

# **Medical Diagnostic Imaging**

**ECE16511/411**

**Spring 2007**

Signals and Systems

Prof Mufeed MahD

UMASS LOWELL

# **Introduction**

## **Signals**

mathematical functions of one or more independent variables, capable of modeling a variety of physical processes.

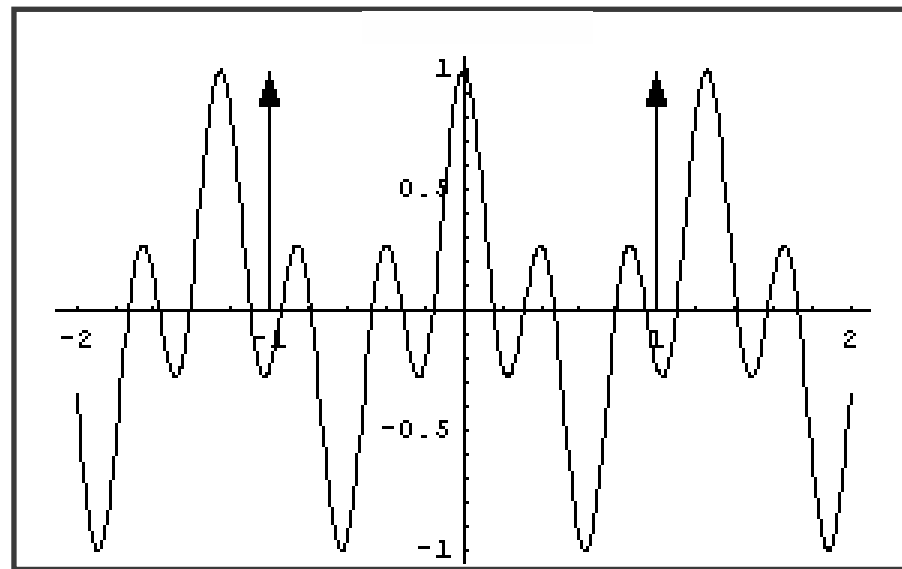
## **Systems**

respond to signals by producing new signals.

**Signals can be classified into three categories:**

**1) Continuous Signal:**

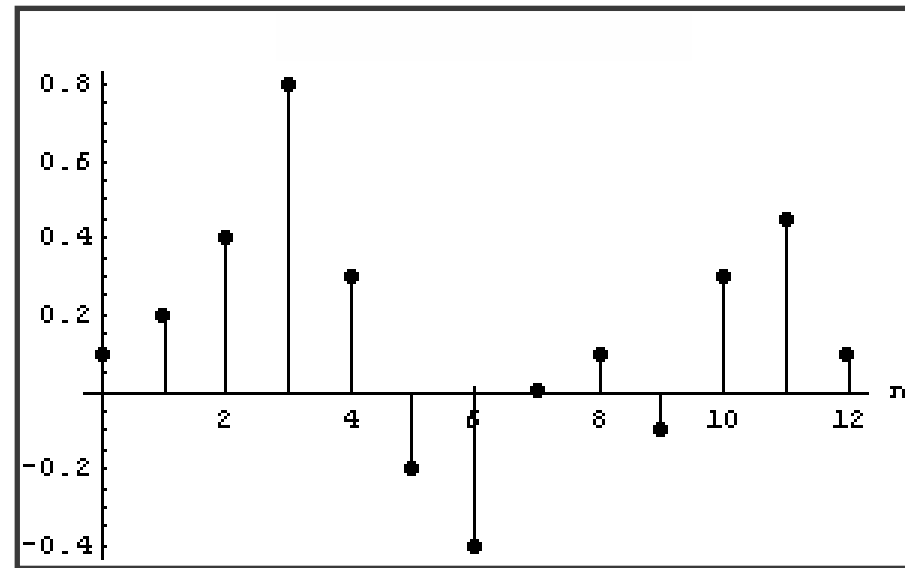
**is a function of independent variables that range over a continuum of values.**



**Signals can be classified into three categories:**

**2) Discrete signal:**

**is a function of independent variables that range over discrete values**



**3) Mixed signal:**

**is a function of some continuous and some discrete independent variables**

**Systems can also be classified into three categories.**

**1) Continuous-to-continuous system:**

**responds to a continuous signal by producing a continuous signal**

**2) Continuous-to-discrete system:**

**responds to a continuous signal by producing a discrete signal.**

**3) Discrete-to-discrete system:**

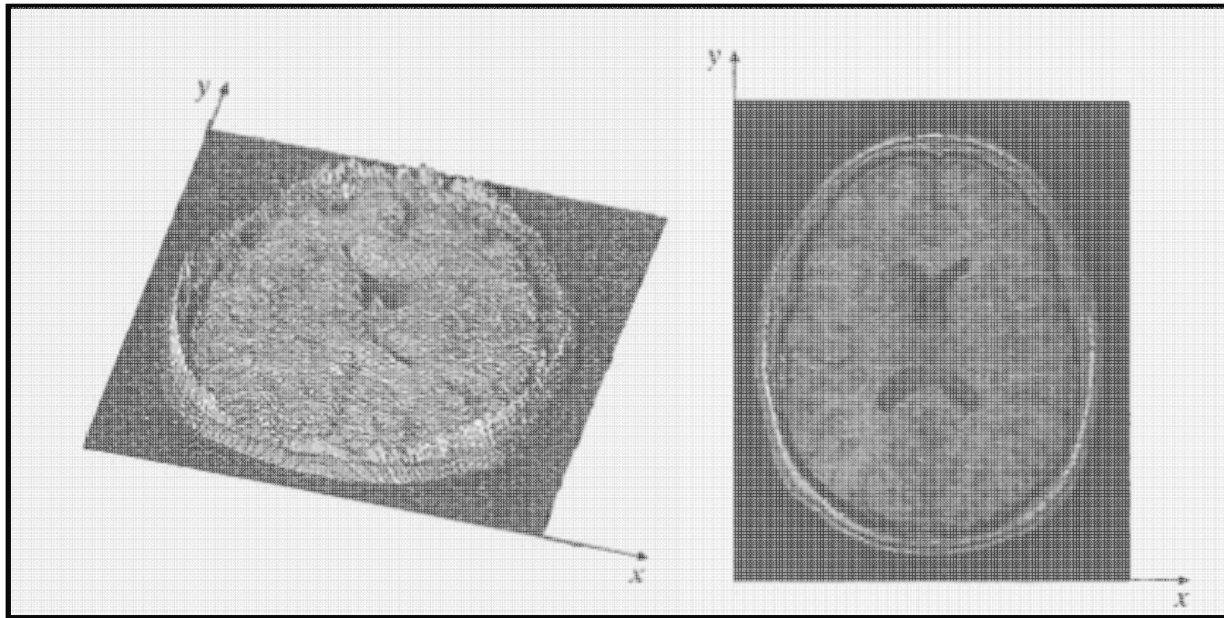
**responds to a discrete signal by producing a discrete signal.**

# Signals

2-D continuous signal is defined as

$$f(x, y), \quad -\infty \leq x, y \leq \infty$$

This signal can be represented (visualized) in two different ways.

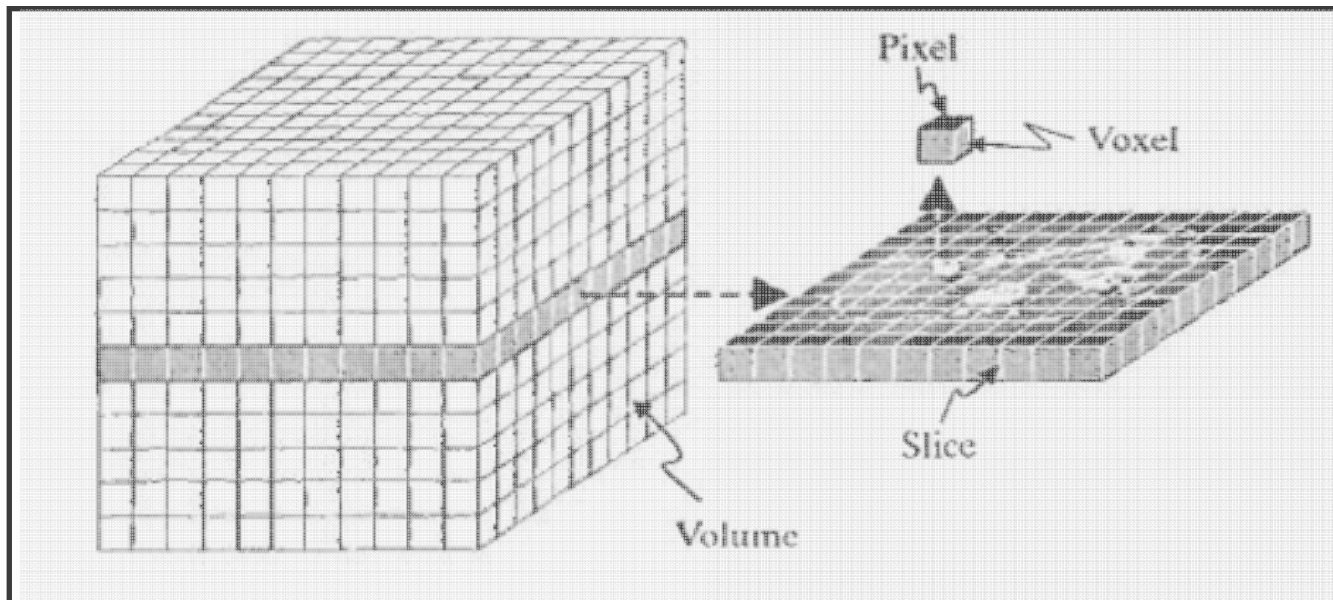


Functional Plot

Image display

# Signals

2D and 3D display



## Point Impulse

Point Impulse is used to model the concept of a point source, which is used in the characterization of imaging system resolution.

