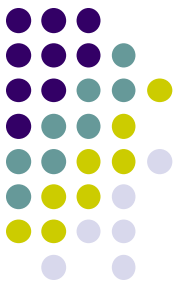


Digital Image Processing:





Introduction to Fourier Transform

- Fourier Transform pair :

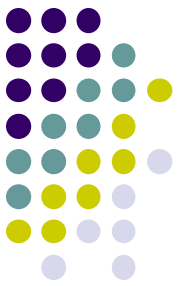
$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi ux} dx$$

$$f(x) = \int_{-\infty}^{\infty} F(u) e^{j2\pi ux} du, j = \sqrt{-1}$$

Fourier spectrum:

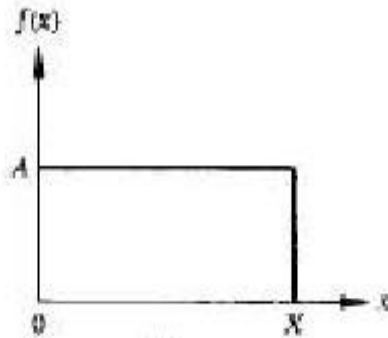
$$|F(u)| = [R^2(u) + I^2(u)]^{1/2}$$

$$e^{-j2\pi ux} = \cos(2\pi ux) - j \sin(2\pi ux)$$

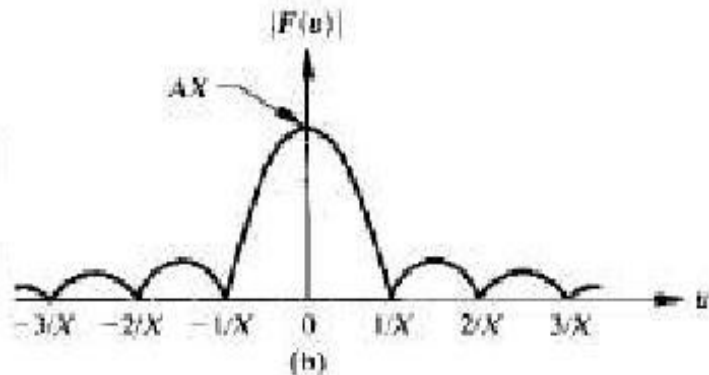


Introduction to Fourier Transform(cont.)

- Example: $f(x)=A, 0<x<X$



(a)



(b)

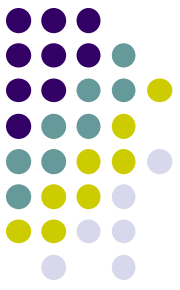
$$F(u) = \int_{-\infty}^{\infty} f(x)e^{-j2\pi ux} dx = \int_0^X Ae^{-j2\pi ux} dx$$

$$= \frac{-A}{j2\pi u} [e^{-j2\pi ux}]_0^X = \frac{-A}{j2\pi u} (e^{-j2\pi uX} - 1)$$

$$= \frac{A}{j2\pi u} (e^{j\pi uX} - e^{-j\pi uX}) e^{-j\pi uX} = \frac{A}{\pi u} \sin(\pi uX) e^{-j\pi uX}$$

$$|F(u)| = AX \left| \frac{\sin(\pi uX)}{\pi uX} \right|$$

A simple function and its Fourier spectrum.



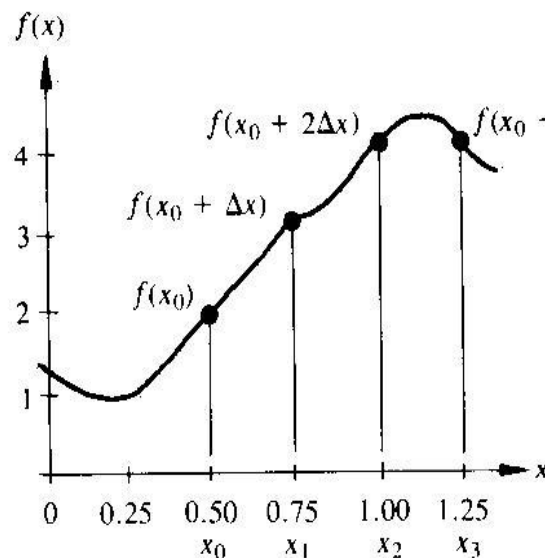
Discrete Fourier Transform

- Discrete Fourier Transform pair

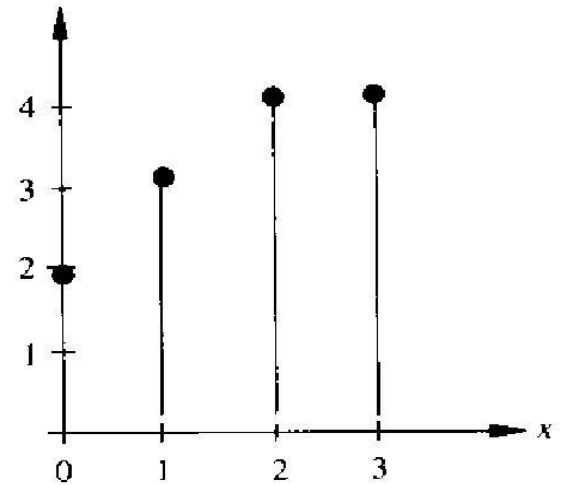
$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) e^{-j2\pi x/N}, u = 0, 1, 2, \dots, N-1$$

$$f(x) = \sum_{u=0}^{N-1} F(u) e^{j2\pi x/N}, x = 0, 1, 2, \dots, N-1$$

- Example:



$$f(x) = f(x_0 + x\Delta x)$$



2-Dimensional Discrete Fourier Transform



For an image of size $M \times N$ pixels

2-D DFT

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

u = frequency in x direction, $u = 0, \dots, M-1$
 v = frequency in y direction, $v = 0, \dots, N-1$

