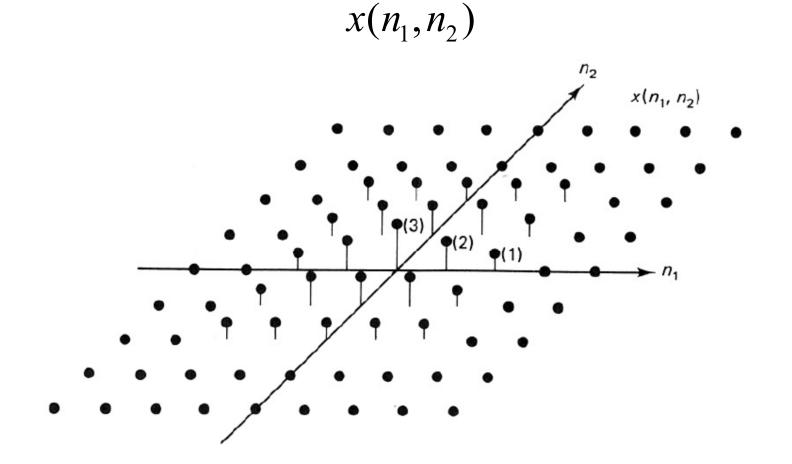
# Video signals

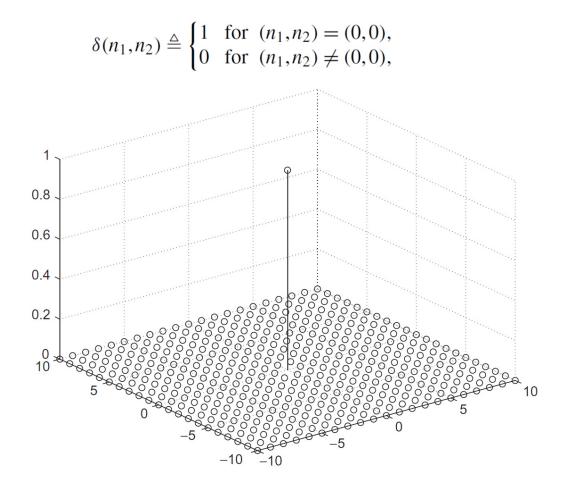
IMAGE AS BIDIMENSIONAL SIGNALS

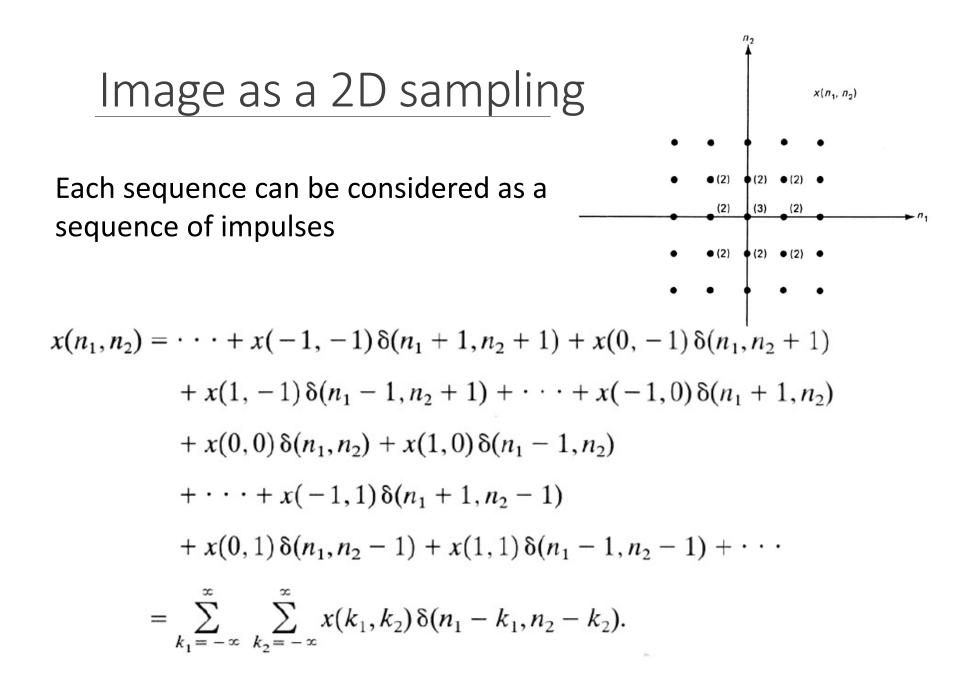
#### Image as a 2D sampling

A digital image can be considered as a 2D discrete signal



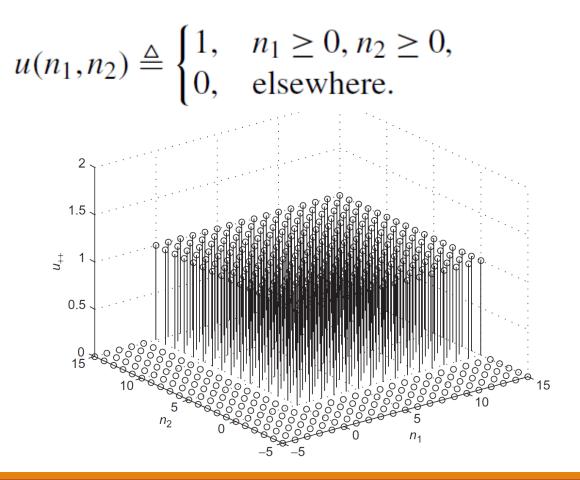
