

Bioengineering 280A
Principles of Biomedical Imaging

Fall Quarter 2004
Lecture 5
Sampling

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Topics

1. Overview of Sampling
2. 1D Sampling
3. 2D Sampling
4. Aliasing

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Analog vs. Digital

The Analog World:

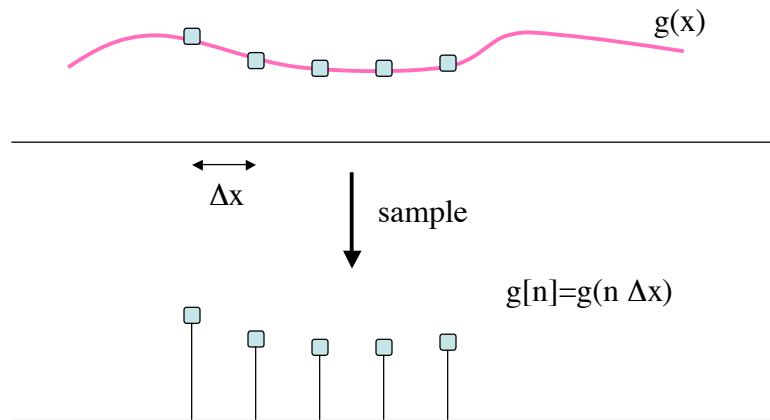
Continuous time/space, continuous valued signals or images, e.g. vinyl records, photographs, x-ray films.

The Digital World:

Discrete time/space, discrete-valued signals or images, e.g. CD-Roms, DVDs, digital photos, digital x-rays, CT, MRI, ultrasound.

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The Process of Sampling



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Questions

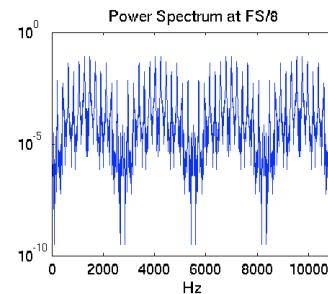
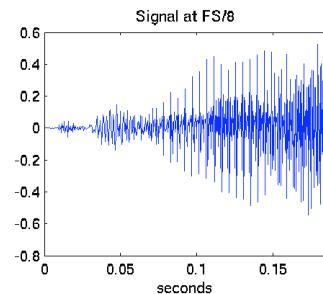
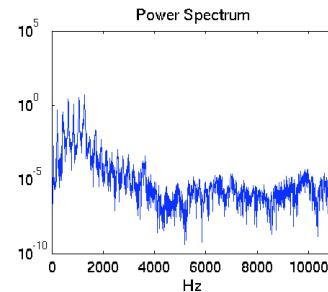
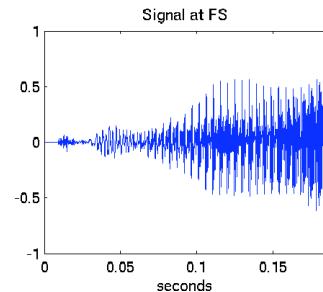
How finely do we need to sample?

What happens if we don't sample finely enough?

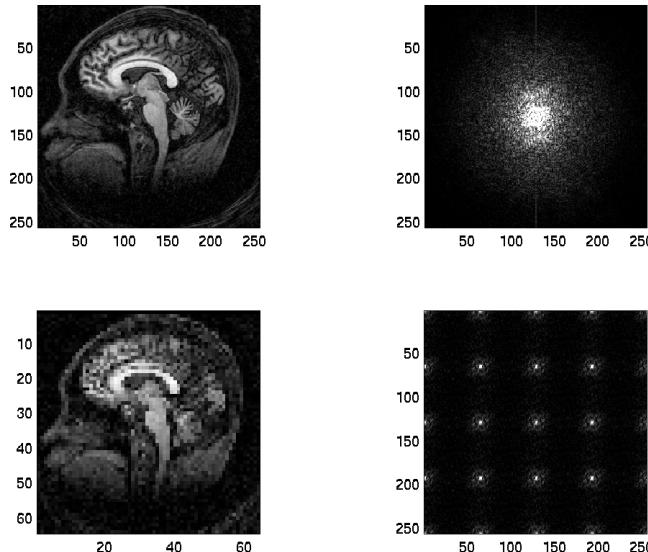
Can we reconstruct the original signal or image from its samples?