The point impulse is also known as the *delta function*, the *Dirac function* and the *impulse function*.

1D point impulse is defined as:

$$\delta(x) = 0, \quad x \neq 0,$$
$$\int_{-\infty}^{\infty} f(x)\delta(x) dx = f(0).$$

2D point impulse is defined as :

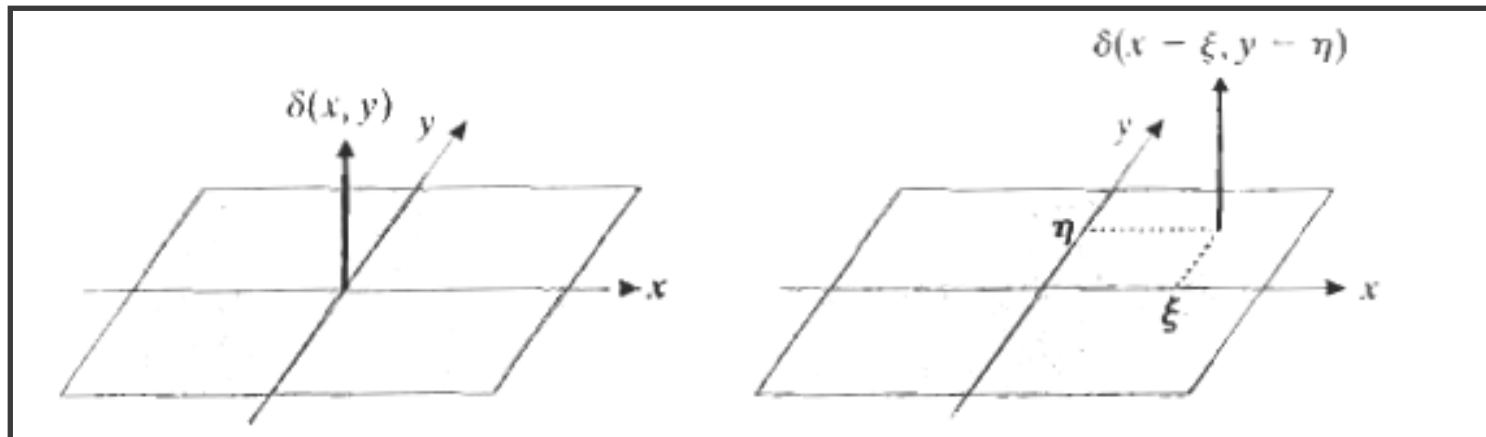
$$\delta(x, y) = 0, \quad (x, y) \neq (0, 0),$$
$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y)\delta(x, y) dx dy = f(0, 0).$$



## Point Impulse

Shifted 2-D point impulse is defined as:

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



## Point Impulse

Properties:

- 1) The product of a function with a point impulse as is another point impulse whose volume is equal to the value of the function at the location of the point impulse.

$$g(x, y) = f(x, y)\delta(x - \xi, y - \eta) = f(\xi, \eta)\delta(x - \xi, y - \eta)$$

- 2) Scaling

$$\delta(ax, by) = \frac{1}{|ab|} \delta(x, y)$$

- 3) Even Function

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

## Line Impulse

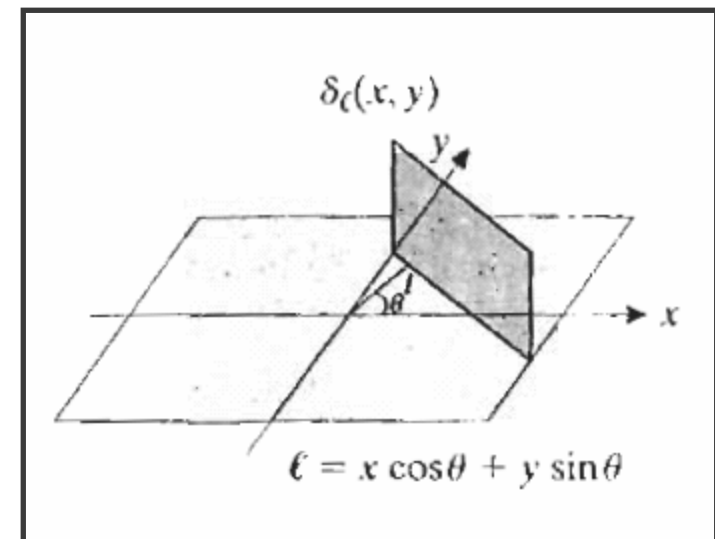
A set of points given by

$$L(\ell, \theta) = \{(x, y) \mid x \cos \theta + y \sin \theta = \ell\}$$

Defines a line whose unit normal is oriented at an angle  $\theta$  relative to the x-axis and is at distance  $\ell$  from the origin in the direction of the unit normal.

The line impulse  $\delta_\ell(x, y)$  associated with line  $L(i, \theta)$  is given by

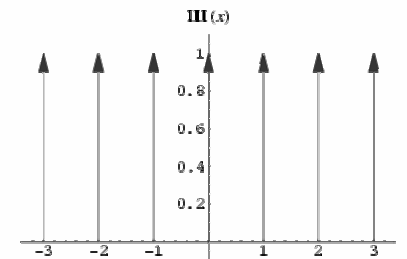
$$\delta_\ell(x, y) = \delta(x \cos \theta + y \sin \theta - \ell)$$



# Comb (Shah) Function

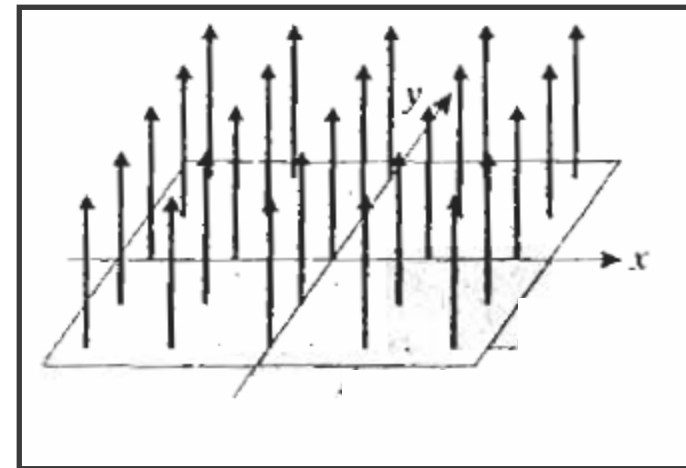
1D Comb function is

$$\text{comb}(x) = \sum_{n=-\infty}^{\infty} \delta(x - n)$$



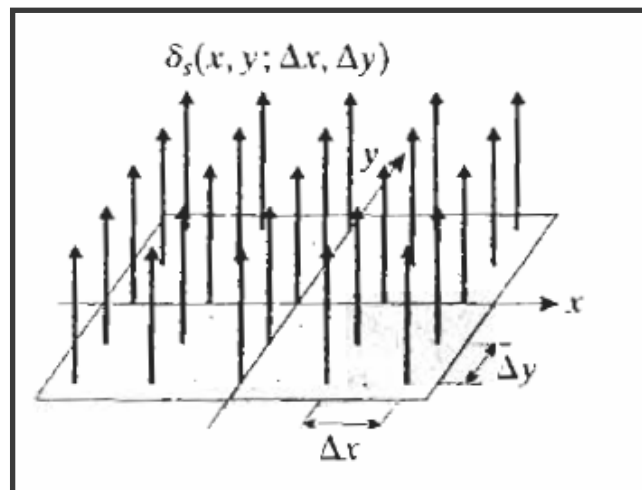
2D Comb function is

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



## The sampling function

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



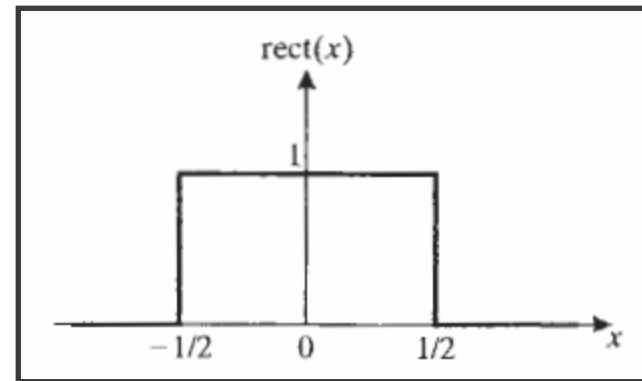
The sampling function related to the comb function is

$$\delta_s(x, y; \Delta x, \Delta y) = \frac{1}{\Delta x \Delta y} \text{comb} \left( \frac{x}{\Delta x}, \frac{y}{\Delta y} \right)$$

