

Friday, January 25, 2019 2:19 PM

Two-dimensional signals + systems

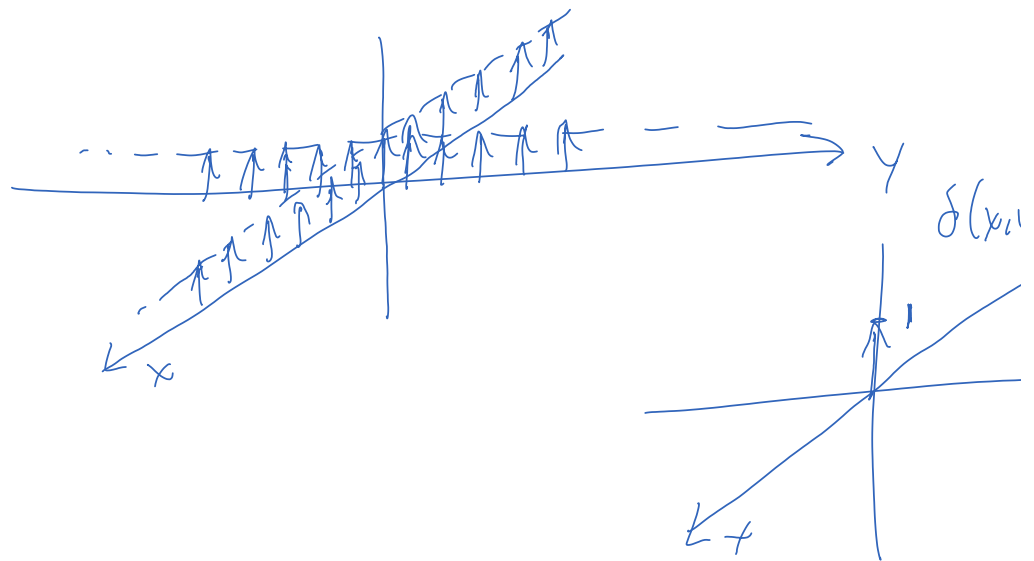
Continuous signals + systems; basic signals

Dirac delta: $\delta(x, y) = 0, x, y \neq 0$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \delta(x - x_0, y - y_0) dx dy = f(x_0, y_0)$$

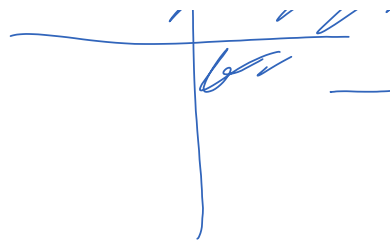
(sifting property)

$$\delta(x, y) = \delta(x) \delta(y)$$



step function:

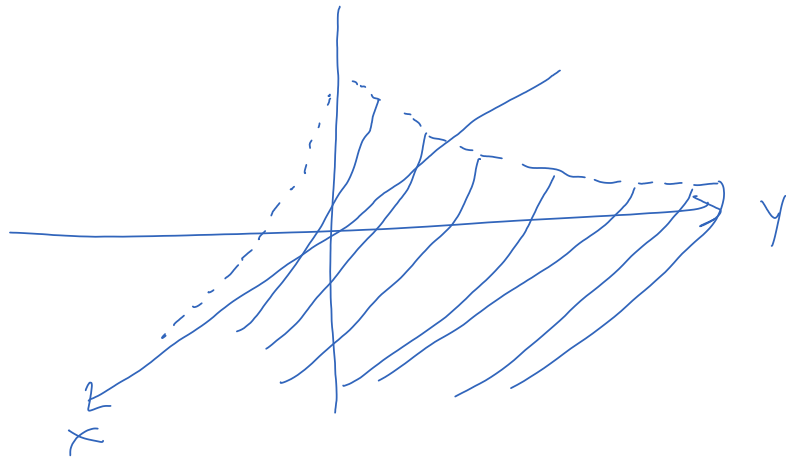
$$u(x, y) = \begin{cases} 1 & , x, y > 0 \\ 0 & , \text{otherwise} \end{cases}$$



$$= u(x) u(y)$$

decaying exponential:

$$f(x, y) = \begin{cases} \exp\{-\alpha x - \beta y\} \\ \alpha, \beta > 0 \end{cases} \quad \text{--- } e$$



sinusoid:

$$f(x, y) = \sin(\omega_x x + \omega_y y)$$

ω_x = frequency along x in

ω_y = " " y in

A separable signal is one that

factored into a product of 1-D's

$$\delta(x, y) = \delta(x) \delta(y)$$

$$u(x, y) = u(x) u(y)$$

$$\exp\{-\alpha x - \beta y\} u(x, y) = [e^{-\alpha x} u(x)] [e^{-\beta y} u(y)]$$

$$\underline{e^{j(\omega_{ox}x + \omega_{oy}y)}} = \left[e^{j\omega_{ox}x} \right]$$

Linear systems

If $T[\cdot]$ is a system, it is linear if

$$T[af(x, y) + bg(x, y)] = aT[f(x, y)] + bT[g(x, y)]$$

for all $a, b, f(x, y), g(x, y)$

- superposition

Ex: film with $T\left[\frac{f(x, y)}{I}\right] = \log(1 + I)$
- not linear

$$T\left[\frac{1}{1} + \frac{1}{1}\right] = \log(1 + 2) = \log 3 \neq$$

$$(1)T[1] + (1)T[1] = 2\log(1 + 1) = \log 4$$

\Rightarrow not linear

Ex: optical system that blurs horizontally

$$T[f(x,y)] = \int_0^1 f(x-x', y) dx'$$

show that system is linear

Must show that superposition holds for all inputs.

$$T[af(x,y) + bg(x,y)] = af(x-x', y) + bg(x-x', y)$$

$$= a \int_0^1 f(x-x', y) dx' + b \int_0^1 g(x-x', y) dx'$$

$$= a T[f(x,y)] + b T[g(x,y)]$$

\Rightarrow linear

Shift-invariant systems

$$g(x,y) = T[f(x,y)]$$

The system is SI iff

$$g(x-x_0, y-y_0) = T[f(x-x_0, y-y_0)]$$

for all shifts (x_0, y_0) and any $f(x,y)$.

Ex: $g(x, y) = T[f(x, y)] = f(x, y) u(x, y)$

Let $f(x, y) = u(x, y)$ Let $(x_0, y_0) = (-1, -1)$

$$T[f(x+1, y+1)] = f(x+1, y+1) u(x, y)$$

$$= u(x+1, y+1) u(x, y)$$

$$= u(x, y) u(x, y) - u(x, y)$$

$$g(x+1, y+1) = u(x+1, y+1)$$

$$\neq u(x, y) \Rightarrow \text{not SI}$$

Linear shift-invariant systems

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x', y') \delta(x-x', y-y') dx' dy'$$

For linear system,

$$g(x, y) = T[f(x, y)]$$

$$= T\left[\int \int f(x', y') \delta(x-x', y-y') dx' dy'\right]$$

$$= \int \int f(x', y') T[\delta(x-x', y-y')] dx' dy'$$

If we define $h(x, y) = T[\delta(x, y)]$, we have

$h(x-x', y-y') = T[\delta(x-x', y-y')]$ for all x', y'
for a shift-invariant system

