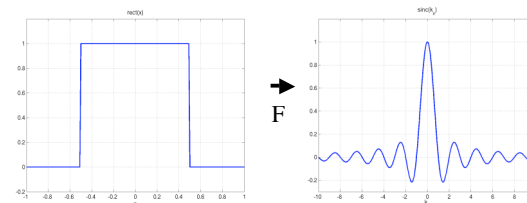


Bioengineering 280A
Principles of Biomedical Imaging

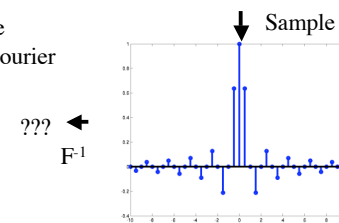
Fall Quarter 2007
MRI Lecture 3

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Fourier Sampling

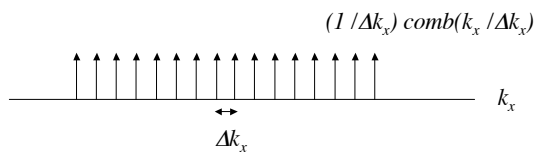


Instead of sampling the signal, we sample its Fourier Transform



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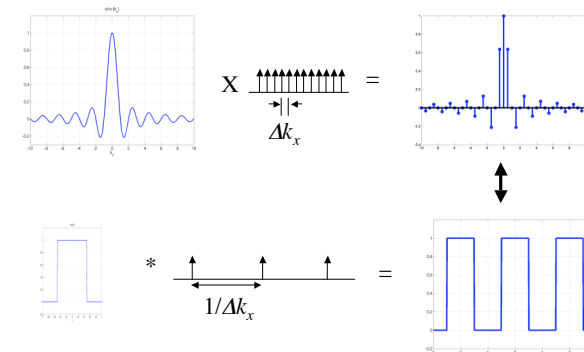
Fourier Sampling



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

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Fourier Sampling -- Inverse Transform



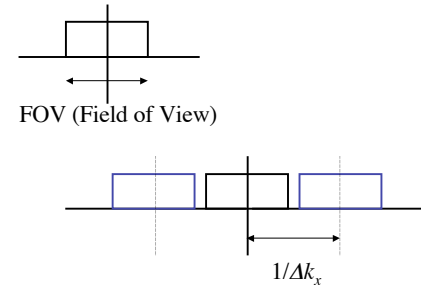
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Fourier Sampling -- Inverse Transform

$$\begin{aligned}
 g_s(x) &= F^{-1}[G_s(k_x)] \\
 &= F^{-1}\left[G(k_x) \frac{1}{\Delta k_x} \text{comb}\left(\frac{k_x}{\Delta k_x}\right)\right] \\
 &= F^{-1}[G(k_x)] * F^{-1}\left[\frac{1}{\Delta k_x} \text{comb}\left(\frac{k_x}{\Delta k_x}\right)\right] \\
 &= g(x) * \text{comb}(x\Delta k_x) \\
 &= g(x) * \sum_{n=-\infty}^{\infty} \delta(x\Delta k_x - n) \\
 &= g(x) * \frac{1}{\Delta k_x} \sum_{n=-\infty}^{\infty} \delta\left(x - \frac{n}{\Delta k_x}\right) \\
 &= \frac{1}{\Delta k_x} \sum_{n=-\infty}^{\infty} g\left(x - \frac{n}{\Delta k_x}\right)
 \end{aligned}$$

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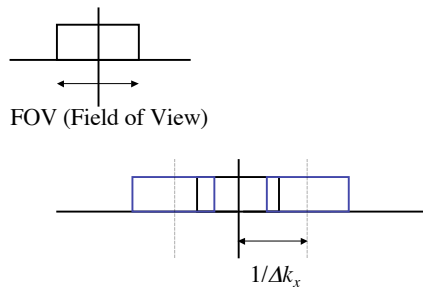
Nyquist Condition



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

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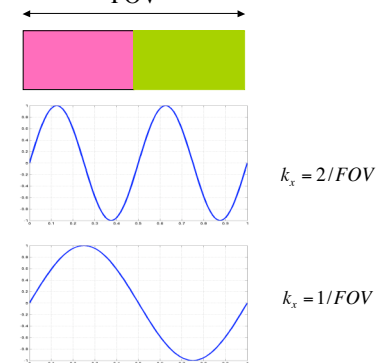
Aliasing