### Impulse definition





# The unit step function

The step function as a combination of impulses.



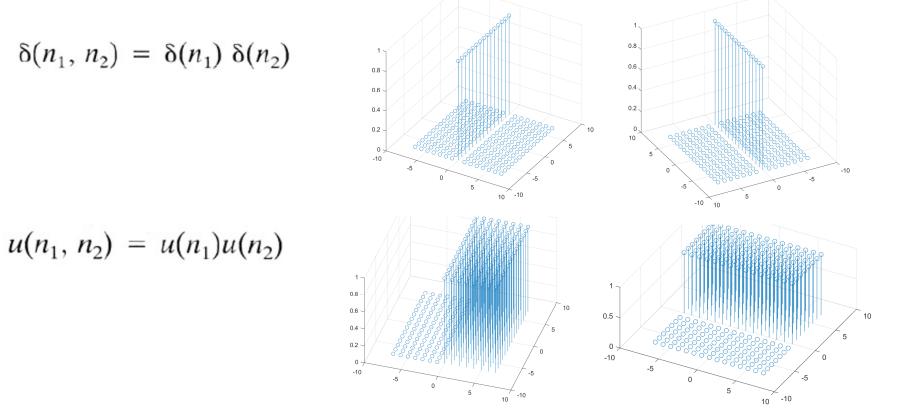
Video Signals

#### Separable sequences

A separable 2D sequence can be written as:

 $x(n_1, n_2) = x_1(n_1)x_2(n_2)$  for all  $n_1$  and  $n_2$ ,

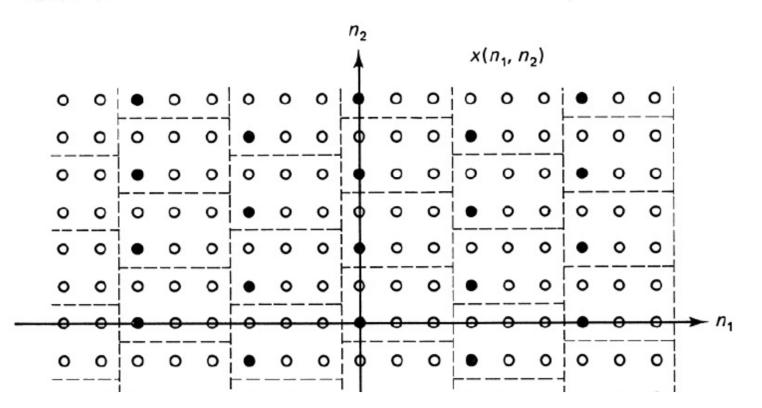
The impulse and step function are separable functions



