

Bioengineering 280A  
Principles of Biomedical Imaging

Fall Quarter 2014  
MRI Lecture 3

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## 2D Fourier Transform

Fourier Transform

$$G(k_x, k_y) = F[g(x, y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) e^{-j2\pi(k_x x + k_y y)} dx dy$$

Inverse Fourier Transform

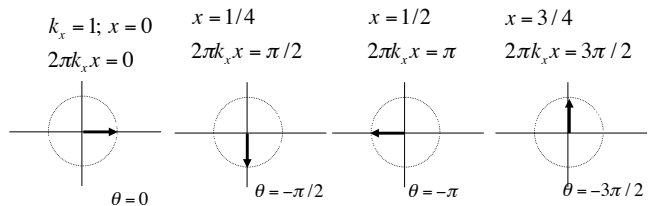
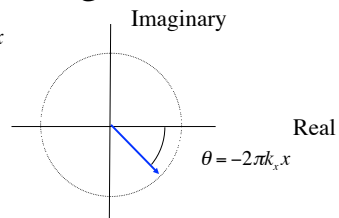
$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G(k_x, k_y) e^{j2\pi(k_x x + k_y y)} dk_x dk_y$$

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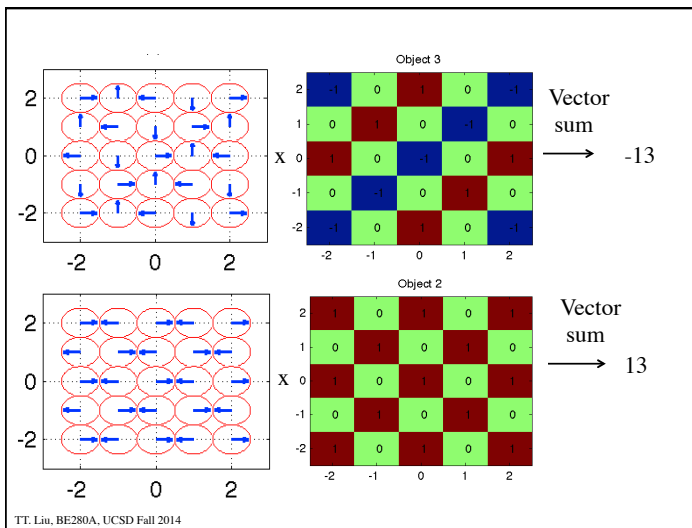
## Phasor Diagram

$$G(k_x) = \int_{-\infty}^{\infty} g(x) \exp(-j2\pi k_x x) dx$$

$\theta = -2\pi k_x x$



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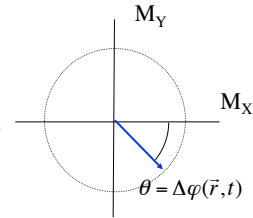


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## Phase

$$M(\vec{r}, t) = M(\vec{r}, 0)e^{j\varphi(\vec{r}, t)}$$

Phase = angle of the magnetization phasor  
 Frequency = rate of change of angle (e.g. radians/sec)  
 Phase = time integral of frequency

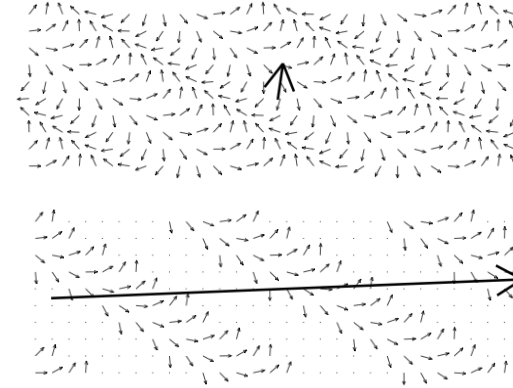


$$\begin{aligned}\varphi(\vec{r}, t) &= -\int_0^t \omega(\vec{r}, \tau) d\tau \\ &= -\omega_0 t + \Delta\varphi(\vec{r}, t)\end{aligned}$$

Where the incremental phase due to the gradients is

$$\begin{aligned}\Delta\varphi(\vec{r}, t) &= -\int_0^t \Delta\omega(\vec{r}, \tau) d\tau \\ &= -\int_0^t \gamma \vec{G}(\vec{r}, \tau) \cdot \vec{r} d\tau\end{aligned}$$

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Hanson 2009

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## Time-Varying Gradient Fields

The transverse magnetization is then given by

$$\begin{aligned}M(\vec{r}, t) &= M(\vec{r}, 0)e^{-t/T_2(\vec{r})}e^{j\varphi(\vec{r}, t)} \\ &= M(\vec{r}, 0)e^{-t/T_2(\vec{r})}e^{-j\omega_0 t} \exp\left(-j\int_0^t \Delta\omega(\vec{r}, \tau) d\tau\right) \\ &= M(\vec{r}, 0)e^{-t/T_2(\vec{r})}e^{-j\omega_0 t} \exp\left(-j\gamma\int_0^t \vec{G}(\tau) \cdot \vec{r} d\tau\right)\end{aligned}$$

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## Signal Equation

Signal from a volume

$$\begin{aligned}s_r(t) &= \int_V M(\vec{r}, t) dV \\ &= \int_x \int_y \int_z M(x, y, z, 0)e^{-t/T_2(\vec{r})}e^{-j\omega_0 t} \exp\left(-j\gamma\int_0^t \vec{G}(\tau) \cdot \vec{r} d\tau\right) dx dy dz\end{aligned}$$

