



DIGITAL IMAGE PROCESSING-1

Unit 2: Lecture 23-24

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Unit 2: Image Transforms

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Last Session

- Discrete Cosine Transform (DCT) Cont..
- Discrete Sine Transform (DST)

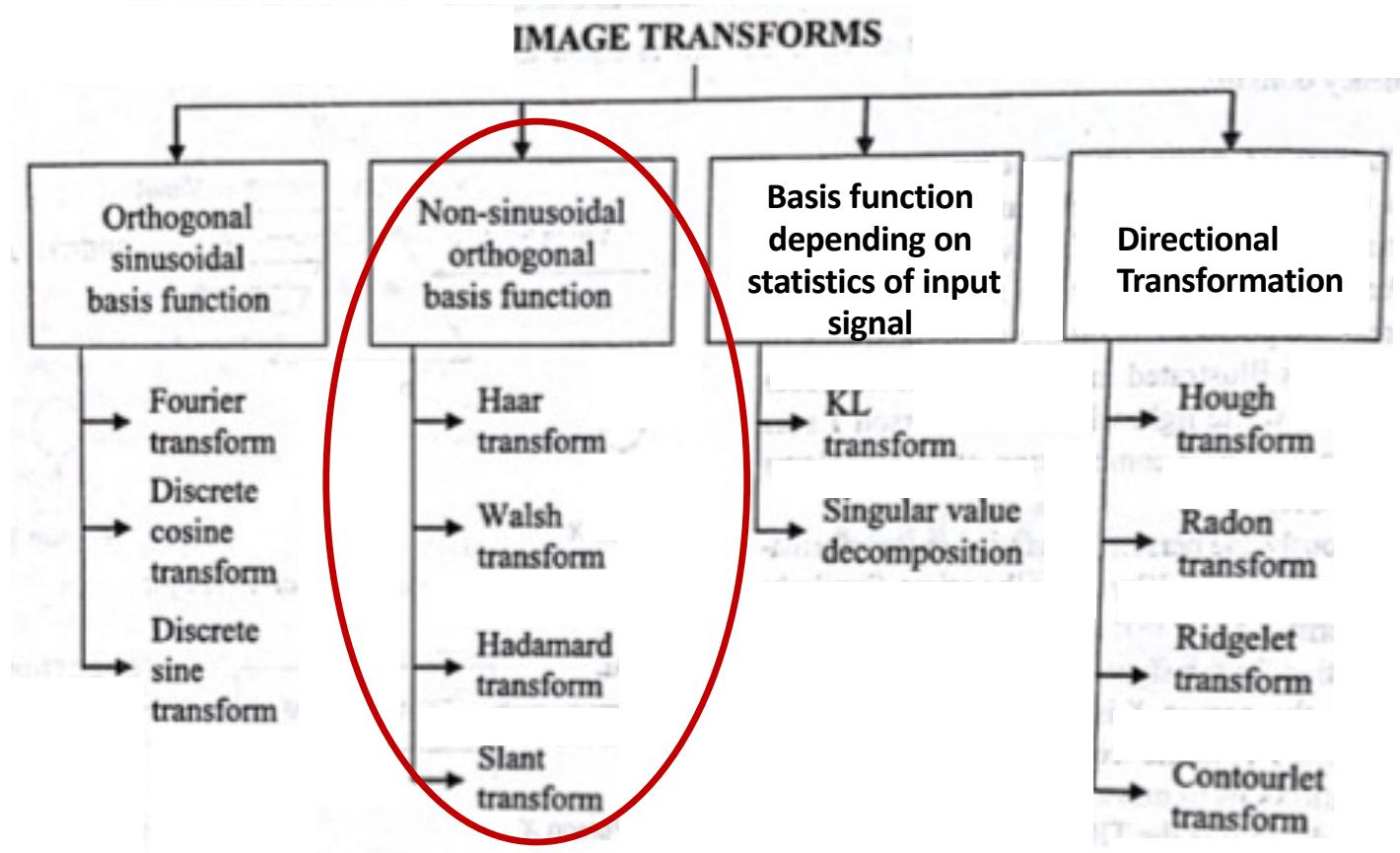
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Today's Session