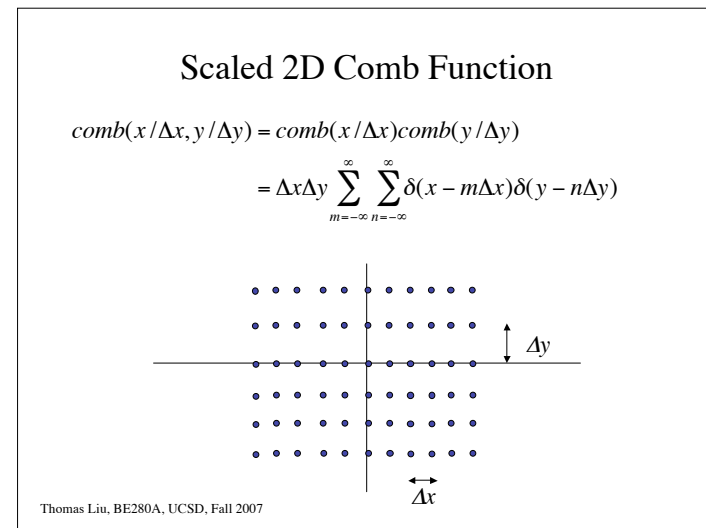
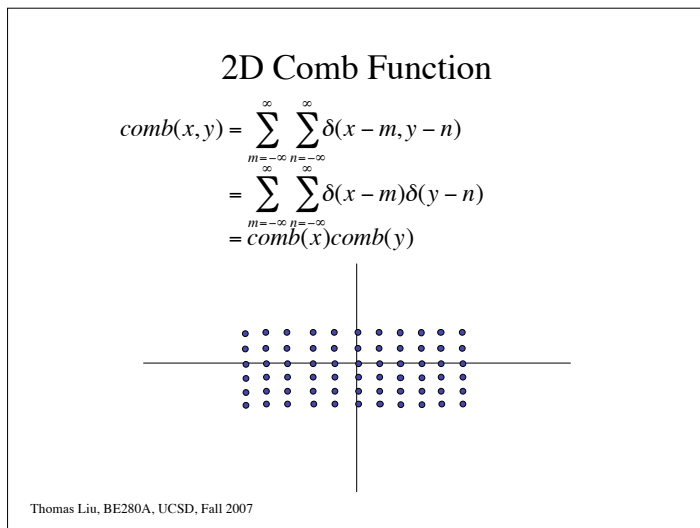
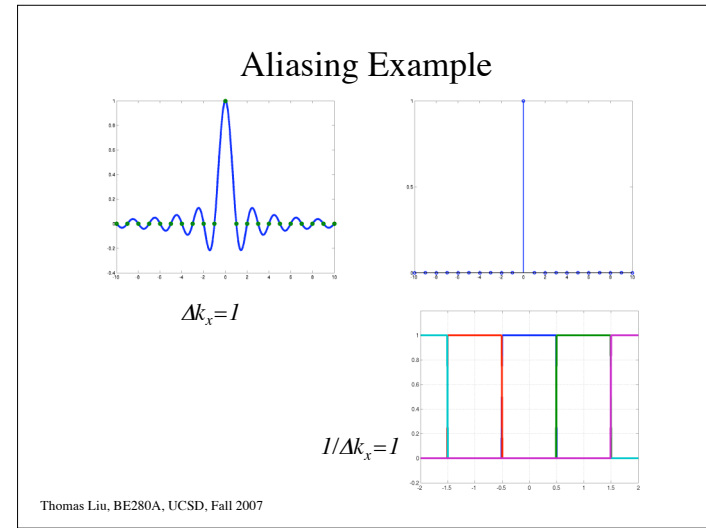
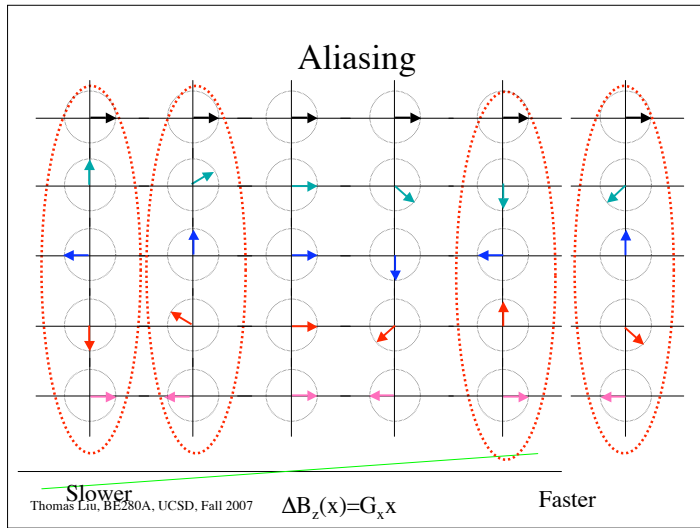
Aliasing occurs when $1/\Delta k_x < \text{FOV}$