#### Periodic sequences

A sequence  $x(n_1, n_2)$  is periodic of period  $N_1 \times N_2$  if:

 $x(n_1, n_2) = x(n_1 + N_1, n_2) = x(n_1, n_2 + N_2)$  for all  $(n_1, n_2)$ 



#### Linear Shift Invariant Systems

Linearity

$$T[ax_1(n_1, n_2) + bx_2(n_1, n_2)] = ay_1(n_1, n_2) + by_2(n_1, n_2)$$

Spatial invariance

$$T[x(n_1 - m_1, n_2 - m_2)] = y(n_1 - m_1, n_2 - m_2)$$

The impulse response

$$y(n_1, n_2) = T[x(n_1, n_2)] = T\left[\sum_{k_1 = -\infty}^{\infty} \sum_{k_2 = -\infty}^{\infty} x(k_1, k_2) \,\delta(n_1 - k_1, n_2 - k_2)\right]$$
$$= \sum_{k_1 = -\infty}^{\infty} \sum_{k_2 = -\infty}^{\infty} x(k_1, k_2) T[\delta(n_1 - k_1, n_2 - k_2)].$$

### Convolution

Defined the impulse response

$$h(n_1, n_2) = T[\delta(n_1, n_2)].$$

The Input/Output relation is given by:

$$y(n_1, n_2) = T[x(n_1, n_2)] = \sum_{k_1 = -\infty}^{\infty} \sum_{k_2 = -\infty}^{\infty} x(k_1, k_2)h(n_1 - k_1, n_2 - k_2).$$

$$y(n_1, n_2) = x(n_1, n_2) * h(n_1, n_2)$$
  
=  $\sum_{k_1 = -\infty}^{\infty} \sum_{k_2 = -\infty}^{\infty} x(k_1, k_2)h(n_1 - k_1, n_2 - k_2).$ 

#### Convolution properties

Commutativity

$$x(n_1, n_2) * y(n_1, n_2) = y(n_1, n_2) * x(n_1, n_2)$$

Associativity

 $(x(n_1, n_2) * y(n_1, n_2)) * z(n_1, n_2) = x(n_1, n_2) * (y(n_1, n_2) * z(n_1, n_2))$ 

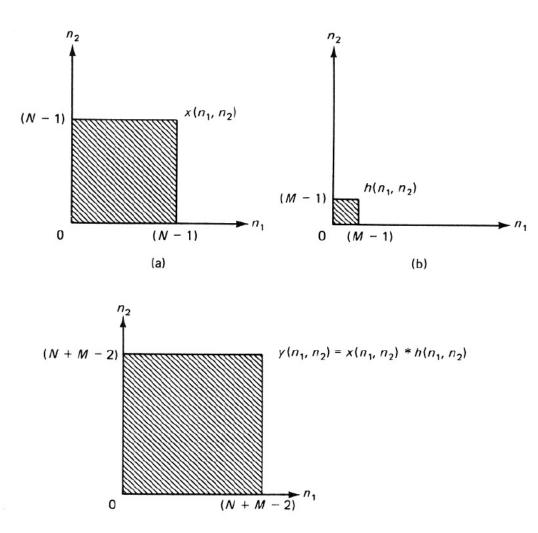
#### Distributivity

$$\begin{aligned} x(n_1, n_2) * (y(n_1, n_2) + z(n_1, n_2)) \\ &= (x(n_1, n_2) * y(n_1, n_2)) + (x(n_1, n_2) * z(n_1, n_2)) \end{aligned}$$

