

Lecture 8:

Recall:

- Image decomposition f

- ① Storage saving
- ② Image processing by modifying transformed image (**coefficient matrix**)
(e.g. Removing coefficients associated to high-frequency
elementary images)

- 2 Separable Image Transformation:

- ① SVD (elementary images not universal and meaningless)
- ② Haar (elementary images universal and meaningful) - **unsmooth**

$$f = \sum_{i=1}^n \sum_{j=1}^n g_{ij} \underbrace{I_{ij}}_{\text{elementary images}}$$

Discrete Fourier Transform:

Definition:

The 2D DFT of a $M \times N$ image $g = (g(k, l))_{k,l}$, where $0 \leq k \leq M-1$,

$0 \leq l \leq N-1$ is defined as:

$$\hat{g}(m, n) = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi \left(\frac{km}{M} + \frac{ln}{N} \right)}$$

(where $j = \sqrt{-1}$, $e^{j\theta} = \cos \theta + j \sin \theta$)

Remark: The inverse of DFT is given by:

$$g(p, q) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \hat{g}(m, n) e^{j2\pi \left(\frac{pm}{M} + \frac{qn}{N} \right)}$$

↑
 (no $\frac{1}{MN}$!) ↑
 DFT of g ↑
 (no -ve sign)

Proof of Inverse DFT:

$$\begin{aligned}
 & \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} e^{j2\pi(\frac{pm}{M} + \frac{qn}{N})} \hat{g}(m, n) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} e^{j2\pi(\frac{pm}{M} + \frac{qn}{N})} \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi(\frac{km}{M} + \frac{ln}{N})} \\
 &= \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} g(k, l) e^{j2\pi\left(\frac{(p-k)m}{M} + \frac{(q-l)n}{N}\right)} \\
 &= \frac{1}{MN} \underbrace{\sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l)}_{(*)} \underbrace{\sum_{m=0}^{M-1} e^{j2\pi\left(\frac{(p-k)m}{M}\right)}}_{\text{if } k \neq p} \underbrace{\sum_{n=0}^{N-1} e^{j2\pi\left(\frac{(q-l)n}{N}\right)}}_{\text{if } l \neq q}
 \end{aligned}$$

Note that: $\sum_{m=0}^{M-1} e^{j2\pi\left(\frac{mt}{M}\right)} = \frac{\left[e^{j2\pi\left(\frac{t}{M}\right)}\right]^M - 1}{e^{j2\pi\left(\frac{t}{M}\right)} - 1} = M \delta(t) := \begin{cases} M & t=0 \\ 0 & t \neq 0 \end{cases}$

$\therefore (*) \text{ becomes: } \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) M \delta(p-k) N \delta(q-l) = g(p, q).$

DFT in Matrix form

Theorem: Consider a $N \times N$ image g , the DFT of g can be written as:

$$\hat{g} = \mathcal{U} g \mathcal{U} \quad (\text{DFT in matrix form})$$

where $\mathcal{U} = (\mathcal{U}_{kl})_{0 \leq k, l \leq N-1} \in M_{N \times N}$ and $\mathcal{U}_{kl} = \frac{1}{N} e^{-j \frac{2\pi k l}{N}}$.

Proof: Need to check $\hat{g}(k, l) = (\mathcal{U} g \mathcal{U})(k, l)$

$$\text{LHS} = \hat{g}(k, l) = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} g(m, n) e^{-j 2\pi \left(\frac{km}{N} + \frac{ln}{N} \right)}$$

$$\text{RHS} : \mathcal{U} g \mathcal{U} = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} g(m, n) \begin{pmatrix} 1 \\ \vec{u}_m \\ \vdots \\ 1 \end{pmatrix} (-\vec{u}_n \rightarrow)$$

$$\begin{pmatrix} 1 & \vec{u}_1 & \cdots & \vec{u}_N \end{pmatrix} \begin{pmatrix} -\vec{u}_1 & -\vec{u}_2 & \cdots & -\vec{u}_N \end{pmatrix}$$