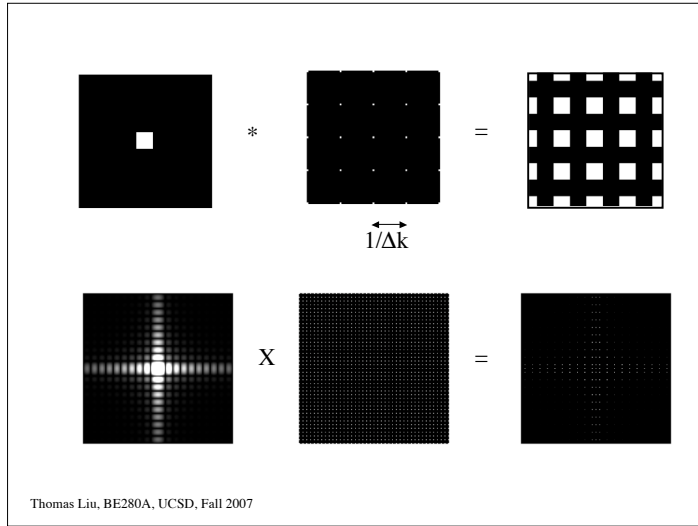
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Intuitive view of Aliasing



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2D k-space sampling

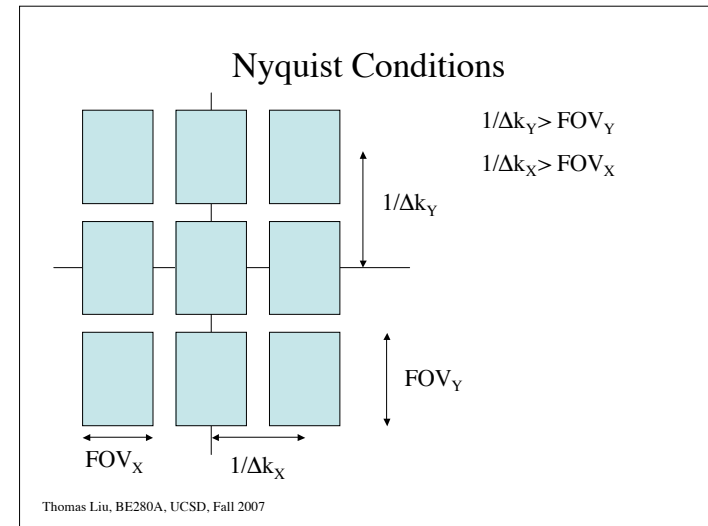
$$\begin{aligned}
 G_S(k_x, k_y) &= G(k_x, k_y) \frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right) \\
 &= G(k_x, k_y) \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(k_x - m\Delta k_x, k_y - n\Delta k_y) \\
 &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty}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

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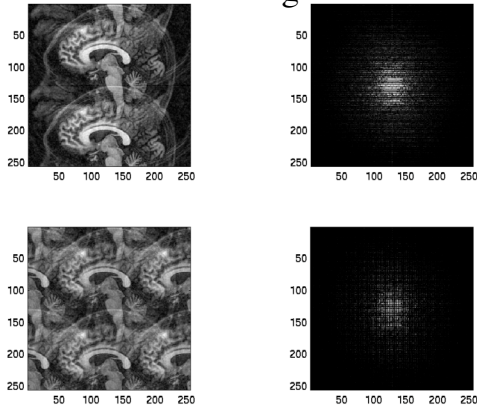
2D k-space sampling

