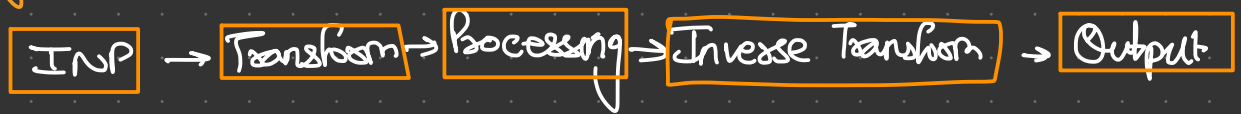


DIP Lecture 9

freq. Domain Processing

→ Frequency Domain:



→ Spatial period:

Minimum no. of pixels b/w identical patterns in periodic image

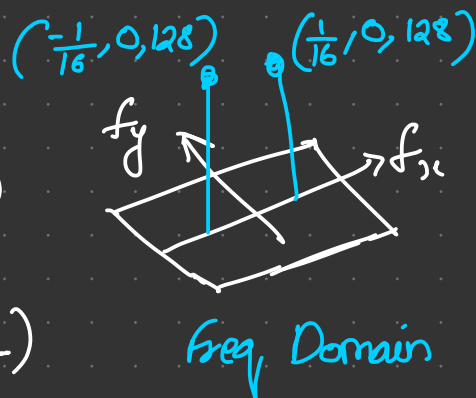


→ Representation of spatial in freq. domain:

→ Expressed in tuples

For $I(x,y) = 128 \cos\left(\frac{2\pi x}{16}\right)$

Now for $I(x,y) = 128 \sin\left(\frac{2\pi x}{4}\right)$

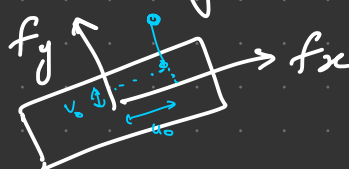


Any compression time/spatial domain is expansion in time domain and vice-versa

If freq. along y -axis then ⁱⁿ freq. domain, impulse are observed in f_y axis

If we have $I(x,y) = \underbrace{\text{constant}}_{\substack{\text{DC component} \\ \text{so} \\ \text{impulse at} \\ \text{origin}}} + \text{periodic}$

$$s(x,y) = \sin[2\pi(u_0x + v_0y)]$$

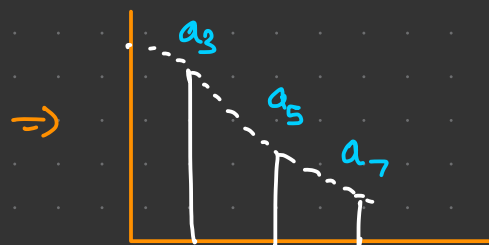
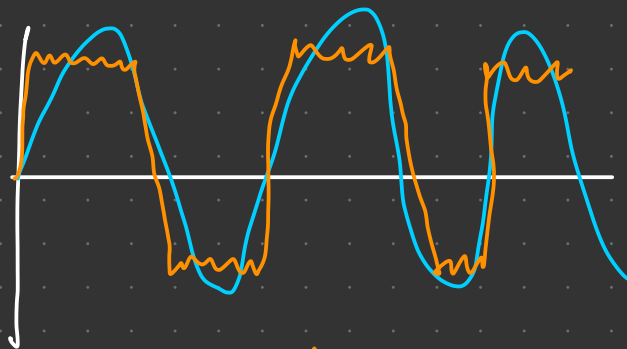


Lots of natural phenomena follow periodic signals

↓
Use Fourier series

Approximate periodic signals with sines & cosines

$$f(x) = \frac{1}{2}a_0 + \sum_{n=1}^{\infty} a_n \cos\left(\frac{n\pi x}{L}\right) + \sum_{n=1}^{\infty} b_n \sin\left(\frac{n\pi x}{L}\right)$$