For now, consider signal from a slice along  $z$  and drop the  $T_2$  term. Define  $m(x, y) \equiv \int_{z_0-\Delta z/2}^{z_0+\Delta z/2} M(\vec{r}, t) dz$

To obtain

$$s_r(t) = \int_x \int_y m(x, y)e^{-j\omega_0 t} \exp\left(-j\gamma\int_0^t \vec{G}(\tau) \cdot \vec{r} d\tau\right) dx dy$$

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## Signal Equation

Demodulate the signal to obtain

$$\begin{aligned}
 s(t) &= e^{j\omega_0 t} s_r(t) \\
 &= \int_x \int_y m(x, y) \exp\left(-j\gamma \int_0^t \vec{G}(\tau) \cdot \vec{r} d\tau\right) dx dy \\
 &= \int_x \int_y m(x, y) \exp\left(-j\gamma \int_0^t [G_x(\tau)x + G_y(\tau)y] d\tau\right) dx dy \\
 &= \int_x \int_y m(x, y) \exp\left(-j2\pi(k_x(t)x + k_y(t)y)\right) dx dy
 \end{aligned}$$

Where

$$\begin{aligned}
 k_x(t) &= \frac{\gamma}{2\pi} \int_0^t G_x(\tau) d\tau \\
 k_y(t) &= \frac{\gamma}{2\pi} \int_0^t G_y(\tau) d\tau
 \end{aligned}$$

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## MR signal is Fourier Transform

$$\begin{aligned}
 s(t) &= \int_x \int_y m(x, y) \exp\left(-j2\pi(k_x(t)x + k_y(t)y)\right) dx dy \\
 &= M(k_x(t), k_y(t)) \\
 &= F[m(x, y)]_{k_x(t), k_y(t)}
 \end{aligned}$$

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## K-space

At each point in time, the received signal is the Fourier transform of the object

$$s(t) = M(k_x(t), k_y(t)) = F[m(x, y)]_{k_x(t), k_y(t)}$$

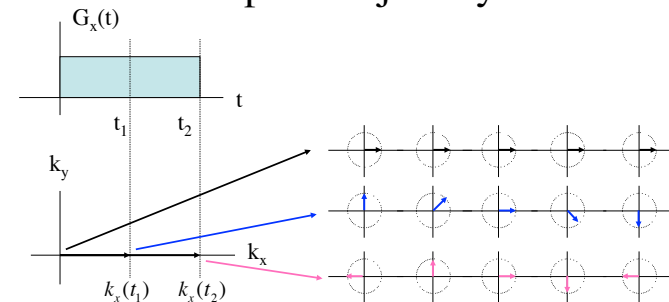
evaluated at the spatial frequencies:

$$\begin{aligned}
 k_x(t) &= \frac{\gamma}{2\pi} \int_0^t G_x(\tau) d\tau \\
 k_y(t) &= \frac{\gamma}{2\pi} \int_0^t G_y(\tau) d\tau
 \end{aligned}$$

Thus, the gradients control our position in k-space. The design of an MRI pulse sequence requires us to efficiently cover enough of k-space to form our image.

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## K-space trajectory



$$k_x(t) = \frac{\gamma}{2\pi} \int_0^t G_x(\tau) d\tau$$

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## Units

Spatial frequencies ( $k_x, k_y$ ) have units of 1/distance.  
Most commonly, 1/cm

Gradient strengths have units of (magnetic field)/  
distance. Most commonly G/cm or mT/m

$\gamma/(2\pi)$  has units of Hz/G or Hz/Tesla.

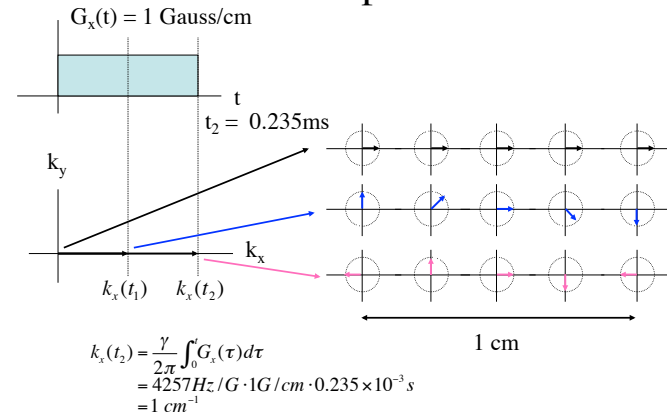
$$k_x(t) = \frac{\gamma}{2\pi} \int_0^t G_x(\tau) d\tau$$

$$= [\text{Hz/Gauss}][\text{Gauss/cm}][\text{sec}]$$

$$= [1/\text{cm}]$$

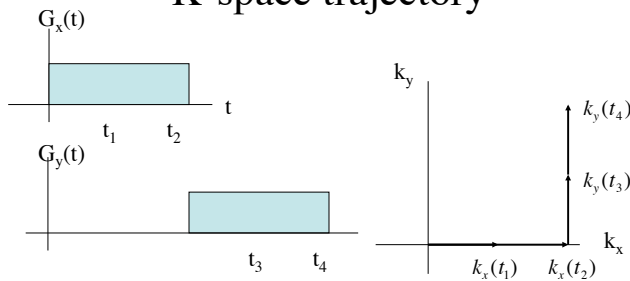
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## Example

