

2D DTFT

For 2D signals, DTFT Analysis and Synthesis Relations are,

$$(S) \quad x(n_1, n_2) = \left(\frac{1}{2\pi}\right)^2 \int_{-\pi}^{\pi} \int_{-\pi}^{\pi} X(\omega_1, \omega_2) e^{j(\omega_1 n_1 + \omega_2 n_2)} d\omega_1 d\omega_2$$

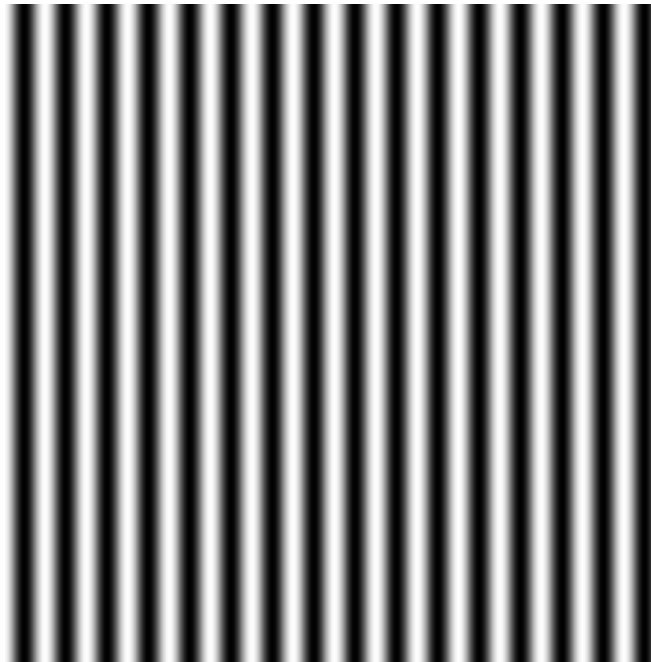
$$(A) \quad X(\omega_1, \omega_2) = \sum_{n_1} \sum_{n_2} x(n_1, n_2) e^{-j(\omega_1 n_1 + \omega_2 n_2)}$$

(S) represents the function $x(n_1, n_2)$ as a linear combination of complex exponentials $\exp(-j(\omega_1 n_1 + \omega_2 n_2))$ with weights $(4\pi^2)^{-1} X(\omega_1, \omega_2) d\omega_1 d\omega_2$.

Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

$$\omega = \pi/8 \quad \theta = 0$$



Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

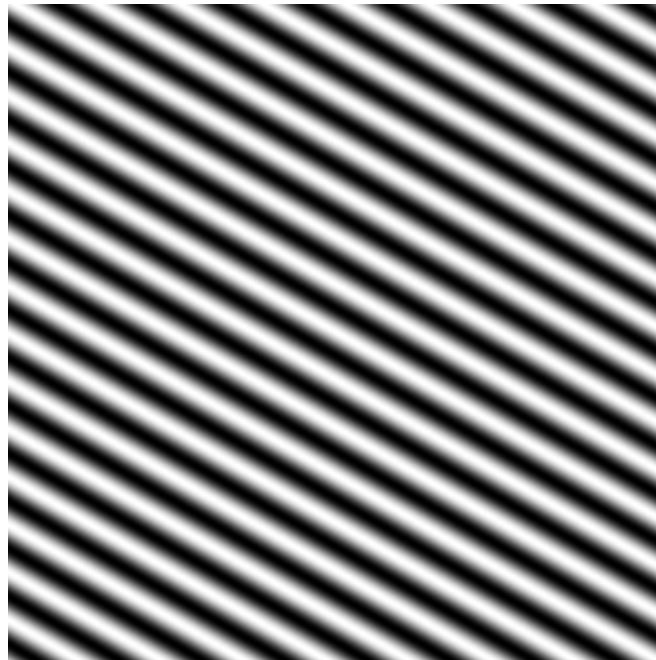
$$\omega = \pi/8 \quad \theta = \pi/6$$



Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

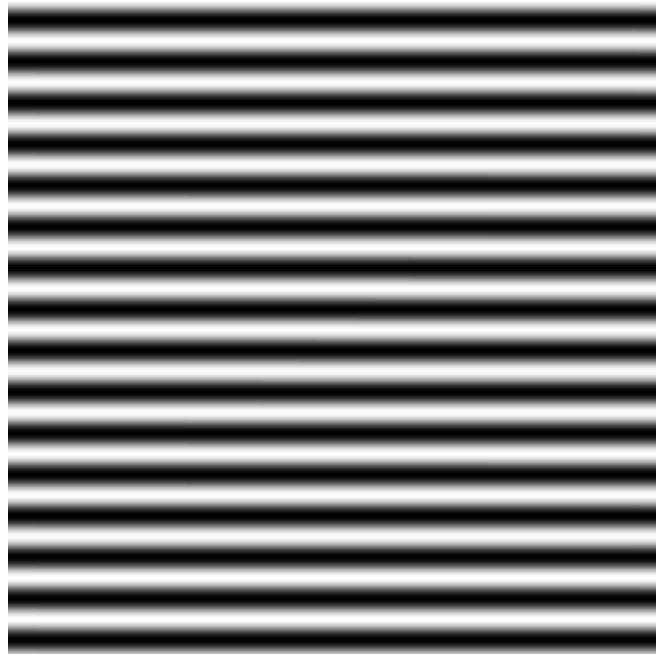
$$\omega = \pi/8 \quad \theta = 2\pi/6$$



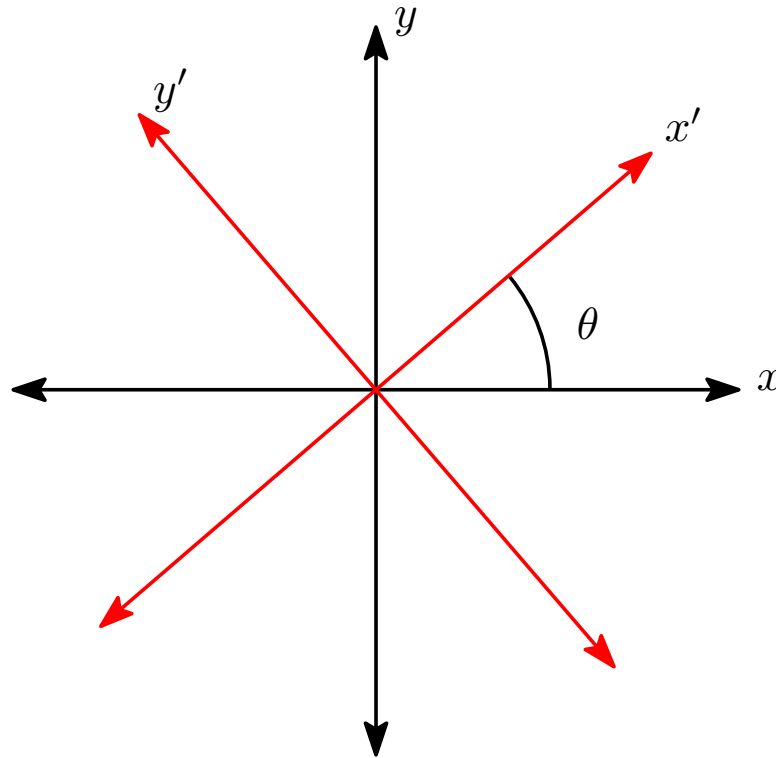
Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