2-D IDFT

$$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(ux/M + vy/N)}$$

$x = 0, \dots, M-1$

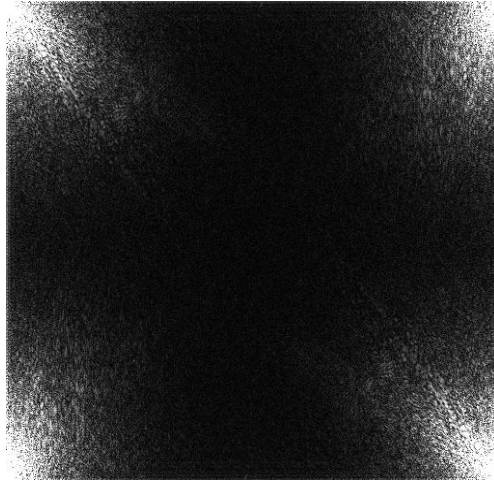
$y = 0, \dots, N-1$



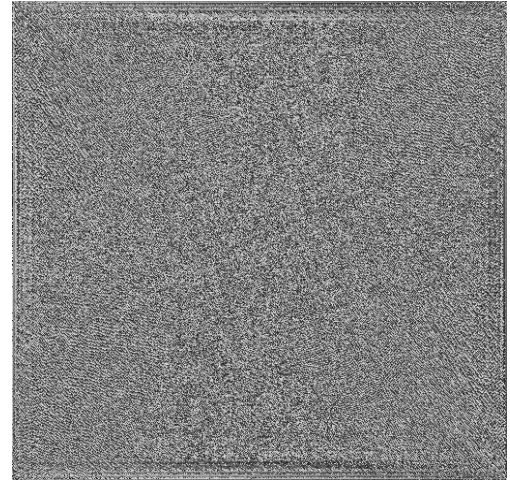
- 2-D DFT (cont.)
 - example



(a) Original Image



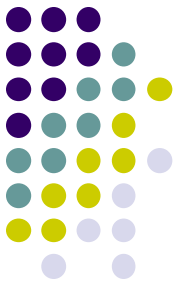
(b) Magnitude



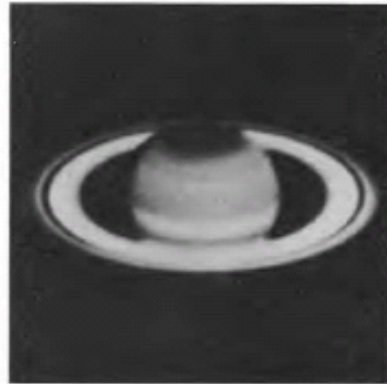
(c) Phase

2 – dim DFT of a 512×512 Lena image

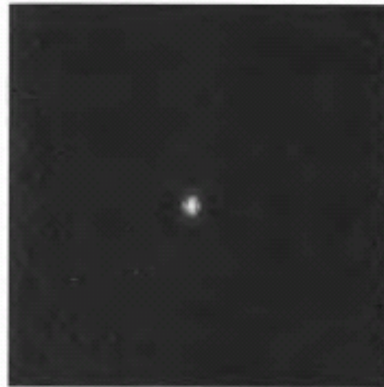
2-D FFT



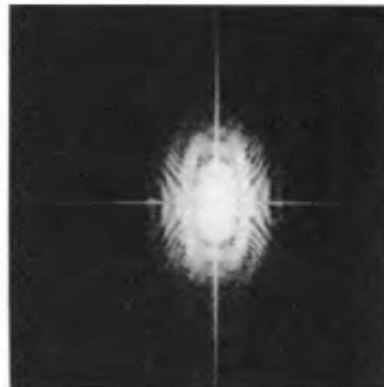
$$D(u, v) = c \log[1 + |F(u, v)|]$$



(a)



(b)



(c)

Figure 3.6 (a) A picture of the planet Saturn; (b) display of $|F(u, v)|$; (c) display of $\log[1 + |F(u, v)|]$ scaled to 8 bits (i.e., 0 to 255 gray levels).

2-D DFT (cont.)