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x', y') h(x-x', y-y') dx' dy'$$

$$= f(x, y) * h(x, y) \quad : \text{2-D convolution}$$

$h(x, y)$ is the impulse response

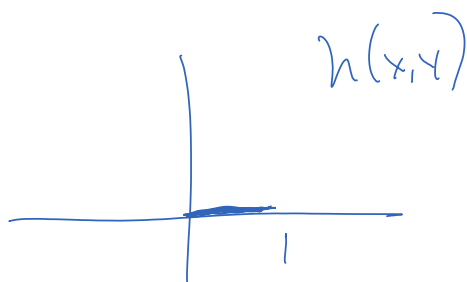
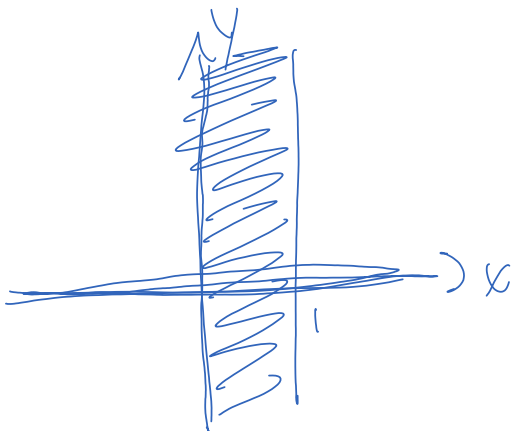
Ex: horizontal blur

$$h(x, y) = \int_0^1 \delta(x-x', y) dx'$$

$$= \int_0^1 \delta(x-x') \delta(y) dx'$$

$$= \delta(y) \int_0^1 \delta(x-x') dx'$$

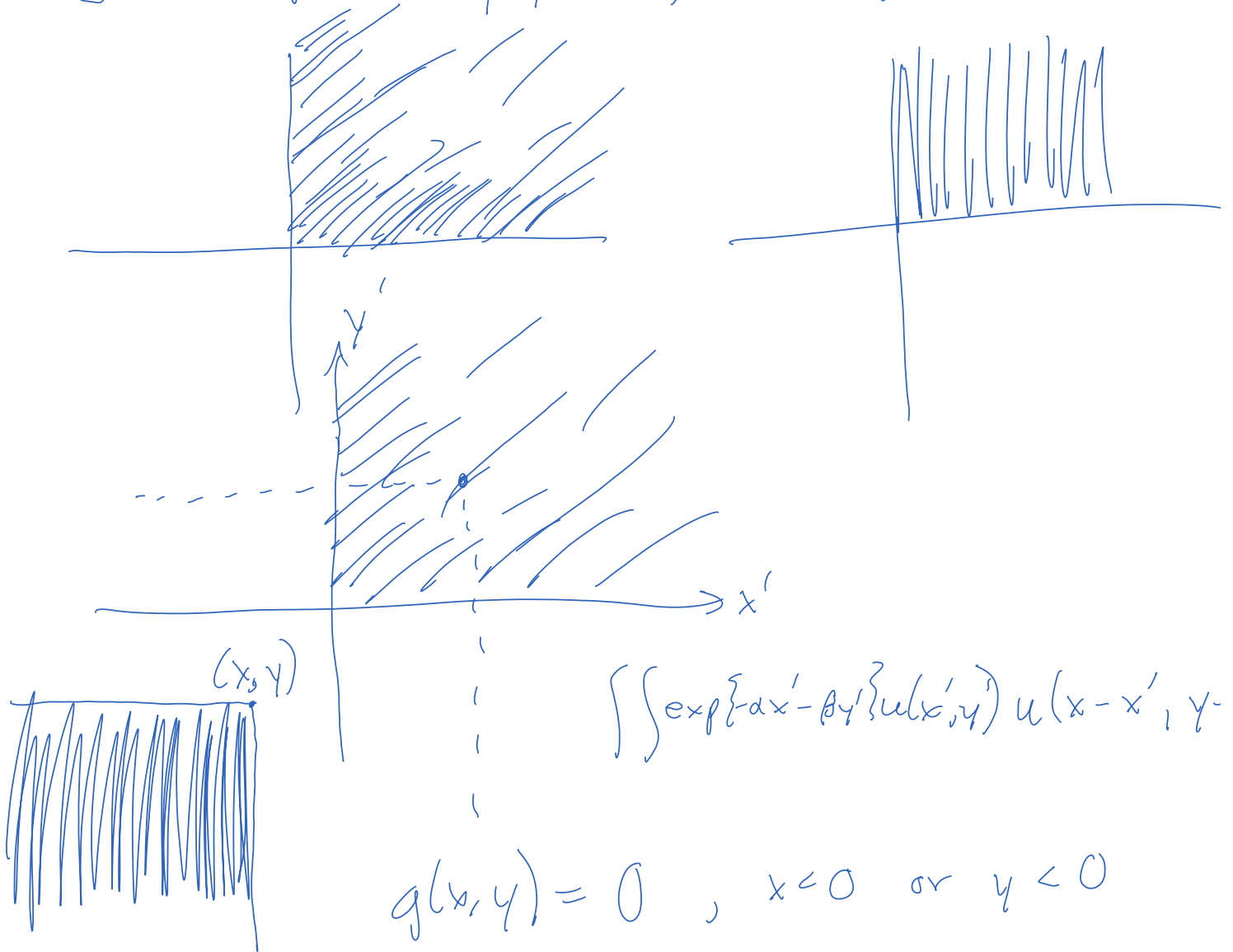
$$= \delta(y) [u(x) - u(x-1)]$$



Read 4.3-4.5

HW #1 - to be posted

Ex: $\exp\{-\alpha x - \beta y\} u(x, y) * u(x, y)$



$$\begin{aligned}
 g(x, y) &= \int_0^y \int_0^x \exp\{-\alpha x' - \beta y'\} dx' dy' \\
 &= \int_0^y e^{-\beta y'} dy' \int_0^x e^{-\alpha x'} dx'
 \end{aligned}$$

$$= -\frac{1}{\beta} e^{-\beta y'} \Big|_0^y - \frac{1}{\alpha} e^{-\alpha x'} \Big|_0^x$$

$$= \frac{1}{\alpha\beta} [e^{-\beta y} - 1][e^{-\alpha x} - 1]$$

$$g(x, y) = \frac{1}{\alpha\beta} [1 - e^{-\alpha x}][1 - e^{-\beta y}] u(x, y)$$

Fourier transform

$$F(\omega_x, \omega_y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp\{-j(\omega_x x + \omega_y y)\} dx dy$$

$$f(x, y) = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(\omega_x, \omega_y) \exp\{j(\omega_x x + \omega_y y)\} d\omega_x d\omega_y$$

$$f(x, y) \longleftrightarrow F(\omega_x, \omega_y)$$

$$\mathcal{F}\{f(x, y)\} = F(\omega_x, \omega_y)$$

Fourier transform properties

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x - x_0, y - y_0) \exp\{-j(\omega_x x + \omega_y y)\} dx dy$$

$$= \exp \{ -j(\omega_x x_0 + \omega_y y_0) \}$$

Linearity

$$af(x,y) + bg(x,y) \longleftrightarrow aF(\omega_x, \omega_y) + bG(\omega_x, \omega_y)$$

Convolution

$$f(x,y) * g(x,y) \longleftrightarrow F(\omega_x, \omega_y) G(\omega_x, \omega_y)$$

Multiplication

$$f(x,y) g(x,y) \longleftrightarrow \frac{1}{4\pi^2} F(\omega_x, \omega_y) * G(\omega_x, \omega_y)$$

Separability

$$F_y(\omega_x, y) = \int_{-\infty}^{\infty} f(x, y) e^{-j\omega_x x} dx$$

$$F(\omega_x, \omega_y) = \int_{-\infty}^{\infty} F_y(\omega_x, y) e^{-j\omega_y y} dy$$

$$= \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} f(x, y) e^{-j\omega_x x} dx \right] e^{-j\omega_y y} dy$$

$$= \iint f(x, y) \exp\{-j(\omega_x x + \omega_y y)\} dx dy$$

FT of separable signals

$$f(x, y) = f_x(x) f_y(y)$$

$$F(\omega_x, \omega_y) = F_x(\omega_x) F_y(\omega_y)$$

- FT is also separable

Shifts

$$f(x - x_0, y - y_0) \longleftrightarrow \exp\{-j(\omega_x x_0 + \omega_y y_0)\} F(\omega_x, \omega_y)$$

$$\exp\{j(\omega_{x0} x + \omega_{y0} y)\} f(x, y) \longleftrightarrow F(\omega_x - \omega_{x0}, \omega_y - \omega_{y0})$$