$$\omega = \pi/8 \quad \theta = 3\pi/6$$



Rotating the Coordinate System

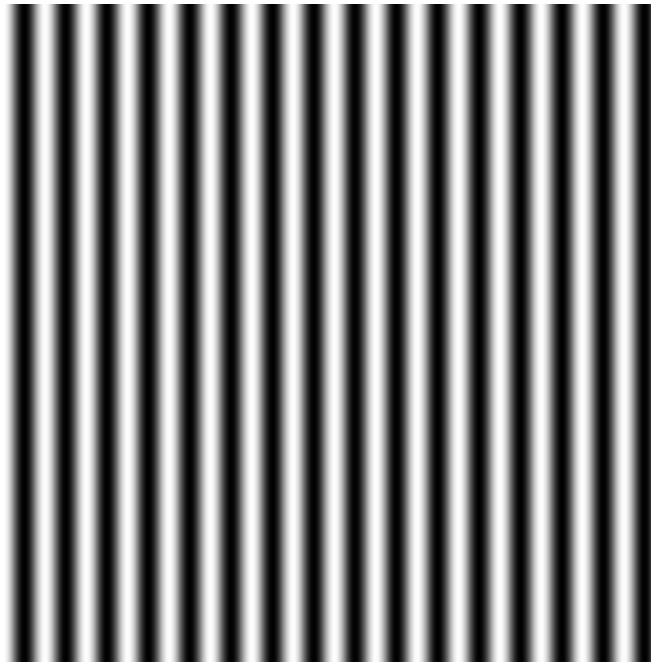


$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

$$\omega = \pi/8 \quad \theta = 0$$



Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

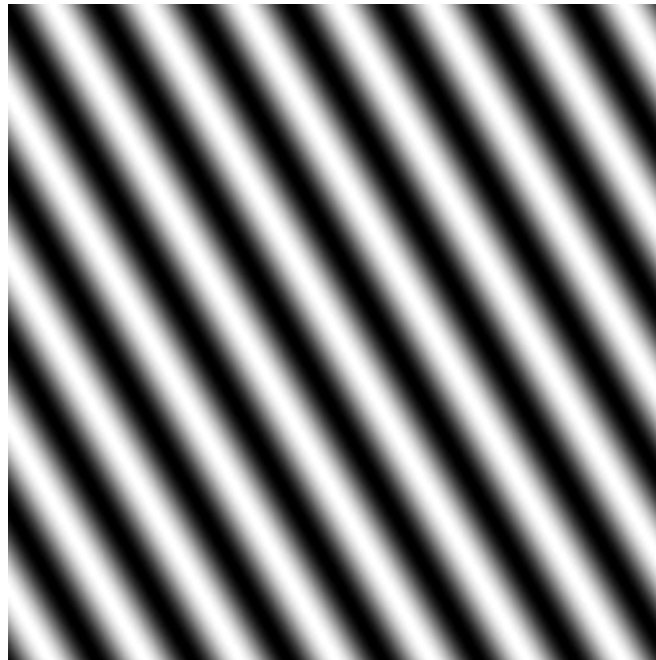
$$\omega = \pi/16 \quad \theta = 0$$



Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

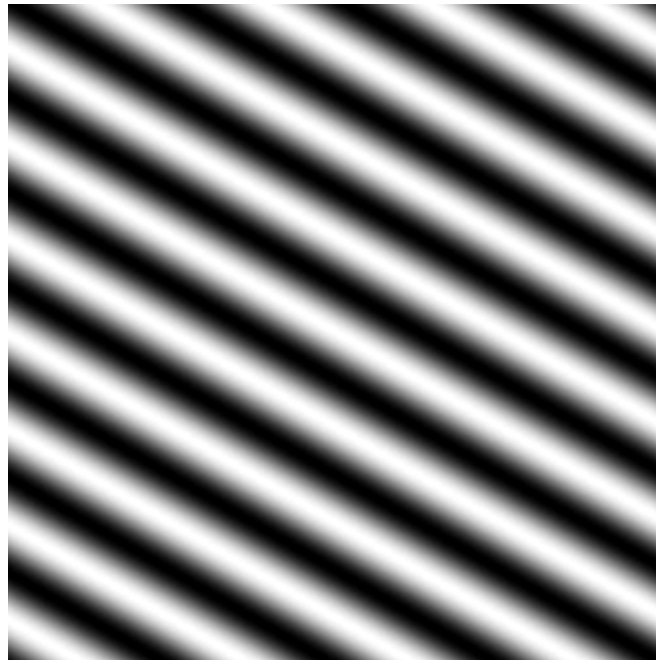
$$\omega = \pi/16 \quad \theta = \pi/6$$



Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

$$\omega = \pi/16 \quad \theta = 2\pi/6$$



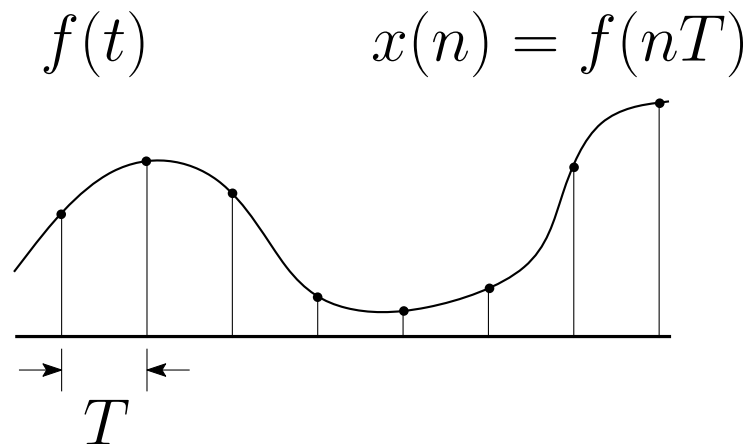
Complex Exponentials

$$\exp\left(j \omega (\cos \theta x + \sin \theta y)\right)$$

$$\omega = \pi/16 \quad \theta = 3\pi/6$$



Sampling – 1D



Suppose we know $f(t)$ only for $t \in \{nT\}_{n \in \mathbb{Z}}$.

Q_1 : Under what conditions can we reconstruct $f(t)$?

Q_2 : How can we reconstruct $f(t)$?

Sampling – 2D

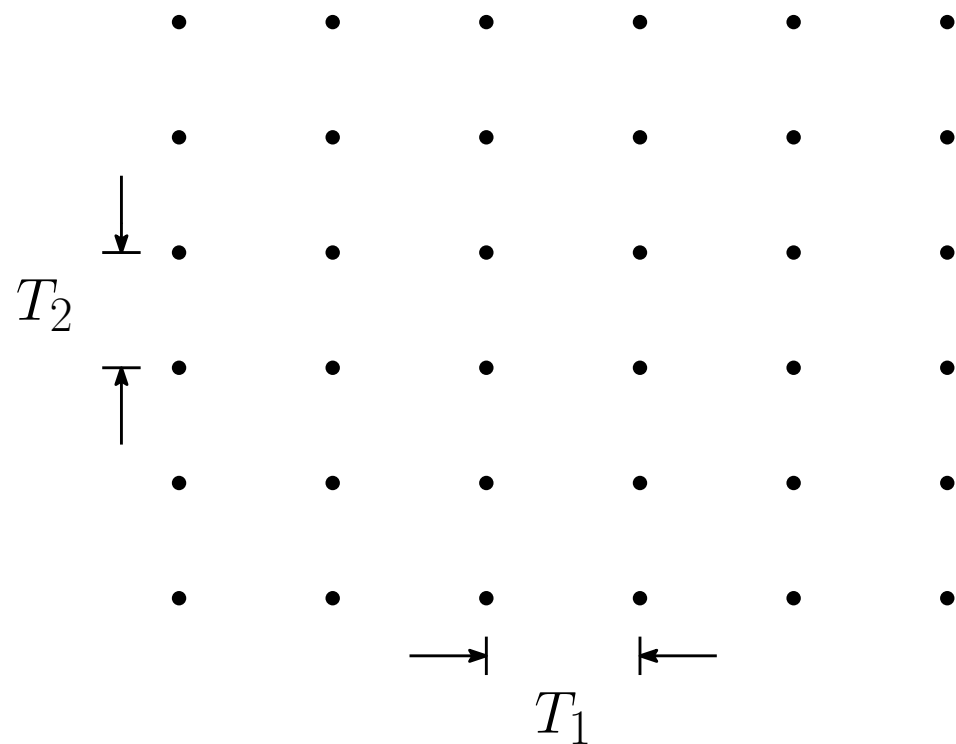


Suppose we know $f(t_1, t_2)$ only the points marked by red dots.

Q_1 : Under what conditions can we reconstruct $f(t_1, t_2)$?

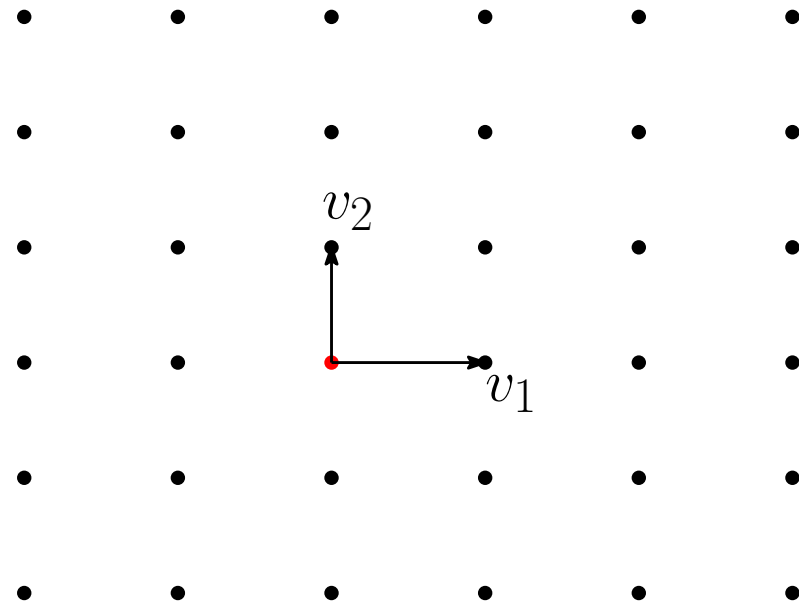
Q_2 : How can we reconstruct $f(t_1, t_2)$?

Rectangular Sampling Lattice



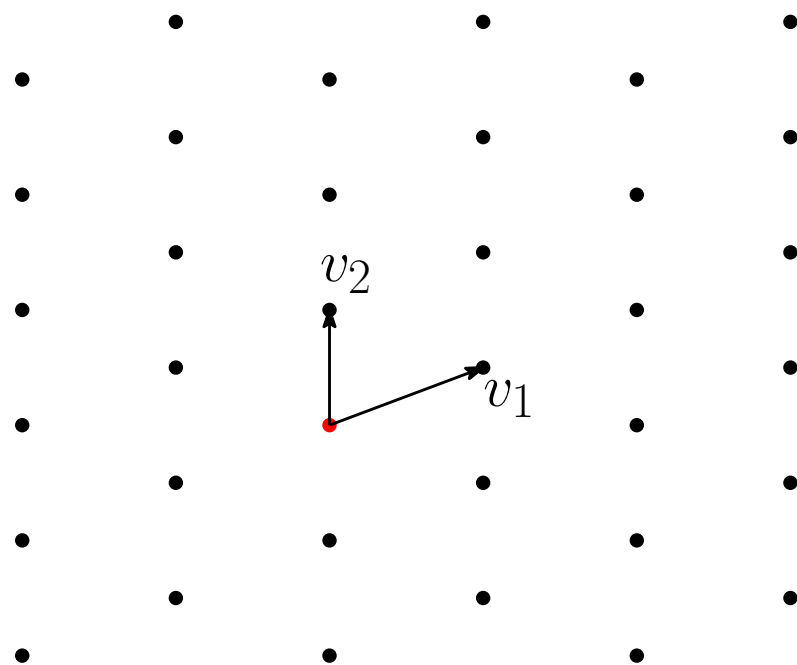
$$\{n_1 T_1, n_2 T_2\}_{n_1, n_2 \in \mathbb{Z}} = \left\{ \begin{bmatrix} T_1 & 0 \\ 0 & T_2 \end{bmatrix} \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} \right\}_{n_1, n_2 \in \mathbb{Z}}$$