Harmonics
↓
where f is the fundamental

As no. of terms increases, the approximation improves

To simplify this,

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

→ unit circle in counter-clockwise

So, $e^{j2\pi f_0 t} = \cos(2\pi f_0 t) + j\sin(2\pi f_0 t)$

$$\cos(\omega t) = \frac{e^{j\omega t} + e^{-j\omega t}}{2}, \quad \sin(\omega t) = \frac{e^{j\omega t} - e^{-j\omega t}}{2j}$$

So representing Fourier series in complex coefficients

$$f(t) = \sum_{n=-\infty}^{\infty} c_n e^{j\left(\frac{2\pi n t}{T}\right)}$$

$$c_n = \frac{1}{T} \int_0^T f(t) e^{-j\frac{2\pi n t}{T}} dt$$

$$a_0 = \frac{1}{T} \int_0^T f(t) dt$$

$$a_m = \frac{2}{T} \int_0^T f(t) \cos\left(\frac{2\pi m t}{T}\right) dt$$

$$b_m = \frac{2}{T} \int_0^T f(t) \sin\left(\frac{2\pi m t}{T}\right) dt$$

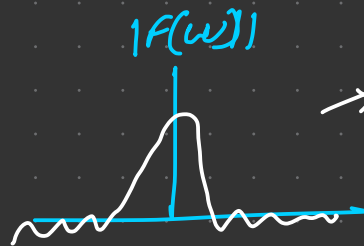
Fourier Transform [Generalisable for non-periodic and periodic]

Approximate non-periodic signals with complex sinusoids

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-j\omega t} dt$$



F.T \Rightarrow



Centrosymmetric
(Symmetric about origin)

Phasor vectors spans all 4 quadrants due to which we get negative frequency

FT:

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-j\omega t} dt$$

$$F(\omega) = \mathcal{F}[f(t)]$$

Inv FT:

$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F(\omega) e^{j\omega t} d\omega$$

$$f(t) = \mathcal{F}^{-1}[F(\omega)]$$

Unit Impulse:

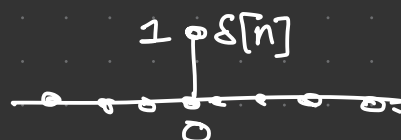
$$\int_{-\infty}^{\infty} \delta(t) dt = 1 \quad \rightarrow \text{Area} = 1$$

$$\delta(t) = \begin{cases} 0, & t \neq 0 \\ \infty, & t = 0 \end{cases}$$

Can be scaled too

Discrete Impulse Function:

$$\delta[n] = \begin{cases} 1, & n = 0 \\ 0, & n \neq 0 \end{cases}$$



properties:

$$1) \int_a^b \delta(t) dt = \begin{cases} 1, & a < 0 < b \\ 0, & \text{otherwise} \end{cases}$$

→ Integral property

$$2) \int_a^b \delta(t) f(t) dt = \int_a^b \delta(t) f(0) dt = f(0) \int_a^b \delta(t) dt$$

↗
Sifting property

$$= \begin{cases} f(0), & a < 0 < b \\ 0, & \text{otherwise} \end{cases}$$

Shifted Unit Impulse

$$\delta(t-\lambda) = \begin{cases} \infty, & t=\lambda \\ 0, & t \neq \lambda \end{cases}$$

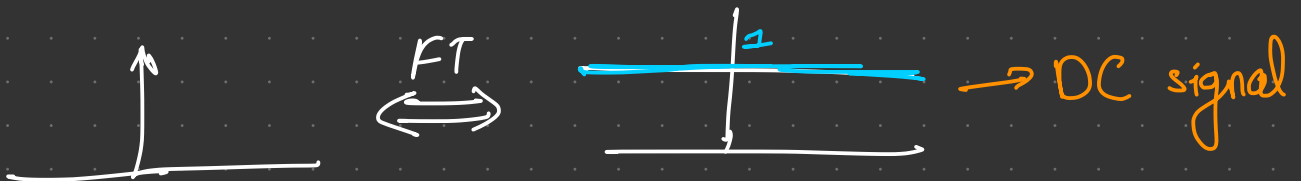


$$\int_a^b \delta(t-\lambda) dt = \begin{cases} 1, & a < \lambda < b \\ 0, & \text{otherwise} \end{cases}$$