- Properties of 2D DFT

- **Separability**

$$F(u, v) = \frac{1}{N} \sum_{x=0}^{N-1} \exp[-j2\pi ux / N] \sum_{y=0}^{N-1} f(x, y) \exp[-j2\pi vy / N], u, v = 0, \dots, N-1$$

$$f(x, y) = \frac{1}{N} \sum_{u=0}^{N-1} \exp[j2\pi ux / N] \sum_{v=0}^{N-1} F(u, v) \exp[j2\pi vy / N], m, = xy = 0, \dots, N-1$$

$$F(u, v) = \frac{1}{N} \sum_{x=0}^{N-1} F(x, v) \exp[-j2\pi ux / N]$$

$$F(x, v) = N \left[\frac{1}{N} \sum_{y=0}^{N-1} f(x, y) \exp[-j2\pi vy / N] \right]$$

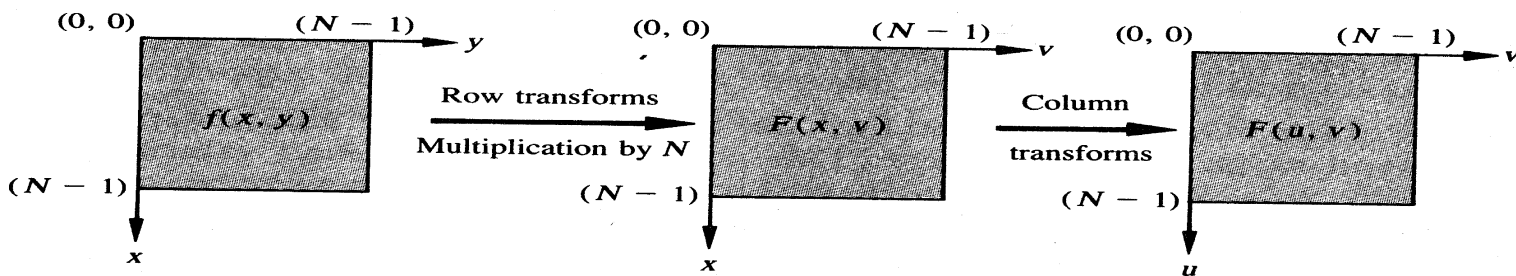
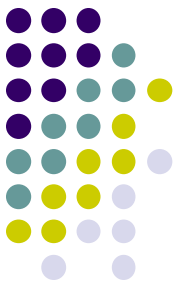
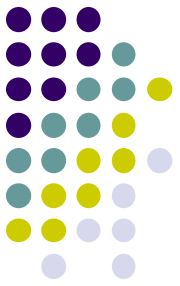


Figure 3.7 Computation of the 2-D Fourier transform as a series of 1-D transforms.

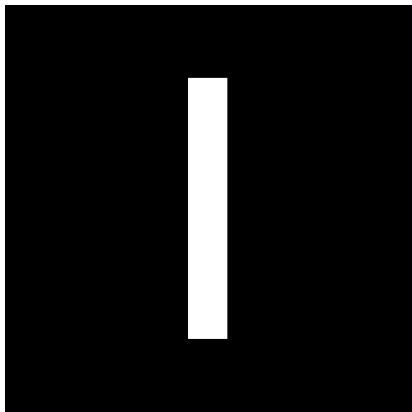




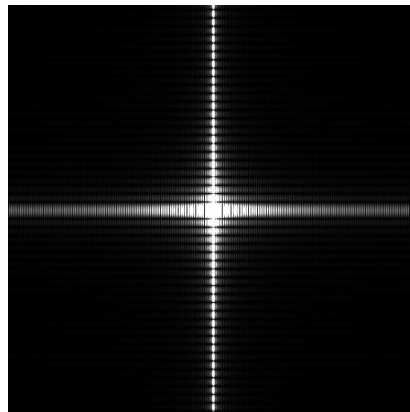
- 2-D DFT (cont.)
 - Properties of 2D DFT (cont.)
 - **Rotation**

$$m = r \cos \theta, \quad n = r \sin \theta, \quad k = \omega \cos \phi, \quad l = \omega \sin \phi$$

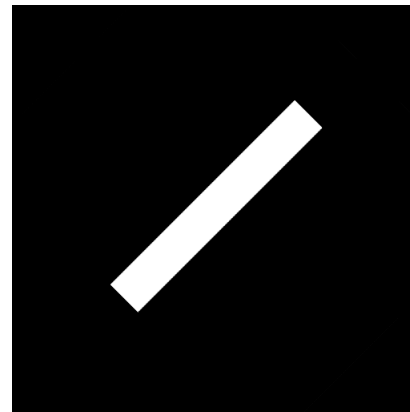
$$f(r, \theta + \theta_0) \Leftrightarrow F(\omega, \phi + \theta_0)$$



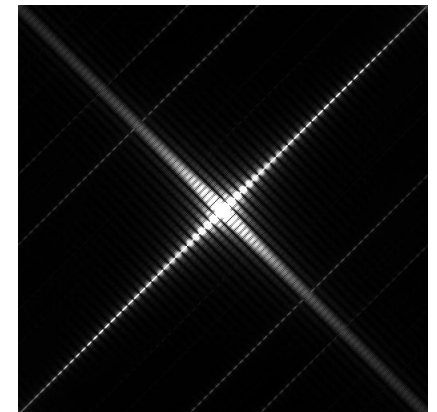
(a) a sample image



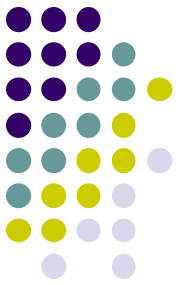
(b) its spectrum



(c) rotated image



(d) resulting spectrum



2-D DFT (cont.)

