

Digital Image Processing

Ch-4 Filtering in the Frequency Domain




Image Enhancement in freq. Domain

- From spatial domain to freq. domain
- then processed
- Inverse transform is applied to bring back into spatial domain
- filters → smoothing & Sharpening
 - removing high & low freq.
- Change → whole image

Types of filters

a) low pass filters
↳ smoother

b) high pass filters
↳ sharper

Ideal
Butterworth
Gaussian

Fourier Transform:

- relation b/w spatial & freq. domain
- image enhancement in freq. domain

(1) Discrete FT

$$x(n) \xleftrightarrow{\text{DFT}} X(k)$$

Spatial domain Freq. domain

$N \rightarrow$ no of samples

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

$$X(k) \xleftrightarrow{\text{IDFT}} x(n)$$

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

Fourier Spectrum

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

Phase angle $\phi(u) = \tan^{-1} [I(u)/R(u)]$

Power spectrum $P(u) = |F(u)|^2 = R^2(u) + I^2(u)$

2D Discrete Fourier Transform [2D DFT]

$$f(x, y) \xrightarrow{2D \text{ DFT}} F(u, v)$$

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

$$F(u, v) \xrightarrow{2D \text{ IDFT}} f(x, y)$$

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

u & $v \rightarrow$ transform or freq. variables

x & $y \rightarrow$ spatial or image variables