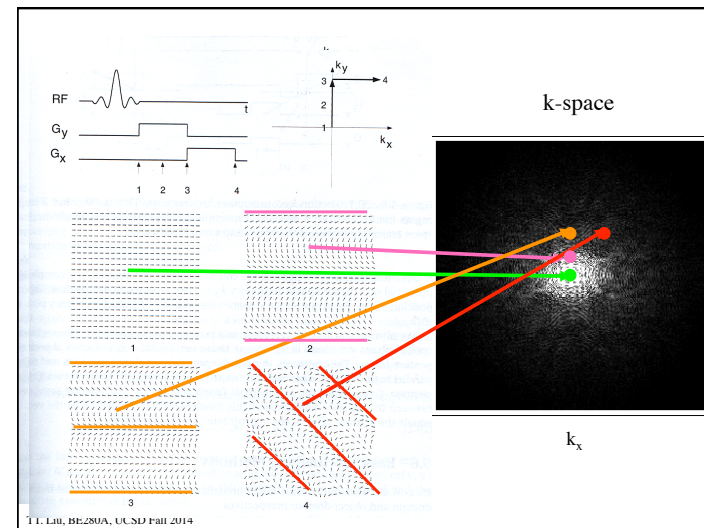


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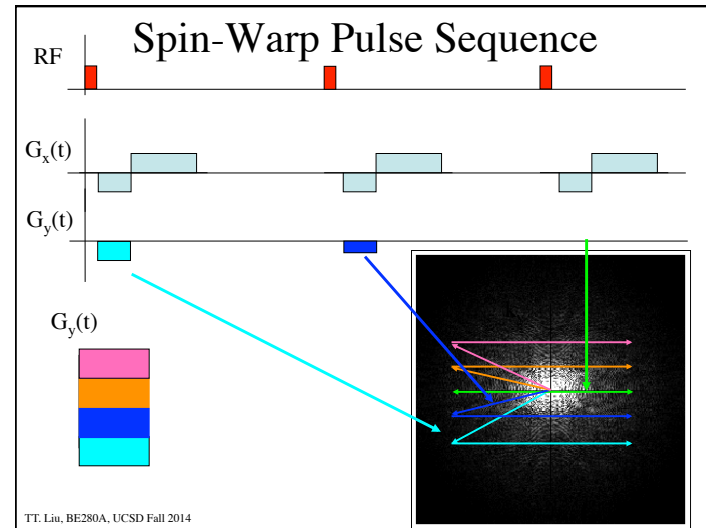
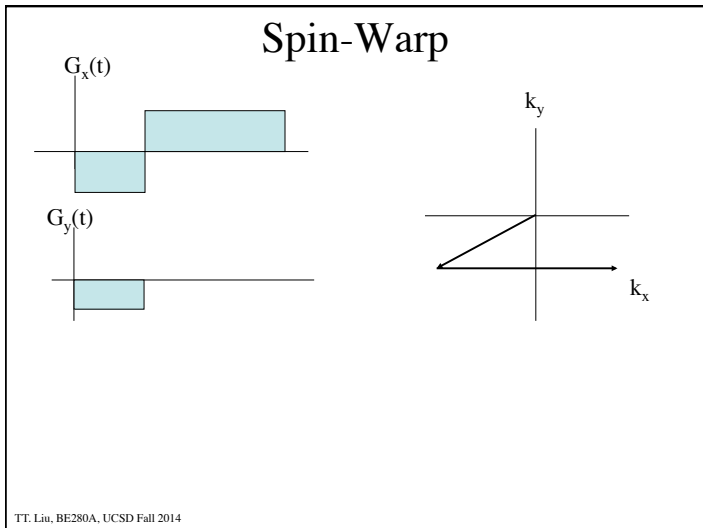
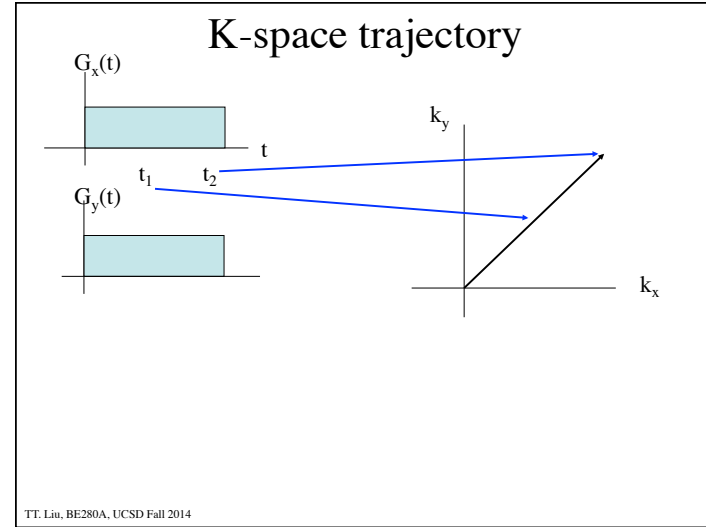
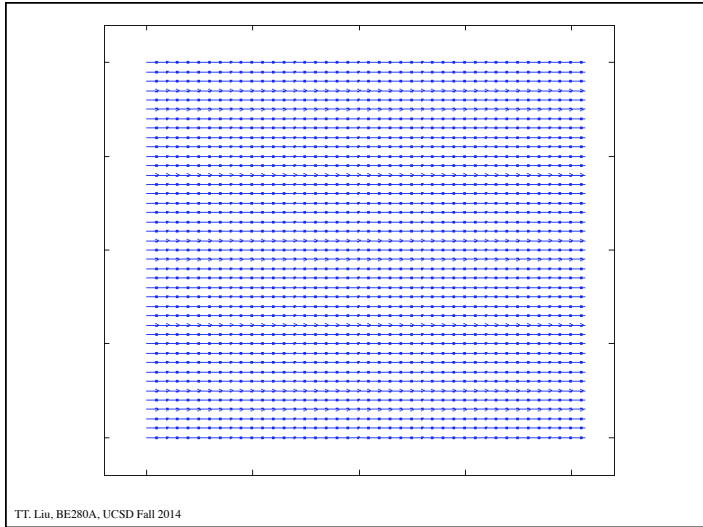
## K-space trajectory

