

DSIPModule 3 :- Image Transformations.Discrete Fourier Transform.

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi k n}{N}}$$

$X(k) \rightarrow$  Transformed Image.

$x(n) \rightarrow$  Input Image.

$N \rightarrow$  Matrix/Samples.

Inverse DFT (IDFT).

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j \frac{2\pi k n}{N}} \quad (0 \leq n \leq N-1)$$

DFT Kernel for  $N=4$ .  $\Rightarrow$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix}$$

Properties :-

- 1] Linearity
- 2] Periodicity
- 3] Convolution
- 4] Multiplicative
- 5] Time Reversal
- 6] Time Shift
- 7] Conjugation.

## \* Fast Fourier Transform

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi k n}{N}}$$

Twiddle Factors :-

$$W_2^0 = 1$$

$$W_4^0 = 1$$

$$W_4^1 = \frac{1}{2} - j$$

$$W_8^0 = 1$$

$$W_8^1 = \frac{1}{\sqrt{2}} - j \frac{1}{\sqrt{2}}$$

$$W_8^2 = -j$$

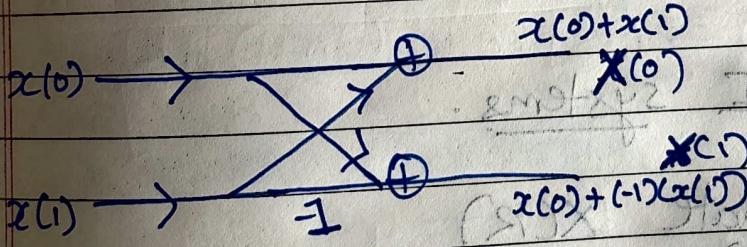
$$W_8^3 = -\frac{1}{\sqrt{2}} - j \frac{1}{\sqrt{2}}$$