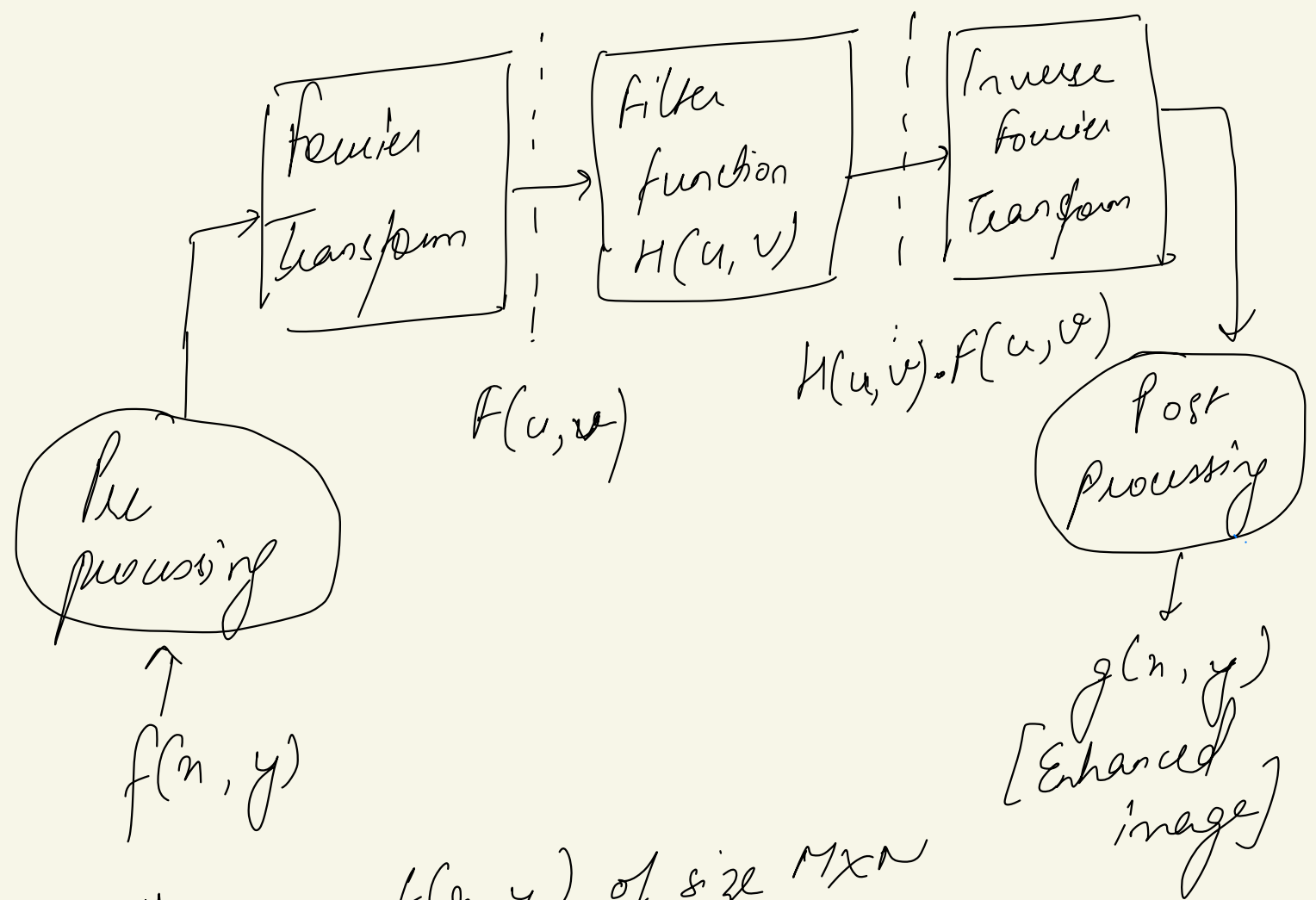
Fourier Spectrum

$$|F(u, v)| = \left[R^2(u, v) + I^2(u, v) \right]^{1/2}$$

Phase angle $\phi(u, v) = \tan^{-1} \left[\frac{I(u, v)}{R(u, v)} \right]$

Power spectrum $P(u, v) = |F(u, v)|^2 = R^2(u, v) + I^2(u, v)$

Steps for filtering in freq. domain:



1) Image $f(x, y)$ of size $M \times N$

2) $f(x, y) (-1)^{x+y}$

3) $F(u, v) \rightarrow$ F.T of $f(x, y)$

4) $H(u, v) \rightarrow$ Filter in freq. domain

$$G(u, v) = H(u, v) \cdot F(u, v)$$

$$g(x, y) = \mathcal{F}^{-1}[G(u, v)](-1)^{x+y}$$

Image Smoothing & Sharpening using freq. domain filters:

a) Low pass filter
→ Smoother

i) Ideal low pass filter

$$H(u,v) = \begin{cases} 1 & ; D(u,v) \leq D_0 \\ 0 & ; D(u,v) > D_0 \end{cases}$$

$D_0 \Rightarrow$ non -ve quantity

$D(u,v) \rightarrow$ distance from pt. (u,v)

$f(x,y) \rightarrow M \times N$ $D(u,v) = \left[u - \frac{M}{2}\right]^2 + \left[v - \frac{N}{2}\right]^2$

ii) Butterworth LPF

$$H(u,v) = \frac{1}{1 + \left[D(u,v) / D_0 \right]^2}$$

Advantage: useful in defining the edges

iii) Gaussian LPF

$$- D^2(u,v) / 2 D_0^2$$

$$H(u,v) = e$$

removes low freq. noise

b) High pass filter

→ Sharper

i) Ideal High pass filter

$$H(u,v) = \begin{cases} 0 & ; D(u,v) \leq D_0 \\ 1 & ; D(u,v) > D_0 \end{cases}$$