Convolution

$$f(x) * g(x) = \int_{-\infty}^{\infty} f(\alpha)g(x-\alpha)d\alpha$$

$$f(x) * g(x) = \begin{cases} x/2 & 0 \leq x \leq 1 \\ 1-x/2 & 1 \leq x \leq 2 \\ 0 & \textit{otherwise} \end{cases}$$

- Convolution Theorem

$$f(x) * g(x) \Leftrightarrow F(u)G(u)$$

$$f(x)g(x) \Leftrightarrow F(u) * G(u)$$

Other separable image transforms:



A 2-dimensional transform with a separable kernel can be computed using a series of 1-dimensional transforms as follows:

$$T(x, v) = \sum_{y=0}^{N-1} f(x, y) g_2(y, v)$$

and then,

$$T(u, v) = \sum_{x=0}^{N-1} T(x, v) g_1(x, u)$$

This procedure can be done in reverse order which produces the same result.

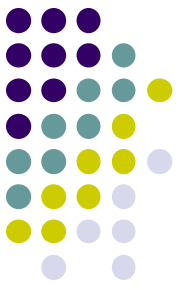
Now realize that if the kernel is separable, we can write it as follows:

$$T(u, v) = \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} g_1(x, u) f(x, y) g_1(y, v)$$

also notice that T , g_1 , and f are two-dimensional $N \times N$ matrices, therefore the above equation can be reduced to

$$\underline{T} = \underline{A} \underline{F} \underline{A}$$

[A is a $N \times N$ matrix with elements $a_{ij} = g_1(i, j)$]



Other Separable Image Transforms:

- Regular Form :
$$T(u) = \sum_{x=0}^{N-1} f(x)g(x,u)$$
$$f(x) = \sum_{u=0}^{N-1} T(u)h(x,u)$$

$g(x, u)$ is called forward transformation kernel, and $h(x, u)$ is called inverse transformation kernel.

- Walsh Transform

$$g(x, u) = \frac{1}{N} \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}, N = 2^n$$

$$n = 3, z = 6 = (110)_2$$

$$W(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}$$

$$b_0(z) = 0$$

$$b_1(z) = 1$$

$$b_2(z) = 1$$

Other Separable Image Transforms(cont.)



- Walsh Transform(cont.)
 - Example : $N=4=2^2$

$$W(0) = \frac{1}{4} \sum_{x=0}^3 f(x) \prod_{i=0}^{2-1} (-1)^{b_i(x)b_{2-1-i}(0)} \quad W(2) = \frac{1}{4} \sum_{x=0}^3 f(x) \prod_{i=0}^{2-1} (-1)^{b_i(x)b_{2-1-i}(2)}$$

$$= \frac{1}{4} [f(0) + f(1) + f(2) + f(3)] \quad = \frac{1}{4} [f(0) - f(1) + f(2) - f(3)]$$

$$W(1) = \frac{1}{4} \sum_{x=0}^3 f(x) \prod_{i=0}^{2-1} (-1)^{b_i(x)b_{2-1-i}(1)} \quad W(3) = \frac{1}{4} \sum_{x=0}^3 f(x) \prod_{i=0}^{2-1} (-1)^{b_i(x)b_{2-1-i}(3)}$$

$$= \frac{1}{4} [f(0) + f(1) - f(2) - f(3)] \quad = \frac{1}{4} [f(0) - f(1) - f(2) + f(3)]$$

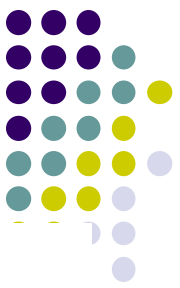
Values of the 1-D Walsh Transformation Kernel for N=8

$\frac{1}{8} g(x, u)$

$u \backslash x$	0	1	2	3	4	5	6	7
0	+	+	+	+	+	+	+	+
1	+	+	+	+	-	-	-	-
2	+	+	-	-	+	+	-	-
3	+	+	-	-	-	-	+	+
4	+	-	+	-	+	-	+	-
5	+	-	+	-	-	+	-	+
6	+	-	-	+	+	-	-	+
7	+	-	-	+	-	+	+	-

Here we calculate the matrix of Walsh coefficients

Two-Dimensional Walsh Transform



Note that the only difference between the forward and inverse transform is $1/N$ constant. Now, we go to 2-dimensional transformation we will have:

$$g(x, y, u, v) = \frac{1}{N} \prod_{i=0}^{n-1} (-1)^{[b_i(x)b_{n-1-i}(u) + b_i(y)b_{n-1-i}(v)]}$$

and

$$h(x, y, u, v) = \frac{1}{N} \prod_{i=0}^{n-1} (-1)^{[b_i(x)b_{n-1-i}(u) + b_i(y)b_{n-1-i}(v)]}$$

