

Discrete Fourier Transform:

Definition: The 1D discrete Fourier Transform (DFT) of a function $f(k)$, defined at discrete points $k=0, 1, 2, \dots, N-1$ is defined as:

$$\hat{f}(m) = \frac{1}{N} \sum_{k=0}^{N-1} f(k) e^{-j \frac{2\pi}{N} mk} \quad (\text{where } j = \sqrt{-1}, e^{j\theta} = \cos\theta + j\sin\theta)$$

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:

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

Remark: The inverse of DFT is given by:

$$g(p, q) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \hat{g}(m, n) e^{j \frac{2\pi}{M} \left(\frac{pm}{M} + \frac{qn}{N} \right)}$$

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

Image decomposition under DFT:

Consider a $N \times N$ image g , the DFT of g :

$$\hat{g}(m, n) = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi(\frac{km+ln}{N})}$$

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

U is clearly symmetric and also:

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

$$\begin{aligned} \text{Note that: } \frac{1}{N^2} \sum_{\alpha=0}^{N-1} e^{-j\left(\frac{2\pi x_1 \alpha}{N}\right)} e^{+j\left(\frac{2\pi x_2 \alpha}{N}\right)} &= \frac{1}{N^2} \sum_{\alpha=0}^{N-1} e^{-j\frac{2\pi(x_2 - x_1)\alpha}{N}} = \frac{1}{N^2} N \delta(x_2 - x_1) \\ &= \frac{1}{N} \delta(x_2 - x_1) \end{aligned}$$

Let $U = \begin{pmatrix} \vec{u}_1 & \vec{u}_2 & \dots & \vec{u}_N \end{pmatrix}$. Then: $\langle \vec{u}_i, \vec{u}_j \rangle = \vec{u}_i^\top \vec{u}_j = \begin{cases} \frac{1}{N} & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$

$\therefore \{\vec{u}_i\}_{i=1}^N$ is orthogonal but NOT orthonormal!

Example Find the DFT of the following 4×4 image

$$g = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

Solution

The matrix U is given by:

$$U = \frac{1}{4} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix}$$

$$U = \left(U_{k,l} \right)_{k,l}$$
$$= \frac{1}{4} \left(e^{-j2\pi \left(\frac{k+l}{4} \right)} \right)$$

$$\therefore \text{DFT of } g = \hat{g} = UgU = \begin{pmatrix} \frac{1}{4} & -\frac{1}{4} & \frac{1}{4} & -\frac{1}{4} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

Why is DFT useful in imaging:

1. DFT of convolution:

Recall:
$$g * w(n, m) = \sum_{n'=0}^{N-1} \sum_{m'=0}^{N-1} g(n-n', m-m') w(n', m')$$

$$(g, w \in M_{N \times N}(\mathbb{R}))$$

Then, the DFT of $g * w = MN \text{DFT}(g) \text{DFT}(w)$

∴ DFT of convolution can be reduced to simple multiplication!

Recall: Shift-invariant image transformation = 2D convolution.

∴ Easy computation/manipulation of shift-invariant transf.
after DFT!!

Proof:

DFT of $g * w$ at (p, q)

$$= \frac{1}{NM} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} g * w(n, m) e^{-j2\pi(\frac{pn}{N} + \frac{qm}{M})}$$

$$= \frac{1}{NM} \sum_{n'=0}^{N-1} \sum_{m'=0}^{M-1} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} g(n-n', m-m') w(n', m') e^{-j2\pi(\frac{pn}{N} + \frac{qm}{M})}$$

$$= \frac{1}{NM} \sum_{n'=0}^{N-1} \sum_{m'=0}^{M-1} w(n', m') e^{-j2\pi(\frac{pn'}{N} + \frac{qm'}{M})}$$

$$\sum_{n''=-n'}^{N-1-n'} \sum_{m''=-m'}^{M-1-m'} g(n'', m'') e^{-j2\pi(\frac{pn''}{N} + \frac{qm''}{M})}$$