(modulation)

Scaling

$$f(ax, by) \longleftrightarrow \frac{1}{|ab|} F\left(\frac{\omega_x}{a}, \frac{\omega_y}{b}\right)$$

Parseval's theorem

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |f(x, y)|^2 dx dy = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |F(\omega_x, \omega_y)|^2 d\omega_x d\omega_y$$

Spatial derivatives

$$\frac{\partial f(x,y)}{\partial x} \longleftrightarrow -j\omega_x F(\omega_x, \omega_y)$$

$$\frac{\partial f(x,y)}{\partial y} \longleftrightarrow -j\omega_y F(\omega_x, \omega_y)$$

Laplacian:

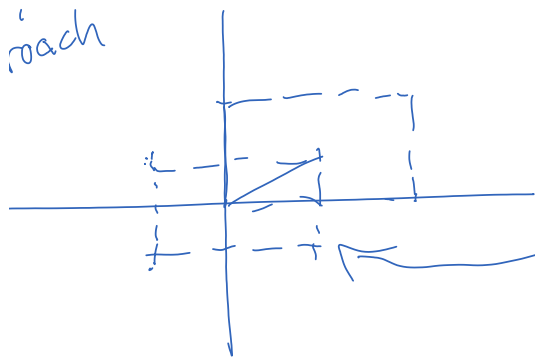
$$\frac{\partial^2 f(x,y)}{\partial x^2} + \frac{\partial^2 f(x,y)}{\partial y^2} \longleftrightarrow -(\omega_x^2 + \omega_y^2) F(\omega_x, \omega_y)$$

$$\text{Ex: } f(x,y) = \begin{cases} 1, & 0 \leq x \leq a, 0 \leq y \leq b \\ 0, & \text{otherwise} \end{cases}$$

$$\begin{aligned} F(\omega_x, \omega_y) &= \int_0^b \int_0^a \exp\{-j(\omega_x x + \omega_y y)\} dx dy \\ &= \left[\int_0^b e^{-j\omega_y y} dy \right] \left[\int_0^a e^{-j\omega_x x} dx \right] \\ &= -\frac{1}{j\omega_y} e^{-j\omega_y y} \Big|_0^b \cdot -\frac{1}{j\omega_x} e^{-j\omega_x x} \Big|_0^a \\ &= -\frac{1}{\omega_x \omega_y} (1 - e^{-j\omega_x a}) (1 - e^{-j\omega_y b}) \\ &= \frac{1}{\omega_x \omega_y} e^{-j\frac{\omega_x a}{2}} \left(e^{j\frac{\omega_x a}{2}} - e^{-j\frac{\omega_x a}{2}} \right) \dots \end{aligned}$$

$$= ab \exp \left\{ -j \left(\frac{a\omega_x}{2} + \frac{b\omega_y}{2} \right) \right\} \frac{\sin \frac{a\omega_x}{2}}{\frac{a\omega_x}{2}} \cdot \frac{\sin \frac{b\omega_y}{2}}{\frac{b\omega_y}{2}}$$

each



2-D centered pulse

⇒ separable

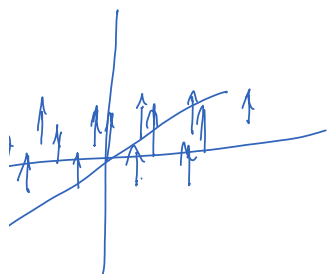
⇒ FT is prod. of
sinc functions
plus linear phase

∴ IFT of $\delta(\omega_x - \omega_{x0}, \omega_y - \omega_{y0})$

$$\begin{aligned} x, y) &= \frac{1}{4\pi^2} \int \int \delta(\omega_x - \omega_{x0}, \omega_y - \omega_{y0}) \exp \{ j(\omega_x x + \omega_y y) \} d\omega_x d\omega_y \\ &= \frac{1}{4\pi^2} \exp \{ j(\omega_{x0} x + \omega_{y0} y) \} \end{aligned}$$

$$1 \longleftrightarrow 4\pi^2 \delta(\omega_x, \omega_y)$$

$$f(x, y) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x - m\Delta x, y - n\Delta y)$$



$$= \sum_m \sum_n \delta(x - m\Delta x) \delta(y - n\Delta y)$$

$$= \left[\sum_m \delta(x - m\Delta x) \right] \left[\sum_n \delta(y - n\Delta y) \right]$$

$$= \left[\sum_m a_m \exp \left\{ j \frac{2\pi m}{\Delta x} x \right\} \right] \left[\sum_n b_n \exp \left\{ j \frac{2\pi n}{\Delta y} y \right\} \right]$$

$$i_m = \frac{1}{\Delta x} \int_{-\frac{\Delta x}{2}}^{\frac{\Delta x}{2}} \delta(x) \exp \left\{ -j \frac{2\pi m}{\Delta x} x \right\} dx$$

$$= \frac{1}{\Delta x} \quad \text{for all } m$$

$$b_n = \frac{1}{\Delta y} \quad \text{for all } n$$

$$\psi) = \left[\sum_m \frac{1}{\Delta x} \exp \left\{ j \frac{2\pi m}{\Delta x} x \right\} \right] \left[\sum_n \frac{1}{\Delta y} \exp \left\{ j \frac{2\pi n}{\Delta y} y \right\} \right]$$

$$= \frac{1}{\Delta x \Delta y} \sum_m \sum_n \exp \left\{ j 2\pi \left(\frac{m}{\Delta x} x + \frac{n}{\Delta y} y \right) \right\}$$

$$\psi_x, \omega_y) = \frac{4\pi^2}{\Delta x \Delta y} \sum_m \sum_n \delta \left(\omega_x - m \frac{2\pi}{\Delta x}, \omega_y - n \frac{2\pi}{\Delta y} \right)$$

$$f(x, y) = \cos(\omega_{x0}x + \omega_{y0}y)$$

$$= \frac{1}{2} \exp(j(\omega_{x0}x + \omega_{y0}y))$$

$$+ \frac{1}{2} \exp(-j(\omega_{x0}x + \omega_{y0}y))$$

$$F(\omega_x, \omega_y) = \frac{4\pi^2}{2} \delta(\omega_x - \omega_{x0}, \omega_y - \omega_{y0})$$

$$+ \frac{4\pi^2}{2} \delta(\omega_x + \omega_{x0}, \omega_y + \omega_{y0})$$

$$f(x, y) = \cos(\omega_{x_0} x + \omega_{y_0} y) \times [\delta(x - x_0, y - y_0) + \delta(x + x_0, y + y_0)]$$

$$= \cos(\omega_{x_0}(x - x_0) + \omega_{y_0}(y - y_0)) + \cos(\omega_{x_0}(x + x_0) + \omega_{y_0}(y + y_0))$$

using conv. thm.:

$$g(x, y) = \delta(x - x_0, y - y_0) + \delta(x + x_0, y + y_0)$$

$$\begin{aligned} \omega_x, \omega_y &= \exp\{-j(\omega_x x_0 + \omega_y y_0)\} + \exp\{j(\omega_x x_0 + \omega_y y_0)\} \\ &= 2 \cos(\omega_x x_0 + \omega_y y_0) \end{aligned}$$

$$\begin{aligned} F_x, \omega_y &= \left[\frac{1}{2} \delta(\omega_x - \omega_{x_0}, \omega_y - \omega_{y_0}) + \frac{1}{2} \delta(\omega_x + \omega_{x_0}, \omega_y + \omega_{y_0}) \right] \\ &\quad \times 2 \cos(\omega_x x_0 + \omega_y y_0) \end{aligned}$$

Sampling

$$s(x, y) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x - m\Delta x, y - n\Delta y)$$

sampled signal:

$$\begin{aligned} f_s(x, y) &= f(x, y) s(x, y) \\ &= f(x, y) \sum_m \sum_n \delta(x - m\Delta x, y - n\Delta y) \end{aligned}$$