$$\begin{aligned}
 g_S(x, y) &= F^{-1}\left[G_S(k_x, k_y)\right] \\
 &= F^{-1}\left[G(k_x, k_y) \frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right)\right] \\
 &= F^{-1}\left[G(k_x, k_y)\right] * F^{-1}\left[\frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right)\right] \\
 &= g(x, y) * \text{comb}(x\Delta k_x) \text{comb}(y\Delta k_y) \\
 &= g(x) * \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x\Delta k_x - m) \delta(y\Delta k_y - n) \\
 &= g(x) * \frac{1}{\Delta k_x \Delta k_y} \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta\left(x - \frac{m}{\Delta k_x}\right) \delta\left(y - \frac{n}{\Delta k_y}\right) \\
 &= \frac{1}{\Delta k_x \Delta k_y} \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} g\left(x - \frac{m}{\Delta k_x}, y - \frac{n}{\Delta k_y}\right)
 \end{aligned}$$

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Aliasing



T1

Windowing

Windowing the data in Fourier space

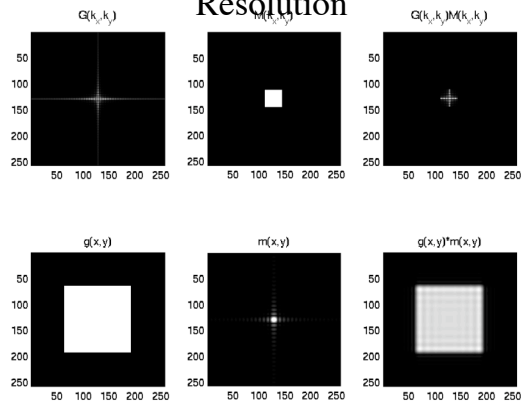
$$G_W(k_x, k_y) = G(k_x, k_y)W(k_x, k_y)$$

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



1

Windowing Example

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

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

$$= W_{k_x} W_{k_y} \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y)$$

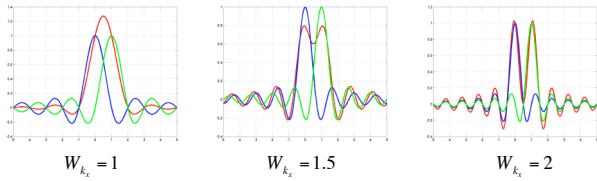
$$g_W(x, y) = g(x, y) * W_{k_x} W_{k_y} \text{sinc}(W_{k_x} x) \text{sinc}(W_{k_y} y)$$

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Windowing Example

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

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



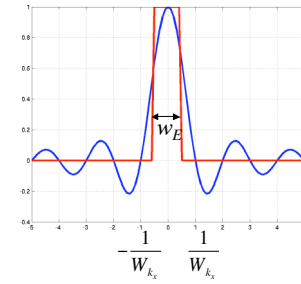
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Effective Width

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

Example

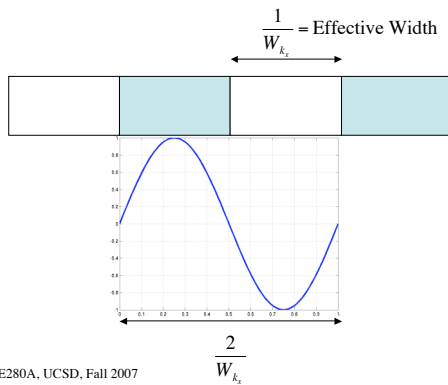
$$\begin{aligned} w_E &= \frac{1}{1} \int_{-\infty}^{\infty} \text{sinc}(W_{k_x} x) dx \\ &= F[\text{sinc}(W_{k_x} x)] \Big|_{k_x=0} \\ &= \frac{1}{W_{k_x}} \text{rect}\left(\frac{k_x}{W_{k_x}}\right) \Big|_{k_x=0} \\ &= \frac{1}{W_{k_x}} \end{aligned}$$



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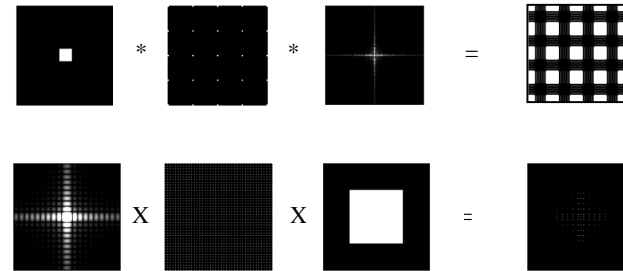
Resolution and spatial frequency