Rectangular Sampling Lattice



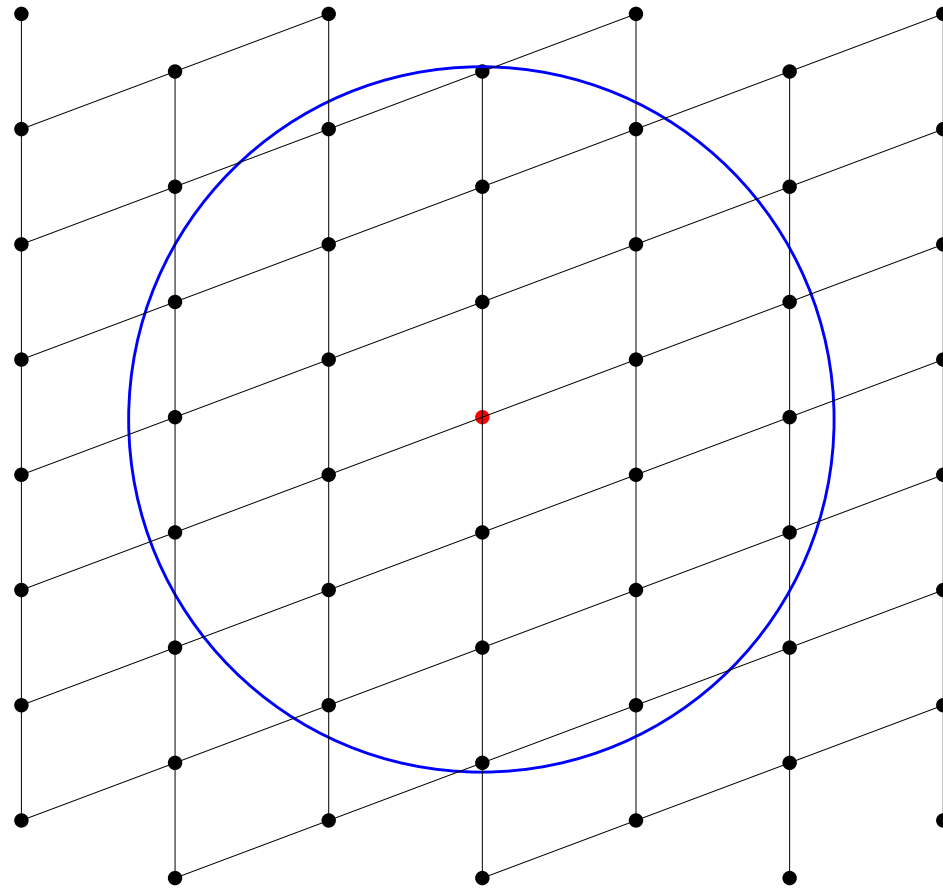
$$\{n_1 T_1, n_2 T_2\}_{n_1, n_2 \in \mathbb{Z}} = \left\{ \begin{bmatrix} T_1 & 0 \\ 0 & T_2 \end{bmatrix} \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} \right\}_{n_1, n_2 \in \mathbb{Z}}$$

General Sampling Lattice



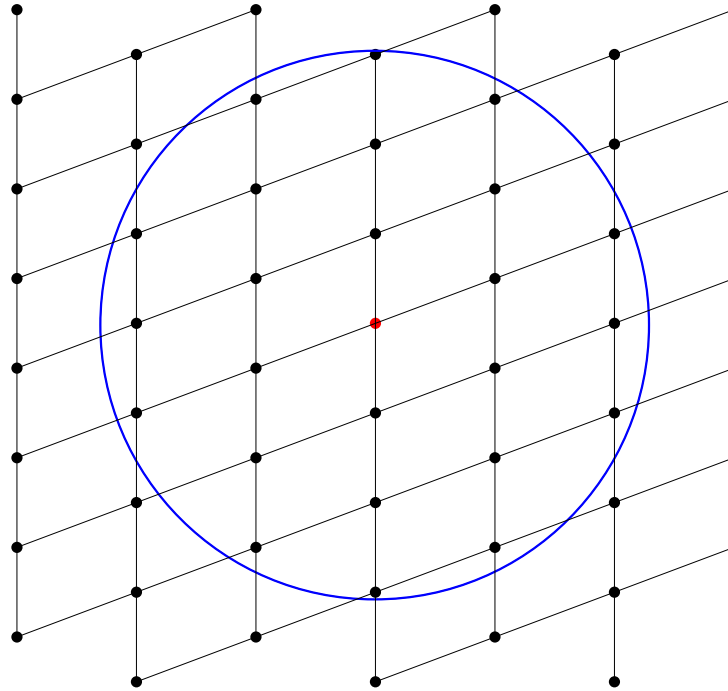
$$\{n_1 \mathbf{v}_1 + n_2 \mathbf{v}_2\}_{n_1, n_2 \in \mathbb{Z}} = \left\{ \underbrace{\begin{bmatrix} \mathbf{v}_1 & \mathbf{v}_2 \end{bmatrix}}_V \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} \right\}_{n_1, n_2 \in \mathbb{Z}}$$

Sampling Density



Sampling Density = Number of Samples per Unit Area

Sampling Density



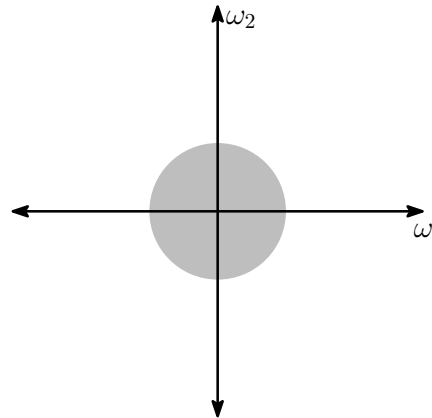
Sampling Density = Number of Samples per Unit Area

$$\rho = \lim_{R \rightarrow \infty} \frac{\text{Number of parallelograms}}{\text{Area of Disk}} \propto \frac{1}{|\det(V)|}$$

Rectangular vs. Hexagonal Sampling

Hexagonal Sampling :

$$V = c \begin{bmatrix} 1 & \sqrt{3}/2 \\ 0 & 1/2 \end{bmatrix}$$



For a function with frequency support restricted to a disk,

$$\rho_{\min}(\text{Rectangular Sampling}) \approx 1.15 \rho_{\min}(\text{Hexagonal Sampling})$$

Irregular Sampling



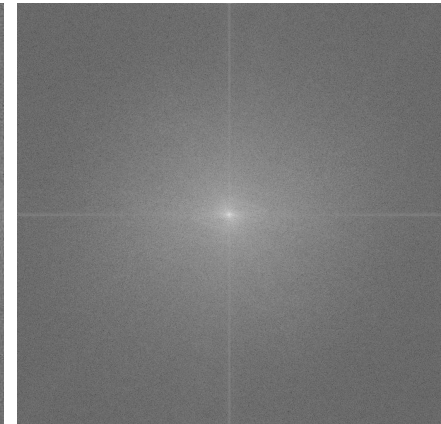
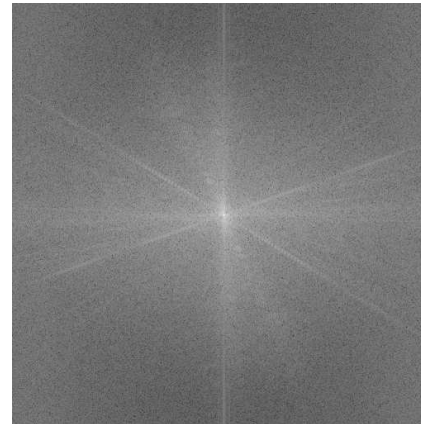
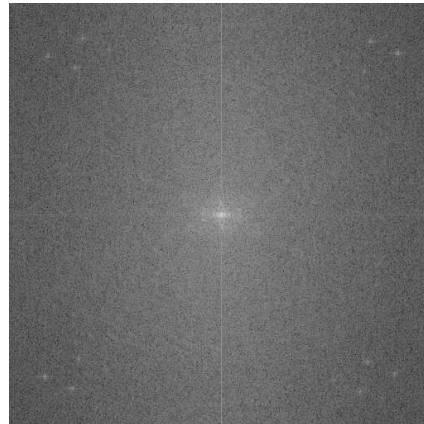
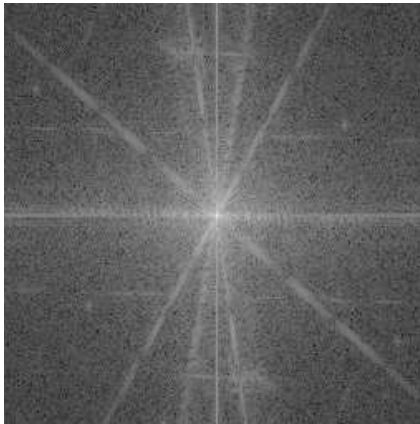
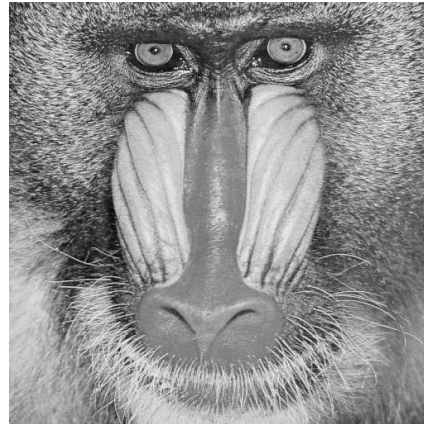
Irregular sampling arises in biomedical imaging problems like MRI, CT, diffraction tomography etc.