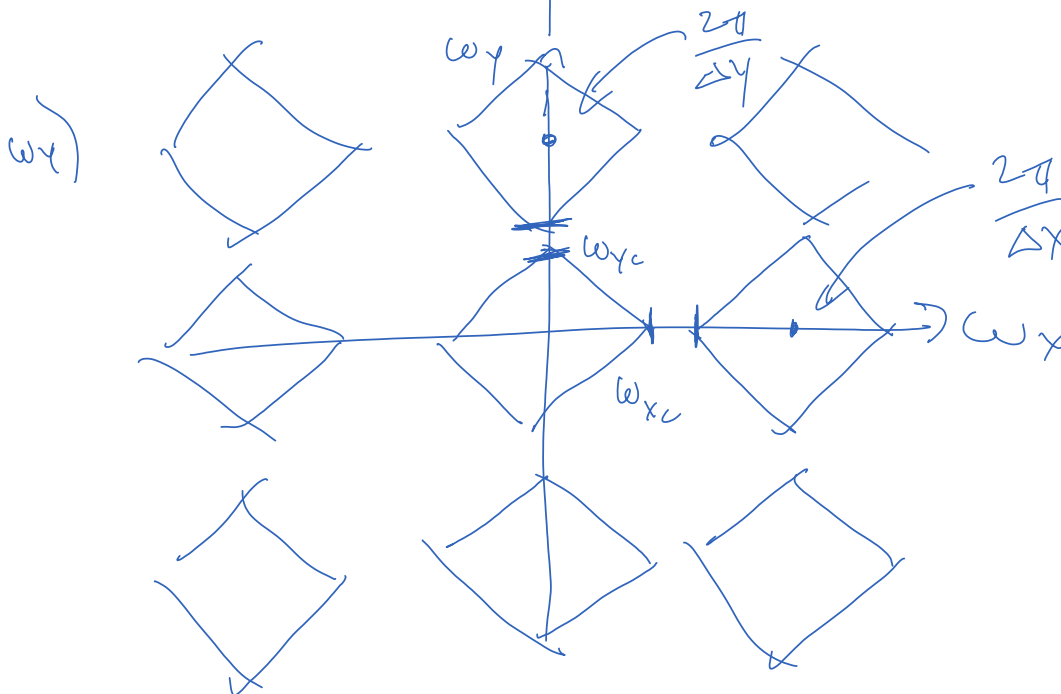
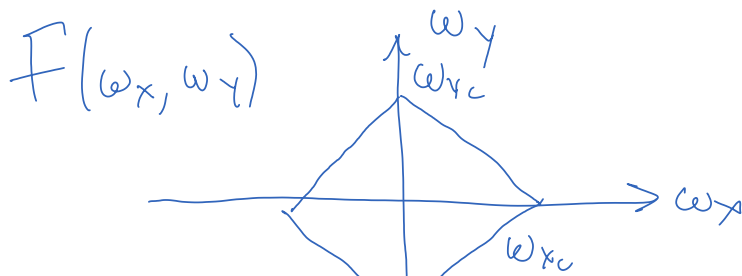
$$= \sum_m \sum_n f(x, y) \delta(x - m\Delta x, y - n\Delta y)$$

$$= \sum_m \sum_n f(m\Delta x, n\Delta y) \delta(x - m\Delta x, y - n\Delta y)$$

$$u_x, \omega_y) = \frac{1}{4\pi^2} F(\omega_x, \omega_y) * S(\omega_x, \omega_y)$$

$$= \frac{1}{4\pi^2} F(\omega_x, \omega_y) * \frac{4\pi^2}{\Delta x \Delta y} \sum_m \sum_n \delta(\omega_x - m \frac{2\pi}{\Delta x}, \omega_y - n \frac{2\pi}{\Delta y})$$

$$= \frac{1}{\Delta x \Delta y} \sum_m \sum_n F(\omega_x - m \frac{2\pi}{\Delta x}, \omega_y - n \frac{2\pi}{\Delta y})$$



$$\frac{2\pi}{\Delta x} - \omega_{xc} > \omega_{xc}$$

$$\frac{2\pi}{\Delta x} > 2\omega_{xc}$$

$$\frac{2\pi}{\Delta y} > 2\omega_{yc}$$

can recover $F(\omega_x, \omega_y)$ if the copies
 it overlap. If Δx and Δy become too
 large, then copies overlap. The original copy
 can then no longer be recovered.

\Rightarrow this is called spatial aliasing

Sampling Theorem

A bandlimited image $f(x, y)$ sampled on a
 uniform rectangular grid with spacing $\Delta x, \Delta y$
 can be recovered from the sample values
 $f(m\Delta x, n\Delta y)$ if

$$\frac{2\pi}{\Delta x} > 2\omega_{xc}$$

and

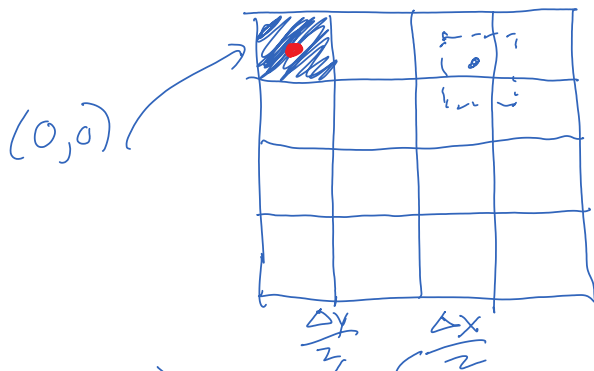
$$\frac{2\pi}{\Delta y} > 2\omega_{yc}$$

4.6-4.7

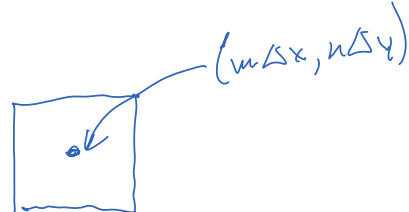
at 2 due Fri.

deal sampling

more realistic sampling model
presents the sampling process as
integrating intensity over rectangular
stches.



$$f(m\Delta x, n\Delta y) = \int_{-\frac{\Delta y}{2}}^{\frac{\Delta y}{2}} \int_{-\frac{\Delta x}{2}}^{\frac{\Delta x}{2}} f(m\Delta x - x', n\Delta y - y') dx' dy'$$



$$P(x, y) = \begin{cases} 1, & -\frac{\Delta x}{2} < x \leq \frac{\Delta x}{2}, \quad -\frac{\Delta y}{2} < y \leq \frac{\Delta y}{2} \\ 0, & \text{otherwise} \end{cases}$$

$$f(m\Delta x, n\Delta y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} P(x', y') f(m\Delta x - x', n\Delta y - y') dx' dy'$$

$$\begin{aligned}
 x, y) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \Pi(x', y') f(x-x', y-y') dx' dy' \\
 &= f(x, y) * \Pi(x, y)
 \end{aligned}$$

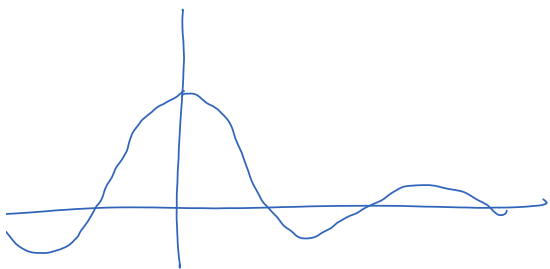
\uparrow
 original scene

$$i(x, y) = f_i(x, y) s(x, y)$$

$$= \sum_m \sum_n f_i(m\Delta x, n\Delta y) \delta(x-m\Delta x, y-n\Delta y)$$

\Rightarrow filtering followed by ideal sampling

$$\begin{aligned}
 \tilde{i}(\omega_x, \omega_y) &= F(\omega_x, \omega_y) \mathcal{F}\{\Pi(x, y)\} \\
 &= F(\omega_x, \omega_y) \frac{\sin \frac{\Delta x}{2} \omega_x}{\frac{\Delta x}{2} \omega_x} \cdot \frac{\sin \frac{\Delta y}{2} \omega_y}{\frac{\Delta y}{2} \omega_y}
 \end{aligned}$$