## Rect Functions

The 1D rect function is given by

$$\text{rect}(x) = \begin{cases} 1, & \text{for } |x| < \frac{1}{2} \\ 0, & \text{for } |x| > \frac{1}{2} \end{cases}.$$



The 2D rect function is given by

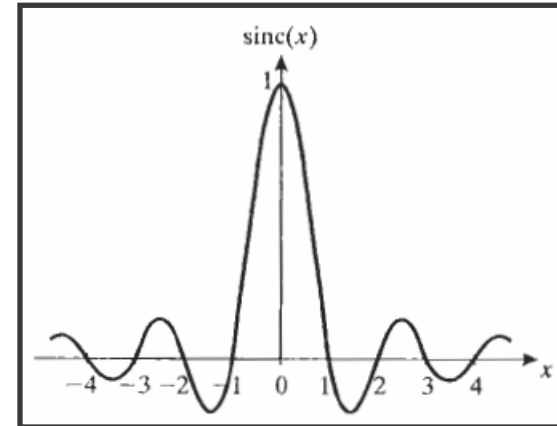
$$\text{rect}(x, y) = \begin{cases} 1, & \text{for } |x| < \frac{1}{2} \text{ and } |y| < \frac{1}{2} \\ 0, & \text{for } |x| > \frac{1}{2} \text{ or } |y| > \frac{1}{2} \end{cases}$$

$$\text{rect}(x, y) = \text{rect}(x) \text{rect}(y)$$

## Sinc Functions

The 1D sinc function is given by

$$\text{sinc}(x) = \frac{\sin(\pi x)}{\pi x}$$



The 2D *sinc function* is given by

$$\text{sinc}(x, y) = \begin{cases} 1, & \text{for } x = y = 0 \\ \frac{\sin(\pi x) \sin(\pi y)}{\pi^2 xy}, & \text{otherwise.} \end{cases}$$

$$\text{sinc}(x, y) = \text{sinc}(x) \text{sinc}(y)$$

## Exponential and Sinusoidal Signals

The complex exponential signal is

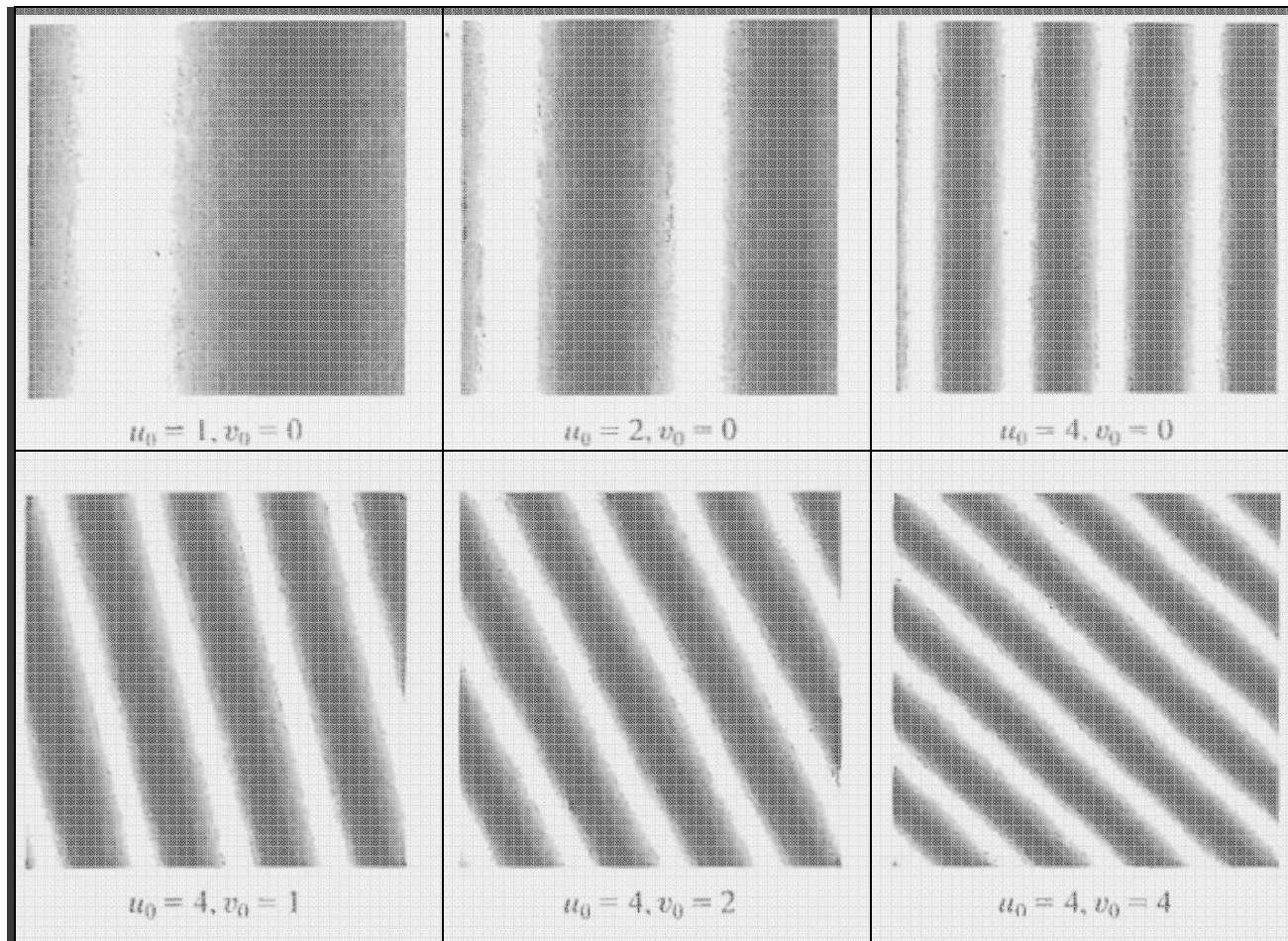
$$\begin{aligned} e(x, y) &= e^{j2\pi(u_0x + v_0y)} \\ &= \cos[2\pi(u_0x + v_0y)] + j \sin[2\pi(u_0x + v_0y)] \\ &= c(x, y) + js(x, y). \end{aligned}$$

where

$$s(x, y) = \sin[2\pi(u_0x + v_0y)] \quad \text{and} \quad c(x, y) = \cos[2\pi(u_0x + v_0y)]$$



## Exponential and Sinusoidal Signals



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

## Separable Signals

A 2D signal  $f(x, y)$  is *separable* if there exist two 1D signals  $f_1(x)$  and  $f_2(y)$  such that

$$f(x, y) = f_1(x)f_2(y)$$

$$\text{rect}(x, y) = \text{rect}(x) \text{rect}(y)$$

$$\text{sinc}(x, y) = \text{sinc}(x) \text{sinc}(y)$$

Operating on separable signals is much simpler than operating on purely 2D signals.

## Periodic Signals

A signal  $f(x, y)$  is a *periodic signal* if there exist two positive constants  $X$  and  $Y$  such that

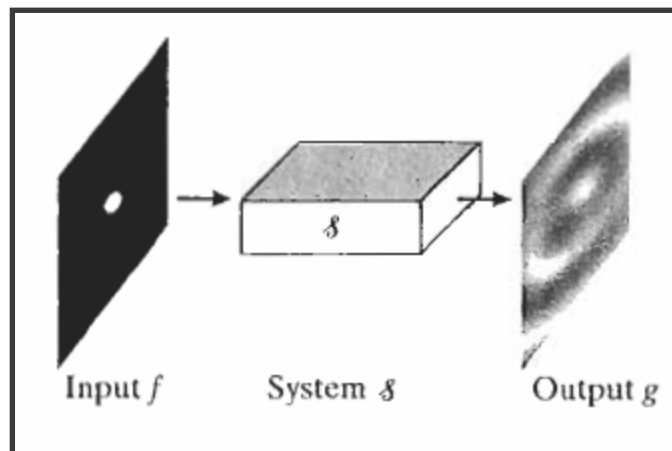
$$f(x, y) = f(x + X, y) = f(x, y + Y)$$

- The sampling function  $\delta_s(x, y; \Delta x, \Delta y)$  is a periodic signal with periods  $X = \Delta x$  and  $Y = \Delta y$ .
- The exponential and sinusoidal signals are periodic with periods  $X = 1/u_o$  and  $Y = 1/v_o$ .

## Systems

A *continuous-to-continuous* (or simply *continuous*) system is defined as a transformation  $\mathcal{L}$  of an *input* continuous signal  $f(x,y)$  to an *output* continuous signal  $g(x, y)$ .

$$g(x, y) = \mathcal{L}[f(x, y)]$$



## Linear Systems

A system  $\mathcal{E}$  is a *linear system* if, when the input consists of a weighted summation of several signals, the output will also be a weighted summation of the responses of the system to each individual input signal.

$$\mathcal{E} \left[ \sum_{k=1}^K w_k f_k(x, y) \right] = \sum_{k=1}^K w_k \mathcal{E} [f_k(x, y)]$$

It is a system that satisfy the superposition principle.