Magnitude and Phase

Fourier Transform = Magnitude $\times e^{j \text{Phase}}$

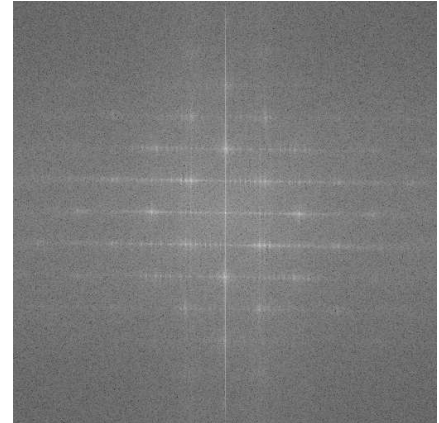
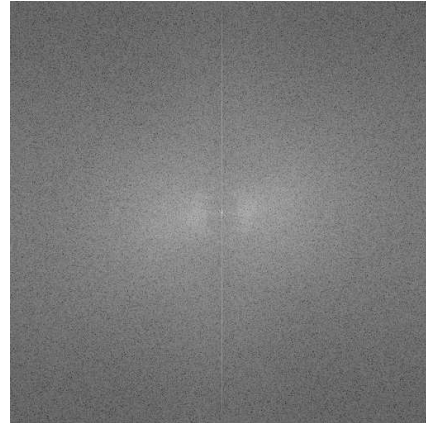
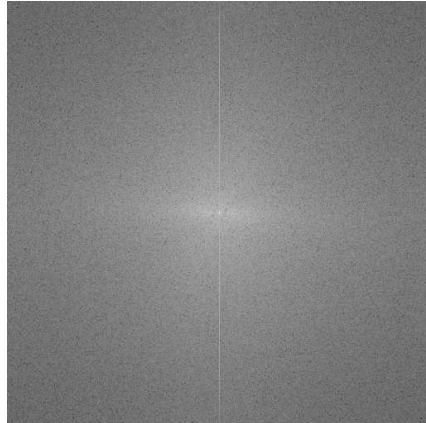
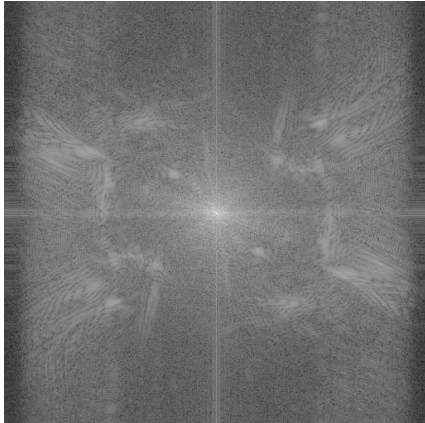
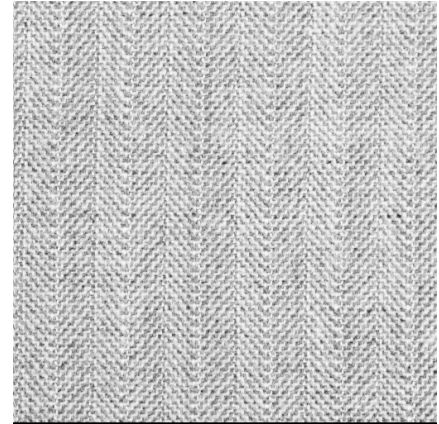
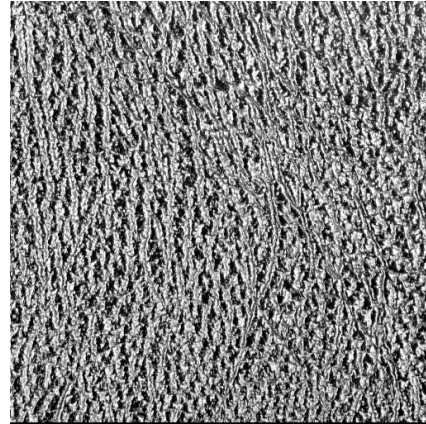
$$X(\omega_1, \omega_2) = |X(\omega_1, \omega_2)| \exp(j \angle X(\omega_1, \omega_2))$$

Typical Fourier Transform Magnitude of Natural Images

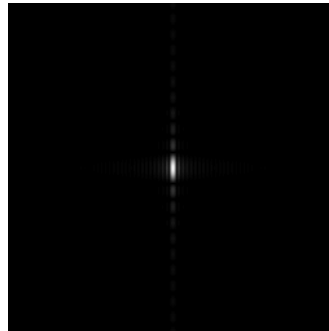
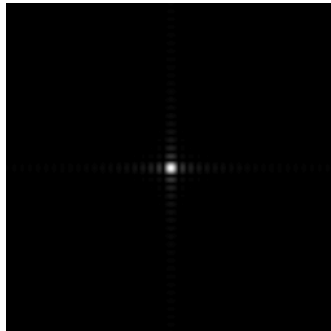
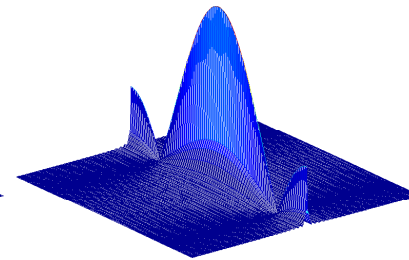
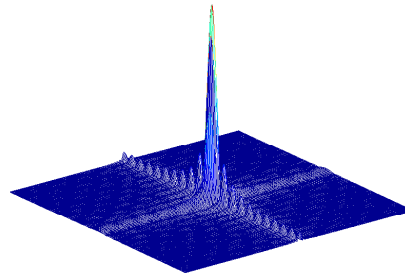
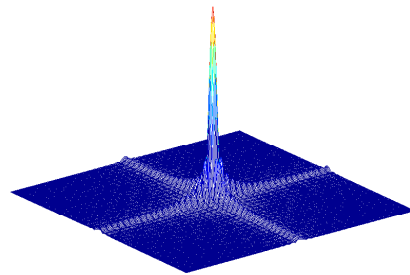
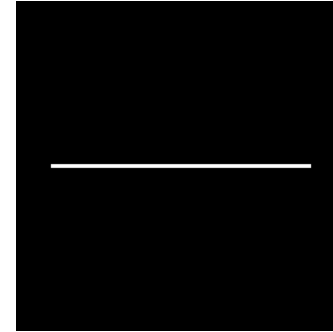
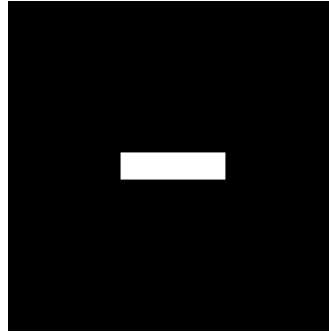
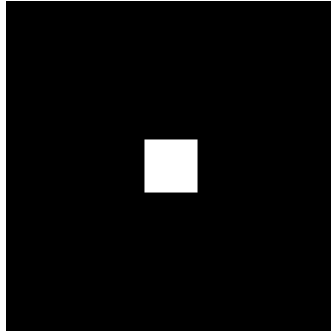


Dominant lowpass behavior...

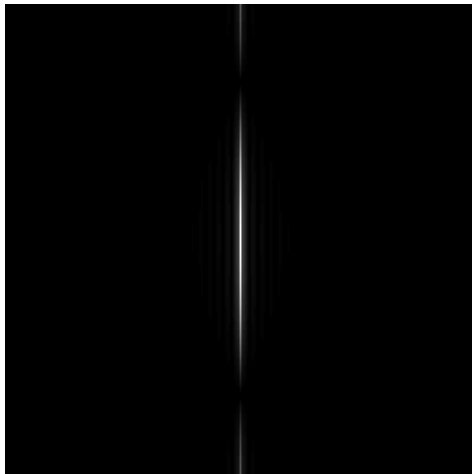
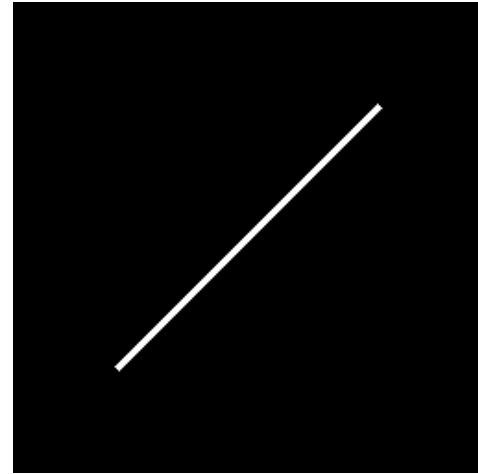
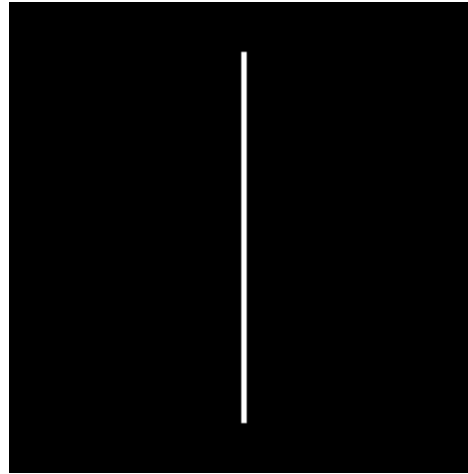
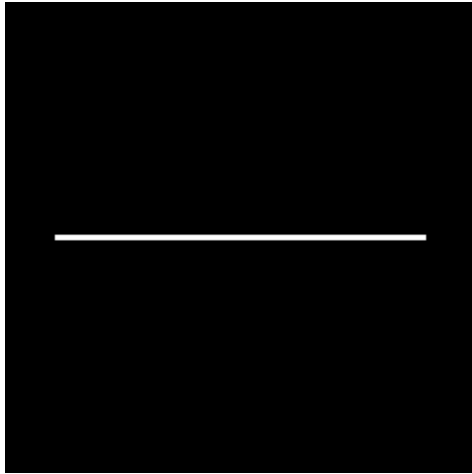
Texture



