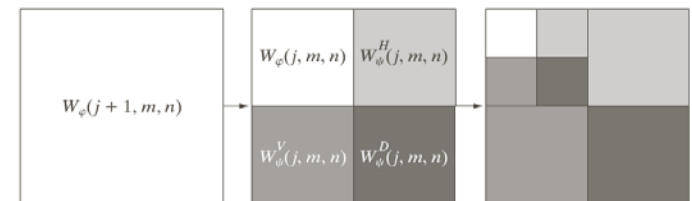
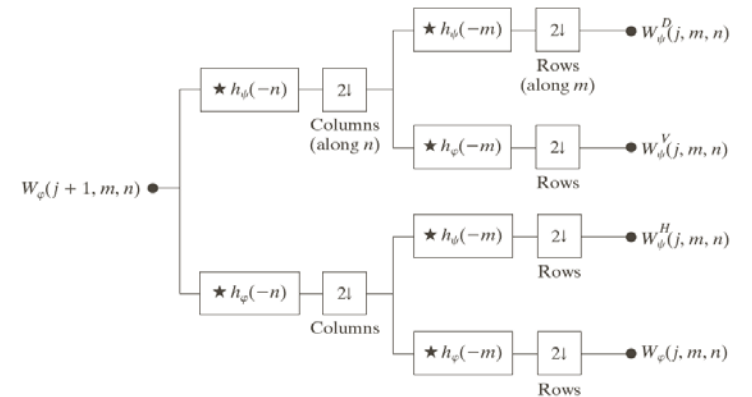
## 2D DWT

$$W_\varphi(j_0, m, n) = \frac{1}{\sqrt{MN}} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \varphi_{j_0, m, n}(x, y)$$

$$W_\psi(j, m, n) = \frac{1}{\sqrt{MN}} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \psi_{j, m, n}^i(x, y) \quad i = \{H, V, D\}$$

## 2D IDWT

$$f(x, y) = \frac{1}{\sqrt{MN}} \sum_m \sum_n W_\varphi(j_0, m, n) \varphi_{j_0, m, n}(x, y) + \frac{1}{\sqrt{MN}} \sum_{i=\{H,V,D\}} \sum_{j=j_0}^{\infty} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} W_\psi(j, m, n) \psi_{j, m, n}^i(x, y)$$