Two-dimensional Walsh

$$W(u, v) =$$

$$\frac{1}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) \prod_{i=0}^{n-1} (-1)^{[b_i(x)b_{n-1-i}(u) + b_i(y)b_{n-1-i}(v)]}$$

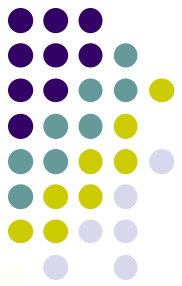
and

Inverse Two-dimensional Walsh

$$f(x, y) =$$

$$\frac{1}{N} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} W(u, v) \prod_{i=0}^{n-1} (-1)^{[b_i(x)b_{n-1-i}(u) + b_i(y)b_{n-1-i}(v)]}$$

Other Separable Image Transforms(cont.)



white:+1
black:-1

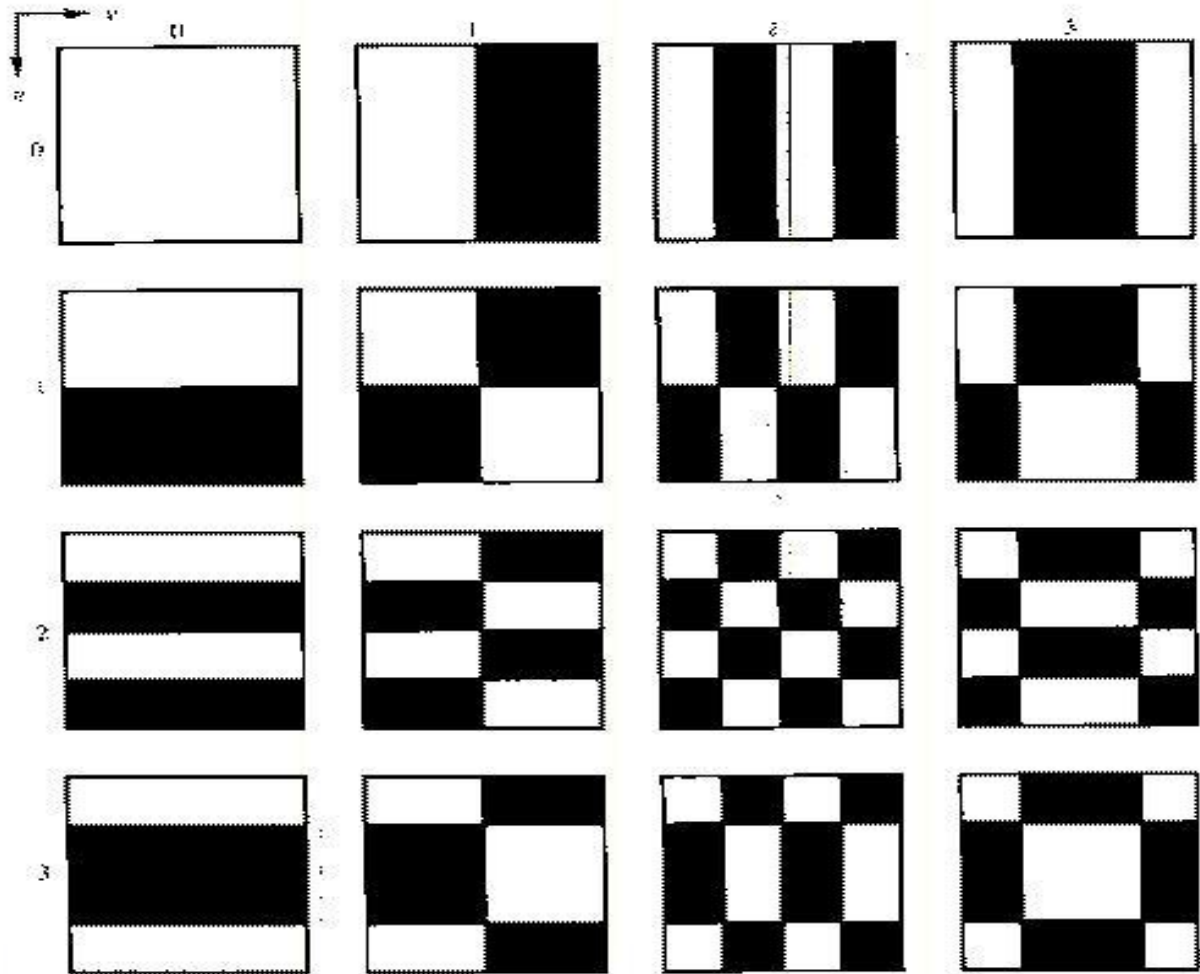


Figure 3.25 Walsh basis functions for $N = 4$. Each block consists of 4×4 elements, corresponding to x and y varying from 0 to 3. The origin of each block is at its top left. White and black denote $+1$ and -1 , respectively.