$\hat{w}(p, q)$

Change of variables:

$$n \rightarrow n'' = n - n'$$

$$m \rightarrow m'' = m - m'$$

$\hat{w}(p, q)$

Note that: g and w are periodically extended.

$$\therefore g(n-N, m) = g(n, m) \text{ and } g(n, m-M) = g(n, m)$$

$$\therefore T = \sum_{m''=-m'}^{M-1-m'} e^{-j2\pi \frac{qm''}{M}} \sum_{n''=-n'}^{-1} g(n'', m'') e^{-j2\pi \frac{pn''}{N}} + \sum_{m''=-m'}^{M-1-m'} e^{-j2\pi \frac{qm''}{M}} \sum_{n''=0}^{N-1-n'} g(n'', m'') e^{-j2\pi \frac{pn''}{N}}$$

$$\text{Consider } \sum_{n''=-N'}^{\frac{-1}{N}} g(n'', m'') e^{-j 2\pi \frac{pn''}{N}} \stackrel{n'''=N+n''}{=} \sum_{n'''=N-n'}^{N-1} g(n'''-N, m'') e^{-j 2\pi \left(\frac{pn''}{N}\right)} e^{j 2\pi p} \\ \text{We can do similar thing for index } m''.$$

$$\therefore T = \sum_{m''=0}^{M-1} \sum_{n''=0}^{N-1} g(n'', m'') e^{-j 2\pi \left(\frac{pn''}{N} + \frac{qm''}{M}\right)} = MN \hat{g}(p, q)$$

$$\therefore \hat{g * w}(p, q) = MN \hat{g}(p, q) \hat{w}(p, q)$$

Remark: Conversely, if $x(n, m) = g(n, m) w(n, m)$

$$\text{Then, } \hat{x}(k, l) = \sum_{p=0}^{N-1} \sum_{q=0}^{M-1} \hat{g}(p, q) \hat{w}(k-p, l-q) \quad (\text{Convolution of } g \text{ and } w)$$

2. Average value of image

$$\text{Average value of } g = \bar{g} = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(k, l) = \underbrace{\frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi(\phi)}}_{\hat{g}(0, 0)}$$

3. DFT of a rotated image

Consider a $N \times N$ image g .

$$\text{Then: } \hat{g}(m, n) = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi\left(\frac{km+ln}{N}\right)}$$

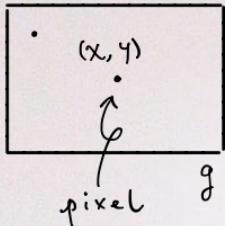
Write k and l in polar coordinates:

$$k \equiv r \cos \theta ; \quad l \equiv r \sin \theta$$

Similarly, write $m \equiv w \cos \phi ; \quad n \equiv w \sin \phi$.

$$\text{Note that: } km + ln = rw (\cos \theta \cos \phi + \sin \theta \sin \phi) = rw \cos(\theta - \phi).$$

Denote $\mathcal{P}(g) = \{(r, \theta) : (r \cos \theta, r \sin \theta) \text{ is a pixel of } g\}$
(Polar coordinate set of g)



If $\begin{cases} x = r \cos \theta \\ y = r \sin \theta \end{cases}$, then $(r, \theta) \in \mathcal{P}(g)$.

Then: $\hat{g}(m, n) = \hat{g}(\omega, \phi) = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(r, \theta) e^{-j2\pi \left(\frac{rw \cos(\theta - \phi)}{N} \right)}$

Identify $\hat{g}(m, n)$ with $\hat{g}(\omega, \phi)$
 Identify $g(k, l)$ with $g(r, \theta)$

Consider a rotated image $\tilde{g}(r, \theta) = g(r, \theta + \theta_0)$ where θ is defined between $-\theta_0$ to $\frac{\pi}{2} - \theta_0$.

\therefore image g is rotated clockwise by θ_0 .