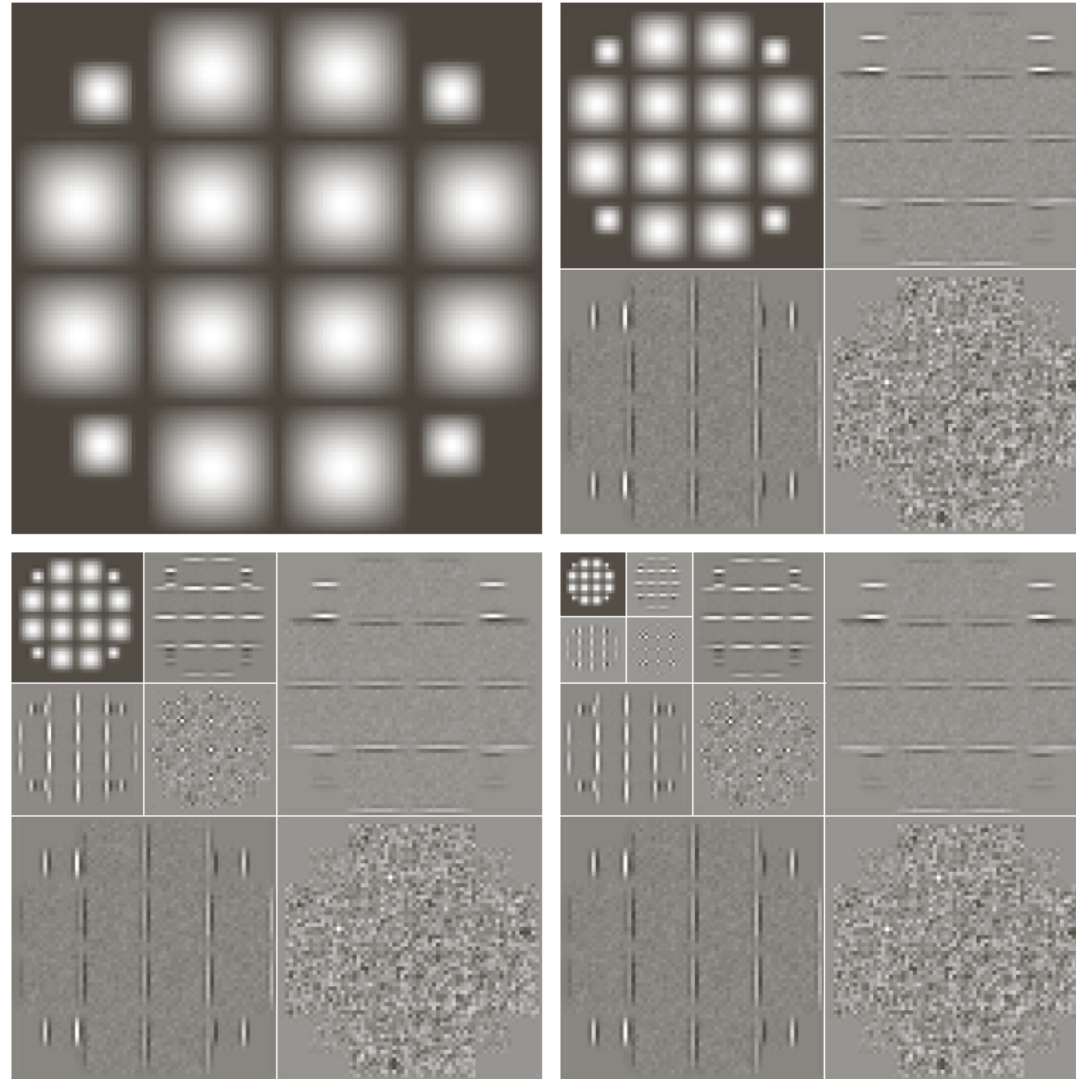
a  
b  
c

**FIGURE 7.24** The 2-D fast wavelet transform: (a) the analysis filter bank; (b) the resulting decomposition; and (c) the synthesis filter bank.

# 2D DWT

a b  
c d

**FIGURE 7.25**  
Computing a 2-D  
three-scale FWT:  
(a) the original  
image; (b) a one-  
scale FWT; (c) a  
two-scale FWT;  
and (d) a three-  
scale FWT.

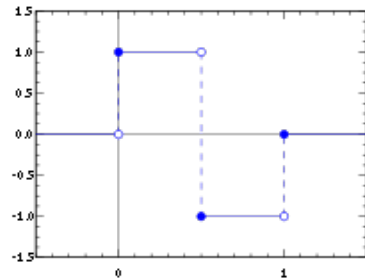


# Mother Wavelet (母小波)

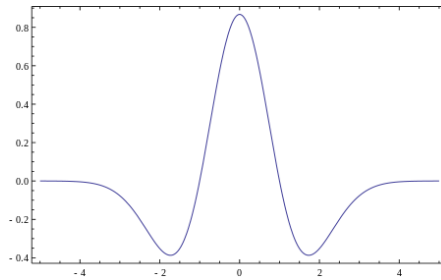
➤ Mother Wavelet should satisfy

- $\int_{-\infty}^{\infty} |\psi(t)|^2 dt = 1$
- $\int_{-\infty}^{\infty} |\psi(t)| dt < \infty$
- $\int_{-\infty}^{\infty} \psi(t) dt = 0$

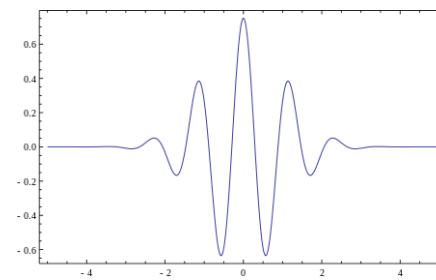
Haar



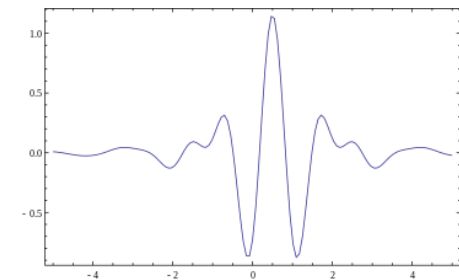
Mexican Hat



Morlet



Meyer



# Image Compression (图像压缩)

- Fundamentals (基础知识)
  - Coding Redundancy (编码冗余)
  - Spatial and Temporal Redundancy (空间和时间冗余)
  - Irrelevant Information (不相关信息)
- Measuring Image Information (信息量)
- Fidelity Criteria (保真度准则)
- Image Compression Model (图像压缩模型)
  - Source coding (信源编码)
  - Channel coding (信道编码)
- Image Formats, Containers and Compression Standards (图像格式、容器和压缩标准)