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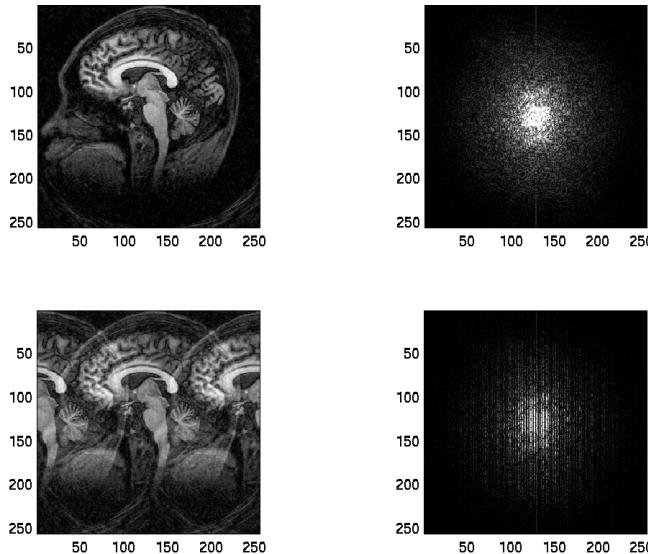
Sampling in the Time Domain



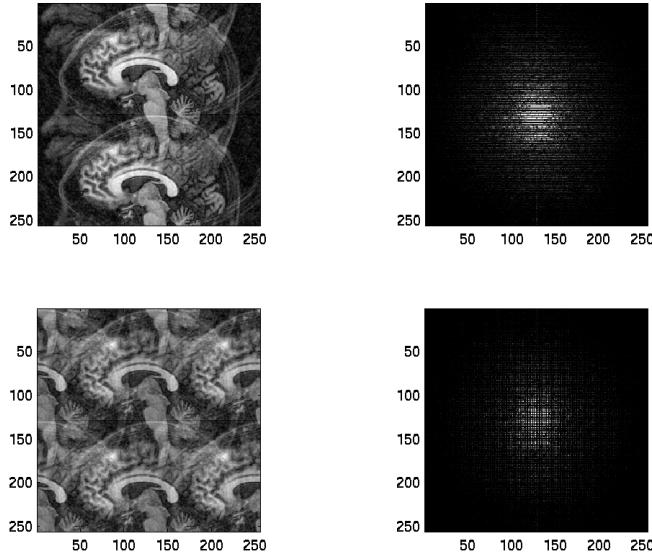
Sampling in Image Space



Sampling in k-space



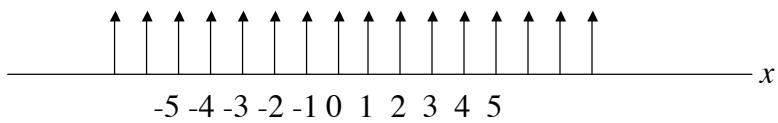
Sampling in k-space



↑

Comb Function

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

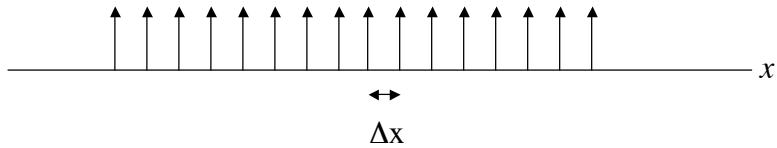


Other names: Impulse train, bed of nails, shah function.