## Impulse Response

The point spread function (PSF), or the impulse response function of a system is the output of a system to a point impulse.

$$h(x, y; \xi, \eta) = \mathcal{S}[\delta_{\xi\eta}(x, y)]$$

For Linear systems

$$\begin{aligned} g(x, y) &= \mathcal{S}[f(x, y)] \\ &= \mathcal{S}\left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\xi, \eta) \delta(x - \xi, y - \eta) d\xi d\eta\right] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathcal{S}[f(\xi, \eta) \delta_{\xi\eta}(x, y)] d\xi d\eta \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\xi, \eta) \mathcal{S}[\delta_{\xi\eta}(x, y)] d\xi d\eta \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\xi, \eta) h(x, y; \xi, \eta) d\xi d\eta \end{aligned}$$

## Shift Invariance

A system is shift-invariant if an arbitrary translation of the input results in an identical translation in the output.

For a SIS the application of a shifted input  
Gives a shifted output

$$f_{x_0 y_0}(x, y) = f(x - x_0, y - y_0)$$

$$g(x - x_0, y - y_0) = \mathcal{S}[f_{x_0 y_0}(x, y)]$$

For a Linear Shift Invariant System (LSI)

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\xi, \eta) h(x - \xi, y - \eta) d\xi d\eta = h(x, y) * f(x, y)$$

## Example1

Check if the system  $g(x, y) = 2f(x, y)$  is LSI

Solution

I. Linearity

If  $g'(x, y)$  is the response of the system to input  $\sum_{k=1}^K w_k f_k(x, y)$ , then

$$\begin{aligned} g'(x, y) &= 2 \left( \sum_{k=1}^K w_k f_k(x, y) \right), \\ &= \sum_{k=1}^K w_k 2f_k(x, y), \\ &= \sum_{k=1}^K w_k g_k(x, y), \end{aligned}$$

II. Shift Invariance

$$g'(x, y) = 2f(x - x_0, y - y_0) = g(x - x_0, y - y_0)$$



## Example2

Check if the system  $g(x, y) = xyf(x, y)$  is LSI

Solution

I. Linearity

If  $g'(x, y)$  is the response of the system to input  $\sum_{k=1}^K w_k f_k(x, y)$ , then

$$g'(x, y) = xy \left( \sum_{k=1}^K w_k f_k(x, y) \right),$$

$$= \sum_{k=1}^K w_k xy f_k(x, y),$$

$$g'(x, y) = \sum_{k=1}^K w_k g_k(x, y),$$

II. Shift Invariance

$$g'(x, y) = xyf(x - x_0, y - y_0),$$

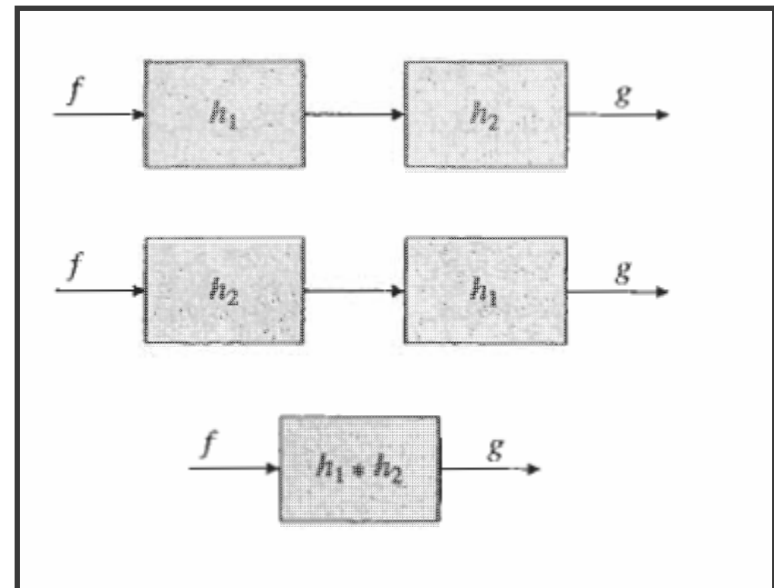
$$\neq (x - x_0)(y - y_0)f(x - x_0, y - y_0).$$

$$g'(x, y) \neq g(x - x_0, y - y_0),$$

# Connections of LSI Systems

## Cascade (Serial)

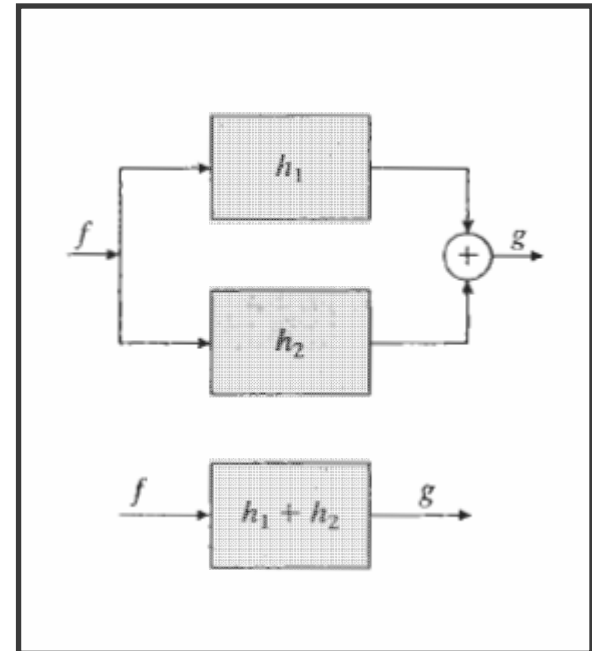
$$\begin{aligned} g(x, y) &= h_2(x, y) * [h_1(x, y) * f(x, y)], \\ &= h_1(x, y) * [h_2(x, y) * f(x, y)], \\ &= [h_1(x, y) * h_2(x, y)] * f(x, y), \end{aligned}$$



## Connections of LSI Systems

### Parallel

$$\begin{aligned} g(x, y) &= h_1(x, y) * f(x, y) + h_2(x, y) * f(x, y), \\ &= [h_1(x, y) + h_2(x, y)] * f(x, y), \end{aligned}$$



### Example3

Find the point spread function for the LSI system  $h_1(x,y)*h_2(x,y)$  where

$$h_1(x, y) = \frac{1}{2\pi\sigma_1^2} e^{-(x^2+y^2)/2\sigma_1^2} \quad \text{and} \quad h_2(x, y) = \frac{1}{2\pi\sigma_2^2} e^{-(x^2+y^2)/2\sigma_2^2}$$

Solution

$$\begin{aligned} h(x, y) &= h_1(x, y) * h_2(x, y), \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h_2(\xi, \eta) h_1(x - \xi, y - \eta) d\xi d\eta, \\ &= \frac{1}{4\pi^2\sigma_1^2\sigma_2^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-(\xi^2+\eta^2)/2\sigma_2^2} e^{-[(x-\xi)^2+(y-\eta)^2]/2\sigma_1^2} d\xi d\eta, \\ &= \frac{1}{4\pi^2\sigma_1^2\sigma_2^2} \int_{-\infty}^{\infty} e^{-\xi^2/2\sigma_2^2 - (x-\xi)^2/2\sigma_1^2} d\xi \\ &\quad \int_{-\infty}^{\infty} e^{-\eta^2/2\sigma_2^2 - (y-\eta)^2/2\sigma_1^2} d\eta. \end{aligned}$$

### Example3

Solution, cont.

$$\begin{aligned}\int_{-\infty}^{\infty} e^{-\xi^2/2\sigma_2^2 - (x-\xi)^2/2\sigma_1^2} d\xi &= e^{-x^2/2(\sigma_1^2 + \sigma_2^2)} \int_{-\infty}^{\infty} e^{-\frac{\sigma_1^2 + \sigma_2^2}{2\sigma_1^2\sigma_2^2} \left\{ \xi - [\sigma_2^2/(\sigma_1^2 + \sigma_2^2)]x \right\}^2} d\xi, \\ &= e^{-x^2/2(\sigma_1^2 + \sigma_2^2)} \int_{-\infty}^{\infty} e^{-\frac{\sigma_1^2 + \sigma_2^2}{2\sigma_1^2\sigma_2^2} \tau^2} d\tau, \\ &= \frac{\sqrt{2\pi}\sigma_1\sigma_2}{\sqrt{\sigma_1^2 + \sigma_2^2}} e^{-x^2/2(\sigma_1^2 + \sigma_2^2)},\end{aligned}$$

$$h(x, y) = \frac{1}{2\pi(\sigma_1^2 + \sigma_2^2)} \exp \left\{ \frac{-(x^2 + y^2)}{2(\sigma_1^2 + \sigma_2^2)} \right\}$$

where,

$$\tau \rightarrow \xi - [\sigma_2^2/(\sigma_1^2 + \sigma_2^2)]x$$

$$\int_{-\infty}^{\infty} e^{-a^2\tau^2} d\tau = \frac{\sqrt{\pi}}{a}, \quad \text{for } a \neq 0$$