FFT - faster than DFT

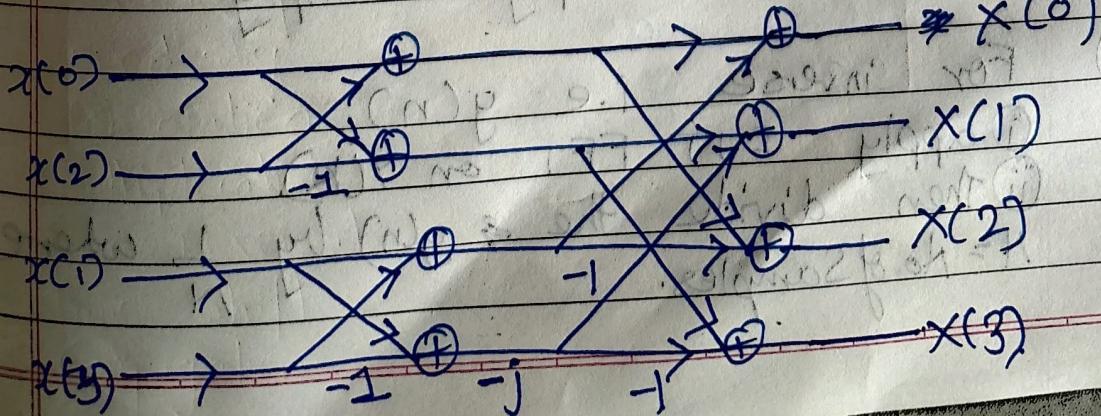
computationally efficient

divide & conquer approach

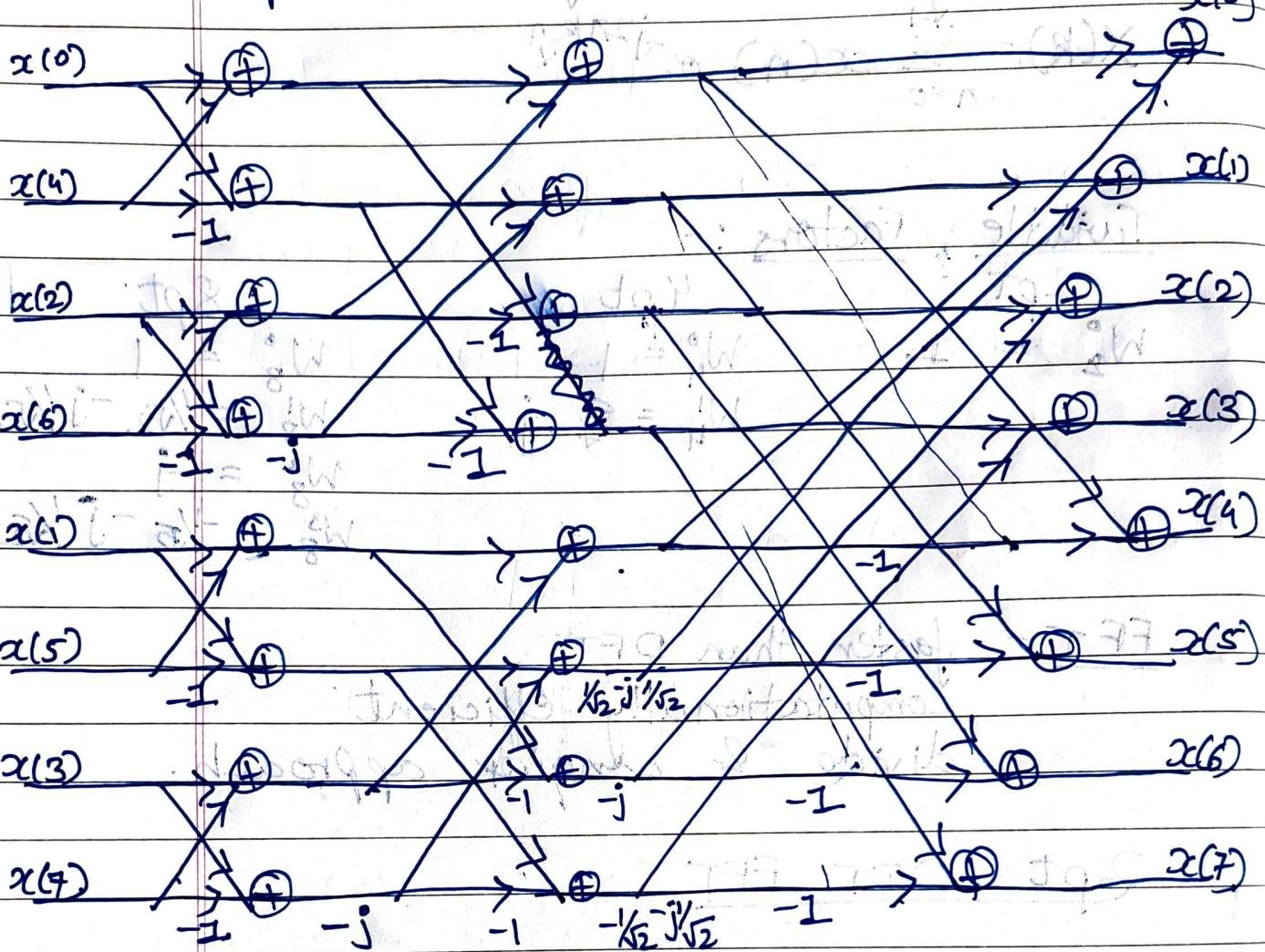
## 2pt DIT FFT



## 4pt DIT FFT



## 8pt DIT FFT



For LTI systems.

- ① First calculate  $X(k)$
- ② Then  $h(k)$
- ③ Final  $Y(k) \Rightarrow X(k) \cdot h(k)$ .  
Simply Multiply

- ④ For inverse i.e.  $y(n)$ ;  
  - ① apply DIT FFT on  $Y(k)$
  - ② Then divide the  $y(n)$  by  $\frac{1}{N}$ , where  
 $N = \text{No. of Samples.}$

## \* Need for Transformation.

### i] Mathematical Convenience:

Complex convolution in time domain is equal to multiplication operation in frequency domain.

### ii] To extract more information.

## \* Unitary Transform.

Preserves the signal energy.

Packs large fraction of the energy of the image into relatively few of the transform coefficients.

A discrete linear transform is unitary if its transform matrix conforms to the unitary condition i.e.  $A \times A^H = I$   
 $A^* \xrightarrow{\text{conjugate}} A^H \xrightarrow{\text{transpose}}$

## \* Orthogonal DFT

$A$  is orthogonal if  $A \cdot A^T = I$

2.0	2.0	2.0	2.0
2.0	2.0	2.0	2.0
2.0	2.0	2.0	2.0

## \* Discrete Cosine Transform (DCT)

DCT is a technique for converting a signal into elementary frequency components and is widely used in image compression.

$$X[k] = \alpha(k) \sum_{n=0}^{N-1} x(n) \cos \left[ \frac{(2n+1)\pi k}{2N} \right] \quad (i)$$

$$\alpha(k) = \begin{cases} 1 & \text{if } k=0 \text{ or primitive} \\ \sqrt{\frac{2}{N}} & \text{if } k \neq 0 \text{ and non-primitive} \end{cases}$$