Hadamard Transform

Transformation kernel is

$$g(x, u) = \frac{1}{N} (-1)^{\sum_{i=0}^{n-1} b_i(x) b_i(u)}$$

Hence;

$$H(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) (-1)^{\sum_{i=0}^{n-1} b_i(x) b_i(u)}$$

where $N = 2^n$

The Hadamard kernel is an orthogonal matrix \Rightarrow

$$h(x, u) = (-1)^{\sum_{i=0}^{n-1} b_i(x) b_i(u)} \quad \text{and}$$

$$f(x) = \sum_{u=0}^{N-1} H(u) (-1)^{\sum_{i=0}^{n-1} b_i(x) b_i(u)}$$

Two-dimensional Hadamard transforms are

$$H(u, v) = \frac{1}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) (-1)^{\sum_{i=0}^{n-1} [b_i(x) b_i(u) + b_i(y) b_i(v)]}$$

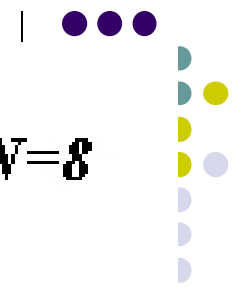


and,

$$f(x, y) = \frac{1}{N} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} H(u, v) (-1)^{\sum_{i=0}^{n-1} [b_i(x)b_i(u) + b_i(y)b_i(v)]}$$

Again notice that coefficient $1/N^2$ is divided between the pair so, one can use the transform for inverse transform and also just like other orthogonal functions kernel is separable.

$$\begin{aligned} g(x, y, u, v) &= g_1(x, u) g_1(y, v) = h_1(x, u) h_1(y, v) \\ &= \left[\frac{1}{\sqrt{N}} (-1)^{\sum_{i=0}^{n-1} b_i(x)b_i(u)} \right] \left[\frac{1}{\sqrt{N}} (-1)^{\sum_{i=0}^{n-1} b_i(y)b_i(v)} \right] \end{aligned}$$



Values of the 1-D Hadamard Transformation Kernel for $N=8$

$$\frac{1}{\sqrt{8}} g_H(x, u)$$

$u \backslash x$	0	1	2	3	4	5	6	7
0	+	+	+	+	+	+	+	+
1	+	-	+	-	+	-	+	-
2	+	+	-	-	+	+	-	-
3	+	-	-	+	+	-	-	+
4	+	+	+	+	-	-	-	-
5	+	-	+	-	-	+	-	+
6	+	+	-	-	-	-	+	+
7	+	-	-	+	-	+	+	-

The Hadamard matrix of lowest order ($N = 2$) is

$$H_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Then letting H_N represent the matrix of order N , the recursive relationship is given by the expression:

$$H_{2N} = \begin{bmatrix} H_N & H_N \\ H_N & -H_N \end{bmatrix}$$

Using matrix format $H = AFA$ suggests that $A = \frac{1}{\sqrt{N}} H_N$.

Fact! The number of sign changes along a column of the Hadamard matrix is often called the *sequency* of that column. For instance, H_8 has a sequency of

0, 7, 3, 4, 1, 6, 2, 5

- It is important to order the sequency such that it increase as u increases.
- For one-dimensional Hadamard this ordering can be accomplished through the following relation:

$$g(x, u) = \frac{1}{N} (-1)^{\sum_{i=0}^{n-1} b_i(x) p_i(u)}$$





Where,

$$p_0(u) = b_{n-1}(u)$$

$$p_1(u) = b_{n-1}(u) + b_{n-2}(u)$$

$$p_2(u) = b_{n-2}(u) + b_{n-3}(u)$$



$$p_{n-1}(u) = b_1(u) + b_0(u)$$

and recursive formula is $\Rightarrow p_i(u) = b_{n-i}(u) + b_{n-i-1}(u)$

$$\left\{ \begin{array}{l} H(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) (-1)^{\sum_{i=0}^{n-1} b_i(u) p_i(u)} \\ f(x) = \frac{1}{N} \sum_{u=0}^{N-1} H(u) (-1)^{\sum_{i=0}^{n-1} b_i(u) p_i(u)} \end{array} \right.$$

In the same manner, the 2-D transform pair can be written as :

$$\left\{ \begin{aligned} H(u, v) &= \frac{1}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) (-1)^{\sum_{i=0}^{n-1} [b_i(x) p_i(u) + b_i(y) p_i(v)]} \\ f(x, y) &= \frac{1}{N} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} H(u, v) (-1)^{\sum_{i=0}^{n-1} [b_i(x) p_i(u) + b_i(y) p_i(v)]} \end{aligned} \right.$$

How does the ordering look for one dimensional kernel?