Convolution with Shifted Impulse

$$x(n_1, n_2) * \delta(n_1 - m_1, n_2 - m_2) = x(n_1 - m_1, n_2 - m_2)$$

### Convolution examples



# The 2D Fourier Transform

The analysis and synthesis formulas for the 2D continuous Fourier transform are as follows:

Analysis

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

Synthesis

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

#### Separability of 2D Fourier Transform

The 2D analysis formula can be written as a 1D analysis in the *x* direction followed by a *1D* analysis in the *y* direction:

$$F(u,v) = \int_{-\infty}^{\infty} \left[ \int_{-\infty}^{\infty} f(x,y) e^{-j2\pi u x} dx \right] e^{-j2\pi v y} dy$$

The 2D synthesis formula can be written as a 1D synthesis in the *x* direction followed by a 1D synthesis in *y* direction:

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

### Separability Theorem

$$f(x,y) = f(x)g(y) \xrightarrow{\mathcal{F}} F(u,v) = F(u)G(v)$$

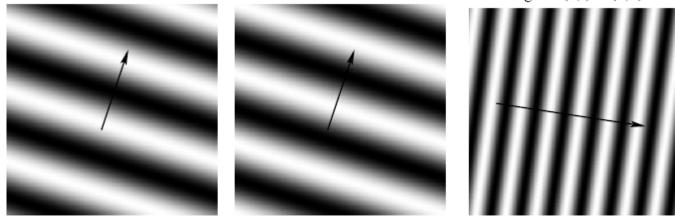
Proof:

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

#### **2D Fourier Basis Functions**

Grating for (k,l) = (1,-3)

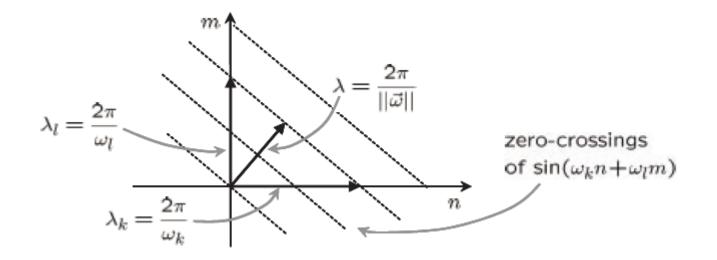
Grating for (k,l) = (7,1)



Real

Imag

Real



The 2D Discrete Fourier Transform for periodic signals

The analysis and synthesis formulas for the 2D discrete Fourier transform are as follows:

Analysis

$$\hat{F}(k,\ell) = \frac{1}{\sqrt{MN}} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} F(m,n) e^{-j2\pi \left(k\frac{m}{M} + \ell\frac{n}{N}\right)}$$

$$F(m,n) = \frac{1}{\sqrt{MN}} \sum_{k=0}^{M-1} \sum_{\ell=0}^{N-1} \hat{F}(k,\ell) e^{j2\pi \left(k\frac{m}{M} + \ell\frac{n}{N}\right)}$$

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$$\frac{\text{Separability of 2D DFT}}{\hat{F}(k,\ell)} = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} \left[ \frac{1}{\sqrt{M}} \sum_{m=0}^{M-1} F(m,n) e^{-j2\pi \left(k\frac{m}{M}\right)} \right] e^{-j2\pi \left(\ell\frac{n}{N}\right)}$$

The 2D forward DFT (for a squared input matrix of size SxS)can be written in matrix notation:

 $\hat{\mathbf{F}} = (\mathbf{W}^* \mathbf{F}) \, \mathbf{W}^*$ 

Where *r* and *c* are row and column indexes starting from zero.

$$W_{rc}^* = \frac{1}{\sqrt{S}} e^{-j2\pi \frac{rc}{S}}$$

# Separability of 2D DFT

And

$$F(m,n) = \frac{1}{\sqrt{N}} \sum_{\ell=0}^{N-1} \left[ \frac{1}{\sqrt{M}} \sum_{k=0}^{M-1} \hat{F}(k,\ell) e^{j2\pi \left(k\frac{m}{M}\right)} \right] e^{j2\pi \left(\ell\frac{n}{N}\right)}.$$