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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Sampling and Windowing



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Sampling and Windowing

Sampling and windowing the data in Fourier space

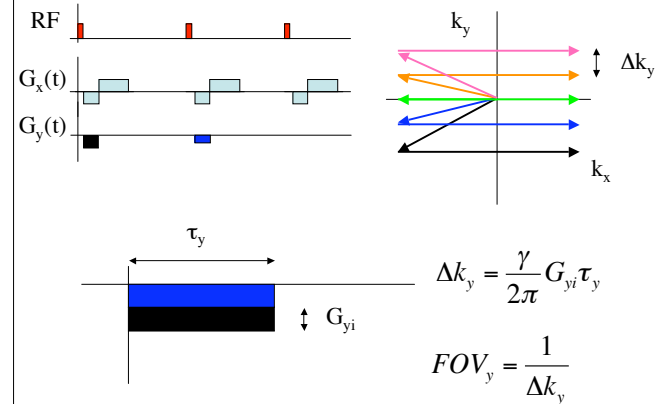
$$G_{SW}(k_x, k_y) = G(k_x, k_y) \frac{1}{\Delta k_x \Delta k_y} \text{comb}\left(\frac{k_x}{\Delta k_x}, \frac{k_y}{\Delta k_y}\right) \text{rect}\left(\frac{k_x}{W_{k_x}}, \frac{k_y}{W_{k_y}}\right)$$

Results in replication and convolution in object space.

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

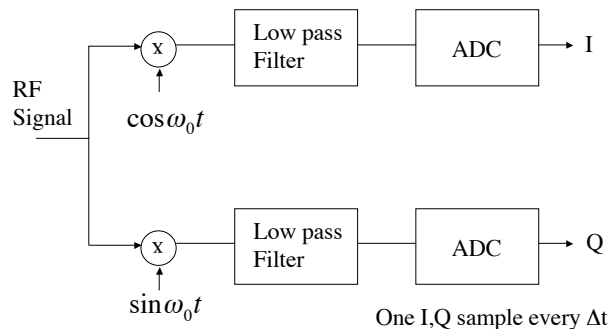
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Sampling in k_y



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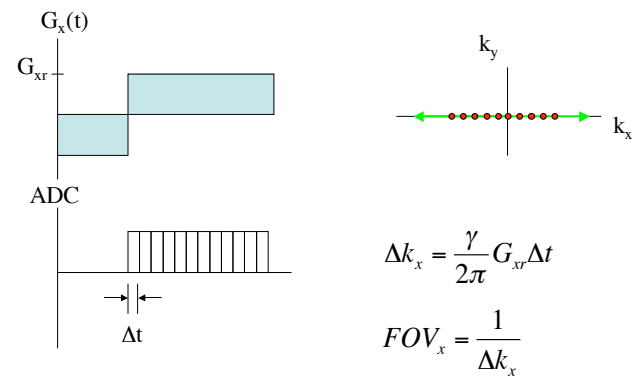
Sampling in k_x



Note: In practice, there are number of ways of implementing this processing.
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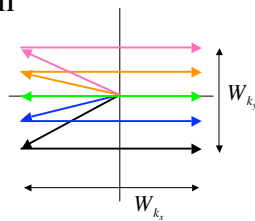
One I,Q sample every Δt
 $M = I + jQ$

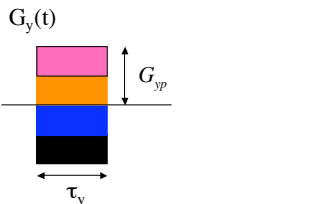
Sampling in k_x



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Resolution

$$\delta_x = \frac{1}{W_{k_x}} = \frac{1}{2k_{x,\max}} = \frac{1}{\frac{\gamma}{2\pi} G_{xr} \tau_x}$$


$$\delta_y = \frac{1}{W_{k_y}} = \frac{1}{2k_{y,\max}} = \frac{1}{\frac{\gamma}{2\pi} 2G_{yp} \tau_y}$$


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Example

Goal:
 $FOV_x = FOV_y = 25.6 \text{ cm}$
 $\delta_x = \delta_y = 0.1 \text{ cm}$

Readout Gradient :