* We're sampling the filtered image instead of the original

* filtered signal has higher frequencies suppressed

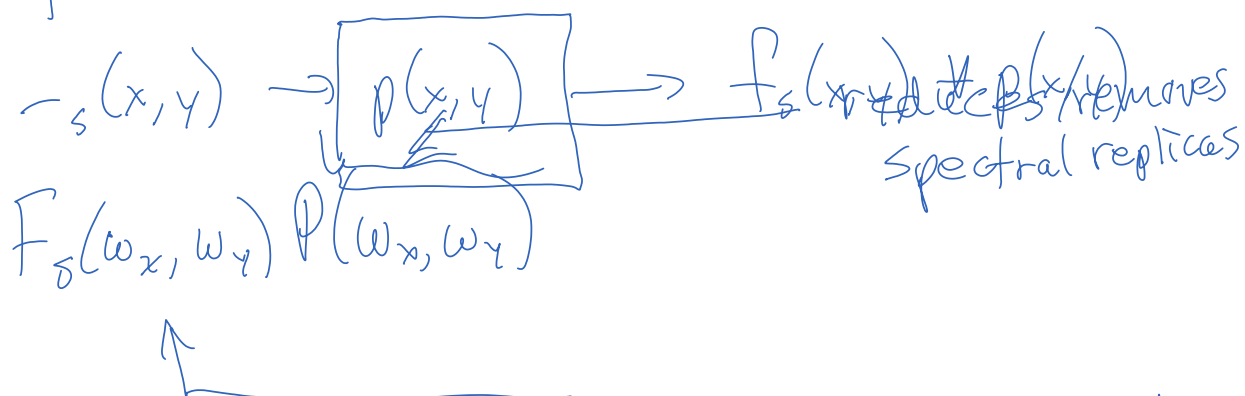
- aliasing will be reduced but not eliminated
- image will be slightly blurred

lay/reconstruction

DSP, reconstruction is done in concept by taking an impulse train from a sequence, then lowpass filtering. This is implemented using electronics.

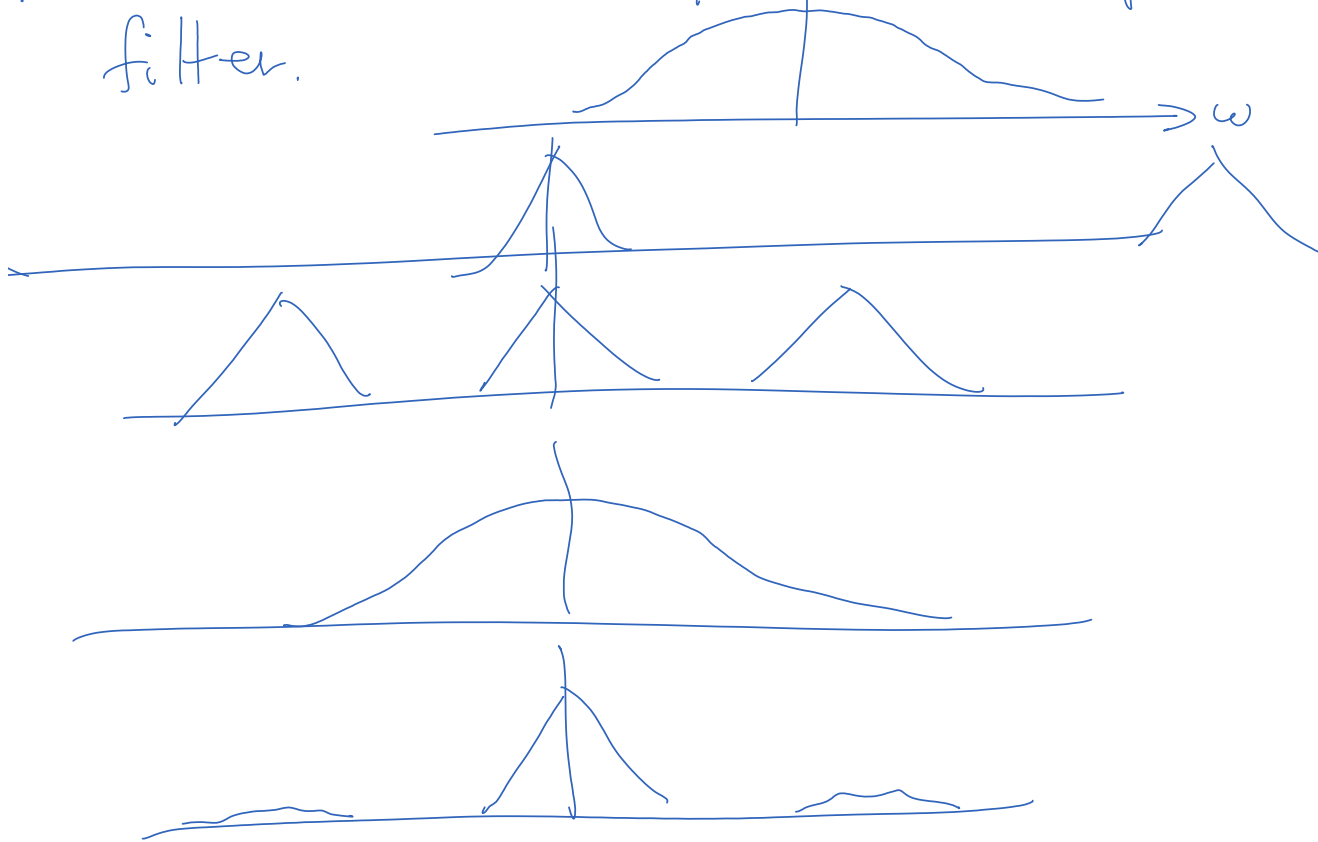
image processing, the display is the filter. (The HVS is also a filter.) Samples are projected as tiny rectangular patches (LCD display), gaussian spots (CRT), etc.

Optical display "filter" can be modeled as —



periodically replicated

Think of human visual system as a lowpass filter.



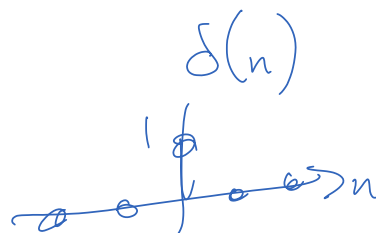
sig signals

delta (impulse, unit sample)

Kronecker delta, not Dirac

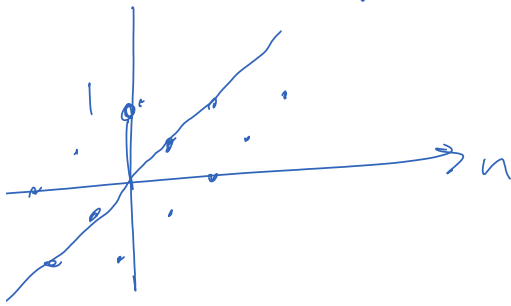
$$\delta(m, n) = \begin{cases} 1 & , \quad m = n = 0 \\ 0 & , \quad \text{otherwise} \end{cases}$$