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Scaled Comb Function

$$\begin{aligned}
 comb\left(\frac{x}{\Delta x}\right) &= \sum_{n=-\infty}^{\infty} \delta\left(\frac{x}{\Delta x} - n\right) \\
 &= \sum_{n=-\infty}^{\infty} \delta\left(\frac{x - n\Delta x}{\Delta x}\right) \\
 &= \Delta x \sum_{n=-\infty}^{\infty} \delta(x - n\Delta x)
 \end{aligned}$$



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1D spatial sampling

$$\begin{aligned}
 g_s(x) &= g(x) \frac{1}{\Delta x} comb\left(\frac{x}{\Delta x}\right) \\
 &= g(x) \sum_{n=-\infty}^{\infty} \delta(x - n\Delta x) \\
 &= \sum_{n=-\infty}^{\infty} g(n\Delta x) \delta(x - n\Delta x)
 \end{aligned}$$

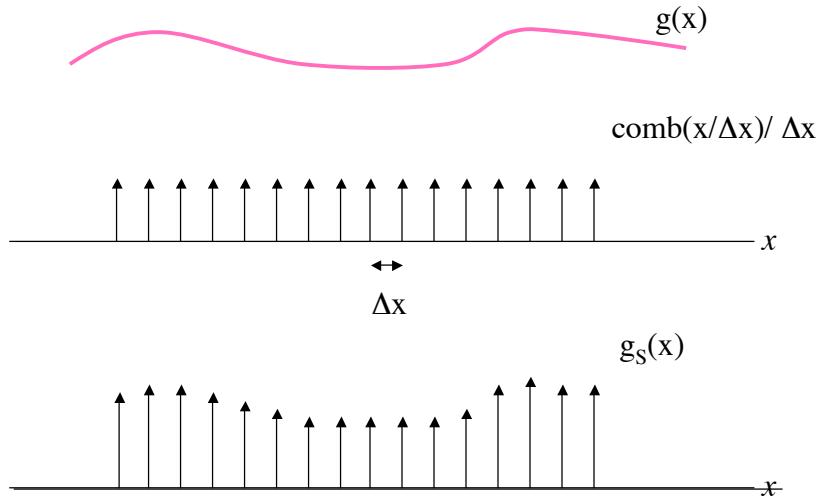
Recall the sifting property $\int_{-\infty}^{\infty} g(x)\delta(x - a) = g(a)$

But we can also write $\int_{-\infty}^{\infty} g(a)\delta(x - a) = g(a) \int_{-\infty}^{\infty} \delta(x - a) = g(a)$

So, $g(x)\delta(x - a) = g(a)\delta(x - a)$

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1D spatial sampling



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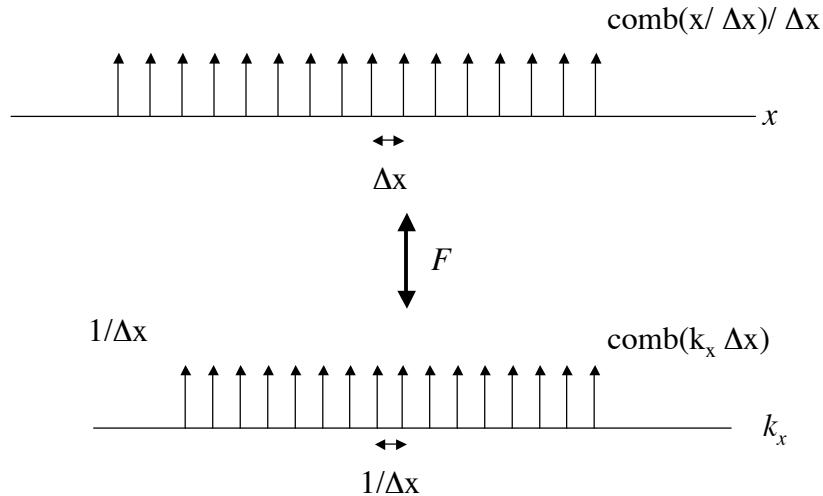
Fourier Transform of $\text{comb}(x)$

$$\begin{aligned} F[\text{comb}(x)] &= \text{comb}(k_x) \\ &= \sum_{n=-\infty}^{\infty} \delta(k_x - n) \end{aligned}$$

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

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Fourier Transform of $\text{comb}(x/\Delta x)$