Standard Trivial Functions for Hadamard



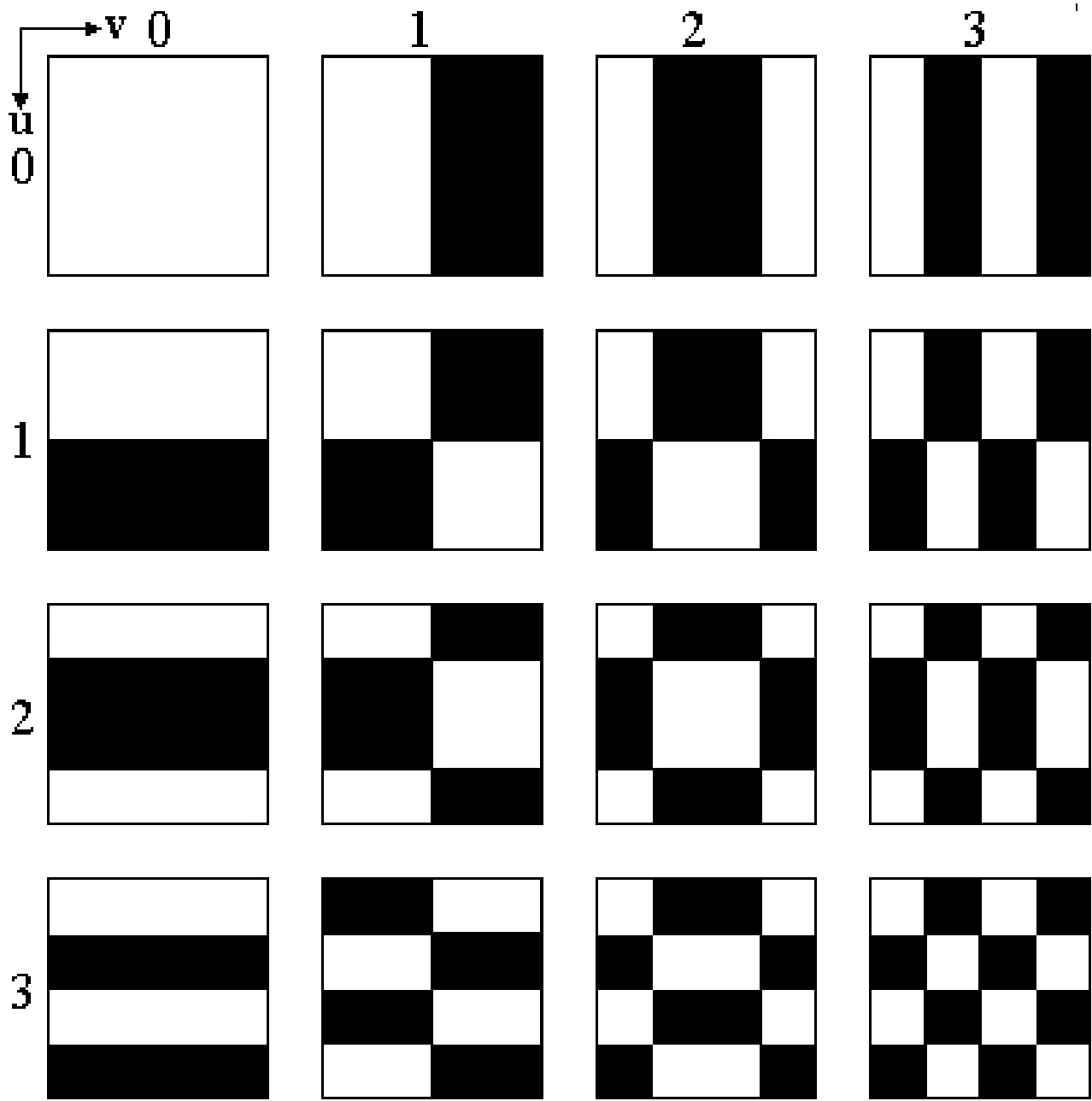
Values of the 1-D Ordered Hadamard Kernel for $N=8$

$u \backslash x$	0	1	2	3	4	5	6	7
0	+	+	+	+	+	+	+	+
1	+	+	+	+	-	-	-	-
2	+	+	-	-	-	-	+	+
3	+	+	-	-	+	+	-	+
4	+	-	-	+	+	-	-	+
5	+	-	-	+	-	+	+	-
6	+	-	+	-	-	+	-	+
7	+	-	+	-	+	-	+	-

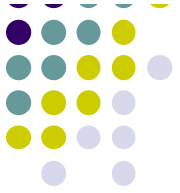
One change

two changes

Ordered Hadamard basis functions for $N = 4$. Each block consist of 4×4 elements, corresponding to x and y varying from 0 to 3.



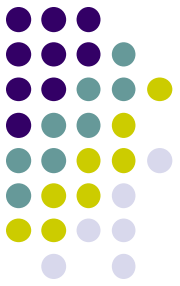
Haar Transform:



- Haar transform
 - Haar function (1910, Haar) : periodic, orthonormal, complete

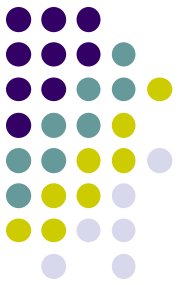
$$h_0(z) \stackrel{\Delta}{=} h_{00}(z) = \frac{1}{\sqrt{N}}, \quad z \in [0,1]$$

$$h_k(z) \stackrel{\Delta}{=} h_{pq}(z) = \frac{1}{\sqrt{N}} \begin{cases} 2^{p/2} & \frac{q-1}{2^p} \leq z < \frac{q-\frac{1}{2}}{2^p} \\ -2^{p/2} & \frac{q-\frac{1}{2}}{2^p} \leq z < \frac{q}{2^p} \\ 0 & \text{elsewhere for } z \in [0,1] \end{cases}$$



$$H_4 = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ \sqrt{2} & -\sqrt{2} & 0 & 0 \\ 0 & 0 & \sqrt{2} & -\sqrt{2} \end{bmatrix}$$