## Separable Systems

2D LSI system with PSF  $h(x,y)$  is separable if there exist two 1D systems with PSF  $h_1(x)$  and  $h_2(y)$

$$h(x, y) = h_1(x)h_2(y)$$

A 2D convolution

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\xi, \eta) h(x - \xi, y - \eta) d\xi d\eta$$

can be computed as two 1D convolution

$$\begin{aligned} \text{Compute } w(x, y) &= \int_{-\infty}^{\infty} f(\xi, y) h_1(x - \xi) d\xi, \text{ for every } y. \\ \text{Compute } g(x, y) &= \int_{-\infty}^{\infty} w(x, \eta) h_2(y - \eta) d\eta, \text{ for every } x. \end{aligned}$$

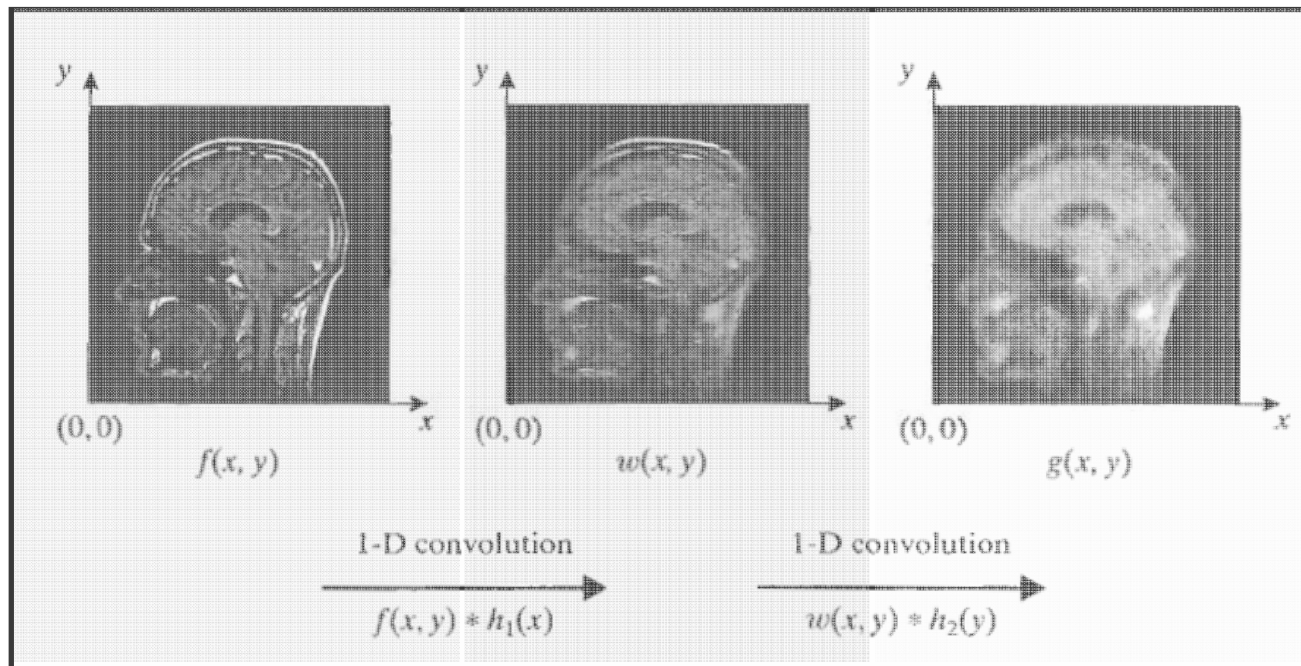
## Example 4

The 2 D convolution of the PSF,  $h(x,y)$ , with an image  $f(x,y)$

$$h(x, y) = \frac{1}{2\pi\sigma^2} e^{-(x^2+y^2)/2\sigma^2}$$

Can be calculated as a two 1D convolution of

$$h_1(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-x^2/2\sigma^2} \quad \text{and} \quad h_2(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-y^2/2\sigma^2}$$



## Stable Systems

A system is bounded input bounded output stable if and only if when the input is bounded signal

$$|f(x, y)| \leq B < \infty$$

For some finite B, there exists a finite B' such that

$$|g(x, y)| = |h(x, y) * f(x, y)| \leq B' < \infty$$

A LSI system is BIBO if and only if its PSF is absolutely integrable.

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |h(x, y)| \, dx \, dy < \infty$$



## The Fourier Transform

The Fourier transform provides a different perspective on how signals and systems interact.

It leads to alternative tools for system analysis and implementation

The 2D Fourier Transform Pair are:

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy,$$

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{j2\pi(ux+vy)} du dv.$$

## The Fourier Transform

A sufficient condition for the existence of Fourier Transform:

- $f(x,y)$  is continuous,
- $f(x,y)$  has a finite number of discontinuities,
- $f(x,y)$  is absolutely integrable

$$|F(u, v)| = \sqrt{F_R^2(u, v) + F_I^2(u, v)},$$

$$\angle F(u, v) = \tan^{-1} \left( \frac{F_I(u, v)}{F_R(u, v)} \right),$$

$$F(u, v) = |F(u, v)| e^{j\angle F(u, v)}.$$

$$F(u, v) = F_R(u, v) + jF_I(u, v)$$

## The Fourier Transform

Example: Find the FT for  $\delta(x,y)$

$$\begin{aligned}\mathcal{F}_{2D}(\delta)(u, v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x, y) e^{-j2\pi(ux+vy)} dx dy, \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x, y) e^{-j2\pi(u0+v0)} dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x, y) dx dy = 1\end{aligned}$$

Example: Find the FT for  $f(x,y)=e^{j2\pi(u_0x+v_0y)}$

$$\begin{aligned}\mathcal{F}_{2D}(f)(u, v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy, \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{j2\pi(u_0x+v_0y)} e^{-j2\pi(ux+vy)} dx dy, \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-j2\pi[(u-u_0)x+(v-v_0)y]} dx dy, \\ &= \delta(u - u_0, v - v_0),\end{aligned}$$

where

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-j2\pi(ux+vy)} dx dy = \delta(u, v),$$