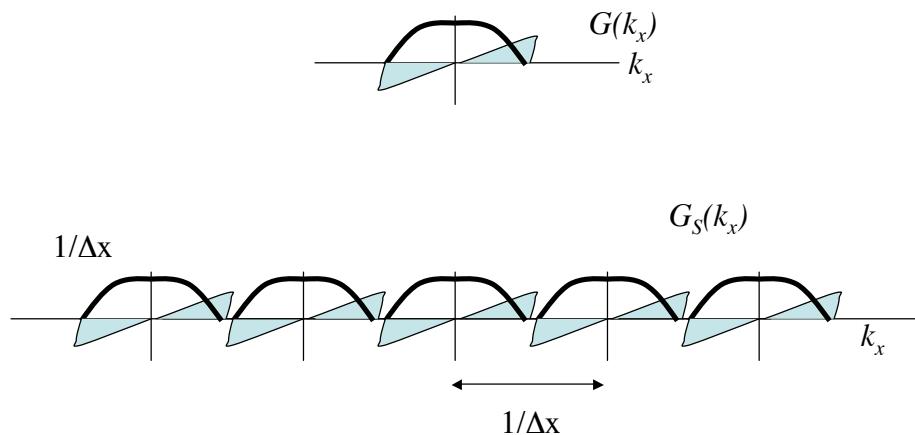
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Fourier Transform of $g_S(x)$

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

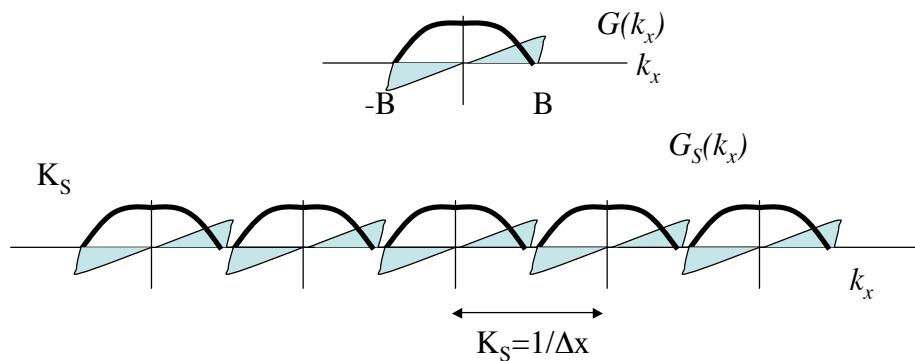
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Fourier Transform of $g_S(x)$



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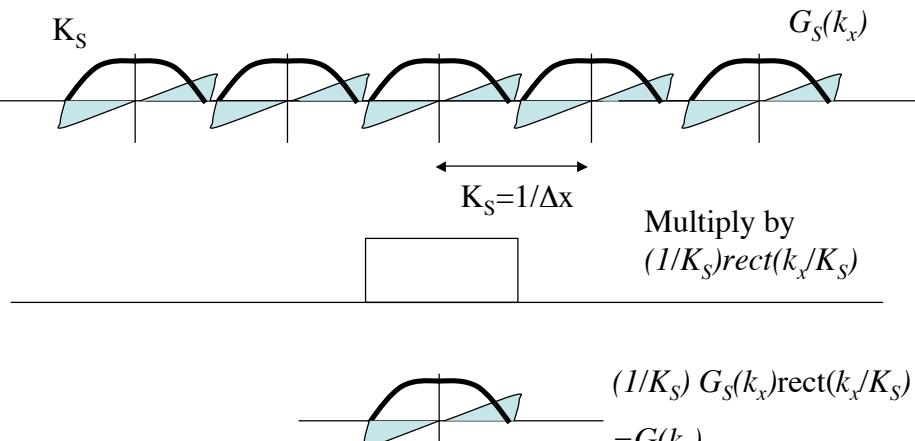
Nyquist Condition



To avoid overlap, we require that $1/\Delta x > 2B$ or
 $K_S > 2B$ where $K_S = 1/\Delta x$ is the sampling frequency