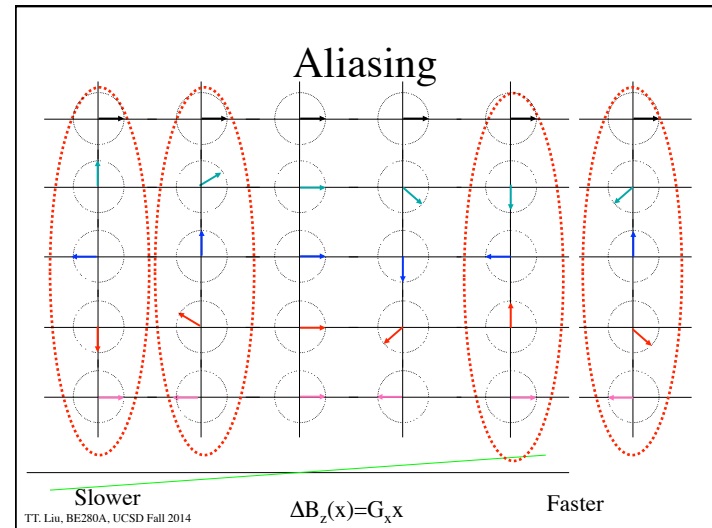
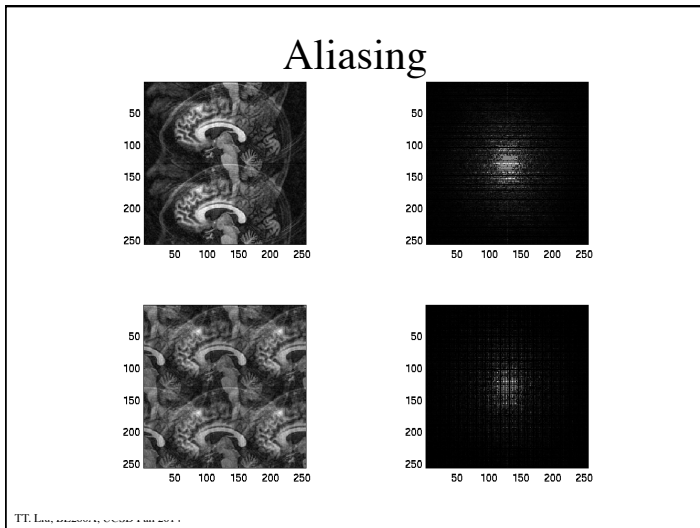
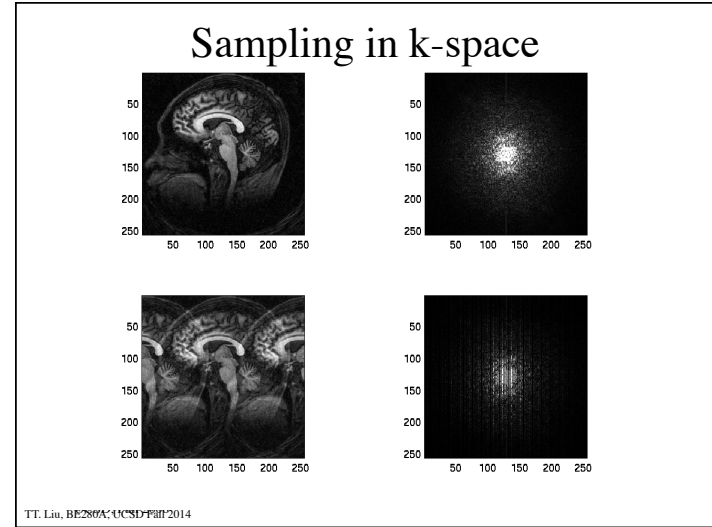
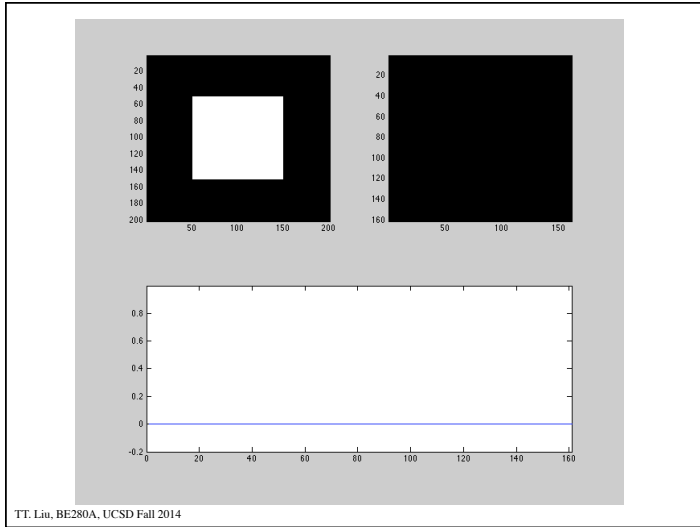


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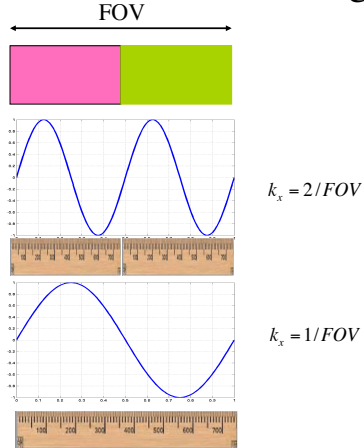