# Fundamentals of Image Compression

- Coding Redundancy (编码冗余)
- Spatial and Temporal Redundancy (空间和时间冗余)
- Irrelevant Information (不相关信息)



**FIGURE 8.1** Computer generated  $256 \times 256 \times 8$  bit images with (a) coding redundancy, (b) spatial redundancy, and (c) irrelevant information. (Each was designed to demonstrate one principal redundancy but may exhibit others as well.)



# Measuring Image Information (信息量)

➤ Information Unit:

$$I(E) = \log \frac{1}{P(E)} = -\log P(E)$$

Where  $P(E)$  is the probability of a random event  $E$ .

➤ Entropy (熵)

$$H = - \sum_{j=1}^J P(a_j) \log P(a_j)$$

Calculate from Histogram

$$\tilde{H} = - \sum_{k=0}^{L-1} p_r(r_k) \log_2 p_r(r_k)$$

# Fidelity Criteria (保真度准则)

## Objective Fidelity Criteria (客观保真度准则)

### ➤ Root Mean Square Error (均方根误差)

$$e_{\text{rms}} = \left\{ \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} [\hat{f}(x, y) - f(x, y)]^2 \right\}^{1/2}$$

Where  $f(x, y)$  is the original image, and  $\hat{f}(x, y)$  is an approximation.

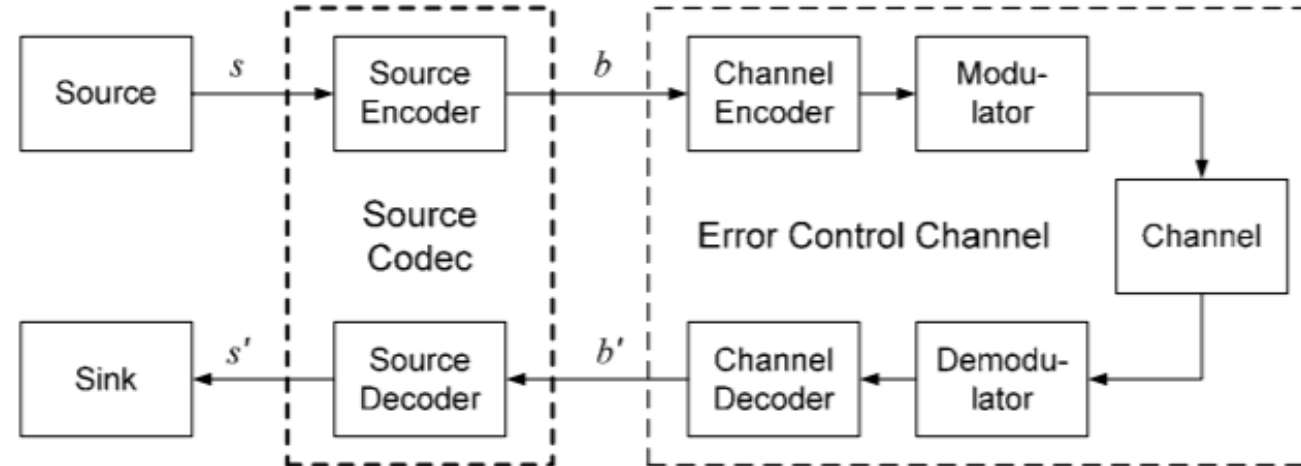
### ➤ Mean-square Signal-to-noise ratio (均方信噪比)

$$\text{SNR}_{\text{ms}} = \frac{\sum_{x=0}^{M-1} \sum_{y=0}^{N-1} [f(x, y)]^2}{\sum_{x=0}^{M-1} \sum_{y=0}^{N-1} [\hat{f}(x, y) - f(x, y)]^2}$$