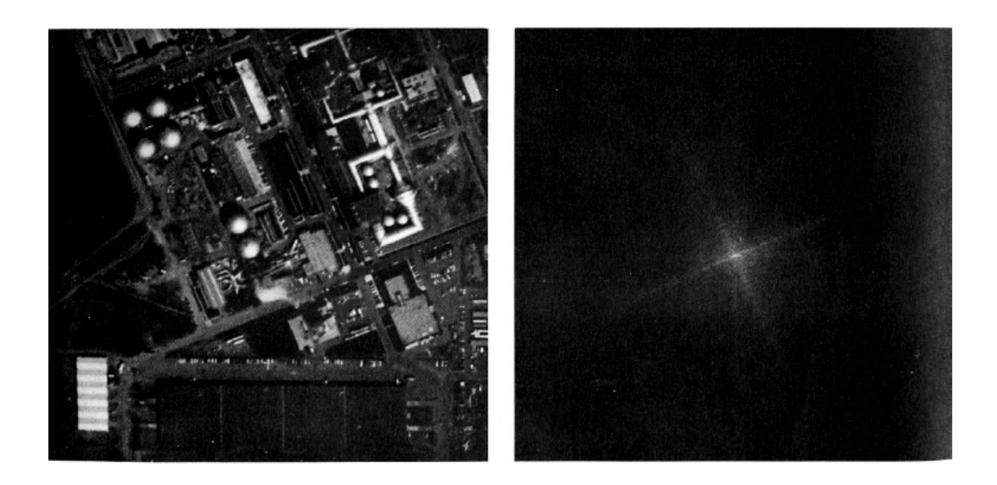
The 2D inverse DFT can be written in matrix notation:  $\mathbf{F} = \left(\mathbf{W}\hat{\mathbf{F}}
ight)\mathbf{W}$ 

Where the matrix elements are

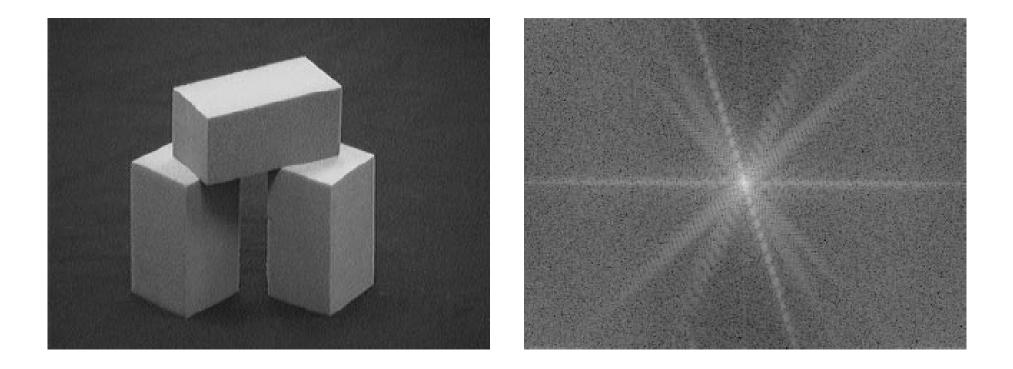
$$W_{rc} = \frac{1}{\sqrt{A}} e^{j2\pi \frac{rc}{A}}$$