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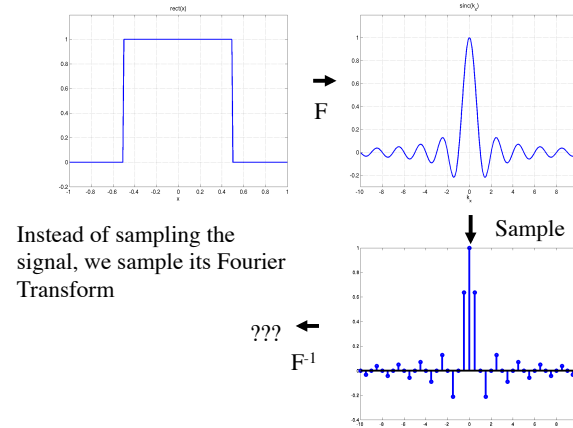


## Intuitive view of Aliasing



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## Fourier Sampling



Instead of sampling the signal, we sample its Fourier Transform

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## Fourier Sampling

$$(1/\Delta k_x) \text{comb}(k_x/\Delta k_x)$$

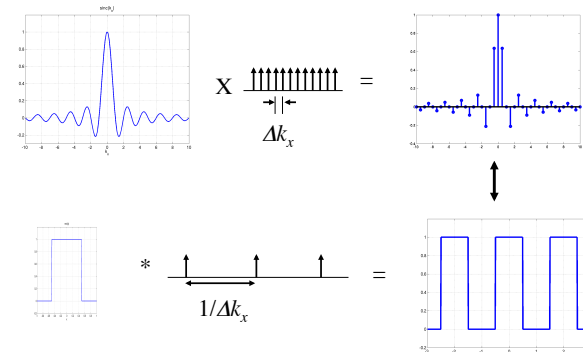
$$G_S(k_x) = G(k_x) \frac{1}{\Delta k_x} \text{comb}\left(\frac{k_x}{\Delta k_x}\right)$$

$$= G(k_x) \sum_{n=-\infty}^{\infty} \delta(k_x - n\Delta k_x)$$

$$= \sum_{n=-\infty}^{\infty} G(n\Delta k_x) \delta(k_x - n\Delta k_x)$$

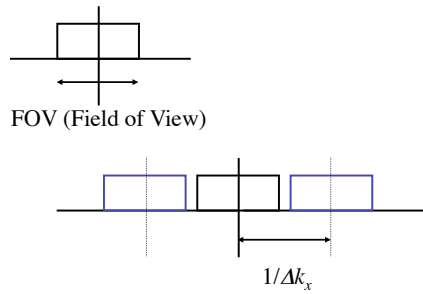
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## Fourier Sampling -- Inverse Transform



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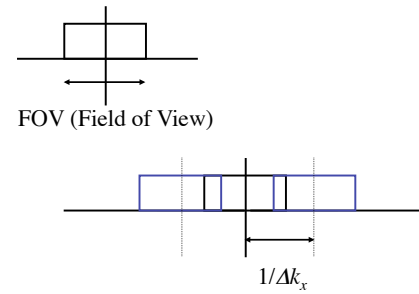
## Nyquist Condition



To avoid overlap,  $1/\Delta k_x > \text{FOV}$ , or equivalently,  $\Delta k_x < 1/\text{FOV}$

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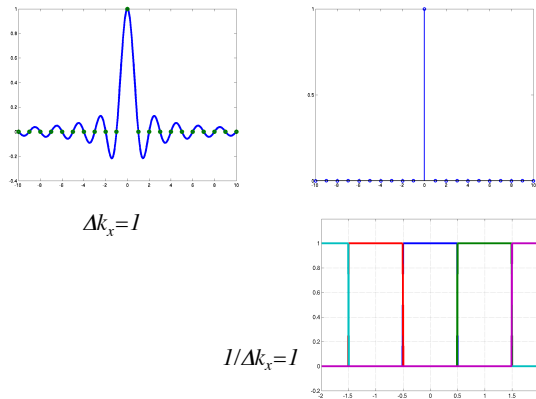
## Aliasing



Aliasing occurs when  $1/\Delta k_x < \text{FOV}$

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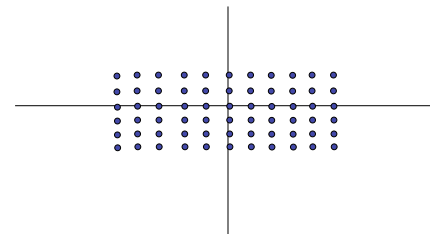
## Aliasing Example



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## 2D Comb Function

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



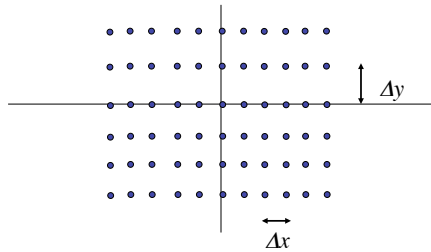
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## Scaled 2D Comb Function

$$\text{comb}(x/\Delta x, y/\Delta y) = \text{comb}(x/\Delta x)\text{comb}(y/\Delta y)$$