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Reconstruction from Samples



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Reconstruction from Samples

If the Nyquist condition is met, then

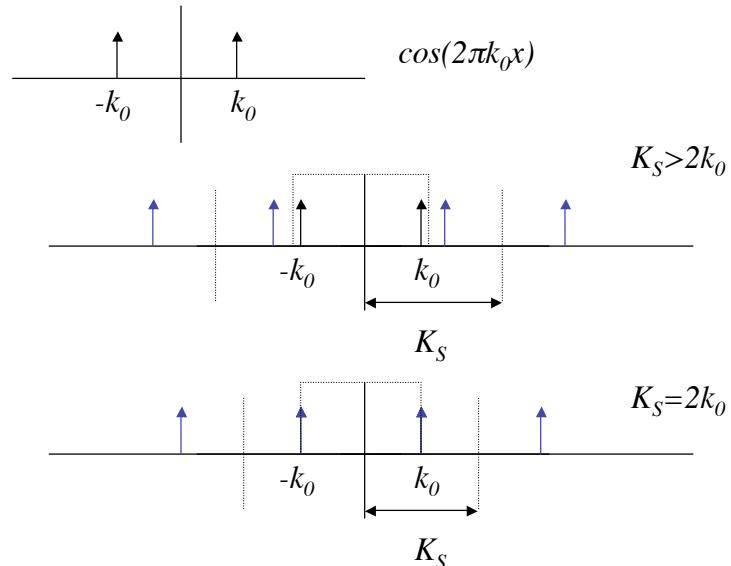
$$\hat{G}_S(k_x) = \frac{1}{K_S} G_S(k_x) \text{rect}(k_x/K_S) = G(k_x)$$

And the signal can be reconstructed by convolving the sample with a sinc function

$$\begin{aligned}\hat{g}_S(x) &= g_S(x) * \text{sinc}(K_s x) \\ &= \sum_{n=-\infty}^{\infty} g(n\Delta X) \delta(x - n\Delta X) * \text{sinc}(K_s x) \\ &= \sum_{n=-\infty}^{\infty} g(n\Delta X) \text{sinc}(K_s(x - n\Delta x))\end{aligned}$$

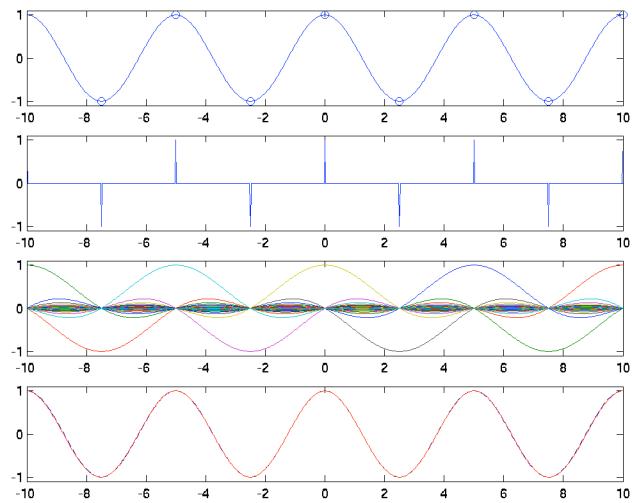
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Example Cosine Reconstruction



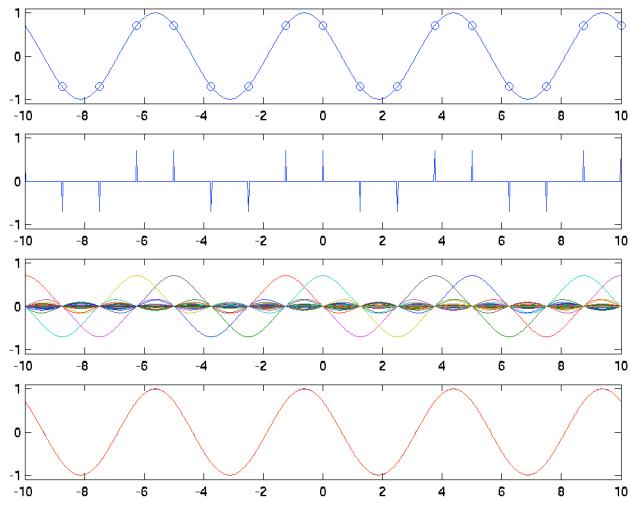
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Cosine Example with $K_S=2k_0$