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

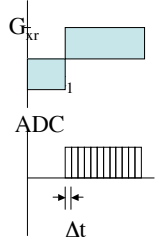
Pick $\Delta t = 32 \text{ } \mu\text{sec}$

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

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

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

1 Gauss = 1×10^{-4} Tesla



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Example

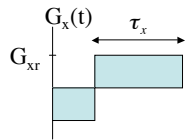
Readout Gradient :

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

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

$$= 8.192 \text{ ms}$$

where

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


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Example

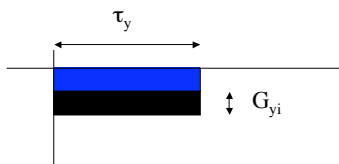
Phase - Encode Gradient :

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

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

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

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

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


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Example

Phase - Encode Gradient :

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

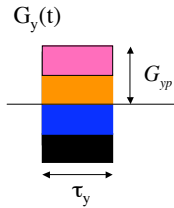
$$G_{yp} = \frac{1}{\delta_y 2 \frac{\gamma}{2\pi} \tau_y} = \frac{1}{(0.1\text{cm})(4257\text{ G}^{-1}\text{s}^{-1})(4.096 \times 10^{-3}\text{s})}$$

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

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

where

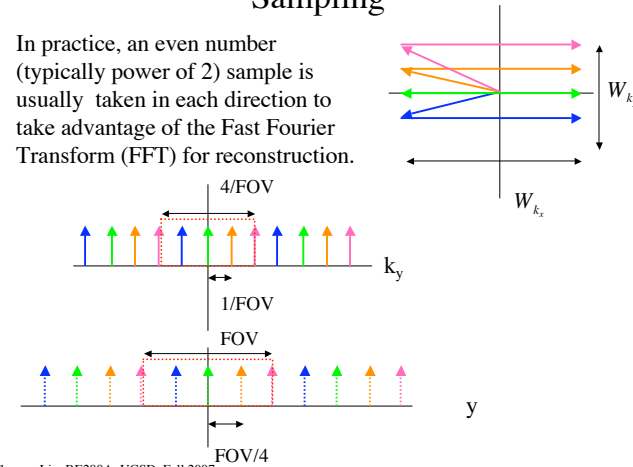
$$N_p = \frac{FOV_y}{\delta_y} = 256$$



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Sampling

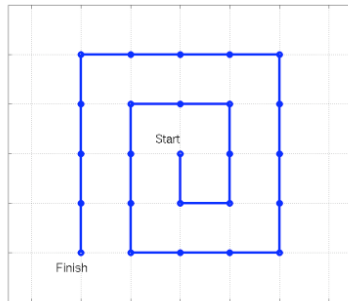
In practice, an even number (typically power of 2) sample is usually taken in each direction to take advantage of the Fast Fourier Transform (FFT) for reconstruction.



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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 k-space trajectory. 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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SCAN TIMING