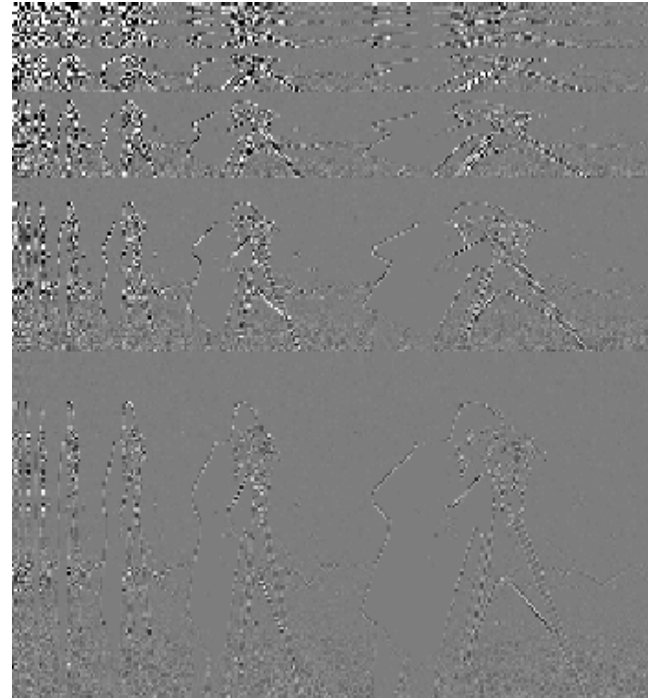
$$H_8 = \frac{1}{\sqrt{8}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ \sqrt{2} & \sqrt{2} & -\sqrt{2} & -\sqrt{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \sqrt{2} & \sqrt{2} & -\sqrt{2} & -\sqrt{2} \\ 2 & -2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & -2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & -2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2 & -2 \end{bmatrix}$$



Haar transform example:



Original Cameraman
256x256



256x256 Haar transform
of Cameraman

Slant transform:

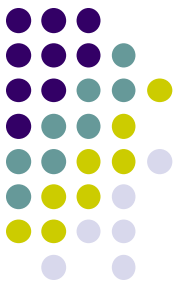


1971, Enomoto and Shibata

$$S_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$S_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 & 1 & 0 \\ a_2 & b_2 & -a_2 & b_2 \\ 0 & 1 & 0 & -1 \\ -b_2 & a_2 & b_2 & a_2 \end{bmatrix} \begin{bmatrix} S_1 & 0_2 \\ 0_2 & S_1 \end{bmatrix}$$

Slant transform contd..:



$$S_N = \frac{1}{2^{\frac{1}{2}}} \begin{bmatrix} 1 & 0 & & 1 & 0 & & & & \\ a_N & b_N & & -a_N & b_N & & & & \\ & & & & & & & & \\ 0 & & I_{(N/2)-2} & & 0 & & I_{(N/2)-2} & & \\ & & & & & & & & \\ 0 & 1 & & 0 & -1 & & & & \\ -b_N & a_N & & b_N & a_N & & & & \\ & & & & & & & & \\ 0 & & I_{(N/2)-2} & & 0 & & -I_{(N/2)-2} & & \end{bmatrix} \begin{bmatrix} S_{N/2} & 0 \\ & 0 & S_{N/2} \end{bmatrix}$$

$$a_N = \left(\frac{3N^2}{4N^2 - 1} \right)^{\frac{1}{2}}$$

$$b_N = \left(\frac{N^2 - 1}{4N^2 - 1} \right)^{\frac{1}{2}}$$

Slant transform contd..:



$$S_2 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{-1}{\sqrt{5}} & \frac{-3}{\sqrt{5}} \\ 1 & -1 & -1 & 1 \\ \frac{1}{\sqrt{5}} & \frac{-3}{\sqrt{5}} & \frac{3}{\sqrt{5}} & \frac{-1}{\sqrt{5}} \end{bmatrix}$$

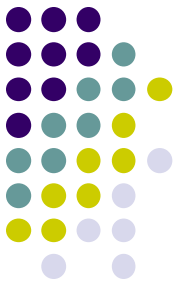
Properties

1. The Slant transform is real and orthogonal

$$S = S^*, \quad S^{-1} = S^t$$

2. fast algorithm : $O(N \log_2 N)$ for $N \times 1$ vector

3. very good energy compaction



DCT:

- Discrete Cosine Transform

$$C(u) = a(u) \sum_{x=0}^{N-1} f(x) \cos[(2x+1)u\pi / 2N]$$

$$u = 0, 1, 2, \dots, N-1$$

$$f(x) = \sum_{u=0}^{N-1} a(u) C(u) \cos[(2x+1)u\pi / 2N]$$

$$x = 0, 1, 2, \dots, N-1$$

$$a(u) = \begin{cases} \frac{1}{\sqrt{N}} & u = 0 \\ \frac{2}{\sqrt{N}} & u = 1, 2, \dots, N-1 \end{cases}$$