Disadvantage: blurred edges

ii) Butterworth HPF

$$H(u,v) = \frac{1}{1 + \left[D_0 / D(u,v) \right]^2}$$

removes high freq. noise

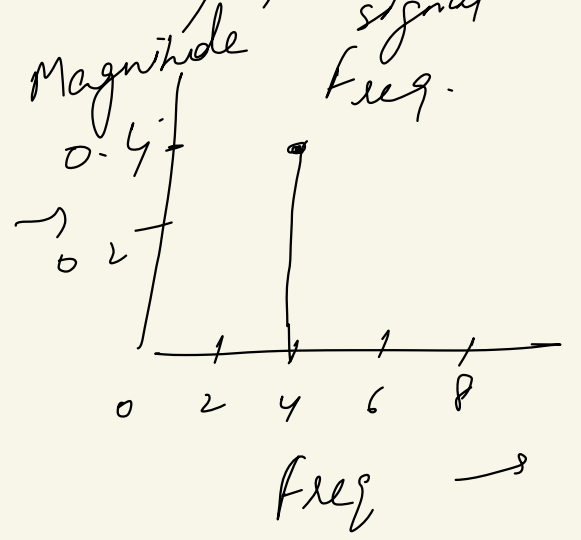
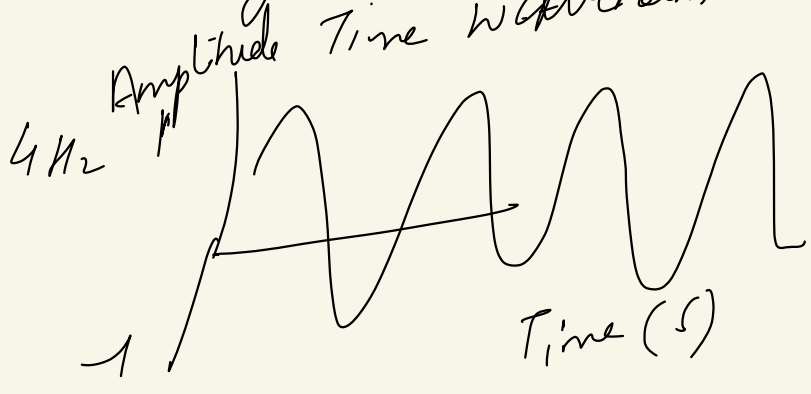
iii) Gaussian HPF

$$H(u,v) = 1 - e^{-D^2(u,v) / 2 D_0^2}$$

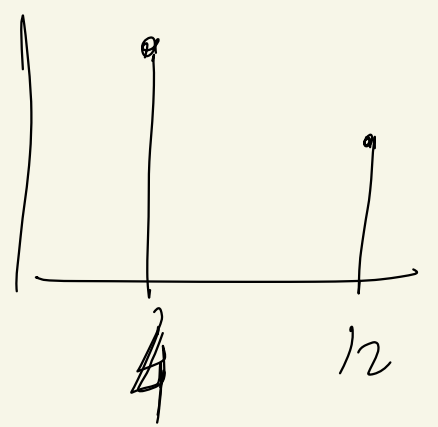
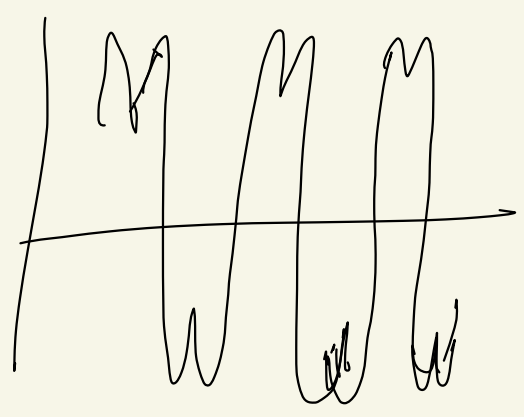
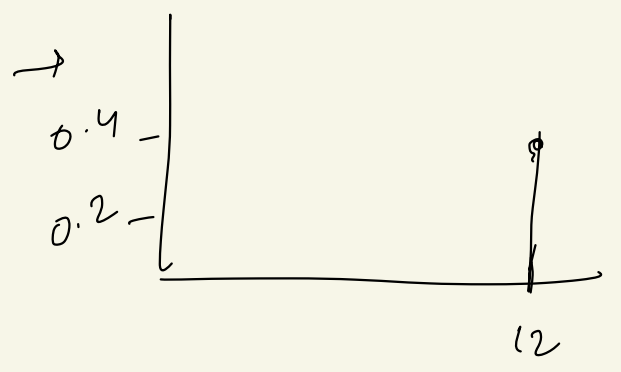
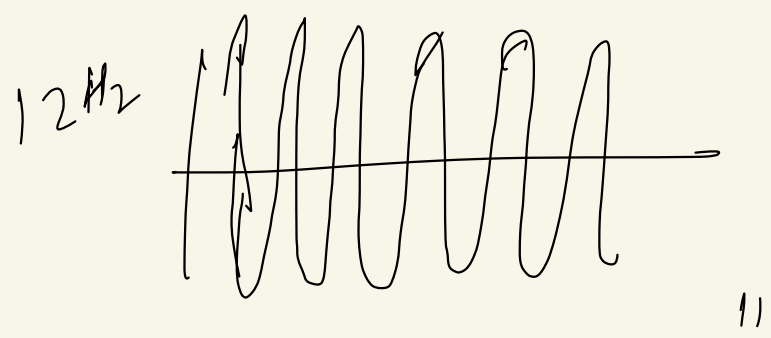
removes high freq. noise

What is FT?