- Walsh transforms
- Walsh - Hadamard transforms
- Slant Transform

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Classification of Image Transforms



Discrete Walsh Transform (DWT)

- The transforms discussed so far have been based on Sine and Cosine functions
- Transforms based on pulse like waveforms take only ± 1 (*simple & fast to compute*)
 - More appropriate for representation of waveforms which contain discontinuities (images)
- Discrete Walsh transform is based on set of harmonically related rectangular waveforms known as **Walsh functions**

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1D Discrete Walsh Transform (DWT)

- Forward transformation kernel for 1D DWT is given by:

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

where

N=No. of samples

n=No. of bits needed to represent x/u

$b_k(z)$ = kth bit in digital/binary representation of z

For ex. $b_2(8)=$ 2nd bit of 8 i.e., $\overset{\textcolor{red}{\downarrow}}{1000}$

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

1D Discrete Walsh Transform (DWT)

Inverse transformation Kernel is given by:

$$h(x, u) = \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}$$

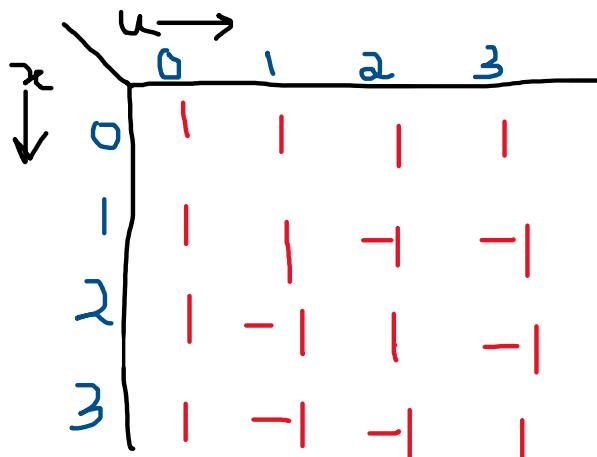
$$f(x) = \sum_{x=0}^{N-1} W(u) \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}.$$

- Also the kernel for forward and inverse transform is same except the scaling factor $1/N$
(positive aspect of Walsh transform)

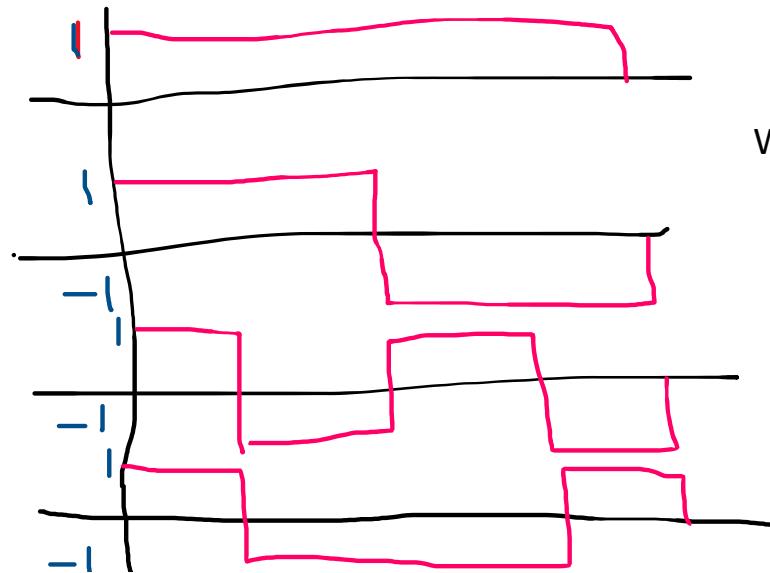
1D Discrete Walsh Transform (DWT)