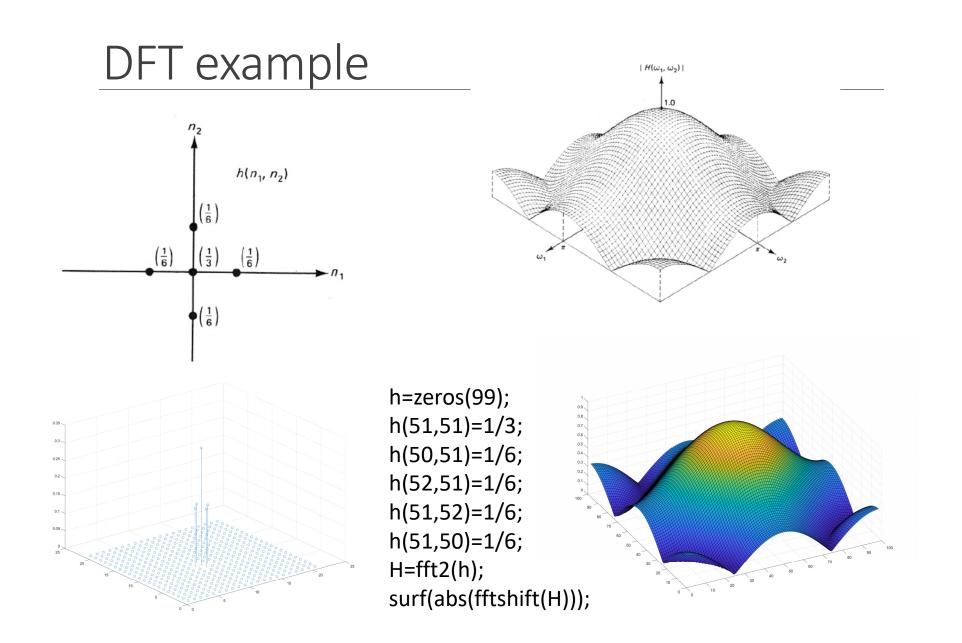


# Transform examples

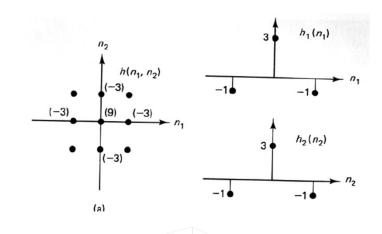


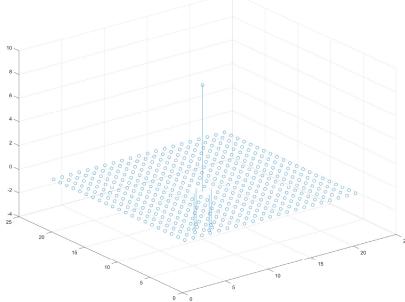
# 2D DFT Example

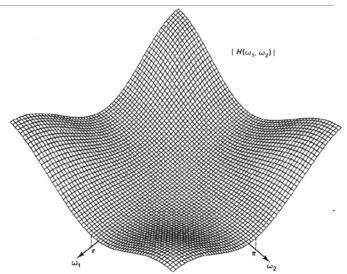




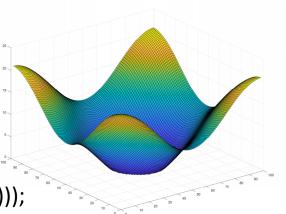
#### Hi-Pass Filter example



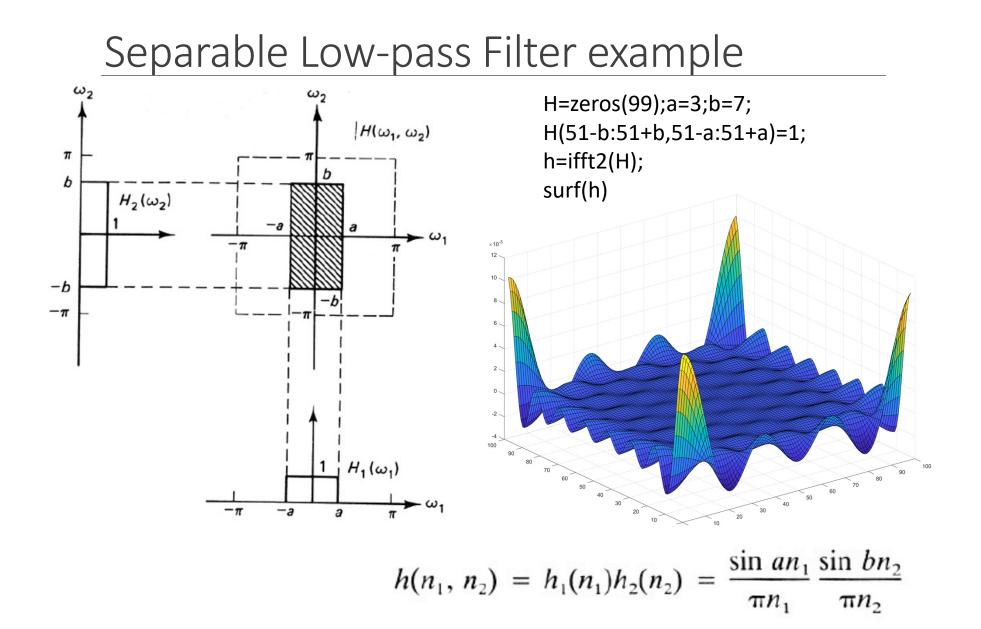




h=zeros(99); h(51,51)=9; h(50,51)=-3; h(52,51)=-3; h(51,50)=-3; h(51,52)=-3; H=fft2(h); surf(abs(fftshift(H)));