Converting time domain signal into freq domain signal

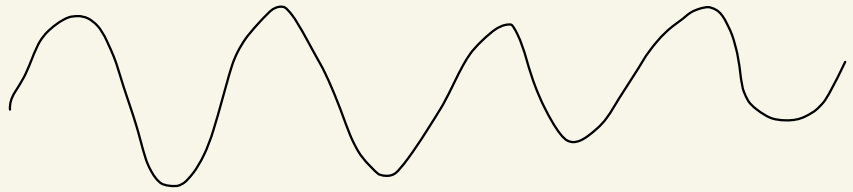


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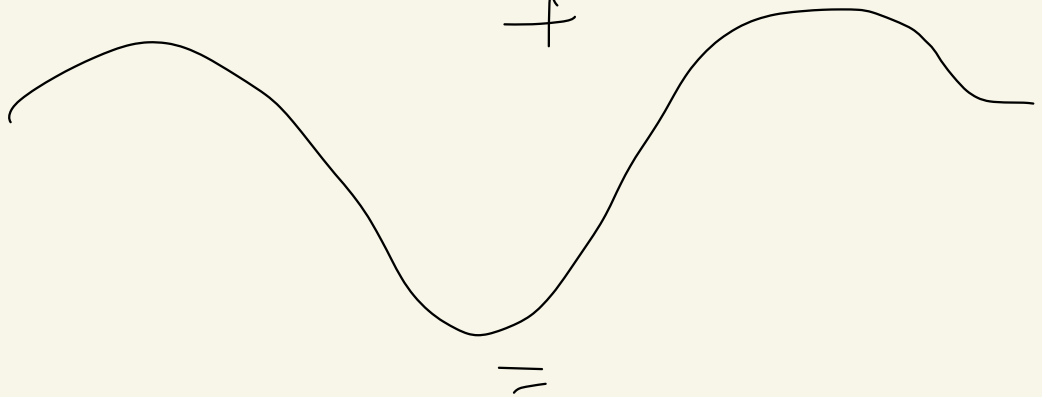




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⇓

What freq bands individual signals

this has?

In time domain we can't predict

F.T. is an important image processing tool which is used to decompose an image into its sine & cosine components.

As we are only concerned with digital images, we will restrict to DFT.

For image size $N \times N$

2D DFT:

$$F(u, v) = \sum_{n=0}^{N-1} \sum_{y=0}^{N-1} f(n, y) e^{-j2\pi \left(\frac{un}{N} + \frac{vy}{N} \right)}$$

IDFT:

$$f(n, y) = \frac{1}{N^2} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi \left(\frac{un}{N} + \frac{vy}{N} \right)}$$

$$e^{j\theta} = \cos \theta + j \sin \theta$$

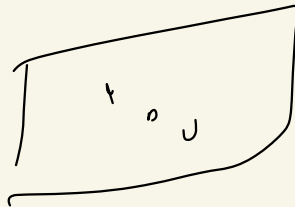
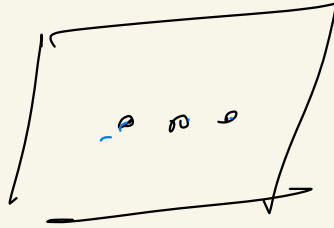
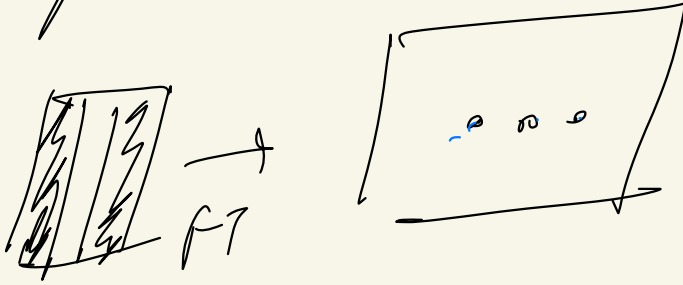
$$e^{-j\theta} = \cos \theta - j \sin \theta$$

$f(n, y) \rightarrow$ spatial domain

Exponential term is the basis fn.