### IDCT

$$x(n) = \alpha(k) \sum_{k=0}^{N-1} X[k] \cos \left[ \frac{(2n+1)\pi k}{2N} \right]$$

### DCT Matrix for N=4

$$\alpha \begin{bmatrix} 1 & 1 & 1 & 1 \\ \sqrt{2}/2 & \sqrt{2}/2 & -\sqrt{2}/2 & -\sqrt{2}/2 \\ \sqrt{3}/2 & -\sqrt{3}/2 & \sqrt{3}/2 & -\sqrt{3}/2 \\ \sqrt{6}/4 & \sqrt{6}/4 & -\sqrt{6}/4 & -\sqrt{6}/4 \end{bmatrix}$$

Q) Compute DCT matrix for  $N=4$ .

$$\Rightarrow X[k] = \alpha(k) \sum_{n=0}^{N-1} x(n) \cos \left\{ \frac{(2n+1)\pi k}{2N} \right\}$$

$$\alpha(k) = \begin{cases} \sqrt{\frac{1}{N}}, & k=0 \\ \sqrt{\frac{2}{N}}, & k \neq 0 \end{cases}$$

$$N=4: x(0)=0.1x_0 + 0.2x_1 + 0.3x_2 + 0.4x_3$$

For  $k=0$ :

$$X[0] = \sqrt{\frac{1}{4}} \sum_{n=0}^3 x(n) \cos \left\{ \frac{(2n+1)\pi(0)}{8} \right\}$$

$$X[0] = \sqrt{\frac{1}{4}} \sum_{n=0}^3 x(n) \cos 0 \quad \because \cos 0 = 1$$

$$X[0] = \sqrt{\frac{1}{4}} \sum_{n=0}^3 x(n)$$

$$X[0] = \frac{1}{2} \{ x(0) + x(1) + x(2) + x(3) \}$$

② For  $k=1$ :

$$X[1] = \sqrt{\frac{2}{4}} \sum_{n=0}^3 x(n) \cos \left\{ \frac{(2n+1)\pi}{8} \right\}$$

$X[1]$

$$X[1] = \frac{1}{\sqrt{2}} \sum_{n=0}^3 x(n) \cdot \cos\left(\frac{(2n+1)\pi}{8}\right)$$

$$X[1] = \frac{1}{\sqrt{2}} \left\{ x(0) \cos \frac{\pi}{8} + x(1) \cdot \cos \frac{3\pi}{8} + x(2) \cos \frac{5\pi}{8} + x(3) \cos \frac{7\pi}{8} \right\}$$

$$X[1] = 0.707 \left\{ x(0) \cdot 0.9239 + x(1) \cdot 0.3827 + x(2) \cdot (-0.3827) + x(3) \cdot (-0.9239) \right\}$$

$$X[1] = 0.6532 x(0) + 0.2706 x(1) + -0.2706 x(2) + 0.6532 x(3)$$

③ For  $R = 2$ .

$$X[2] = \frac{1}{\sqrt{2}} \sum_{n=0}^3 x(n) \cdot \cos\left(\frac{(2n+1)\pi}{8}\right)$$

$$X[2] = 0.707 \left\{ x(0) \cos\left(\frac{\pi}{4}\right) + x(1) \cdot \cos \frac{3\pi}{4} + x(2) \cos \frac{5\pi}{4} + x(3) \cos \frac{7\pi}{4} \right\}$$

$$X[2] = 0.707 \left\{ 0.707 x(0) + -0.707 x(1) - 0.707 x(2) + 0.707 x(3) \right\}$$

$$X[2] = 0.5 x(0) - 0.5 x(1) - 0.5 x(2) + 0.5 x(3)$$

(4)

For R = 3

$$X[3] = \frac{1}{\sqrt{2}} \left\{ x(0) \cos \frac{3\pi}{8} + x(1) \cos \frac{9\pi}{8} + x(2) \cos \frac{15\pi}{8} + x(3) \cos \frac{21\pi}{8} \right\}$$

$$X[3] = 0.707 \left\{ 0.3827x(0) - 0.9238x(1) + 0.9238x(2) - 0.3827x(3) \right\}$$

(T-N) constraint boundary

$$X[3] = 0.2706x(0) - 0.6532x(1) + 0.6532x(2) - 0.2706x(3)$$

∴ DCTN matrix for N=4 is:

$$\begin{bmatrix} 0.5 & 0.5 & 0.5 & 0.5 \\ 0.6532 & 0.2706 & -0.2706 & -0.6532 \\ 0.5 & -0.5 & -0.5 & 0.5 \\ 0.2706 & \cancel{0.6532} & 0.6532 & -0.2706 \end{bmatrix}$$