Inverse Walsh transform kernel for N=4 and n=2 bits:

$$h(x, u) = \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}$$



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



Walsh Functions

1D Discrete Walsh Transform (DWT)

- Discrete Walsh transform is based on set of harmonically related rectangular waveforms known as **Walsh functions**

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

Also the kernel for forward and inverse transform is same except the scaling factor 1/N (positive aspect of Walsh transform)

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

$$h(x, u) = \prod_{i=0}^{n-1} (-1)^{b_i(x)b_{n-1-i}(u)}$$

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

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1D Discrete Walsh Transform (DWT)

- This transform also has energy compaction but not as strong as DCT
- It is separable & symmetric
 - Perform 1D DWT row-wise
 - Perform 1D DWT column-wise of the resultant

2D Discrete Walsh Transform (DWT)

- Forward transformation kernel for 2D DWT is given by:
- $g(x, y, u, v) = \frac{1}{N} \prod_{i=0}^{n-1} (-1)^{\{b_i(x)b_{n-1-i}(u)+b_i(y)b_{n-1-i}(v)\}}$
- $\therefore W(u, v) = \frac{1}{N} \sum_{y=0}^{N-1} \sum_{x=0}^{N-1} f(x, y) \prod_{i=0}^{n-1} (-1)^{\{b_i(x)b_{n-1-i}(u)+b_i(y)b_{n-1-i}(v)\}}$
- Inverse transformation kernel is:

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

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

Discrete Hadamard Transform (DHT)

- The Hadamard transform is also called as Walsh-Hadamard transform (WHT) and is basically same a Walsh transform **but recursive (???)**
- Forward transform is given by

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

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

- Inverse transform is given by

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

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

2D Discrete Hadamard Transform (DHT)

- Forward transformation kernel for 2D DWT is given by:

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

- Inverse transformation kernel is:

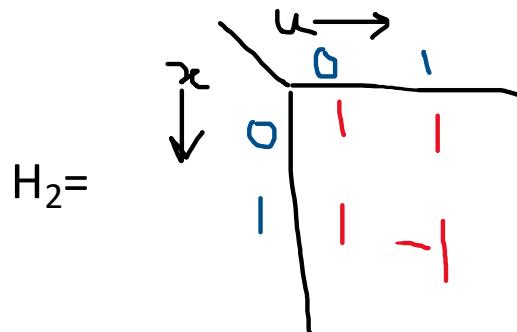
$$h(x, y, u, v) = (-1)^{\sum_{i=0}^{n-1} \{b_i(x)b_i(u)+b_i(y)b_i(v)\}}.$$

g and h are separable and symmetric and can be implemented row-wise and column-wise

1D Discrete Hadamard Transform (DHT)

Inverse Hadamard transform kernel for N=2 and n=1 bit:

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



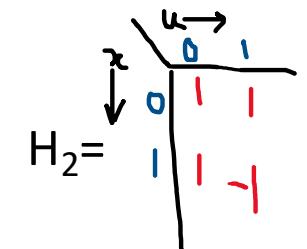
$$\begin{aligned} h(0,0) &= 1 \\ h(0,1) &= 1 \\ h(1,0) &= 1 \\ h(1,1) &= -1 \end{aligned}$$

1D Discrete Hadamard Transform (DHT)

Inverse Hadamard transform kernel for N=4 and n=2 bit:

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

$$H_4 = \begin{matrix} & \begin{matrix} u \rightarrow & 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} x \downarrow & \end{matrix} & \begin{matrix} 0 & | & | & | & | \\ | & - & | & | & | \\ 2 & | & - & | & | \\ 3 & | & - & - & | \end{matrix} \end{matrix} = \begin{bmatrix} H_2 & H_2 \\ H_2 & H_2 \end{bmatrix}$$



Hence H_4 can be computed from H_2 (hence recursive) or H_{2N} can be computed from H_N