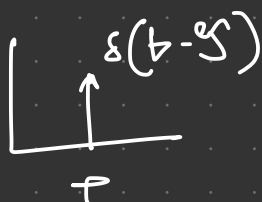
$$\int_a^b \delta(t-\lambda) f(t) dt = \begin{cases} f(\lambda), & a < \lambda < b \\ 0, & \text{otherwise} \end{cases}$$

FT of impulse functions

what freq components are present



FT of shifted impulses



$$F(\omega) = e^{-j\omega\tau}$$

$$\int_a^b \delta(t-\tau) x(t) dt = \begin{cases} x(\tau), & a < \tau < b \\ 0, & \text{otherwise} \end{cases}$$

Duality property:

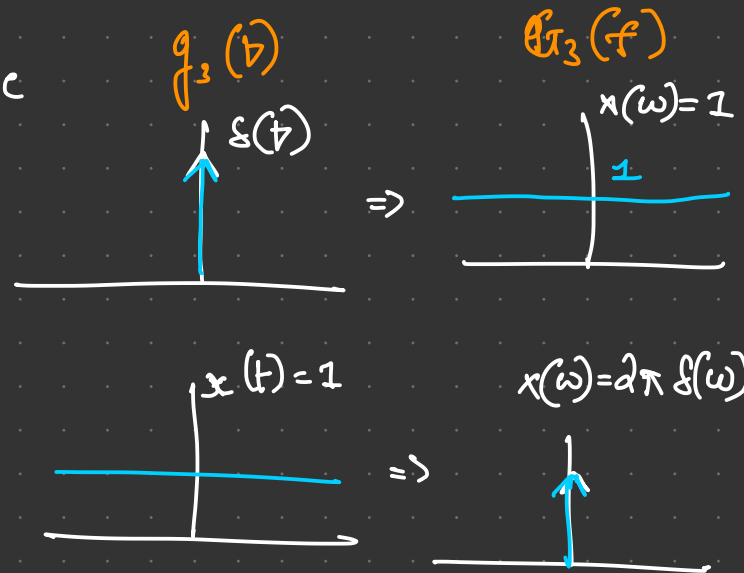
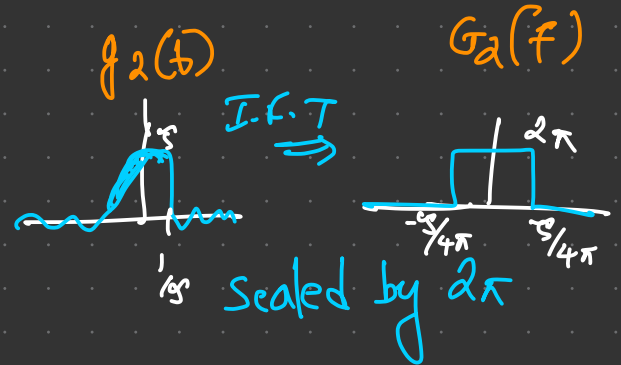
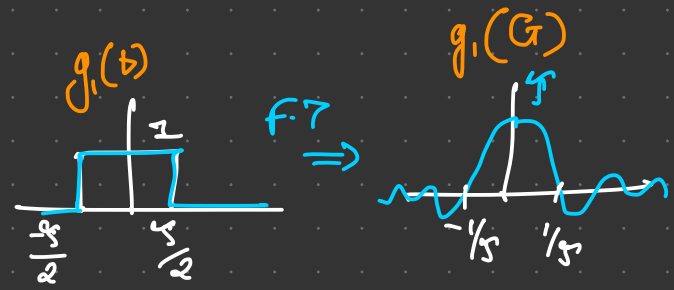
$$F[F(t)] = F(\omega)$$

$$F[F(t)] = 2\pi f(-\omega)$$

Advantages

We only need to know one direction of transform.