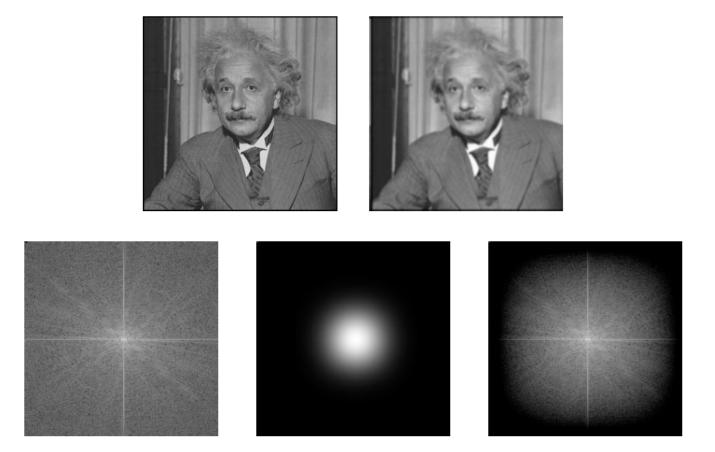
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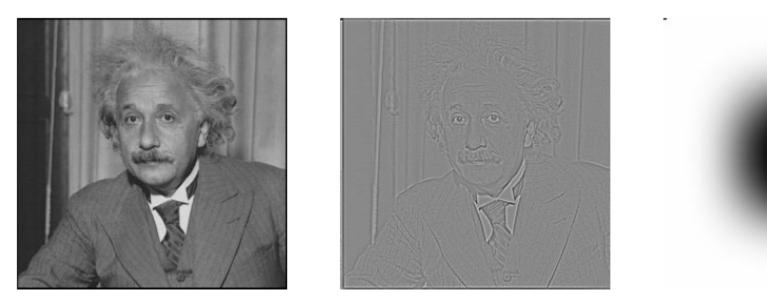
#### low-pass (blurred) version of an image

h(n)=  $\frac{1}{16}$  (1,4,6,4,1) (1D impulse response in both dimensions)



Video Signals

#### Hi-Pass Filter



impulse response is defined by  $\delta(n)$ -h(n,m) where h(n,m) is the separable blurring kernel used in the previous figure