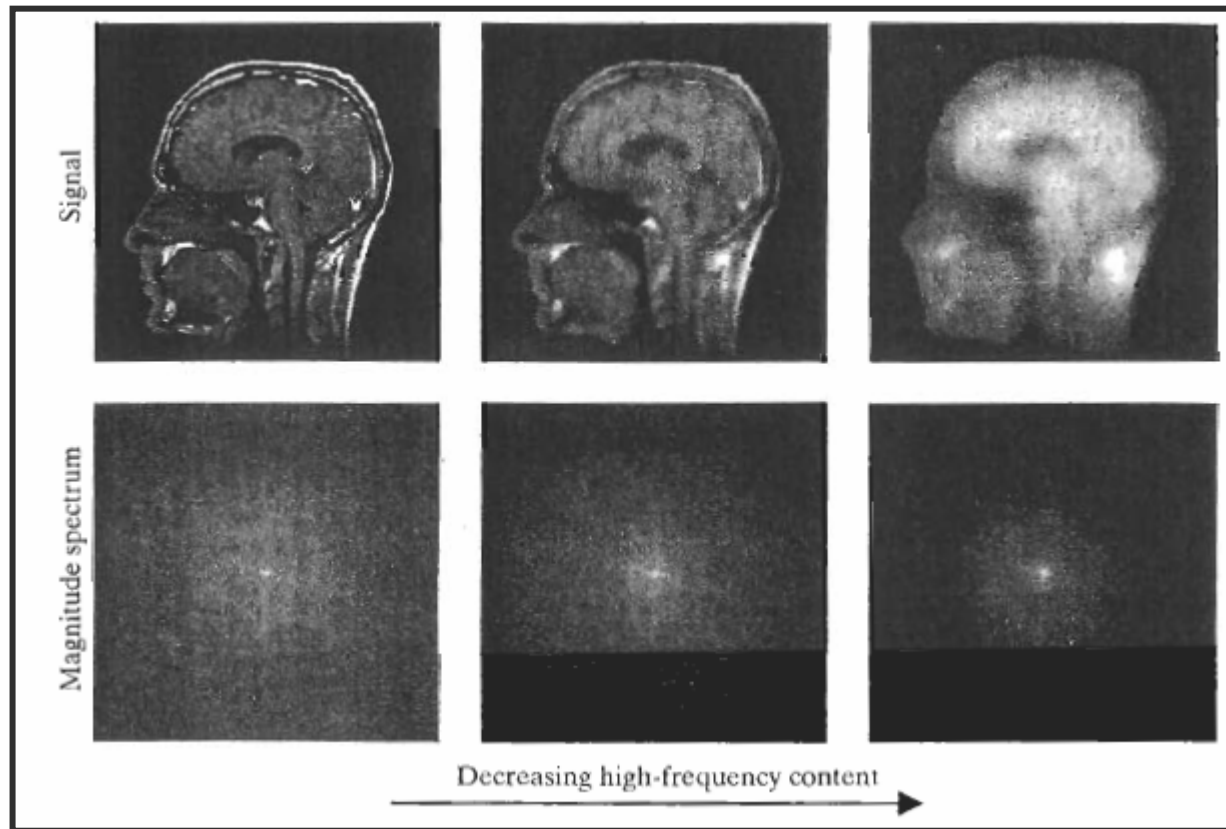
## The Fourier Transform

Basic Fourier transform pairs	
Signal	Fourier Transform
1	$\delta(u, v)$
$\delta(x, y)$	1
$\delta(x - x_0, y - y_0)$	$e^{-j2\pi(ux_0 + vy_0)}$
$\delta_s(x, y; \Delta x, \Delta y)$	$\text{comb}(u\Delta x, v\Delta y)$
$e^{j2\pi(u_0x + v_0y)}$	$\delta(u - u_0, v - v_0)$
$\sin[2\pi(u_0x + v_0y)]$	$\frac{1}{2j} [\delta(u - u_0, v - v_0) - \delta(u + u_0, v + v_0)]$
$\cos[2\pi(u_0x + v_0y)]$	$\frac{1}{2} [\delta(u - u_0, v - v_0) + \delta(u + u_0, v + v_0)]$
$\text{rect}(x, y)$	$\text{sinc}(u, v)$
$\text{sinc}(x, y)$	$\text{rect}(u, v)$
$\text{comb}(x, y)$	$\text{comb}(u, v)$
$e^{-\pi(x^2 + y^2)}$	$e^{-\pi(u^2 + v^2)}$

## The Fourier Transform

### Note:

Slow signal variation in space produces a spectral content that is more concentrated at low frequencies, whereas fast signal variation results in spectral content at high frequencies



## The Fourier Transform

The 1D Fourier Transform Pair are:

$$F(u) = \mathcal{F}_{1D}(f)(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi ux} dx,$$

$$f(x) = \mathcal{F}_{1D}^{-1}(F)(x) = \int_{-\infty}^{\infty} F(u) e^{j2\pi ux} du,$$

Example: Find the FT for  $\text{rect}(x)$

$$\begin{aligned} \mathcal{F}_{1D}(\text{rect})(u) &= \int_{-\infty}^{\infty} \text{rect}(x) e^{-j2\pi ux} dx, \\ &= \int_{-1/2}^{1/2} e^{-j2\pi ux} dx, \\ &= \frac{1}{j2\pi u} e^{-j2\pi ux} \Big|_{-1/2}^{1/2}, \\ &= \frac{1}{\pi u} \frac{e^{j\pi u} - e^{-j\pi u}}{2j}, \\ &= \frac{\sin(\pi u)}{\pi u} = \text{sinc}(u). \end{aligned}$$

# The Fourier Transform

## Properties: Linearity

$$\mathcal{F}_{2D}(a_1f + a_2g)(u, v) = a_1F(u, v) + a_2G(u, v)$$

## Properties: Translation

$$f(x, y) \longrightarrow F(u, v)$$

$$f_{x_0y_0}(x, y) = f(x - x_0, y - y_0) \longrightarrow \mathcal{F}_{2D}(f_{x_0y_0})(u, v) = F(u, v)e^{-j2\pi(ux_0 + vy_0)}$$

$$|\mathcal{F}_{2D}(f_{x_0y_0})(u, v)| = |F(u, v)|$$

$$\angle \mathcal{F}_{2D}(f_{x_0y_0})(u, v) = \angle F(u, v) - 2\pi(ux_0 + vy_0)$$

## The Fourier Transform

### Properties: Conjugation

If  $f(x,y)$  is complex-valued 2D signal, then

$$\mathcal{F}_{2D}(f^*)(u, v) = F^*(-u, -v)$$

### Properties: Conjugate Symmetry

If  $f(x,y)$  is real-valued 2D signal, then

$$F(u, v) = F^*(-u, -v) .$$

$$F_R(u, v) = F_R(-u, -v) \quad \text{and} \quad F_I(u, v) = -F_I(-u, -v) ,$$

$$|F(u, v)| = |F(-u, -v)| \quad \text{and} \quad \angle F(u, v) = -\angle F(-u, -v) .$$



## The Fourier Transform

### Properties: Scaling

$$f(x, y) \longrightarrow F(u, v)$$

$$f_{ab}(x, y) = f(ax, by) \xrightarrow{\text{scaling}} \mathcal{F}_{2D}(f_{ab})(u, v) = \frac{1}{|ab|} F\left(\frac{u}{a}, \frac{v}{b}\right)$$

Example: Find the FT for

$$f(x, y) = \text{rect}\left(\frac{x - x_0}{\Delta x}, \frac{y - y_0}{\Delta y}\right)$$

The FT for the rect function is

$$\mathcal{F}_{2D}(\text{rect})(u, v) = \text{sinc}(u, v).$$

Scaling property

$$\mathcal{F}_{2D}\left\{\text{rect}\left(\frac{x}{\Delta x}, \frac{y}{\Delta y}\right)\right\} = \Delta x \Delta y \text{sinc}(\Delta x u, \Delta y v)$$

Translation property

$$\mathcal{F}_{2D}(f)(u, v) = \Delta x \Delta y \text{sinc}(\Delta x u, \Delta y v) e^{-j2\pi(ux_0 + vy_0)}$$

## The Fourier Transform

### Properties: Rotation

$$f_{\theta}(x, y) = f(x \cos \theta - y \sin \theta, x \sin \theta + y \cos \theta)$$

$$\mathcal{F}_{2D}(f_{\theta})(u, v) = F(u \cos \theta - v \sin \theta, u \sin \theta + v \cos \theta)$$

### Properties: Convolution

$$\mathcal{F}_{2D}(f * g)(u, v) = F(u, v)G(u, v)$$

## The Fourier Transform

Example:

Find the convolution of  $f(x,y) = \text{sinc}(Ux, Vy)$  and  $g(x,y) = \text{sinc}(Vx, Uy)$  where  $0 < U < V$ .

The FT for the sinc function is  $\mathcal{F}_{2D}\{\text{sinc}(x, y)\} = \text{rect}(u, v)$ .

The scaling property

$$F(u, v) = \mathcal{F}_{2D}(f)(u, v) = \frac{1}{UV} \text{rect}\left(\frac{u}{U}, \frac{v}{V}\right) \quad G(u, v) = \mathcal{F}_{2D}(g)(u, v) = \frac{1}{UV} \text{rect}\left(\frac{u}{V}, \frac{v}{U}\right)$$

The convolution property  $f(x, y) * g(x, y) = \mathcal{F}_{2D}^{-1}\{F(u, v)G(u, v)\}$

$$= \mathcal{F}_{2D}^{-1} \left\{ \frac{1}{(UV)^2} \text{rect}\left(\frac{u}{U}, \frac{v}{V}\right) \text{rect}\left(\frac{u}{V}, \frac{v}{U}\right) \right\}$$

$$= \mathcal{F}_{2D}^{-1} \left\{ \frac{1}{(UV)^2} \text{rect}\left(\frac{u}{V}, \frac{v}{V}\right) \right\}$$

$$= \frac{1}{U^2} \mathcal{F}_{2D}^{-1} \left\{ \frac{1}{V^2} \text{rect}\left(\frac{u}{V}, \frac{v}{V}\right) \right\}$$

$$= \frac{1}{U^2} \text{sinc}(Vx, Vy)$$

## The Fourier Transform

### Properties: Product

$$\begin{aligned}\mathcal{F}_{2D}(fg)(u, v) &= F(u, v) * G(u, v), \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G(\xi, \eta) F(u - \xi, v - \eta) d\xi d\eta\end{aligned}$$

### Properties: Separable Product

$$f(x, y) = f_1(x)f_2(y),$$

$$\mathcal{F}_{2D}(f)(u, v) = F_1(u)F_2(v),$$

## The Fourier Transform

### Properties: Parseval's Theorem

The energy content of a 2D signal in the spatial domain is the same as in the frequency domain.