The basis fns. are sine & cosine waves with increasing frequencies

The response of the FT to periodic patterns in the spatial domain images can be seen very easily in the following artificial images:-



```
I = imread("1.png");
I = MxNx3 uint8
im = fft2(I);
figure, imshow(im);
```

Ques 2D DFT of 4×4 gray scale image

$$f(n, y) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$$F(u, v) = \text{kernel} \times f(n, y) \times \text{kernel}^T$$

DFT basis (kernel) for $N=4$ is

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

for y)

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

$$\begin{bmatrix} 4 & 4 & 4 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$F(u, v) = K \cdot f(x, y) \cdot K^T$$

$$= \begin{bmatrix} 4 & 4 & 4 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -j \\ 1 & j & -1 & -j \end{bmatrix}$$

$$= \begin{bmatrix} 16 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \rightarrow \text{Black}$$

2. $f(x) = \{1, 0, 0, 1\}$
IDFT

$$F[k] = \sum_{n=0}^{N-1} f(n) e^{-j2\pi kn/N}$$

where $k=0, 1, \dots, N-1$

0 → Black
255 → white

Q DFT $f(n) = \{1, 0, 0, 1\}$

$$F[k] = \sum_{n=0}^{N-1} f(n) e^{-j2\pi kn/N}$$

where

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

$$N = 4$$

$$F[k] = \sum_{n=0}^3 f(n) e^{-j2\pi kn/4}$$

$$= f(0) e^0 + f(1) e^{-j2\pi k/4} + f(2) e^{-j4\pi k/4} + f(3) e^{-j6\pi k/4}$$

$$= 1 + 0 + 0 + e^{-j3\pi k/2}$$

$$= 1 + e^{-j3\pi k/2}$$

$$k=0 \quad F[0] = 1 + e^0 = 1 + 1 = 2$$

$$k=1 \quad F[1] = 1 + e^{-j3\pi/2} = 1 + j$$

$$k=2 \quad F[2] = 1 + e^{-j3\pi} = 1 - 1 = 0$$

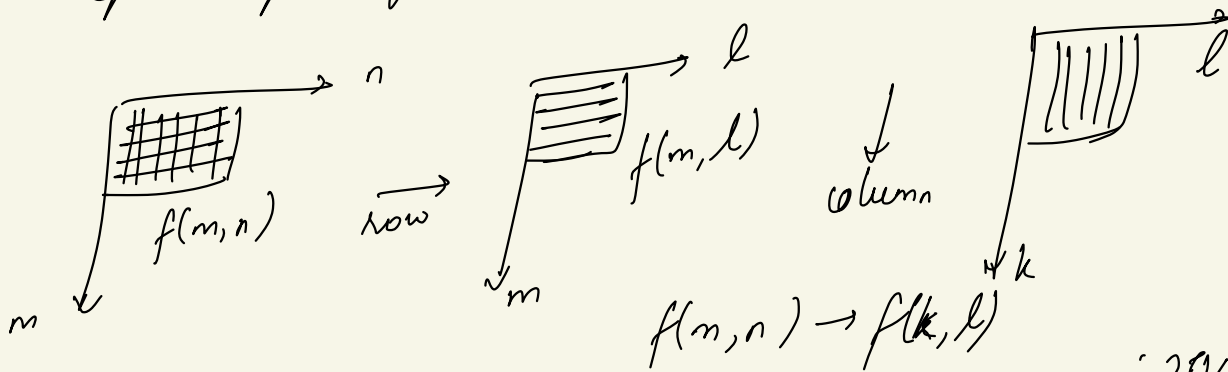
$$k=3 \quad F[3] = 1 + e^{-j9\pi/2} = 1 - j$$