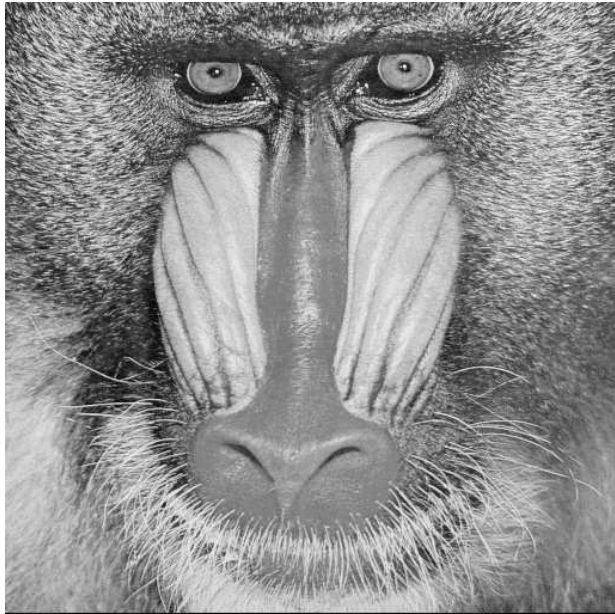
Fourier Transform of a Box / Line



Fourier Transform of a Box / Line



Magnitude vs. Phase



$$X(\omega_1, \omega_2)$$

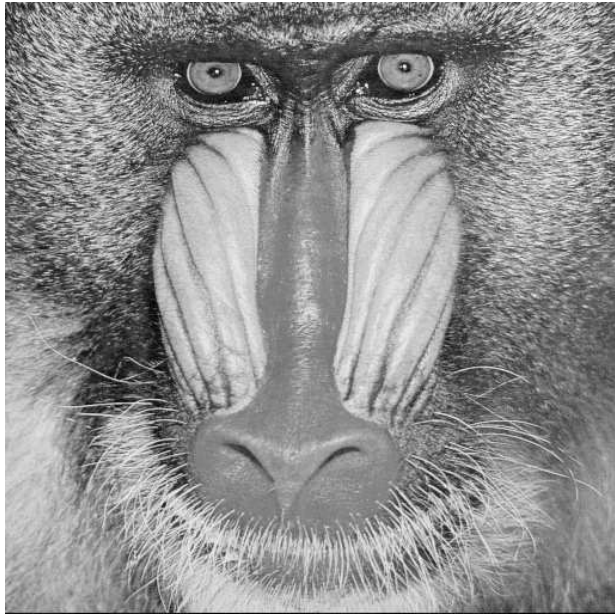


$$Y(\omega_1, \omega_2)$$



$$|X| \exp(j \angle Y)$$

Magnitude vs. Phase



$$X(\omega_1, \omega_2)$$

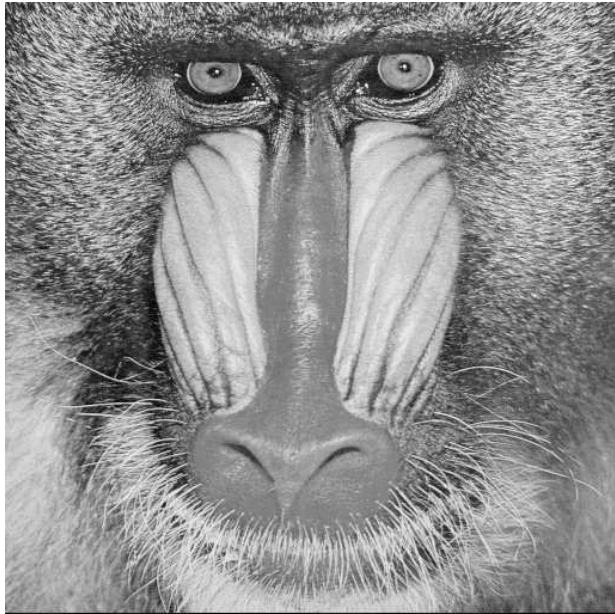


$$Y(\omega_1, \omega_2)$$



$$|X| \exp(j \angle Y)$$

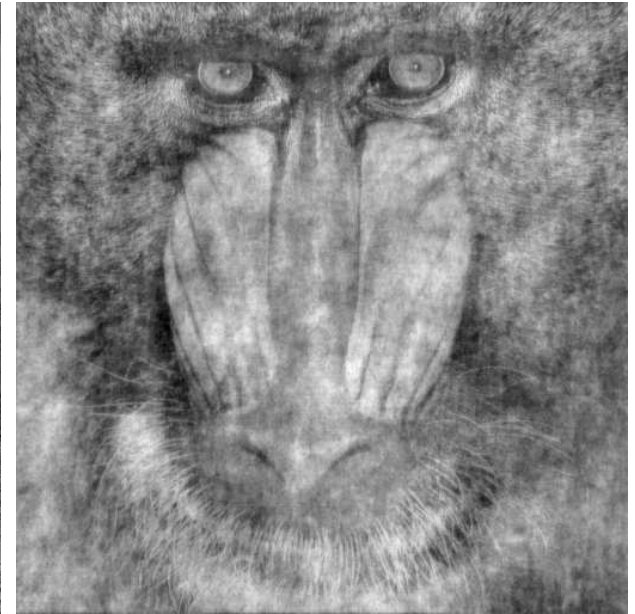
Magnitude vs. Phase



$$X(\omega_1, \omega_2)$$



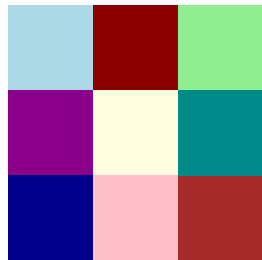
$$Y(\omega_1, \omega_2)$$



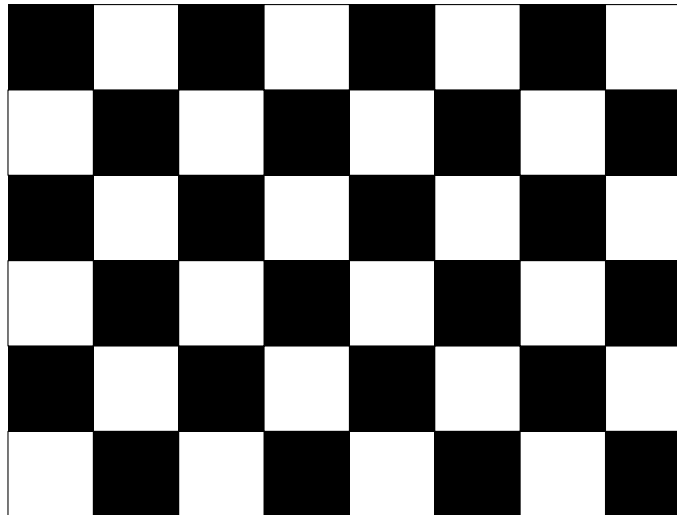
$$|Y| \exp(j \angle X)$$

Convolution

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

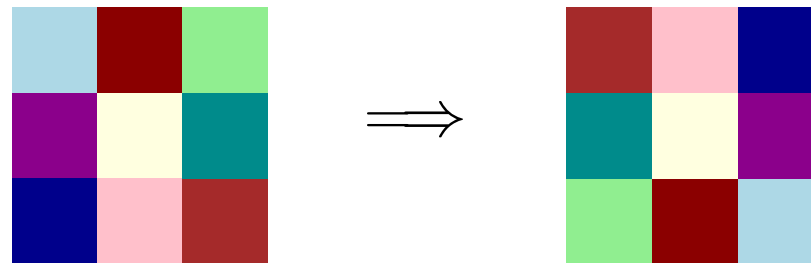


**



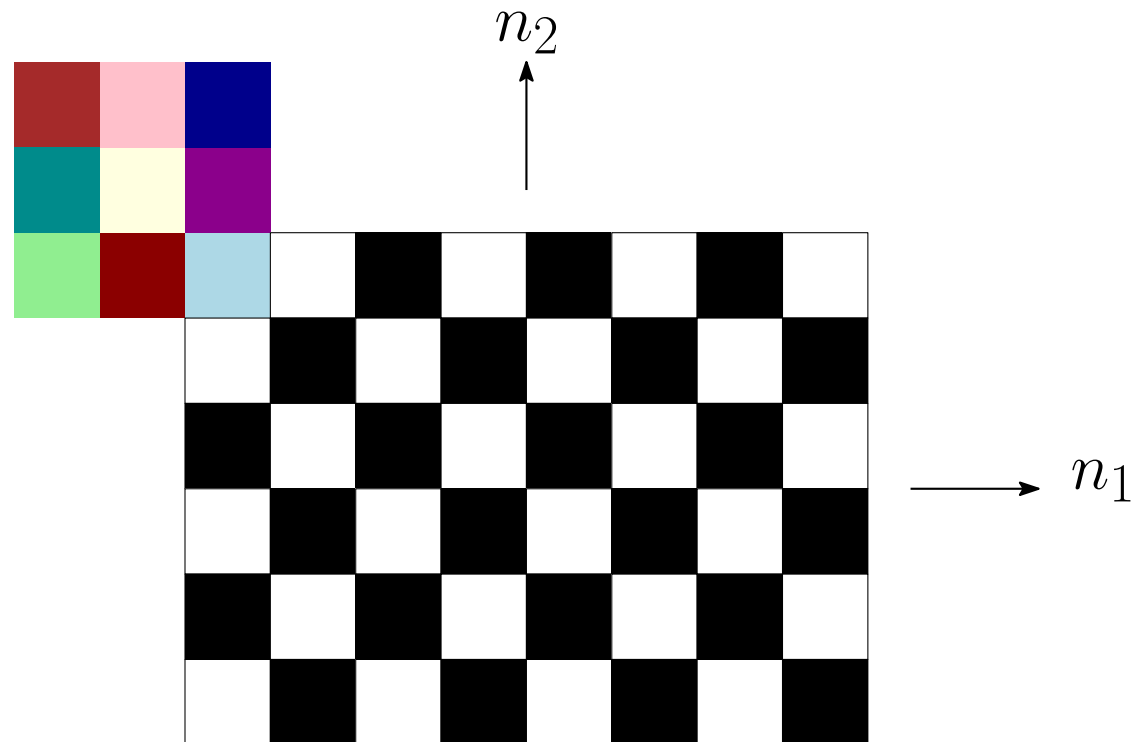
Convolution – Step 1

Flip the kernel in both directions.