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

Kronecker Product

- Kronecker product of two matrices A and B is defined as

$$A \otimes B = \{a(m, n)B\} = \begin{bmatrix} a(1,1)B & a(1,2)B & \dots & a(1, n)B \\ a(2,1)B & a(2,2)B & \dots & a(2, n)B \\ \vdots & \vdots & & \vdots \\ a(m, 1)B & a(m, 2)B & \dots & a(m, n)B \end{bmatrix}$$

1D Discrete Hadamard Transform (DHT)

- Basis functions of Hadamard transform are also non sinusoidal.
- Just like Walsh transform, Hadamard transform matrix also has only ± 1 in their basis functions.
- A $N \times N$ Hadamard transform matrix is generated by iterative rule (where $N=2^n$ or $n = \log_2 N$)

$$H_n = H_1 \otimes H_{n-1}$$

Where \otimes is Kronecker Product

$$H_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \quad n = \log_2 N$$

$$H_n = H_1 \otimes H_{n-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \otimes H_{n-1}$$

$$H_n = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{n-1} & H_{n-1} \\ H_{n-1} & -H_{n-1} \end{bmatrix}$$

Kronecker Product

Example: Prove that $A \otimes B \neq B \otimes A$

$$A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} B = \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix}$$

$$A \otimes B = \begin{bmatrix} 1 \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix} & 1 \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix} \\ 1 \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix} & -1 \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 4 & 3 & 4 & 3 \\ 2 & 1 & 2 & 1 \\ 4 & 3 & -4 & -3 \\ 2 & 1 & -2 & -1 \end{bmatrix}$$

$$B \otimes A = \begin{bmatrix} 4 \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} & 3 \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \\ 2 \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} & 1 \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 4 & 4 & 3 & 3 \\ 4 & -4 & 3 & -3 \\ 2 & 2 & -1 & -1 \\ 2 & -2 & -1 & 1 \end{bmatrix}$$

Therefore $A \otimes B \neq B \otimes A$

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Hadamard Matrix

Ex. Calculate 4×4 Hadamard matrix

Sol:

$$\text{Given } N = 4 \Rightarrow n = \log_2 N = \log_2 4 \Rightarrow n = 2$$

$$H_2 = H_1 \otimes H_1$$

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

$$H_2 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

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Hadamard Matrix

Ex. Calculate 8×8 Hadamard matrix

Solution :

$$\text{Given } N = 8 \Rightarrow n = \log_2 N = \log_2 8 \Rightarrow n = 3$$

$$H_3 = H_1 \otimes H_2$$

$$H_2 = \frac{1}{\sqrt{2}} \cdot \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$H_2 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$H_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \otimes \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$H_3 = \frac{1}{2\sqrt{2}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & -1 & -1 & 1 & 1 & 1 & 1 & -1 \end{bmatrix}$$

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Hadamard Transform

- 1D forward Hadamard Transform

$$V = HU$$

- 1D inverse Hadamard Transform

$$U = H^{*T}V$$

- 2D forward Hadamard Transform

$$V = HUH^T$$

- 2D inverse Hadamard Transform

$$U = H^{*T}VH^*$$

Properties of H

Property 1: H is real

$$H^* = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} = H \Rightarrow H = H^*$$

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

Property 2: H is symmetric

$$H^T = H \Rightarrow H \text{ is symmetric}$$

Property 3: H is unitary

$$HH^{*T} = I$$

Properties of H

Property 3: H is unitary

$$H^* = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

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

$$HH^* = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} H & 0 & 0 & 0 \\ 0 & H & 0 & 0 \\ 0 & 0 & H & 0 \\ 0 & 0 & 0 & H \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = I$$

$\therefore HH^* = I \Rightarrow H$ is unitary matrix.