That way we can get reverse transform



FT of complex exponential

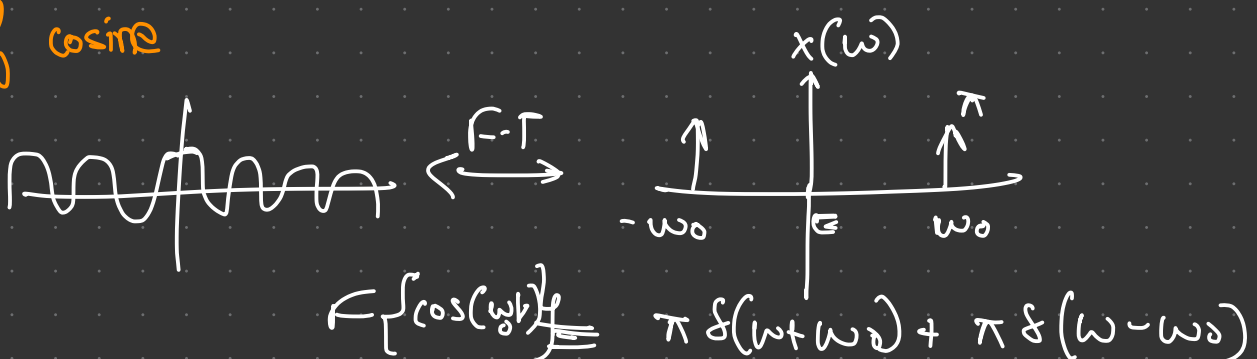
$$e^{i\omega_0 t} \xrightarrow{F} 2\pi \delta(\omega - \omega_0)$$

FT of real even function is also real

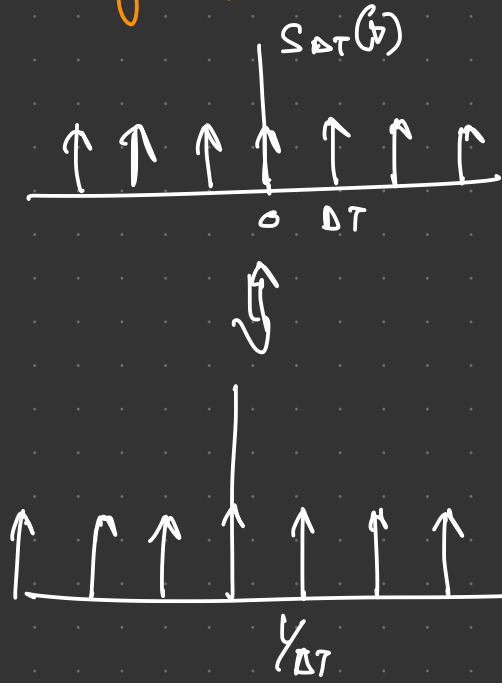
$$F(x_1(t)) + F(x_2(t)) = F(x_1(t) + x_2(t))$$

↓
Not applicable for multiplication

FT of cosine



FT of impulse train



$$s_{\Delta T}(t) = \sum_{n=-\infty}^{\infty} \delta(t - n\Delta T)$$

$$F(s_{\Delta T}(t)) = \sum_{n=-\infty}^{\infty} \delta(u - n/\Delta T)$$

$$\downarrow$$

$$S(u)$$

Impulse train is used for sampling
for $f(t)$,

$$\bar{f}(t) = \sum_{n=-\infty}^{\infty} f_n \delta(t - n\Delta T)$$

↑ Sampled values

for m -samples

$$\bar{f}(t) = \sum_{n=0}^{(m-1)\Delta T} f_n \delta(t - n\Delta T)$$

FT of sampled functions

$$\bar{F}(u) = \frac{1}{\Delta T} \sum_{n=-\infty}^{\infty} F(u - \frac{n}{\Delta T})$$

↑ Continuous copies

Digital processing of freq.

→ Need discrete freq. samples from continuous

→ Need only one period value

$$\rightarrow \bar{F}(u) = \sum_{n=0}^{M-1} f_n e^{-j2\pi u n \Delta T}$$

$$u = \frac{m}{M \Delta T}$$

Take $m = M-1, u = 1/\Delta T$

$$F[m] = \sum_{n=0}^{M-1} f_n e^{-j \frac{2\pi m n}{M}}, m = 0, 1, \dots, (M-1)$$

$m = 0, \dots, M-1$ ↑
Correction