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

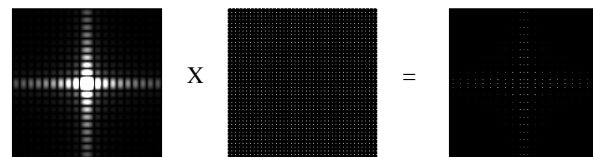


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## 2D k-space sampling

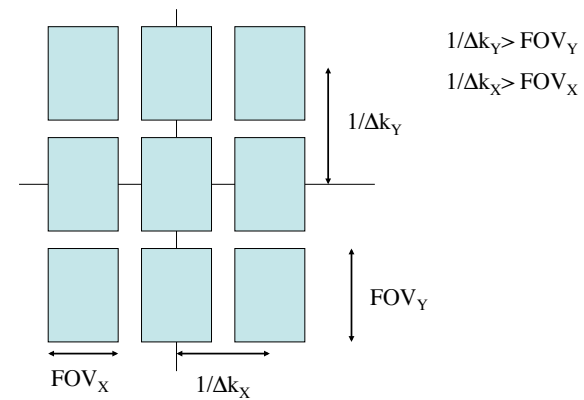
$$\begin{aligned} G_S(k_x, k_y) &= G(k_x, k_y) \frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right) \\ &= G(k_x, k_y) \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(k_x - m\Delta k_x, k_y - n\Delta k_y) \\ &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} G(m\Delta k_x, n\Delta k_y) \delta(k_x - m\Delta k_x, k_y - n\Delta k_y) \end{aligned}$$

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## Nyquist Conditions



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## Windowing

Windowing the data in Fourier space

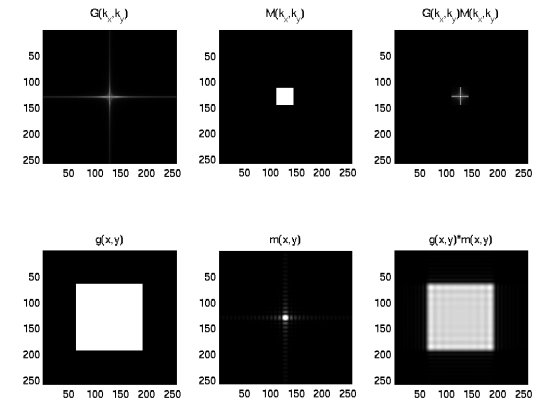
$$G_W(k_x, k_y) = G(k_x, k_y)W(k_x, k_y)$$

Results in convolution of the object with the inverse transform of the window

$$g_w(x, y) = g(x, y) * w(x, y)$$

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## Resolution



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## Windowing Example

$$W(k_x, k_y) = \text{rect}\left(\frac{k_x}{W_{k_x}}\right)\text{rect}\left(\frac{k_y}{W_{k_y}}\right)$$

$$\begin{aligned} w(x, y) &= F^{-1}\left[\text{rect}\left(\frac{k_x}{W_{k_x}}\right)\text{rect}\left(\frac{k_y}{W_{k_y}}\right)\right] \\ &= W_{k_x} W_{k_y} \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y) \end{aligned}$$

$$g_w(x, y) = g(x, y) ** W_{k_x} W_{k_y} \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y)$$

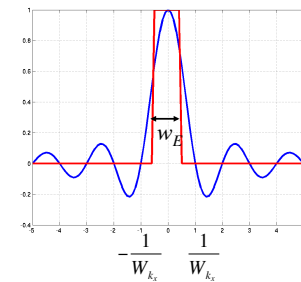
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## Effective Width

$$w_E = \frac{1}{w(0)} \int_{-\infty}^{\infty} w(x) dx$$

Example

$$\begin{aligned} w_E &= \frac{1}{1} \int_{-\infty}^{\infty} \text{sinc}(W_{k_x} x) dx \\ &= F[\text{sinc}(W_{k_x} x)]\Big|_{k_x=0} \\ &= \frac{1}{W_{k_x}} \text{rect}\left(\frac{k_x}{W_{k_x}}\right)\Big|_{k_x=0} \\ &= \frac{1}{W_{k_x}} \end{aligned}$$

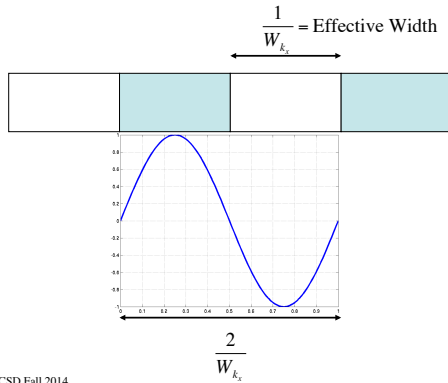


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## Resolution and spatial frequency

With a window of width  $W_{k_x}$ , the highest spatial frequency is  $W_{k_x}/2$ .

This corresponds to a spatial period of  $2/W_{k_x}$ .

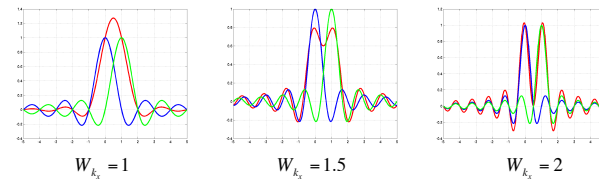


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## Windowing Example

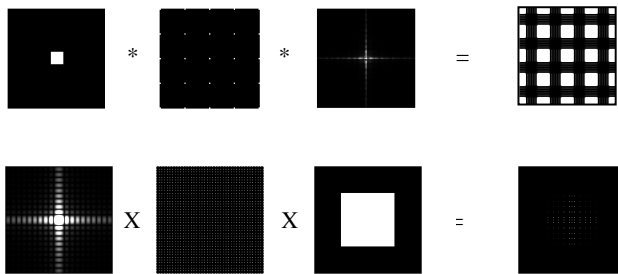
$$g(x, y) = [\delta(x) + \delta(x-1)]\delta(y)$$

$$\begin{aligned} g_w(x, y) &= [\delta(x) + \delta(x-1)]\delta(y) * W_{k_x} W_{k_y} \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y) \\ &= W_{k_x} W_{k_y} ([\delta(x) + \delta(x-1)] * \text{sinc}(W_{k_x} x)) \text{sinc}(W_{k_y} y) \\ &= W_{k_x} W_{k_y} (\text{sinc}(W_{k_x} x) + \text{sinc}(W_{k_x}(x-1))) \text{sinc}(W_{k_y} y) \end{aligned}$$