Properties of H

Property 4: Sequence of H is not in order (no. of sign changes in each row)

$$H_3 = \frac{1}{2\sqrt{2}} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \end{bmatrix} \begin{array}{c} 0 \\ \rightarrow 7 \\ \rightarrow 3 \\ \rightarrow 4 \\ \rightarrow 1 \\ \rightarrow 6 \\ \rightarrow 2 \\ \rightarrow 5 \end{array}$$

Hadamard Transform

Note : From above properties matrix notation of Hadamard transform reduces to following form

$$H = H^* = H^T = H^{*T}$$

- 1D forward Hadamard Transform

$$V = HU$$

- 1D inverse Hadamard Transform

$$U = H^{*T}V = HV$$

- 2D forward Hadamard Transform

$$V = HUH^T = HUH$$

- 2D inverse Hadamard Transform

$$U = H^{*T}VH^* = HVH$$

Hadamard Transform

- It is a fast transform
 - 1D transformation can be implemented in $O(N \log_2 N)$ additions and subtractions
- Since it contains only ± 1 values, no multiplications are required.
 - Useful in situations where minimizing amount of computation is very important
- Energy compaction of WHT is more than Walsh transform but less than DCT
 - Hence if sufficient computation power is available DCT is the choice

Example on Hadamard Transform

1. Find the forward Hadamard transform of the given matrix U.

$$U = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

1D Hadamard Transform is given by

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

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Example on Hadamard Transform

2. Find the forward Hadamard transform of the given matrix U.

$$U = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

2D Hadamard Transform is given by .

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

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Example on Hadamard Transform

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

$$= \frac{1}{\sqrt{4}} \begin{bmatrix} 16 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Examples on Hadamard Transform

Ex 3. Find WHT for $x(n) = [2,3,4,5]$

$$H = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \quad V = HU$$

$$V = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \\ 4 \\ 5 \end{bmatrix} = \begin{bmatrix} 7 \\ -1 \\ -2 \\ 5 \end{bmatrix}$$

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Slant Transform

- Many monochrome images have large areas of uniform intensity and areas of linearly increasing or decreasing brightness.
- With the exception of the discrete sine transform, all of the transforms that we have studied so far include a basis vector for representing efficiently constant gray level areas
- **None has a basis function that is targeted specifically at the representation of linearly increasing or decreasing intensity values.**
- **Slant transform includes such a basis function.**

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Slant Transform

- Slant transform is widely used in image compression
- It is orthogonal and hence very fast
- Its kernel can be generated recursively like Hadamard transform
- Its basis function has sawtooth waveform or slant basis vectors
- A slant basis vector that is monotonically decreasing in consistent steps from maximum to minimum has the sequency property and a fast computational algorithm

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Slant Transform

- Transformation matrix of the slant transform of order $N \times N$ where $N = 2^n$ is generated recursively using

$$\mathbf{A}_{\text{Sl}} = \frac{1}{\sqrt{N}} \mathbf{S}_N$$

where *slant* matrix

$$\mathbf{S}_N = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ a_N & b_N & -a_N & b_N & 0 \\ 0 & \mathbf{I}_{(N/2)-2} & 0 & \mathbf{0} & \mathbf{I}_{(N/2)-2} \\ 0 & 1 & 0 & -1 & 0 \\ -b_N & a_N & b_N & a_N & -\mathbf{I}_{(N/2)-2} \\ 0 & \mathbf{0} & \mathbf{I}_{(N/2)-2} & \mathbf{0} & -\mathbf{I}_{(N/2)-2} \end{bmatrix}$$

Here \mathbf{I}_N , is the identity matrix of order $N \times N$

\mathbf{I}_1 is a 1×1 identity matrix [1] and \mathbf{I}_0 is the empty matrix of size 0×0 .

$$\mathbf{S}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

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Slant Transform

$$\mathbf{S}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

and coefficients a_N and b_N are

$$a_N = \left[\frac{3N^2}{4(N^2 - 1)} \right]^{1/2} \quad \text{and} \quad b_N = \left[\frac{N^2 - 4}{4(N^2 - 1)} \right]^{1/2}$$

for $N > 1$. When $N \geq 8$, matrix \mathbf{S}_N is not sequency ordered, but can be made so