$$= \delta(m) \delta(n)$$



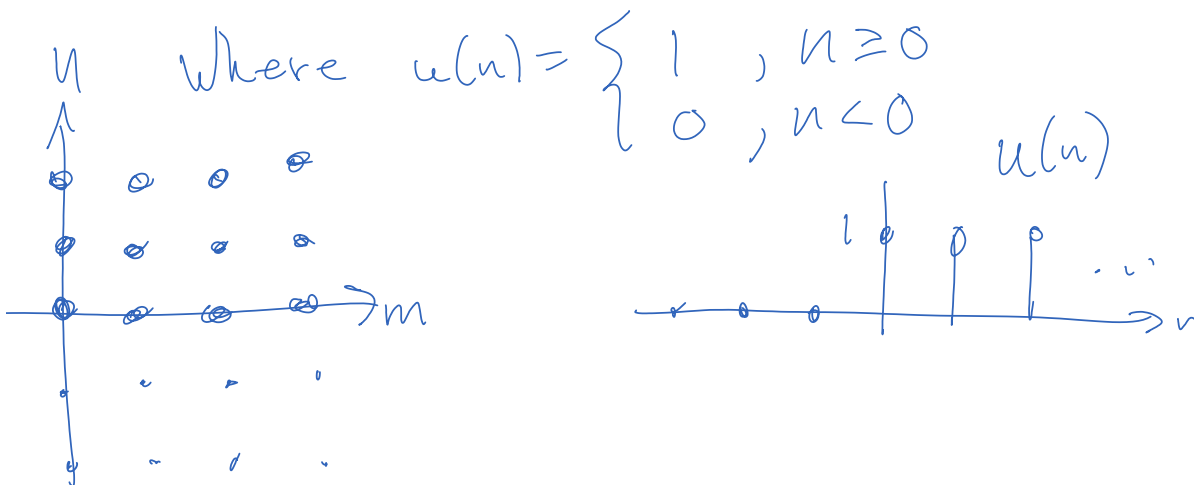
Σ

where $\delta(n) = \begin{cases} 1, & n=0 \\ 0, & n \neq 0 \end{cases}$



$\Rightarrow \rho \quad u(m, n) = \begin{cases} 1, & m, n \geq 0 \\ 0, & \text{otherwise} \end{cases}$

$$= u(m) u(n)$$



4.8-4.9

posted

ponential

$$f(m, n) = \exp \left\{ \underbrace{-\alpha m}_{-\alpha m} - \underbrace{\beta n}_{-\beta n} \right\} u(m, n)$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} u(x, y) \left[\int_{-\infty}^{\infty} u(x, y) \right] dx dy$$

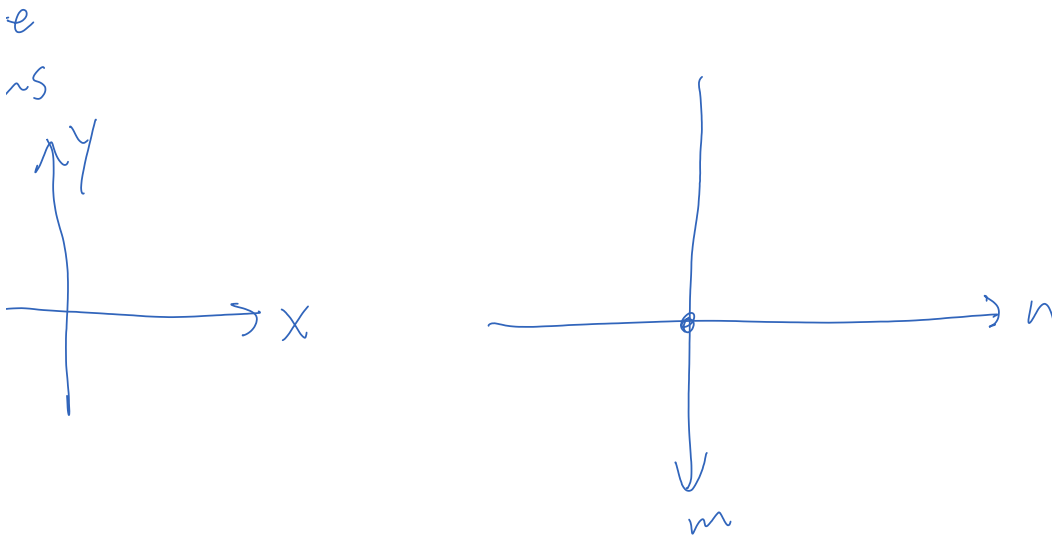
$\alpha, \beta \geq 0$ to prevent blowing up

said

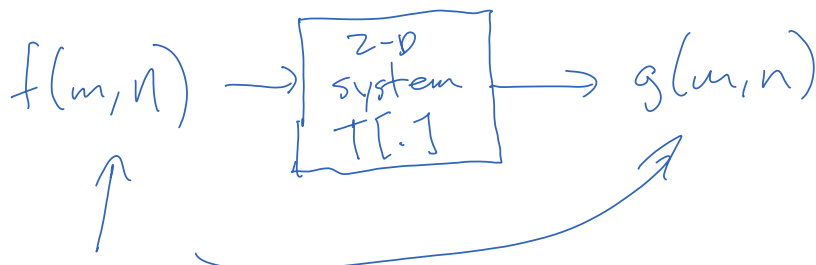
$$f(m, n) = f(x, y) \Big|_{(x, y) = (m \Delta x, n \Delta y)}$$

$$= \sin(\underbrace{\omega_x \Delta x}_m m + \underbrace{\omega_y \Delta y}_n n)$$

$$= \sin(\omega_m m + \omega_n n)$$



1 systems



Integer
index values

-y

$$[af(m,n) + bg(m,n)] = aT[f(m,n)] + bT[g(m,n)]$$

for all $a, b, f(m,n), g(m,n)$

invariance

$$[f(m-k, n-l)] = g(m-k, n-l)$$

for all $k, l, f(m,n)$

(k, l must be integers)