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## Sampling and Windowing



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## Sampling and Windowing

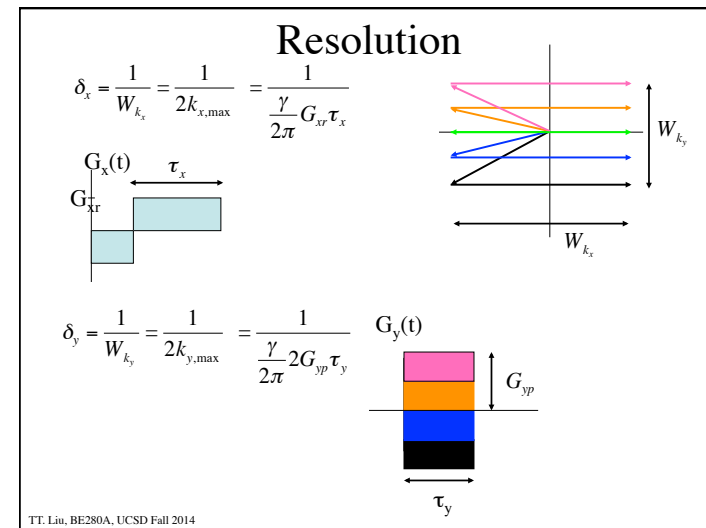
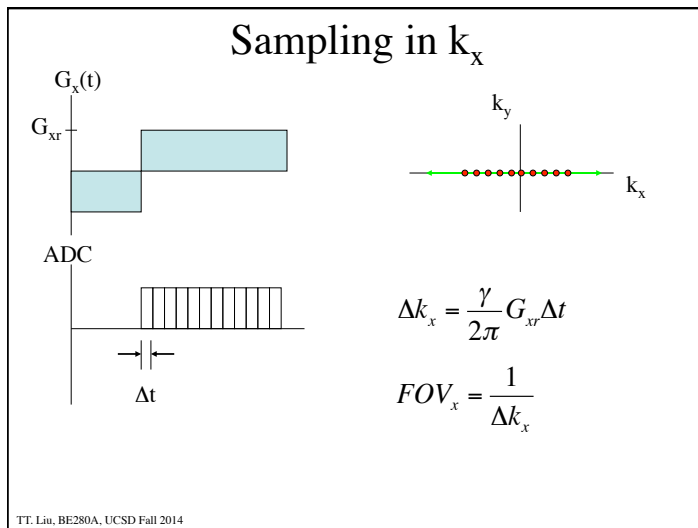
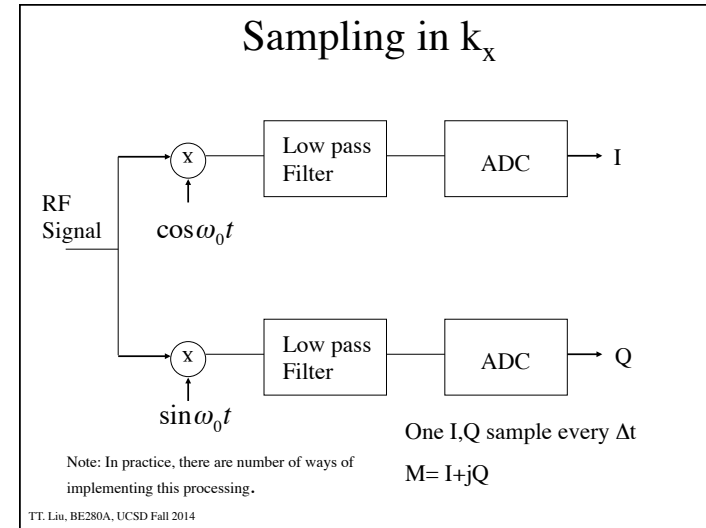
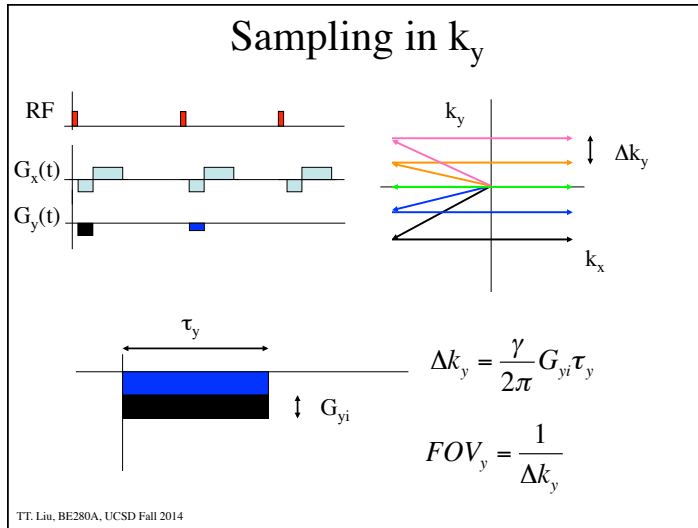
Sampling and windowing the data in Fourier space

$$G_{SW}(k_x, k_y) = G(k_x, k_y) \frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right) \text{rect}\left(\frac{k_x}{W_{k_x}}, \frac{k_y}{W_{k_y}}\right)$$