DFT of \tilde{g} is:

$$\hat{\tilde{g}}(\omega, \phi) = \frac{1}{N^2} \sum_{(r, \theta) \in \mathcal{P}(\tilde{g})} \tilde{g}(r, \theta) e^{-j2\pi \left(\frac{rw \cos(\theta - \phi)}{N} \right)} = \frac{1}{N^2} \sum_{(r, \tilde{\theta}) \in \mathcal{P}(g)} g(r, \tilde{\theta}) e^{-j2\pi \left(\frac{rw \cos(\tilde{\theta} - \theta_0 - \phi)}{N} \right)}$$

$\tilde{g}(r, \underbrace{\theta + \theta_0}_{\tilde{\theta}})$

$$\therefore \hat{\tilde{g}}(\omega, \phi) = \hat{g}(\omega, \phi + \theta_0). \quad (\phi \text{ is also defined between } -\theta_0 \text{ to } \frac{\pi}{2} - \theta_0)$$

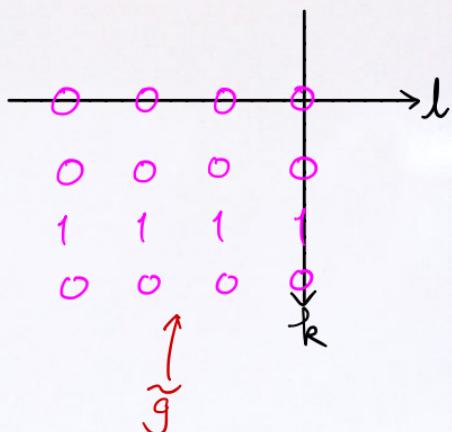
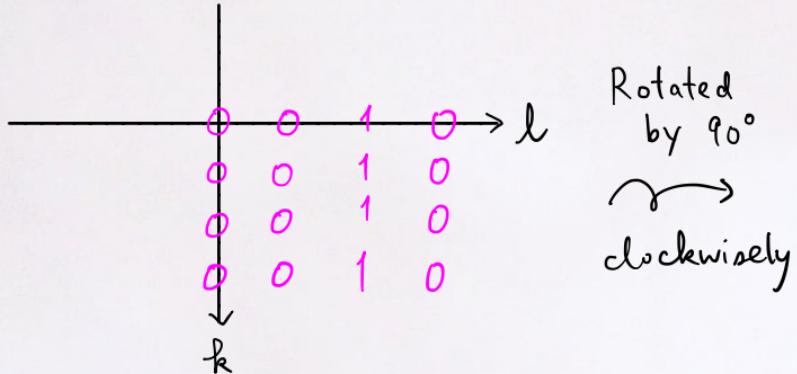
DFT of an image rotated by θ_0 = DFT of the original image rotated by θ_0 .

Example: Let $g = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$. Then: $\hat{g} = \begin{pmatrix} \frac{1}{4} & -\frac{1}{4} & \frac{1}{4} & -\frac{1}{4} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$

$$0 \leq k \leq 3$$

$$0 \leq l \leq 3$$

Note that g in the coordinate system:



Note that indices of \tilde{g} are taken as: $\begin{cases} -3 \leq l \leq 0 \\ 0 \leq k \leq 3 \end{cases}$.

Now, DFT of $\tilde{g} = \hat{\tilde{g}}$ (given by: $\sum_{k=0}^3 \sum_{l=-3}^0 \tilde{g}(k, l) e^{-j2\pi(\frac{km+ln}{4})}$)

$$= \begin{pmatrix} 0 & 0 & 0 & \frac{1}{4} \\ 0 & 0 & 0 & -\frac{1}{4} \\ 0 & 0 & 0 & \frac{1}{4} \\ 0 & 0 & 0 & -\frac{1}{4} \end{pmatrix} \left| \begin{array}{c} 0 \\ 1 \\ 2 \\ 3 \end{array} \right. \quad ; \quad \begin{array}{l} 0 \leq k \leq 3 \\ -3 \leq l \leq 0 \end{array}$$

\xrightarrow{l} \xrightarrow{k}

-3 -2 -1 0