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |f(x, y)|^2 dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |F(u, v)|^2 du dv$$

### Properties: Separability

The FT of a 2D signal  $f(x, y)$  can be calculated as two 1D FT

$$r(u, y) = \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi ux} dx, \text{ for every } y.$$
$$F(u, v) = \int_{-\infty}^{\infty} r(u, y) e^{-j2\pi vy} dy, \text{ for every } u.$$

# The Fourier Transform

Properties of the Fourier transform		
Property	Signal	Fourier Transform
	$f(x, y)$	$F(u, v)$
	$g(x, y)$	$G(u, v)$
Linearity	$a_1 f(x, y) + a_2 g(x, y)$	$a_1 F(u, v) + a_2 G(u, v)$
Translation	$f(x - x_0, y - y_0)$	$F(u, v) e^{-j2\pi(ux_0 + vy_0)}$
Conjugation	$f^*(x, y)$	$F^*(-u, -v)$
Conjugate symmetry	$f(x, y)$ is real-valued	$F(u, v) = F^*(-u, -v)$
		$F_R(u, v) = F_R(-u, -v)$
		$F_I(u, v) = -F_I(-u, -v)$
		$ F(u, v)  =  F(-u, -v) $
		$\angle F(u, v) = -\angle F(-u, -v)$
Signal reversing	$f(-x, -y)$	$F(-u, -v)$
Scaling	$f(ax, by)$	$\frac{1}{ ab } F\left(\frac{u}{a}, \frac{v}{b}\right)$
Rotation	$f(x \cos \theta - y \sin \theta, x \sin \theta + y \cos \theta)$	$F(u \cos \theta - v \sin \theta, u \sin \theta + v \cos \theta)$
Circular symmetry	$f(x, y)$ is circularly symmetric	$F(u, v)$ is circularly symmetric
		$ F(u, v)  = F(u, v)$
		$\angle F(u, v) = 0$
Convolution	$f(x, y) * g(x, y)$	$F(u, v) G(u, v)$
Product	$f(x, y) g(x, y)$	$F(u, v) * G(u, v)$
Separable product	$f(x) g(y)$	$F(u) G(v)$
Parseval's theorem	$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty}  f(x, y) ^2 dx dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty}  F(u, v) ^2 du dv .$	

## The Transfer Function

The transfer function is the FT of the PSF  $h(x,y)$  and is denoted as  $H(u,v)$ .

$$H(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\xi, \eta) e^{-j2\pi(u\xi + v\eta)} d\xi d\eta$$

$$h(x, y) = \iint G(u, v) = H(u, v) F(u, v) e^{j2\pi(ux + vy)} du dv$$

The transfer function is the ratio of the output signal to input signal in the frequency domain.

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

## The Transfer Function

Example: Low Pass Filter (LPF)

The transfer function for a LPF is:

$$H(u, v) = \begin{cases} 1, & \text{for } \sqrt{u^2 + v^2} \leq c \\ 0, & \text{for } \sqrt{u^2 + v^2} > c \end{cases}$$

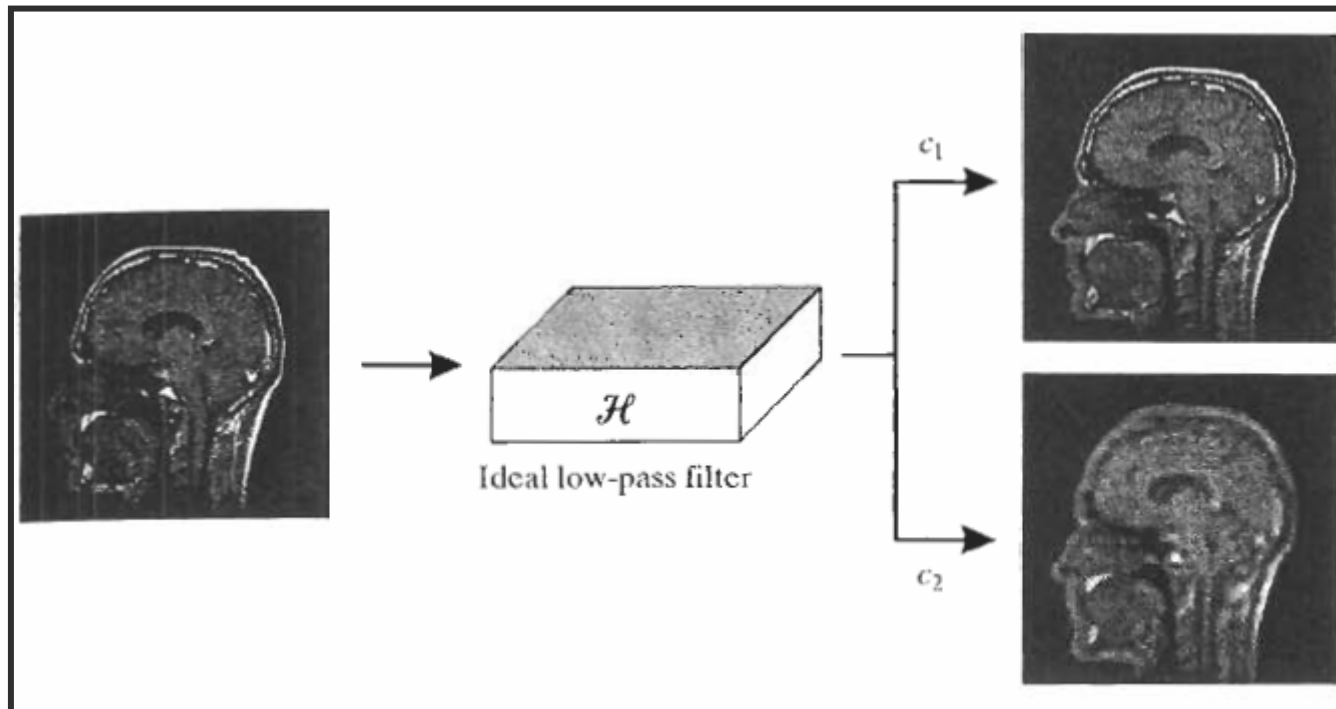
The input output signals are:

$$G(u, v) = \begin{cases} F(u, v), & \text{for } \sqrt{u^2 + v^2} \leq c \\ 0, & \text{for } \sqrt{u^2 + v^2} > c \end{cases}$$



## The Transfer Function

Example, LPF cont.,



## Circular Symmetry:

A 2D signal is said to have a circular symmetry if

$$f_{\theta}(x, y) = f(x, y), \quad \text{for every } \theta$$

where

$$f_{\theta}(x, y) = f(x \cos \theta - y \sin \theta, x \sin \theta + y \cos \theta)$$

The circular symmetry implies  $f(x, y)$  is even, which gives even real  $F(u, v)$

$$|F(u, v)| = F(u, v) \quad \text{and} \quad \angle F(u, v) = 0$$

## Hankel Transform:

Hankel transform is defined for circularly symmetric signals.

Define  $r = \sqrt{x^2 + y^2}$       Then  $f(x, y) = f(r)$ ,

And  $F(u, v) = F(q)$       For  $q = \sqrt{u^2 + v^2}$ .

Hankel Transform is

$$F(q) = \mathcal{H}\{f(r)\}$$

$$F(q) = 2\pi \int_0^{\infty} f(r) J_0(2\pi qr) r dr$$

$$J_n(r) = \frac{1}{\pi} \int_0^{\pi} \cos(nr - r \sin \phi) d\phi, \quad n = 0, 1, 2, \dots,$$

$$J_0(r) = \frac{1}{\pi} \int_0^{\pi} \cos(r \sin \phi) d\phi$$

Hankel Inverse transform is

$$f(r) = 2\pi \int_0^{\infty} F(q) J_0(2\pi qr) q dq$$

## Hankel Transform:

Selected Hankel Transform Pairs	
Signal	Hankel Transform
$\exp\{-\pi r^2\}$	$\exp\{-\pi q^2\}$
$1$	$\delta(q)/\pi q = \delta(u, v)$
$\delta(r - a)$	$2\pi a J_0(2\pi a q)$
$\text{rect}(r)$	$\frac{J_1(\pi q)}{2q}$
$\text{sinc}(r)$	$\frac{2 \text{rect}(q)}{\pi \sqrt{1-4q^2}}$
$\frac{1}{r}$	$\frac{1}{q}$

## Properties: Scaling (a=b)

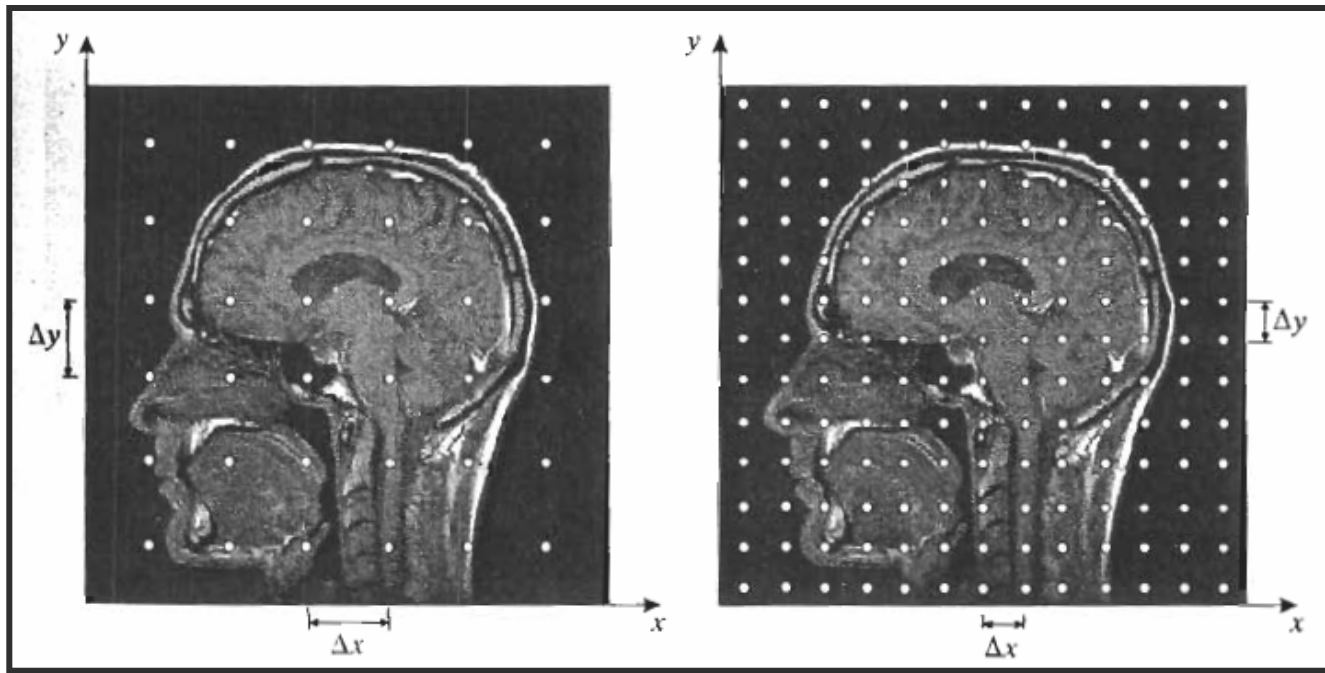
$$\mathcal{H} \{f(ar)\} = \frac{1}{a^2} F(q/a)$$

## Sampling:

A 2D continuous signal is replaced by a discrete signal whose values are the values of the continuous signal at the vertices of a 2D rectangular grid.