## \* Symmetric & Asymmetric

If ~~not~~ Transformation matrix Symmetric then:-

$$F = T f T'$$

If not then :-  $F = T f T'$

## \* Hadamard Transform (H.T)

$$H_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}; H_4 = \begin{bmatrix} H_2 & H_2 \\ H_2 & -H_2 \end{bmatrix}$$

$$H_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}; H_8 = \begin{bmatrix} H_4 & H_4 \\ H_4 & -H_4 \end{bmatrix}$$

$$H_8 = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 \end{bmatrix}$$

•  $H \cdot T$  is orthogonal when normalized by  $\sqrt{2}$

$\frac{1}{\sqrt{2}}$  orthogonal between  $x(n)$  and  $H(T)$

$$\cdot X[K] = \{ H(T) \cdot x(n) \cdot H(T)^T \}$$

~~IDEF~~ Inverse transform  $x(n) = \frac{1}{N} \{ H(T) \cdot X(K) \cdot H(T)^T \}$

•  $H \cdot T$  is symmetric

### \* Walsh Transform :- Symmetric

Walsh Transform is obtained from Hadamard Transform matrix by rearranging the rows in the increasing sign-change order

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

$$F = T f T$$

$$f = \frac{1}{N} T^T F T$$

# \* Haar Transform

Asymmetric, real, orthogonal

$$F = H \cdot f \cdot H^T$$

Very fast transform.

Poor energy compaction for images.

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

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

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

## \* Slant Transform

Unitary transform.

Designed for image coding.

$$S_4 \quad (N=4) = \frac{1}{\sqrt{4}} \begin{bmatrix} 1 & 1 & -1 & 1 \\ 1 & -1 & 1 & 1 \\ 1 & 1 & 1 & -1 \\ 1 & -1 & -1 & -1 \end{bmatrix}$$

$S_4$  is orthogonal & step size is uniform  
 $\therefore$  values of  $a = 2b$ ,  $b = \frac{1}{\sqrt{2}}$

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

Asymmetric Transform  $\therefore F = S_4 \cdot f \cdot S_4'$

## \* KL Transform / Hotelling transform 12

De correlates the data which facilitates high degrees of compression.

Based on statistical properties of image it is helpful for image compression.

Data from neighbouring pixels is highly correlated, so achieving image compression without losing the quality of image is a challenge.

Steps to compute KL/Hotelling /PCA :-

- 1] Find mean vector & covariance matrix of  $X$ .
- 2] Find eigenvalues & their eigen vectors of the covariance matrix.
- 3] Create transformation matrix  $T$ , such that rows of  $T$  are the eigen vectors.
- 4]  $X = T [C_2 - m]$ .

## \* Low-Pass Filtering

Preserves low frequencies, attenuates high frequencies.

Ideal :- ILPF

Butterworth (BLPF) - attenuates high frequency smoothly, less of ringing effect.

$$H(u,v) = \frac{1}{1 + \sqrt{k^2 + l^2}^{2n}} D_0 \rightarrow \text{cutoff frequency.}$$

Gaussian Filter (GLPF) - blurs & smoothes property, with no edges, no ringing effect

$$H(u,v) = e^{-\frac{(u^2+v^2)}{2D_0^2}} D_0 \rightarrow \text{cutoff freq.}$$

Ringing Effect - Sharp cutoff frequency produce an overshoot of image features whose frequency is close to the cutoff frequencies.

## \* High Pass Filtering

Preserves high frequency, attenuates low frequencies.

$$H_{HP}(u,v) = 1 - H_{LP}(u,v)$$



## Homomorphic Filters

Removes shading effects from an image.

Enhances high frequencies

Attenuates low frequencies but preserves fine details

Separates image into two components:

1] Illumination

2] Reflectance

Preserves the sharp edges & features of the image

Illumination  $i(x,y)$  varies slowly & affects low frequency.

Reflectance  $r(x,y)$  varies faster & affects high frequencies.

Steps:  
 $f(x,y) \rightarrow$  Take  $\ln$   $\rightarrow$  Apply FFT  $\rightarrow$  Apply  $H(u,v)$   $\rightarrow$  Take Inverse FFT  $\rightarrow$  Take exp.

$$(u,v) H \rightarrow I \Rightarrow (u,v) H$$