5

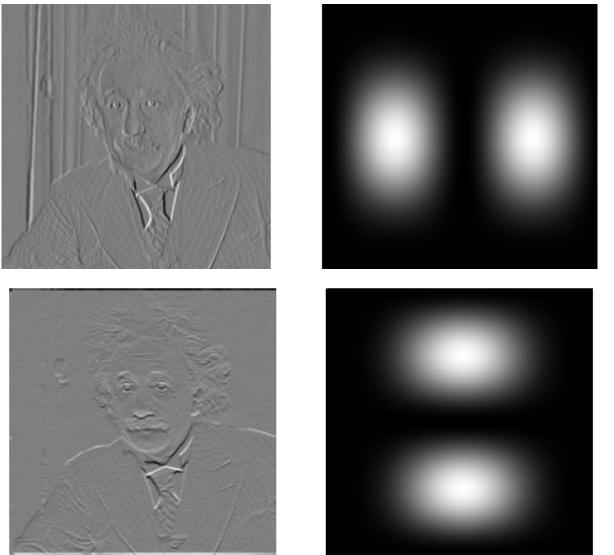
#### Band Pass Filter



a band-pass filtered version of Albert, and the amplitude spectrum of the filter.

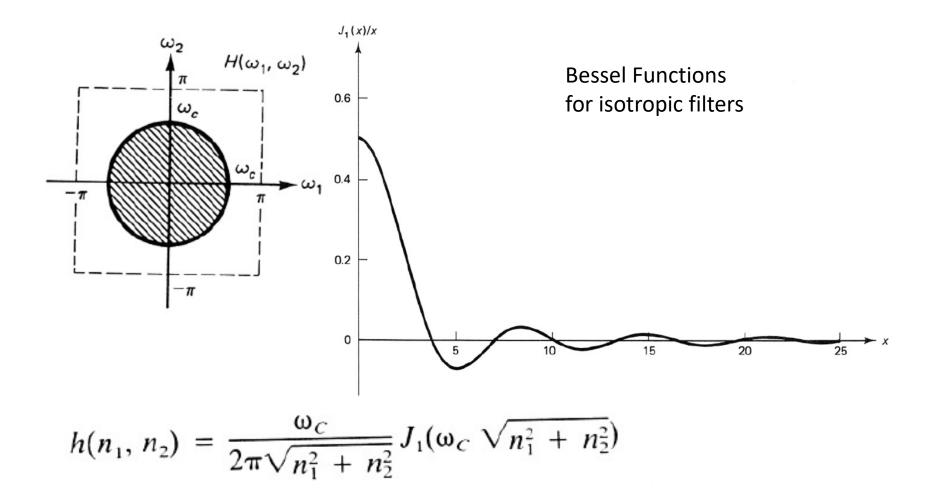
This impulse response is defined by the difference of two low-pass filters.

# Directional filters



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#### Circular filter example



#### Video Signals

# Nyquist Sampling Theorem and Aliasing

Consider a perspective image of an infinite checkerboard.

The signal is dominated by high frequencies in the image near the horizon.

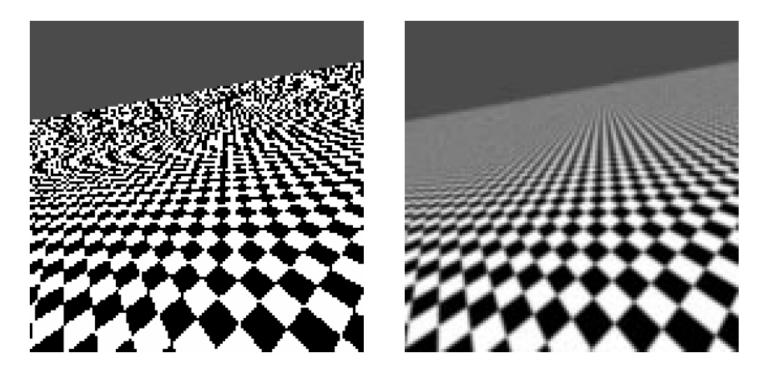
Properly designed cameras blur the signal before sampling using:

- The point spread function due to diffraction
- Imperfect focus
- Averaging the signal over each CCD element.

#### Nyquist Sampling Theorem and Aliasing

These operations attenuate high frequency components in the signal.

Without this (physical) preprocessing, the sampled image can be severely aliased (corrupted):



#### Reconstruction using just phase or intensity



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#### Image superposition combining phases and intesities