$$\mathcal{U} g \mathcal{U}(k, l) = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} g(m, n) \frac{e^{-j \frac{2\pi k m}{N}}}{N} \cdot \frac{e^{-j \frac{2\pi l n}{N}}}{N}$$

$$= \text{LHS}$$

(k-th row, l-th col
of $\mathcal{U} g \mathcal{U}$)

$$\vec{u}_m = \begin{pmatrix} u_{0m} \\ u_{1m} \\ \vdots \\ u_{Nm} \end{pmatrix} = \frac{1}{N} \begin{pmatrix} e^{-j \frac{2\pi (0)m}{N}} \\ e^{-j \frac{2\pi (1)m}{N}} \\ \vdots \\ e^{-j \frac{2\pi (N-1)m}{N}} \end{pmatrix}$$

$$\vec{u}_n = (u_{0n}, u_{1n}, \dots, u_{nn})$$

$$= \frac{1}{N} (e^{-j \frac{2\pi (0)n}{N}}, \dots, e^{-j \frac{2\pi (l)n}{N}}, \dots)$$

Theorem: $U^* U = \frac{1}{N} I$ where $U^* = (\bar{U})^T$ (conjugate transpose)

$$U U^* = \frac{1}{N} I.$$

$$\therefore U^{-1} = (NU)^*$$

$(\overline{a+jb} = a-jb)$
 $(\overline{e^{j\theta}} = \overline{\cos \theta + j \sin \theta} = \cos \theta - j \sin \theta = e^{-j\theta})$

Proof: Consider $(U^* U)_{(k,l)}$ (k -th row, l -th col of $U^* U$)

$$(U^* U)_{(k,l)} = \left(\underbrace{\text{k-th row of } U^*}_{\text{l-col of } U} \right) \left(\text{l-col of } U \right)$$

$$= \overline{(\vec{u}_k^T)} \vec{u}_l$$

$$= \left(\overline{e^{j\frac{2\pi k(0)}{N}}} \cdots, \overline{e^{j\frac{2\pi k(N-1)}{N}}} \right) \left(\begin{array}{c} e^{-j\frac{2\pi l(0)}{N}} \\ \vdots \\ e^{-j\frac{2\pi l(N-1)}{N}} \end{array} \right)$$

$$= \sum_{\alpha=0}^{N-1} \overline{e^{j\frac{2\pi k\alpha}{N}}} \overline{e^{j\frac{2\pi l\alpha}{N}}} = \sum_{\alpha=0}^{N-1} \overline{e^{j\frac{2\pi (k-l)\alpha}{N}}} = \frac{1}{N} \delta(l-k)$$

$$U = \begin{pmatrix} \vec{u}_1 & \vec{u}_2 & \cdots & \vec{u}_N \end{pmatrix}$$

$$\bar{U} = \begin{pmatrix} \vec{u}_1^T \\ \vec{u}_2^T \\ \vdots \\ \vec{u}_N^T \end{pmatrix}$$

$$(\bar{U})^* = \begin{pmatrix} -(\vec{u}_1^T) \\ \vdots \\ -(\vec{u}_N^T) \end{pmatrix}$$

$$\therefore U^*U(k,l) = \begin{cases} \frac{1}{N} & \text{if } k=l \\ 0 & \text{if } k \neq l \end{cases}$$

$$\Rightarrow U^*U = \frac{1}{N} I$$

$$\text{Similarly, } UU^* = \frac{1}{N} I$$

Image decomposition by DFT

Suppose $\hat{g} = \text{DFT}(g) = U g U$

Then: $U U^* = \frac{1}{N} I = U^* U$

$$\therefore g = (NU)^* \hat{g} (NU)^*$$

$$\therefore g = \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} \hat{g}_{kl} \vec{w}_k \vec{w}_l^T$$

Elementary image of DFT

where $\vec{w}_k = k^{\text{th}} \text{ col of } (NU)^*$