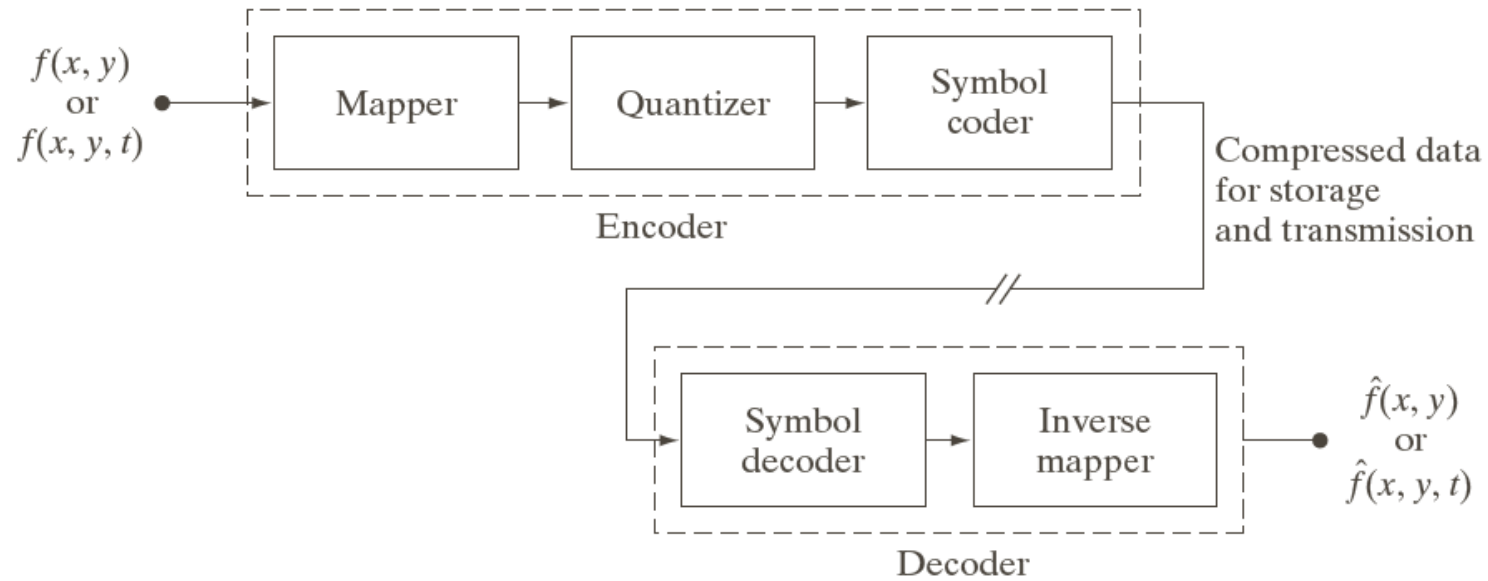
## Subjective Fidelity Criteria (主观保真度准则)

Value	Rating	Description
1	Excellent	An image of extremely high quality, as good as you could desire.
2	Fine	An image of high quality, providing enjoyable viewing. Interference is not objectionable.
3	Passable	An image of acceptable quality. Interference is not objectionable.
4	Marginal	An image of poor quality; you wish you could improve it. Interference is somewhat objectionable.
5	Inferior	A very poor image, but you could watch it. Objectionable interference is definitely present.
6	Unusable	An image so bad that you could not watch it.

# Image Compression Model (图像压缩模型)



# Source coding (信源编码)



**FIGURE 8.5**  
Functional block  
diagram of a  
general image  
compression  
system.

# Channel coding (信道编码)

## Hamming Code

For a 4-bit binary number  $b_3b_2b_1b_0$ , define the 7-bit code word as

$$h_1 = b_3 \oplus b_2 \oplus b_0$$

$$h_2 = b_3 \oplus b_1 \oplus b_0$$

$$h_4 = b_2 \oplus b_1 \oplus b_0$$

$$h_3 = b_3$$

$$h_5 = b_2$$

$$h_6 = b_1$$

$$h_7 = \oplus b_0$$

## Parity (奇偶校验)

$$c_1 = h_1 \oplus h_3 \oplus h_5 \oplus h_7$$

$$c_2 = h_2 \oplus h_3 \oplus h_6 \oplus h_7$$

$$c_4 = h_4 \oplus h_5 \oplus h_6 \oplus h_7$$

# Image Formats, Containers and Compression Standards