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Slant Transform

Example: For N=4 determine the slant matrix A_{SI}

using

$$A_{SI} = \frac{1}{\sqrt{N}} S_N \quad \text{For } N=4 \quad A_{SI} = \frac{1}{\sqrt{4}} S_4$$

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

Using this determine S_4

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Slant Transform

$$\mathbf{S}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \quad a_N = \left[\frac{3N^2}{4(N^2 - 1)} \right]^{1/2} \quad b_N = \left[\frac{N^2 - 4}{4(N^2 - 1)} \right]^{1/2}$$

Step 1: Determine a_4 and b_4

$$a_4 = \left[\frac{3N^2}{4(N^2 - 1)} \right]^{1/2} = \left[\frac{3(4^2)}{4(4^2 - 1)} \right]^{1/2} = \frac{2}{\sqrt{5}}$$

$$b_4 = \left[\frac{N^2 - 4}{4(N^2 - 1)} \right]^{1/2} = \left[\frac{4^2 - 4}{4(4^2 - 1)} \right]^{1/2} = \frac{1}{\sqrt{5}}$$

Step 2: Determine S_4

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Slant Transform

Step 2: Determine S_4

$$a_4 = \frac{2}{\sqrt{5}} \quad b_4 = \frac{1}{\sqrt{5}} \quad A_{SI} = \frac{1}{\sqrt{4}} S_4$$

$$S_4 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ 0 & 1 & 0 & -1 \\ -\frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & -1 \end{bmatrix}$$

$$A_{sl} = \frac{1}{2} S_4 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} & -\frac{1}{\sqrt{5}} & -\frac{3}{\sqrt{5}} \\ 1 & -1 & -1 & 1 \\ \frac{1}{\sqrt{5}} & -\frac{3}{\sqrt{5}} & \frac{3}{\sqrt{5}} & -\frac{1}{\sqrt{5}} \end{bmatrix}$$

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

I_0 = Empty Matrix of size 0

$$S_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Since N= 4, the basis vectors of A_{SI} (and the rows of slant matrix of S_4) are sequency ordered

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Slant Transform

1D Slant transform:-

Forward Slant Transform

$$V = SV.$$

Inverse Slant Transform.

$$\begin{aligned} V &= S^{*T}V \quad (\because S^* = S) \\ &= S^TV \end{aligned}$$

2D Slant Transform:-

Forward Slant Transform.

$$V = SUST$$

Inverse Slant Transform.

$$\begin{aligned} V &= S^{*T}VS^* \quad (\because S^* = S) \\ &= S^TVS \end{aligned}$$

Slant Transform

Example: Determine the slant transform the slant transform of function $f = [2 \ 3 \ 4 \ 5]^T$

$$t_{sl} = A_{sl} f = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ \frac{3}{\sqrt{5}} & \frac{1}{\sqrt{5}} & -\frac{1}{\sqrt{5}} & -\frac{3}{\sqrt{5}} \\ 1 & -1 & -1 & 1 \\ \frac{1}{\sqrt{5}} & -\frac{3}{\sqrt{5}} & \frac{3}{\sqrt{5}} & -\frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 4 \\ 5 \end{bmatrix} \quad \mathbf{t}_{sl} = \mathbf{A}_{sl} \mathbf{f} = [7 \ -2.24 \ 0 \ 0]^T$$

- The transform contains only two nonzero terms, while the Walsh-Hadamard transform in the previous example had three nonzero terms.
- The slant transform represents f more efficiently because f is a linearly increasing function—that is, f is highly correlated with the slant basis vector of sequency one.
- Thus, there are fewer terms in a linear expansion using slant basis functions as opposed to Walsh basis functions.

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Slant Transform

- The slant transform reproduces the linear variations very well
- Its performance at edges is not as optimal as KLT or DCT
- Because of slant nature of the lower order coefficients its effect is to smear the edges

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Next Session

- Haar Transform
- KL Transform
- SVD



THANK YOU

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