— can be represented by a convolution

sum

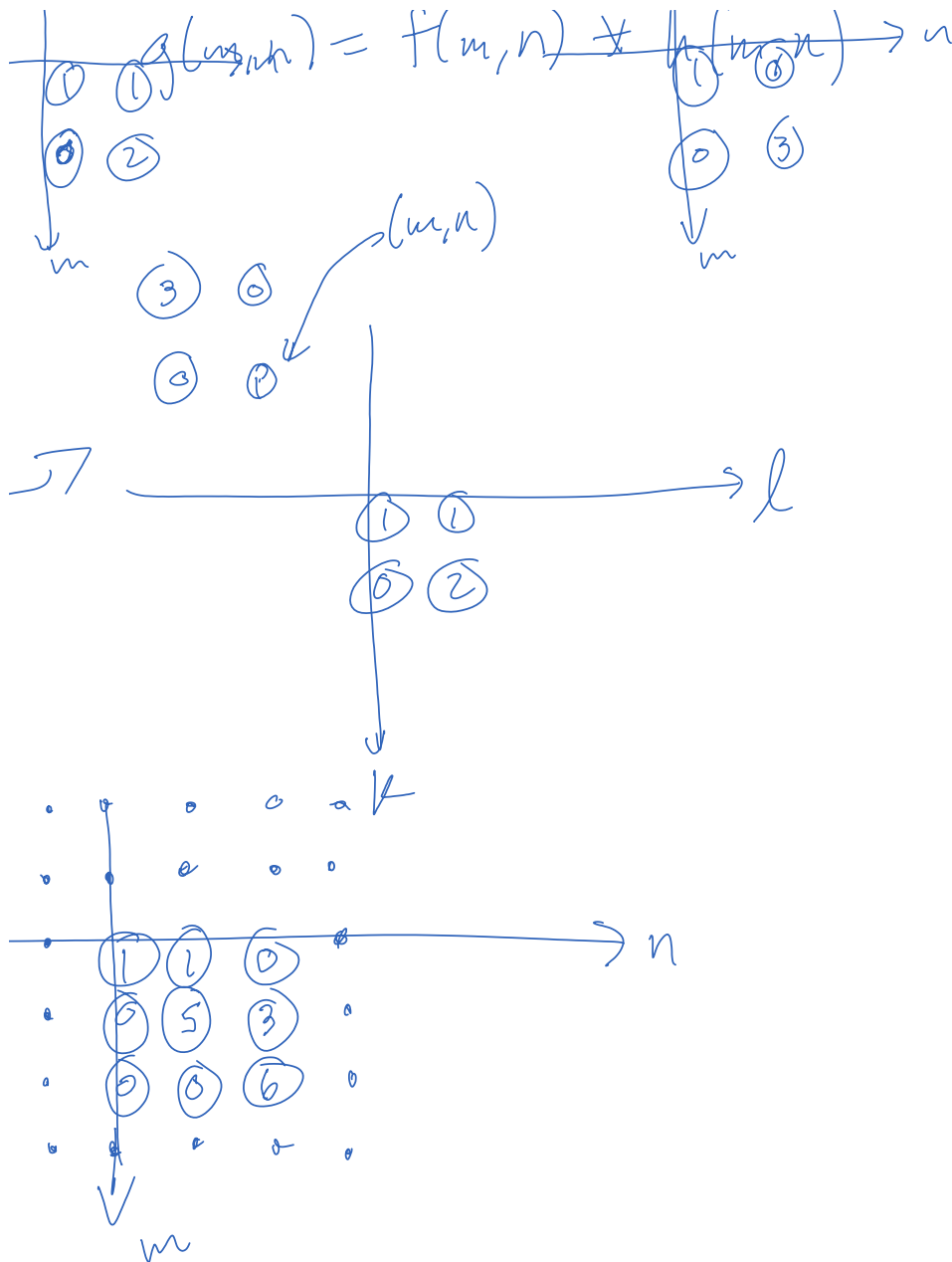
$\infty \infty$

$$f(m,n) = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} f(k,l) \delta(m-k, n-l)$$

$$g(m,n) = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} f(k,l) h(m-k, n-l)$$

where $h(m,n) = T[\delta(m,n)]$

is the impulse response
 $h(m,n)$



Transform of 2-D sequence

$$x(n) = X(\omega) = \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega n}$$

ω is in rad/sample

$$w_n) = \sum_{n=-\infty}^{\infty} x(m,n) e^{-j\omega n} ; \text{ 1-D FT of } x(m,n) \text{ along row } m$$

$$\begin{aligned}
 n) &= \sum_{m=-\infty}^{\infty} X_u(m; \omega_n) e^{-j\omega_m m} e^{-j(\omega_m m + \omega_n n)} \\
 &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} x(m, n) e
 \end{aligned}$$

$$\frac{1}{(2\pi)^2} \int_{-\pi}^{\pi} \int_{-\pi}^{\pi} X(\omega_m, \omega_n) e^{j(\omega_m m + \omega_n n)} d\omega_m d\omega_n$$

$$= X(\omega_m - 2\pi, \omega_n) = X(\omega_m, \omega_n - 2\pi) = X(\omega_m - 2\pi, \omega_n - 2\pi)$$

impulse response gives frequency
of system

$$H(\omega_m, \omega_n) = \mathcal{F}\{h(m, n)\}$$

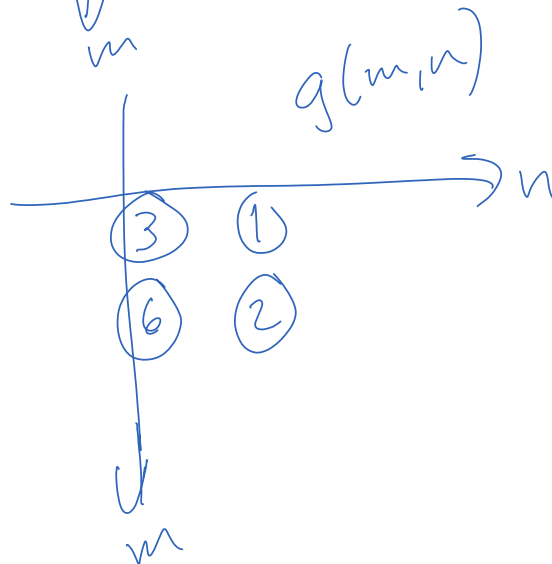
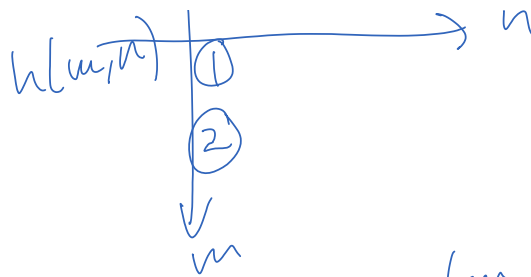
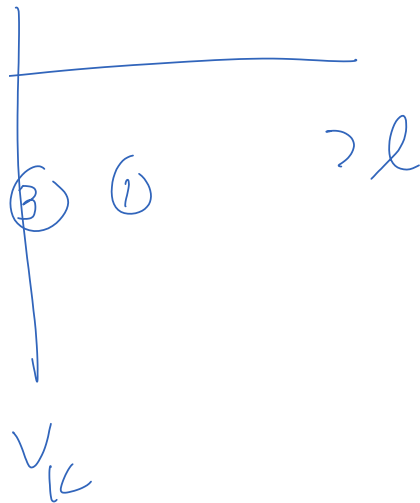
$$Y(\omega_m, \omega_n) = F(\omega_m, \omega_n) H(\omega_m, \omega_n)$$

- convolution theorem

$$x * h(m, n) \longrightarrow F(\omega_m, \omega_n) H(\omega_m, \omega_n)$$

$$t(m, n) = 3\delta(m, n) + \delta(m, n-1)$$

$$g(m, n) = \delta(m, n) + 2\delta(m-1, n)$$



$$\delta(m-l, n-l) e^{-j(\omega_m m + \omega_n n)}$$

$$(\omega_m k + \omega_n l) \quad F(\omega_m, \omega_n) = 3 + e^{-j\omega_n}$$

$$H(\omega_m, \omega_n) = 1 + 2e^{-j\omega_m}$$

$$Y(\omega_m, \omega_n) = F(\omega_m, \omega_n) H(\omega_m, \omega_n)$$

$$= (3 + e^{-j\omega_n}) (1 + 2e^{-j\omega_m})$$

$$= 3 + e^{-j\omega_n} + 6e^{-j\omega_m} + 2e^{-j(\omega_m + \omega_n)}$$

$$3\delta(m, n) + \delta(m, n-1) + 6\delta(m-1, n) + 2\delta(m-1, n-1)$$

: of discrete-space FT:

e — can decompose into $\rightarrow (\omega_m k + \omega_n l)$ two 1-D FTs

$$f(m-k, n-l) \longleftrightarrow F(\omega_m, \omega_n) \otimes$$

$$\text{on } f(m,n)g(m,n) \longleftrightarrow \frac{1}{4\pi^2} F(\omega_m, \omega_n) * G(\omega_m, \omega_n)$$

||

ation thm.

als theorem

$\pi \pi$

$$\int_{-\pi}^{\pi} \int_{-\pi}^{\pi} |F(\omega_m, \omega_n)|^2 d\omega_m d\omega_n$$

$$\sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} |f(m,n)|^2 = \frac{1}{4\pi^2} \int_{-\pi}^{\pi} \int_{-\pi}^{\pi} |F(\omega_m, \omega_n)|^2 d\omega_m d\omega_n$$

i function and can't be stored in a

. Integrals can't be calculated perfectly
computer

ete Fourier transform (DFT)

image,

$$\left(\frac{2\pi k}{M}, \frac{2\pi l}{N} \right)$$

$$.) = F(\omega_m, \omega_n)$$

$$(\omega_m, \omega_n) =$$

$$0 \leq k \leq M-1, \quad 0 \leq l \leq N-1$$

$$= \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m,n) \exp \left\{ j \left(\frac{2\pi k m}{M} + \frac{2\pi l n}{N} \right) \right\} \quad \text{- DFT}$$

$$f(m,n) = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} F(k,l) \exp \left\{ j \left(\frac{2\pi k m}{M} + \frac{2\pi l n}{N} \right) \right\}$$

So:

v

Parseval's theorem

$$\sum \sum |f(m,n)|^2 = \frac{1}{MN} \sum \sum |F(k,l)|^2$$

$$\sum \sum |f(m,n) - g(m,n)|^2 = \frac{1}{MN} \sum \sum |F(k,l) - G(k,l)|^2$$

separability - Can take 1-D DFT of rows
 & then 1-D DFTs of columns (or reverse)

seq. is separable, then DFT is separable.