$\rightarrow \cos \frac{3\pi}{2} - j \sin \frac{3\pi}{2}$
 $= 0 - j \times (-1)$
 $= +j$

$$f[k] = \{1+j, 0, 1-j\}$$

properties 2D DFT

1) separable property: Successive 1D operation on row & column



$$\begin{aligned}
 F(k, l) &= \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m, n) \cdot e^{-j \frac{2\pi}{N} m k} \cdot e^{-j \frac{2\pi}{N} n l} \\
 &= \sum_{m=0}^{N-1} \left(\sum_{n=0}^{N-1} f(m, n) \cdot e^{-j \frac{2\pi}{N} n l} \right) \cdot e^{-j \frac{2\pi}{N} m k} \\
 &= \sum_{m=0}^{N-1} f(m, l) \cdot e^{-j \frac{2\pi}{N} m k} \\
 &= f(k, l)
 \end{aligned}$$

2) Spatial shift property:

$$\begin{aligned}
 f(m, n) &\rightarrow f(m-m_0, n) \\
 f(m-m_0, n) &\xrightarrow{\text{DFT}} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m-m_0, n) e^{j \frac{2\pi}{N} m k} e^{j \frac{2\pi}{N} n l} \\
 &= \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m-m_0, n) e^{j \frac{2\pi}{N} (m-m_0) k} \cdot e^{j \frac{2\pi}{N} m k} \cdot e^{-j \frac{2\pi}{N} m l}
 \end{aligned}$$

$$= e^{-j\frac{2\pi}{N} m_0 k} \left(\sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m-m_0, n) \cdot e^{j\frac{2\pi}{N} nl} \right) e^{j\frac{2\pi}{N} (m-m_0) k}$$

$$= e^{-j\frac{2\pi}{N} m_0 k} \sum_{m=0}^{N-1} f(m-m_0, l) e^{j\frac{2\pi}{N} (m-m_0) k}$$

$$= e^{-j\frac{2\pi}{N} m_0 k} f(k, l)$$

Direction of shift

3) Periodicity $f(k, l) = f(k + pN, l + qN)$

$$f(k + pN, l + qN) = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m, n) e^{-j\frac{2\pi}{N} m(k+pN)} e^{-j\frac{2\pi}{N} n(l+qN)}$$

$$= \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m, n) \cdot e^{j\frac{2\pi}{N} mk} \cdot e^{-j\frac{2\pi}{N} m p N} \cdot e^{-j\frac{2\pi}{N} n l} \cdot e^{-j\frac{2\pi}{N} n q N}$$

$$= \underbrace{e^{-j2\pi m p} \cdot e^{-j2\pi n q}}_{=1} \cdot \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} f(m, n) e^{-j\frac{2\pi}{N} mk} e^{-j\frac{2\pi}{N} nl}$$

$$= \sum_{m=0}^{N-1} \left(\sum_{n=0}^{N-1} f(m, n) e^{-j\frac{2\pi}{N} nl} \right) e^{-j\frac{2\pi}{N} mk}$$

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

4) Scaling \rightarrow to increase / decrease size of image
expansion in 1 domain \approx compression in other domain

$$f(ax, by) \rightarrow \frac{1}{|ab|} F\left(\frac{u}{a}, \frac{v}{b}\right)$$

5) Convolution property \rightarrow Convolution in spatial domain =
Multiplication in freq. domain

$$x(n) * h(n) = \sum_{k=-\infty}^{\infty} x(k) \cdot h(n-k)$$

$$f(x, y) * g(x, y) = F(u, v) \cdot G(u, v)$$

6) Multiplication by exponential in spatial domain
leads to freq. shift

$$f(x, y) \cdot e^{j\pi \frac{x w_0}{X}} \cdot e^{j2\pi \frac{y v_0}{Y}}$$

$$= F[(w - w_0), (v - v_0)]$$

7) Rotation $f(x, y) = f(x \cos \theta, y \sin \theta)$

\downarrow DFT

$$F[R \cos \phi, R \sin \phi]$$

$$\text{DFT}(x \cos(\theta + \theta_0), x \sin(\theta + \theta_0)) \rightarrow F[R \cos(\phi + \phi_0), R \sin(\phi + \phi_0)]$$

i.e. if image is rotated by specific angle,
spectrum is rotated by same angle