Convolution – Step 2

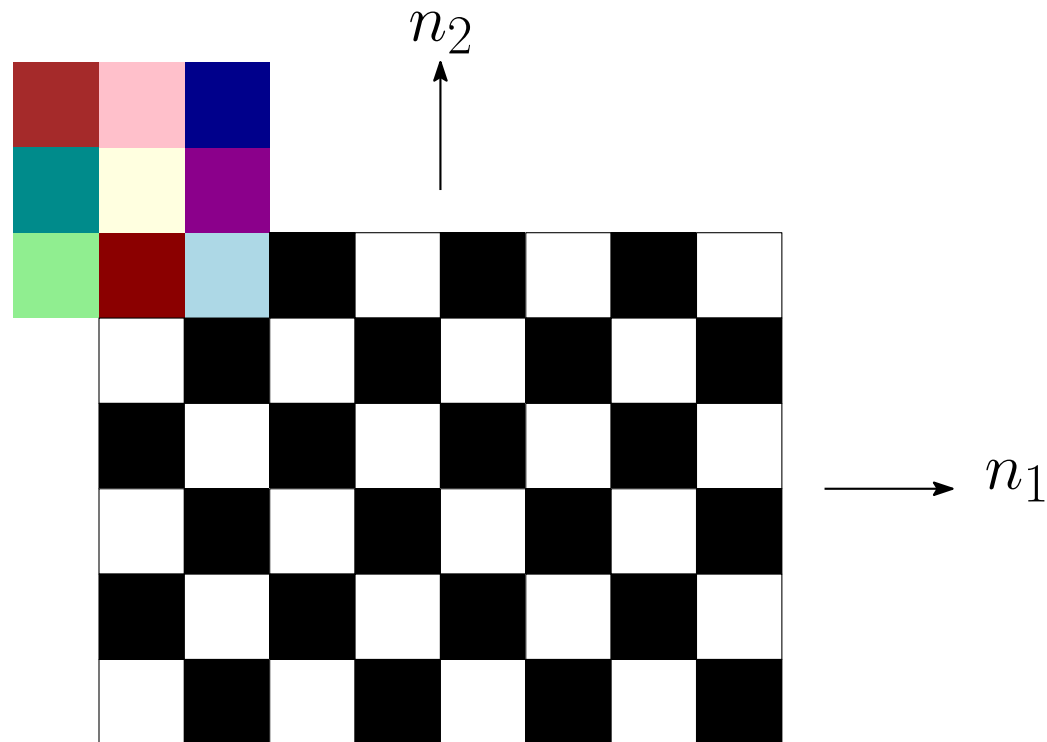
Compute the inner product of the flipped mask with the image.



$$\{h * x\}(m, l)$$

Convolution – Step 2

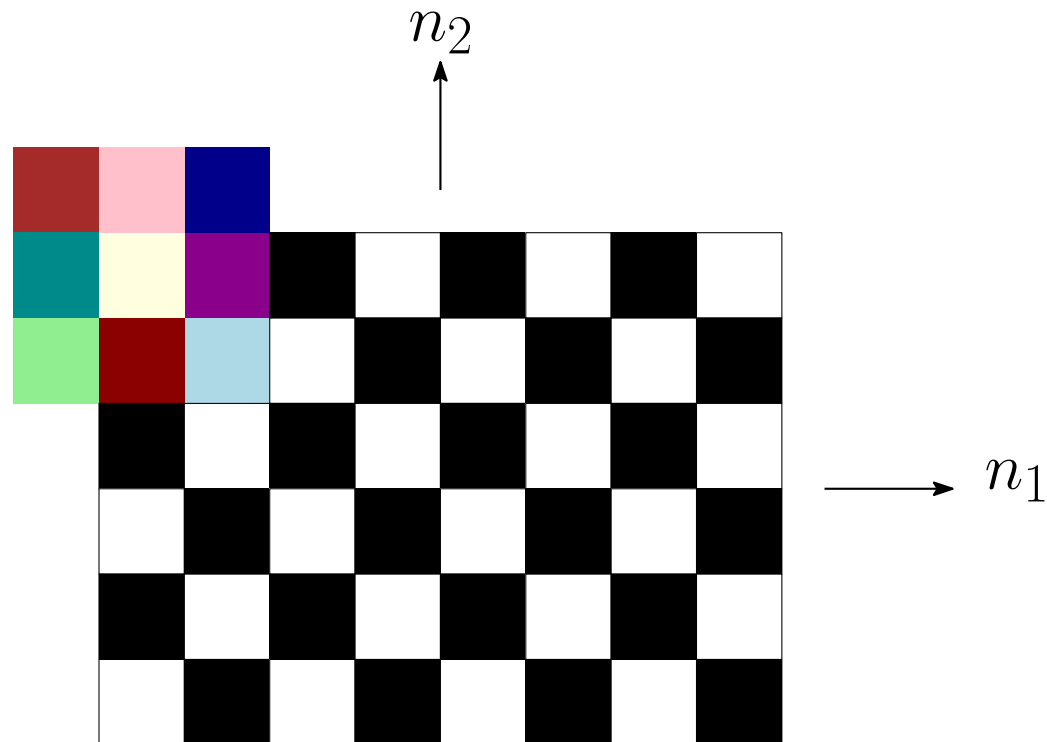
Shift the mask and repeat.



$$\{h * x\}(m + 1, l)$$

Convolution – Step 2

Shift the mask and repeat.



$$\{h * x\}(\cdot, \cdot)$$

Convolution – Some Special Cases

$1/9$	$1/9$	$1/9$
$1/9$	$1/9$	$1/9$
$1/9$	$1/9$	$1/9$

**



= ?

Convolution – Some Special Cases



Convolution – Some Special Cases

$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$

**



= ?

Convolution – Some Special Cases



Convolution – Some Special Cases

1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49

**



= ?

Convolution – Some Special Cases



Convolution – Some Special Cases

$\delta(n_1, n_2)$ –

1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49
1/49	1/49	1/49	1/49	1/49	1/49	1/49

**



= ?

Convolution – Some Special Cases



Convolution – Some Special Cases

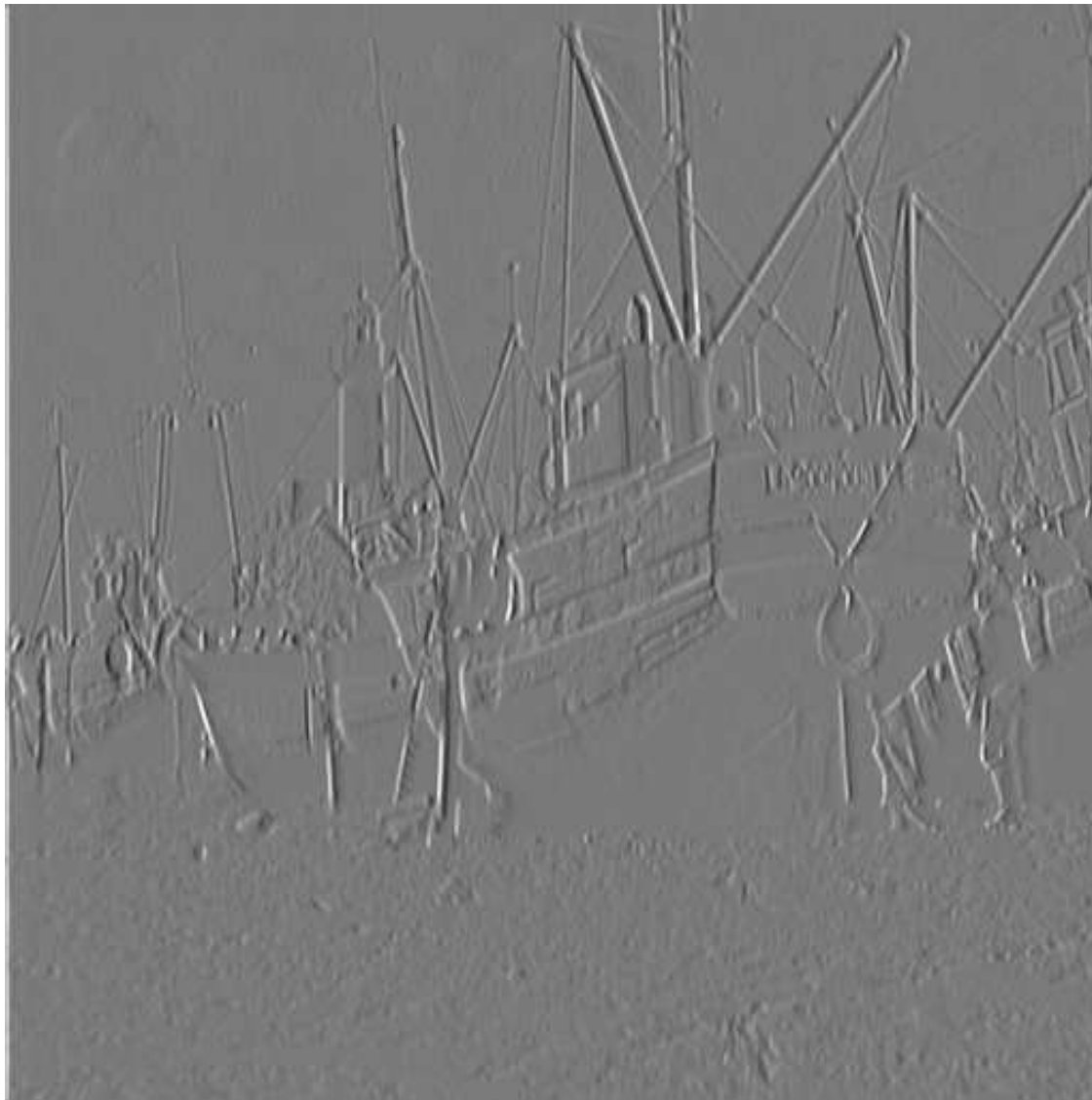
1	0	-1
1	0	-1
1	0	-1

**



= ?