Results in replication and convolution in object space.

$$g_{SW}(x, y) = W_{k_x} W_{k_y} g(x, y) * \text{comb}(\Delta k_x x, \Delta k_y y) * \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y)$$

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## Example

Goal :

$$FOV_x = FOV_y = 25.6 \text{ cm}$$

$$\delta_x = \delta_y = 0.1 \text{ cm}$$

Readout Gradient :

$$FOV_x = \frac{1}{\frac{\gamma}{2\pi} G_{xr} \Delta t}$$

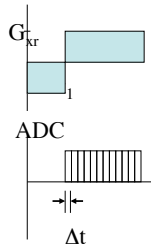
Pick  $\Delta t = 32 \mu\text{sec}$

$$G_{xr} = \frac{1}{FOV_x \frac{\gamma}{2\pi} \Delta t} = \frac{1}{(25.6 \text{ cm})(42.57 \times 10^6 \text{ T}^{-1} \text{ s}^{-1})(32 \times 10^{-6} \text{ s})}$$

$$= 2.8675 \times 10^{-5} \text{ T/cm}$$

$$= .28675 \text{ G/cm}$$

1 Gauss =  $1 \times 10^{-4}$  Tesla



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## Example

Readout Gradient :

$$\delta_x = \frac{1}{\frac{\gamma}{2\pi} G_{xr} \tau_x}$$

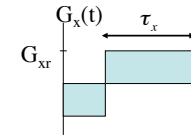
$$\tau_x = \frac{1}{\delta_x \frac{\gamma}{2\pi} G_{xr}} = \frac{1}{(0.1 \text{ cm})(4257 \text{ G}^{-1} \text{ s}^{-1})(0.28675 \text{ G/cm})}$$

$$= 8.192 \text{ ms}$$

$$= N_{\text{read}} \Delta t$$

where

$$N_{\text{read}} = \frac{FOV_x}{\delta_x} = 256$$



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## Example

Phase - Encode Gradient :

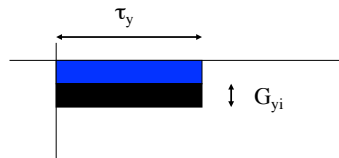
$$FOV_y = \frac{1}{\frac{\gamma}{2\pi} G_{yi} \tau_y}$$

Pick  $\tau_y = 4.096 \text{ msec}$

$$G_{yi} = \frac{1}{FOV_y \frac{\gamma}{2\pi} \tau_y} = \frac{1}{(25.6 \text{ cm})(42.57 \times 10^6 \text{ T}^{-1} \text{ s}^{-1})(4.096 \times 10^{-3} \text{ s})}$$

$$= 2.2402 \times 10^{-7} \text{ T/cm}$$

$$= .00224 \text{ G/cm}$$



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## Example

Phase - Encode Gradient :

$$\delta_y = \frac{1}{\frac{\gamma}{2\pi} 2 G_{yp} \tau_y}$$

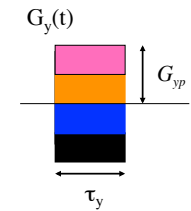
$$G_{yp} = \frac{1}{\delta_y \frac{\gamma}{2\pi} 2 \tau_y} = \frac{1}{(0.1 \text{ cm})(4257 \text{ G}^{-1} \text{ s}^{-1})(4.096 \times 10^{-3} \text{ s})}$$

$$= 0.2868 \text{ G/cm}$$

$$= \frac{N_p}{2} G_{yi}$$

where

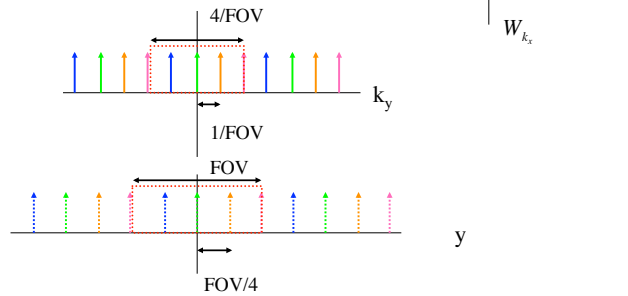
$$N_p = \frac{FOV_x}{\delta_x} = 256$$



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## Sampling

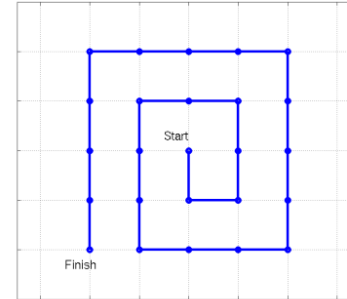
In practice, an even number (typically power of 2) sample is usually taken in each direction to take advantage of the Fast Fourier Transform (FFT) for reconstruction.



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## Example

Consider the k-space trajectory shown below. ADC samples are acquired at the points shown with  $\Delta t = 10 \mu\text{sec}$ . The desired FOV (both x and y) is 10 cm and the desired resolution (both x and y) is 2.5 cm. Draw the gradient waveforms required to achieve the desired FOV and resolution. Label the waveform with the gradient amplitudes required to achieve the desired FOV and resolution. Also, make sure to label the time axis correctly.



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