of Echoes 1 2 4

TE Min Full

TE2

TR 750

Inv Time

T2

Flip Angle

Echo Train Length

Bandwidth 35

Bandwidth2

ACQUISITION TIMING

Freq 352 Freq DIR A/P

Phase 192 Center Pres Water

NEX 2.0 Flow Comp Direction

Phase FOV 0.75 Autoshim Phase Correct

of Acqs Before Pause Agent

SCANNING RANGE

FOV 22 S/A L/R Center P/A Center

Slice Thickness 5.0 Start End

Spacing 2.0 # Slices Table Delta

ACTUAL End

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GE Medical Systems 2003

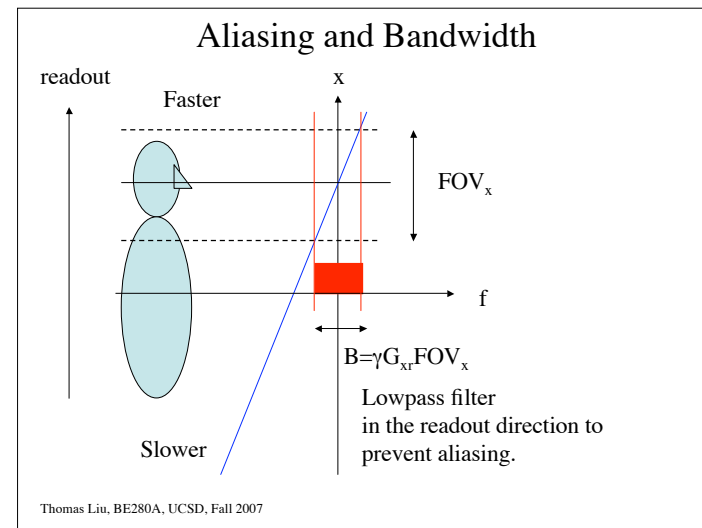
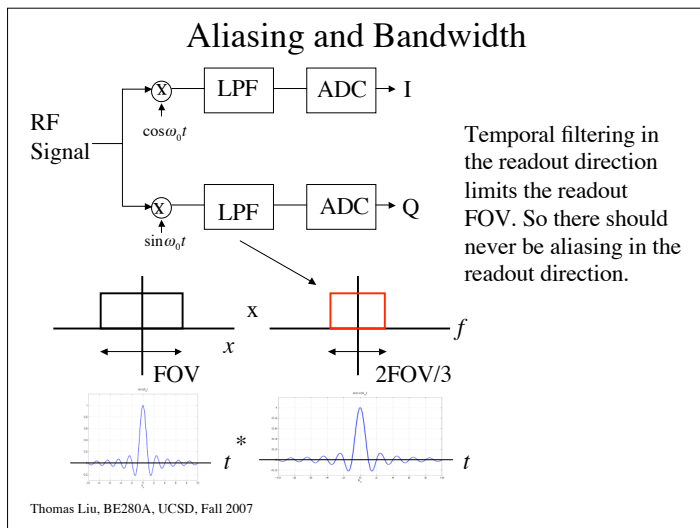
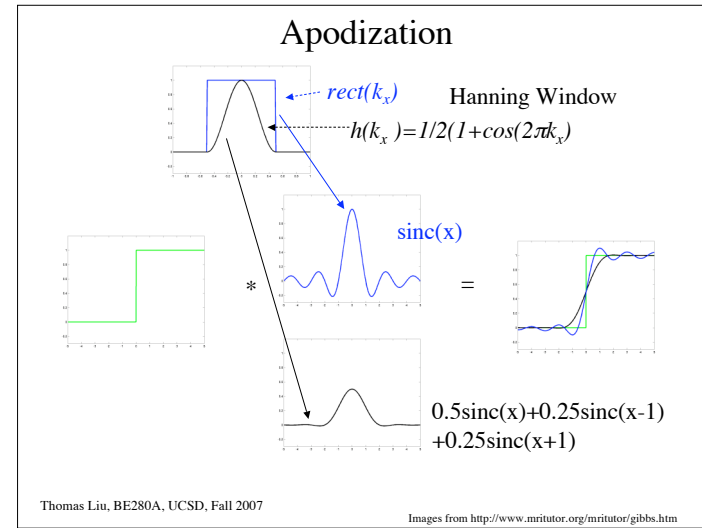
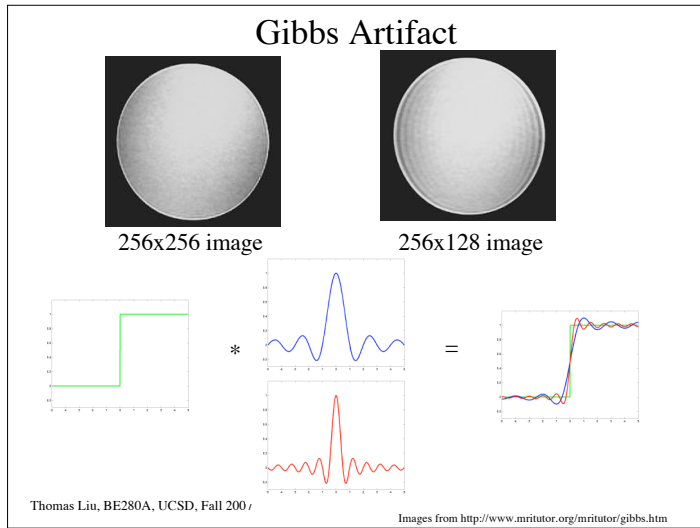


Figure 7-31 Default Axial Directions

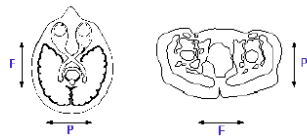


Figure 7-32 Default Sagittal and Coronal Directions