For  $\Delta x$  and  $\Delta y$  sampling periods

$$f_d(m, n) = f(m\Delta x, n\Delta y)$$



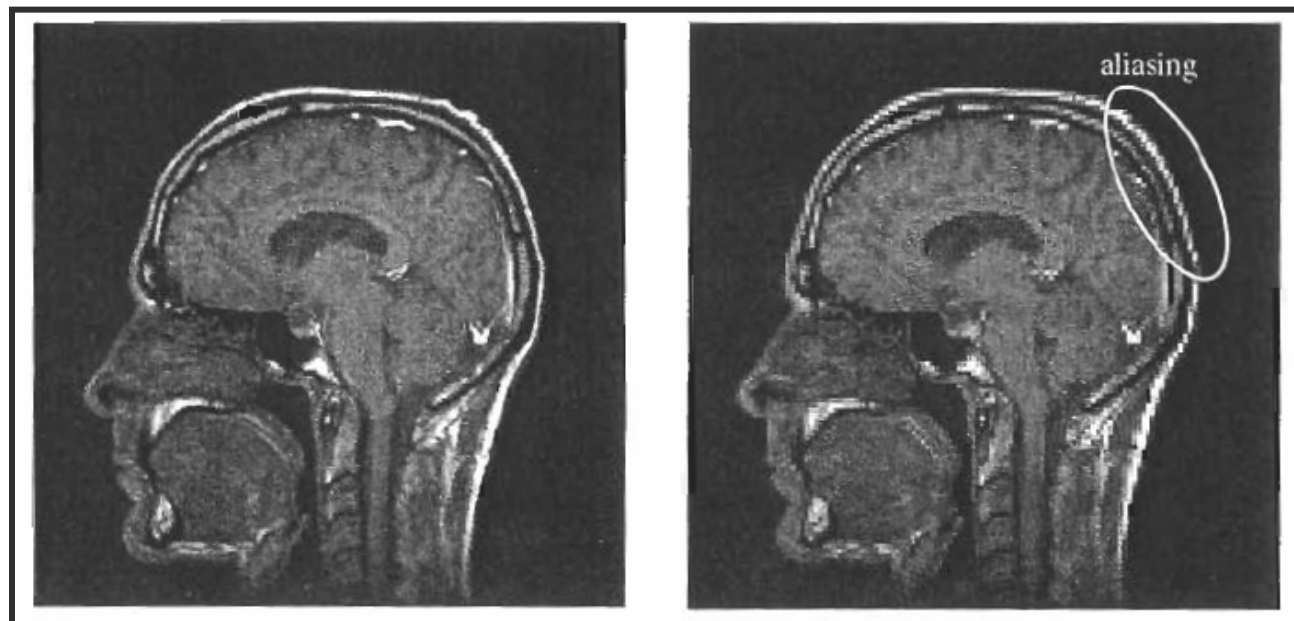
## **Sampling:**

More sampling:

More detectors, signal storage, subsequent processing, etc.

Less sampling:

Aliasing



## Sampling:

The sampled 2D signal in spatial domain is:

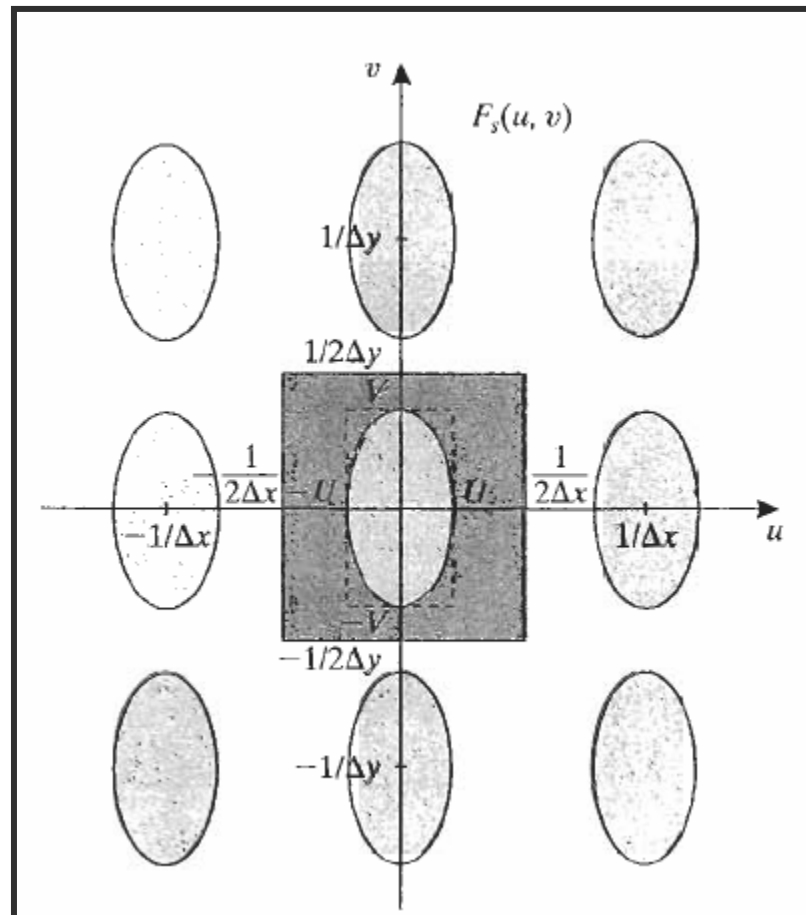
$$\begin{aligned} f_s(x, y) &= f(x, y) \delta_s(x, y; \Delta x, \Delta y) \\ &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} f_d(m, n) \delta(x - m\Delta x, y - n\Delta y) \end{aligned}$$

The sampled 2D signal in frequency domain is:

$$\begin{aligned} F_s(u, v) &= F(u, v) * \text{comb}(u\Delta x, v\Delta y) \\ &= \frac{1}{\Delta x \Delta y} \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} F(u - m/\Delta x, v - n/\Delta y) \end{aligned}$$

## Sampling:

The spectrum of the sampled 2D signal is:



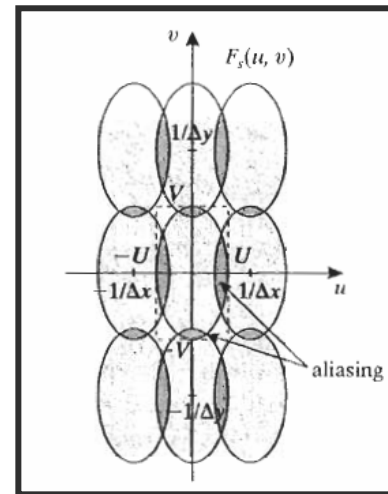
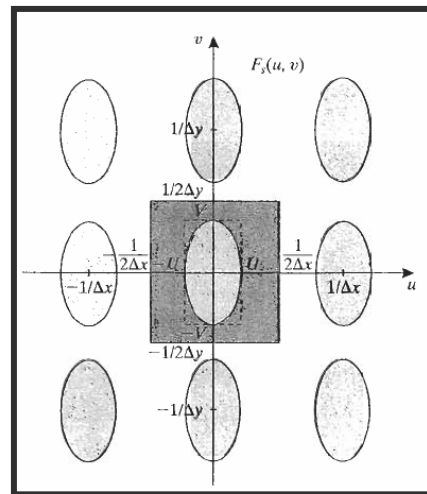


## Sampling:

### Nyquist Sampling Theorem

A 2D continuous band-limited signal  $f(x,y)$  with cutoff frequencies  $U$  and  $V$  can be UNIQUELY determined from its samples  $f_d(m,n)$  if and only if the sampling periods  $\Delta x$  and  $\Delta y$  satisfy:

$$\Delta x \leq \frac{1}{2U} \quad \text{and} \quad \Delta y \leq \frac{1}{2V}$$



**Thanks**