Convolution – Some Special Cases



Convolution – Some Special Cases

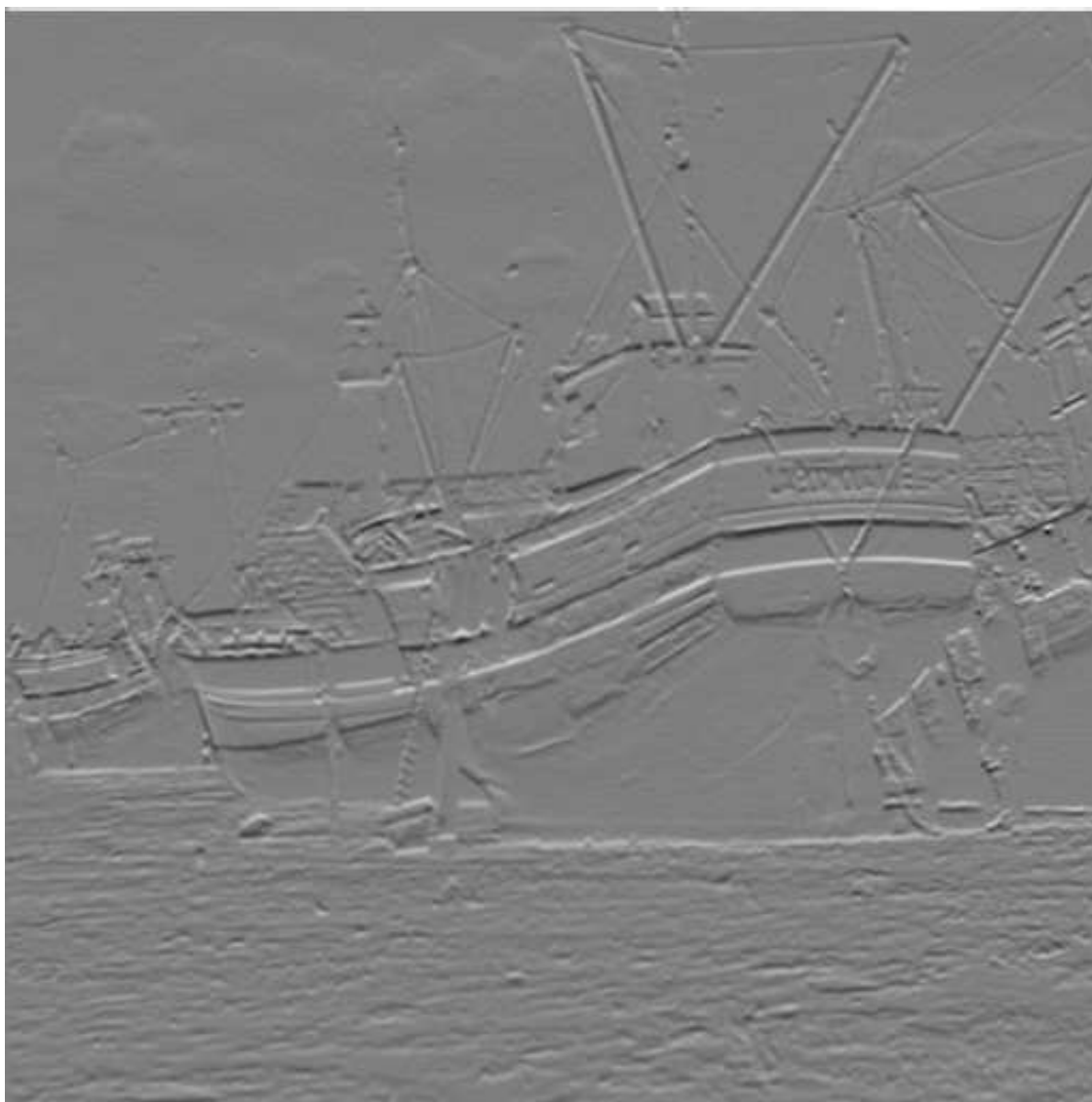
1	1	1
0	0	0
-1	-1	-1

**



= ?

Convolution – Some Special Cases



Convolution Theorem

Convolution in the time domain corresponds to multiplication in the frequency domain.

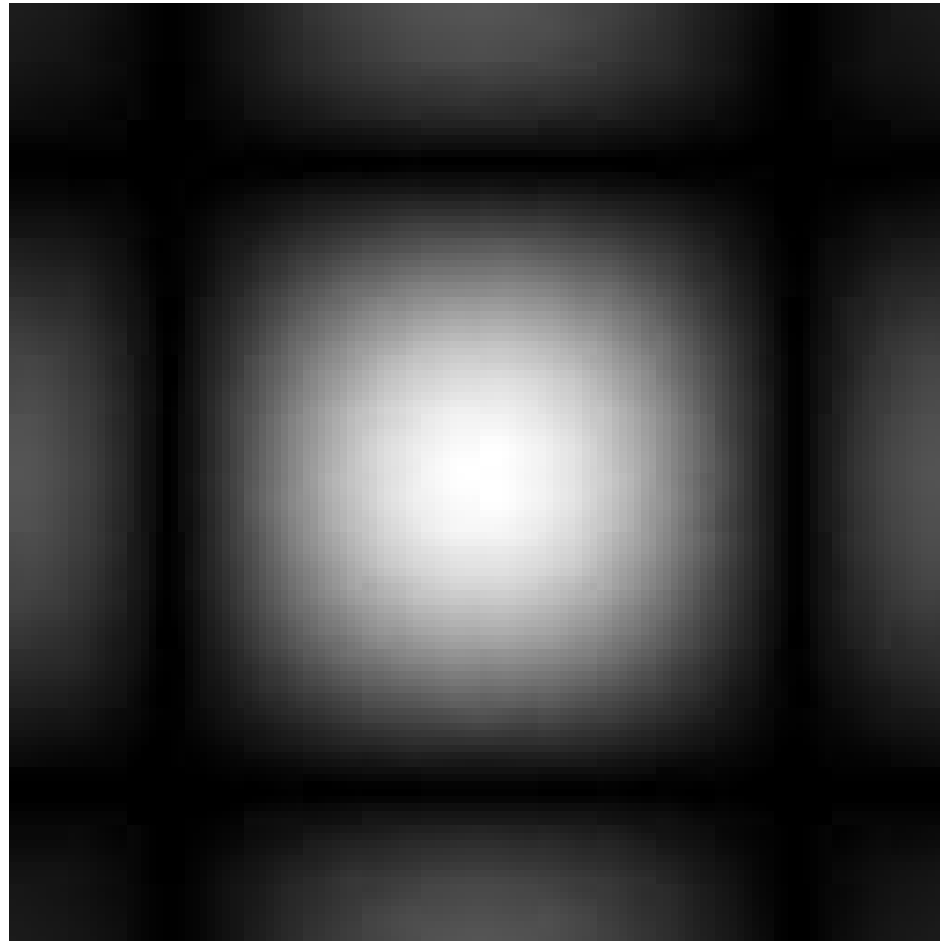
$$y(n_1, n_2) = h(n_1, n_2) * * x(n_1, n_2)$$