$$f(m,n) = f_m(m) f_n(n)$$

periodicity

$$f(k+M, l+N) = F(k, l+N) = F(k+M, l) = F(k, l)$$

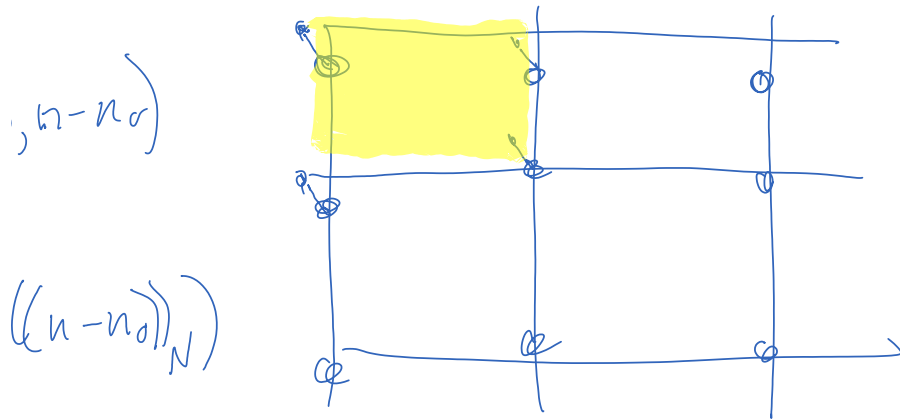
$$f(m+nM, n+N) = f(m, n+N) = f(m+M, n) = f(m, n)$$

Circular shift

$$f((n-n_0)_N) \longleftrightarrow F(k, l) \exp\left\{-j\left(\frac{2\pi k m_0}{M} + \frac{2\pi l n_0}{N}\right)\right\}$$

$$((m))_M = m \bmod M$$

$$\text{exp}\left\{-j\left(\frac{2\pi k m_0}{M} + \frac{2\pi l n_0}{N}\right)\right\}$$



convolution

$$f((m))_M \longleftrightarrow \boxed{F(k, l) H(k, l)}$$

$$f((m))_M = \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} f(k, l) h((m-k)_M, (n-l)_N)$$

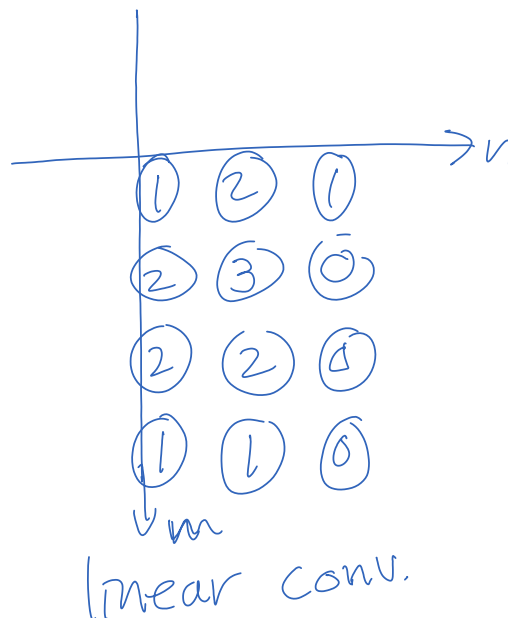
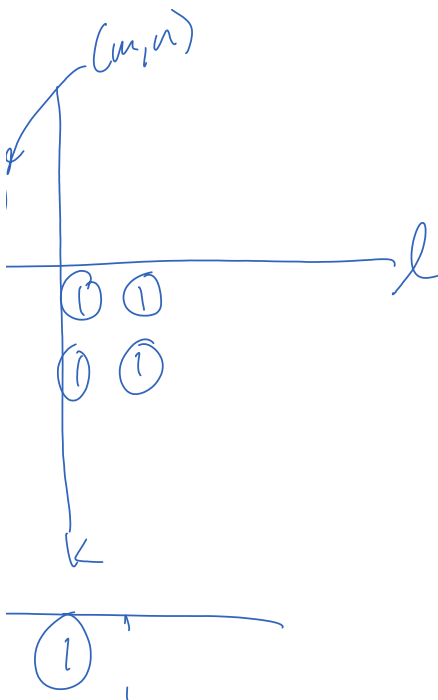
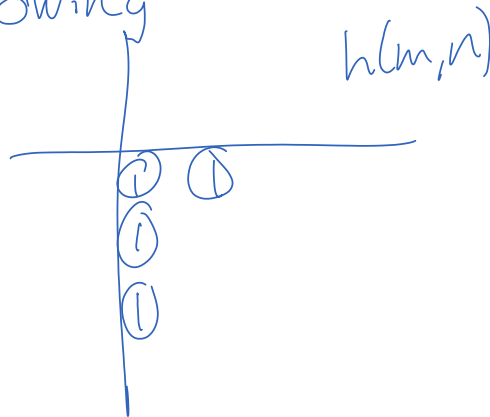
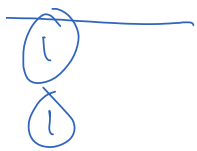
i.e. a linear convolution of $f(m, n)$ with a periodic version of $h(m, n)$

size is $g(m,n)$? $M \times N$

size is $f \times h$? $(M_f + M_n - 1) \times (N_f + N_n - 1)$

implement circular convolution by
 periodic extension of one of the signals

implement by linear convolution
 followed by periodic extension of the
 result, then windowing





Circ. CONV.

convolution w/ DFTs:

let $h(m,n)$ be $M_n \times N_n$

$f(m,n)$ be $M_f \times N_f$

pad $h(m,n)$ and $f(m,n)$ to be
 $\geq (M_f + M_n - 1) \times (N_f + N_n - 1)$

DFTs of padded seg's.

multiply DFTs pointwise

$$H \times F$$

take IDFT of result

Fourier transform (FFT)

efficient DFT implementation

replaced by decomposing DFT into
 sum of small DFTs

- requires $M^2 N^2$ multiplies
- requires $MN \log_2 MN$ multiplies

224 x 1024,

$$\text{DFT} = 10^{12} \text{ mults.}$$

$$\text{FFT} = \frac{10^7}{2^{\frac{2\pi k m}{N}} \exp\left\{-j \frac{2\pi k m}{N}\right\}}$$

(1 day vs. 1 sec)

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

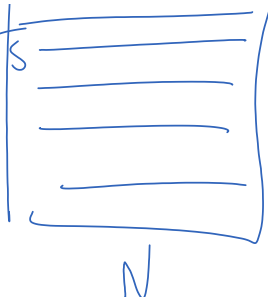
$$\sum_{m=0}^{M-1} \left\{ \sum_{n=0}^{N-1} f(m,n) \exp \left\{ -j \frac{2\pi k n}{N} \right\} \right\}$$

1-D DFT

⇒ row-column decomposition

1-D FFT of all rows

Then 1-D FFT of columns of result

th DFT requires N^2 mults
 es
 row-col w/ DFTs M 

$$+ N \cdot M^2 = NM(M+N)$$

compare to $M^2 N^2$

1-D FFTs

(each requires $\frac{1}{2} N \log_2 N$ mults)

or R-C w/ FFTs

$$\frac{1}{2} MN \log_2 N + \frac{1}{2} NM \log_2 M$$

$$= \frac{1}{2} MN \log_2 MN$$

1×10^24 ,

$$\text{direct DFT} = 2^{40} \approx 10^{12} \text{ mults}$$

$$\text{R-C DFT} = 2^{31} \approx 2 \times 10^9$$

$$\text{R-C FFT} = 10 \times 2^{20} \approx 10^7$$

radix FFT

1) divide-and-conquer strategy)

DFT is divided into successively
aller 2-D DFTs

multiplies is $\frac{3}{8} N^2 \log_2 N^2$

compared to $\frac{1}{2} N^2 \log_2 N^2$ for R-C FFT

* more complex than R-C FFT