$$\Longleftrightarrow$$

$$Y(\omega_1, \omega_2) = H(\omega_1, \omega_2) X(\omega_1, \omega_2)$$

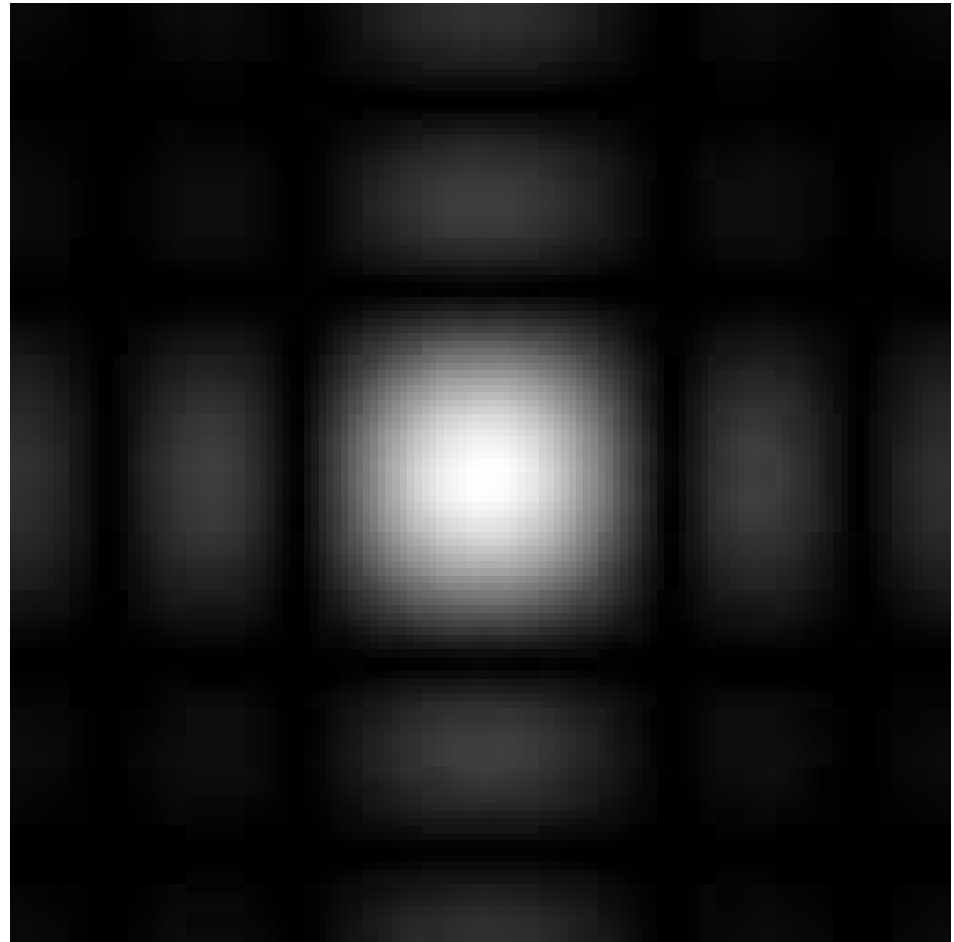
Fourier Transform of Some Masks

$1/9$	$1/9$	$1/9$
$1/9$	$1/9$	$1/9$
$1/9$	$1/9$	$1/9$



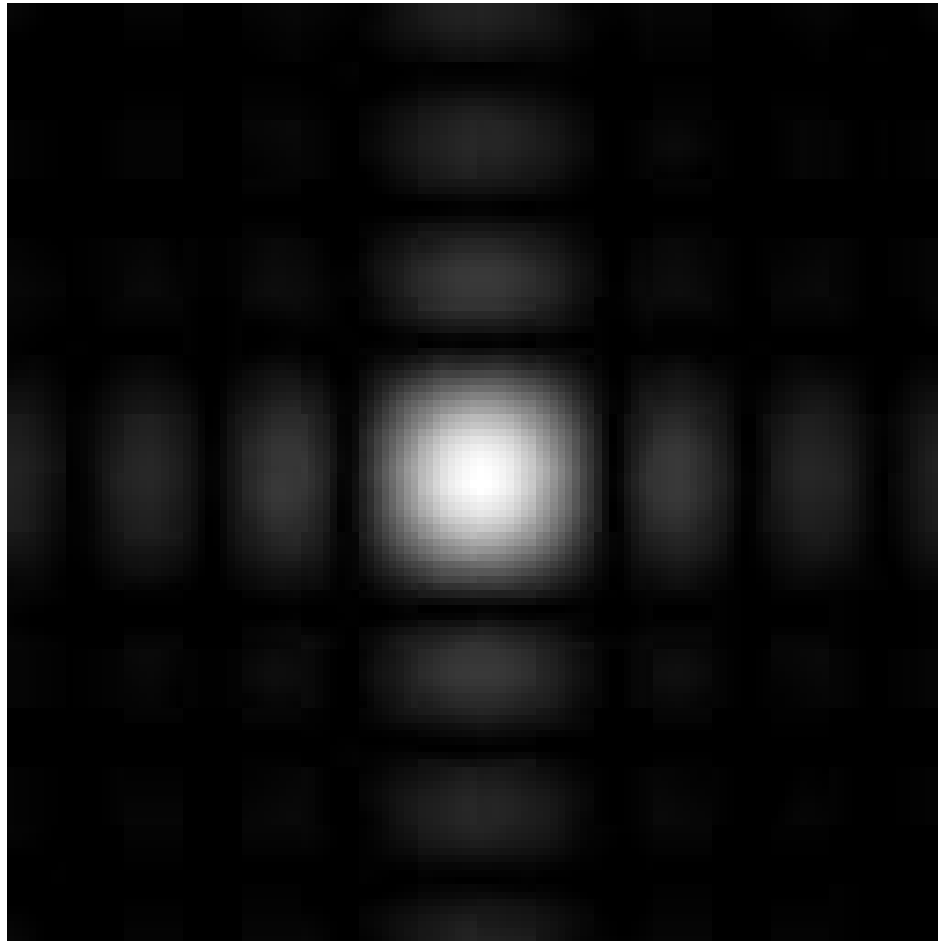
Fourier Transform of Some Masks

$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$
$1/25$	$1/25$	$1/25$	$1/25$	$1/25$



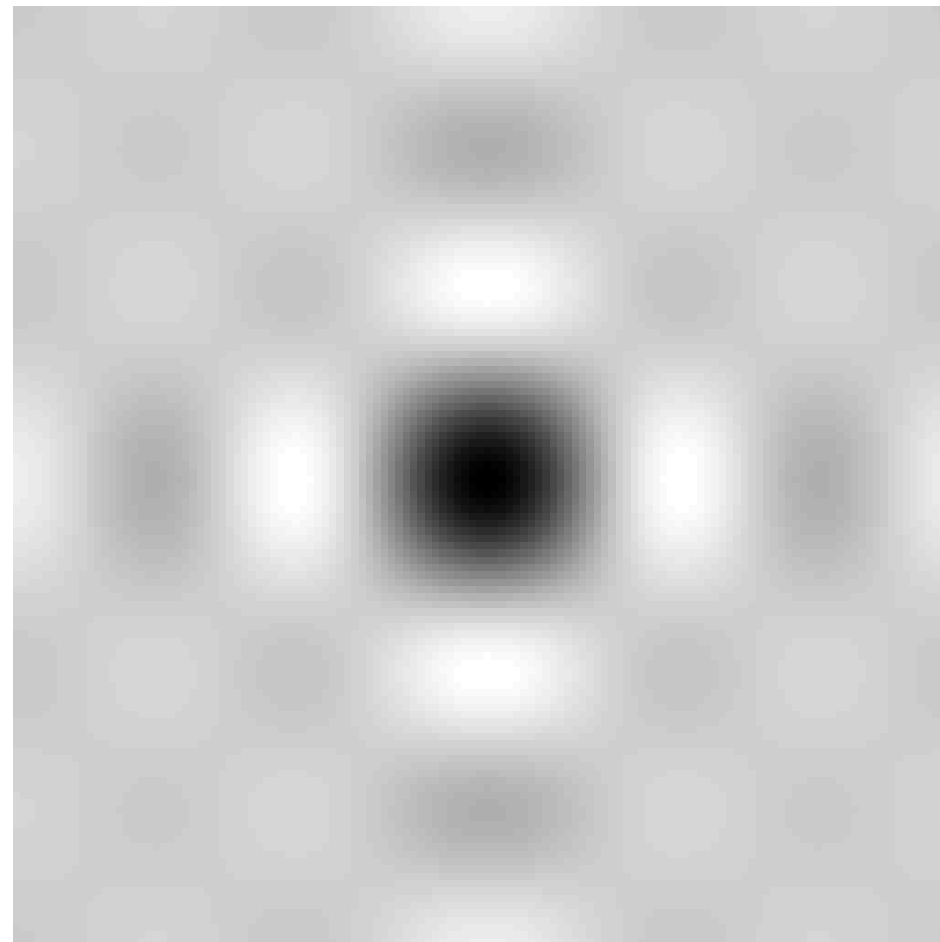
Fourier Transform of Some Masks

$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$
$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$	$1/49$



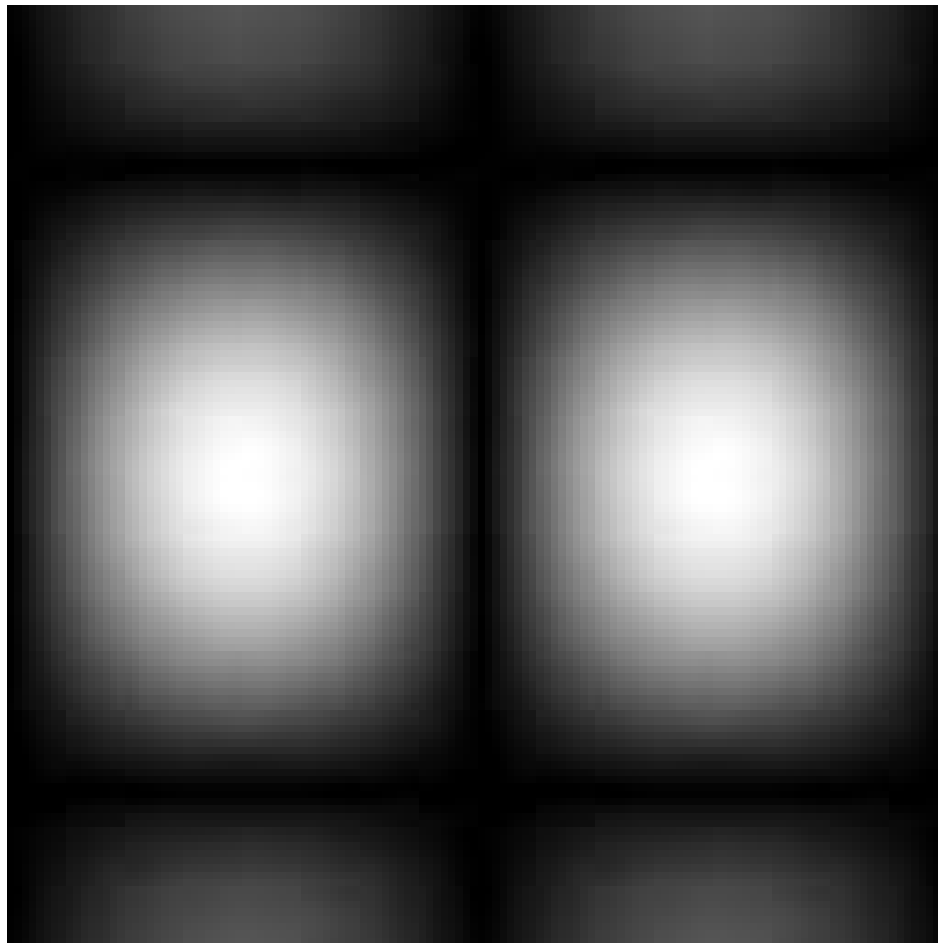
Fourier Transform of Some Masks

$$\delta(n_1, n_2) \rightarrow \begin{pmatrix} 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \\ 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 & 1/49 \end{pmatrix} \Rightarrow$$



Fourier Transform of Some Masks

1	0	-1
1	0	-1
1	0	-1



Fourier Transform of Some Masks

1	1	1
0	0	0
-1	-1	-1