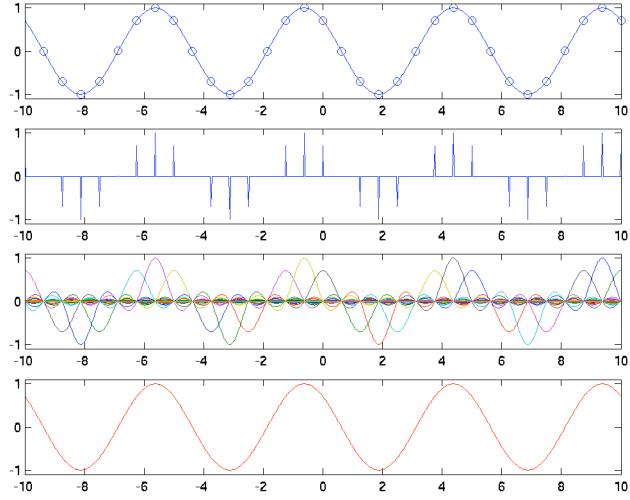
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Example with $K_s=4k_0$



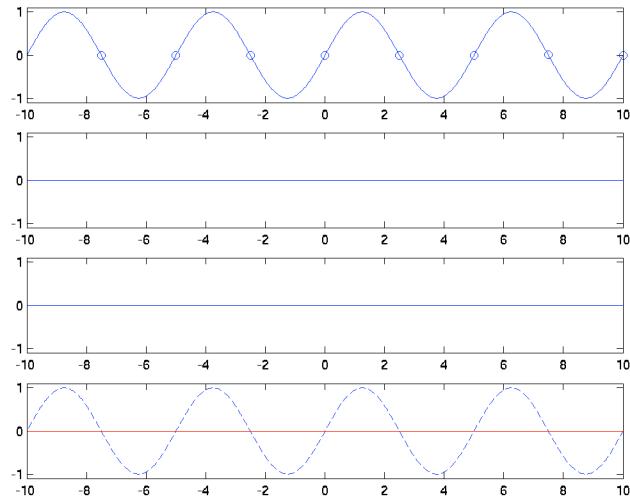
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Example with $K_s=8k_0$



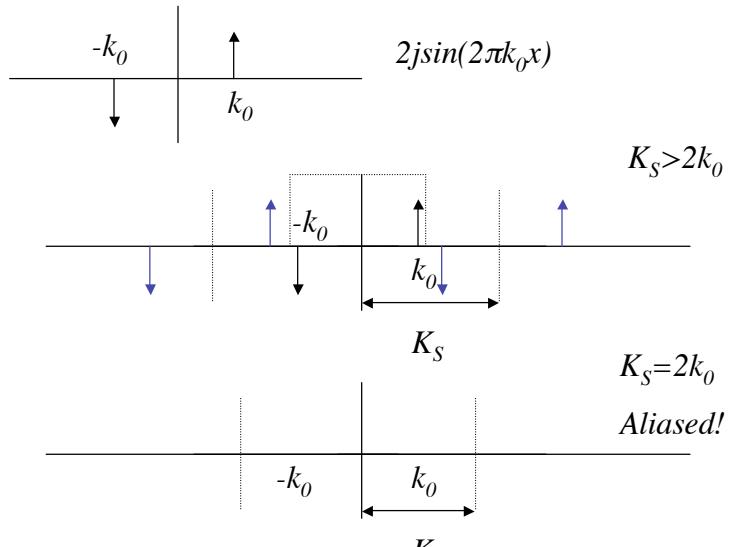
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Sine Example with $K_s=2k_0$



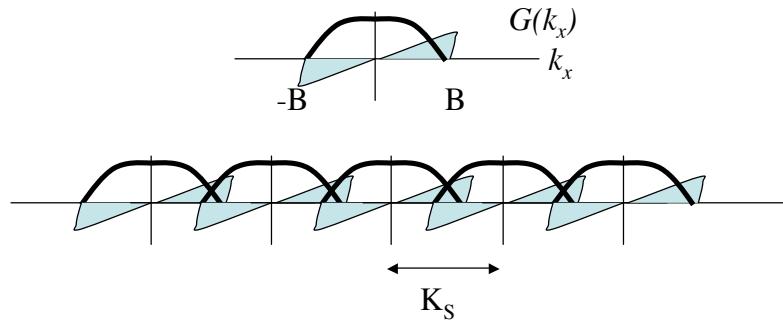
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Example Sine Reconstruction



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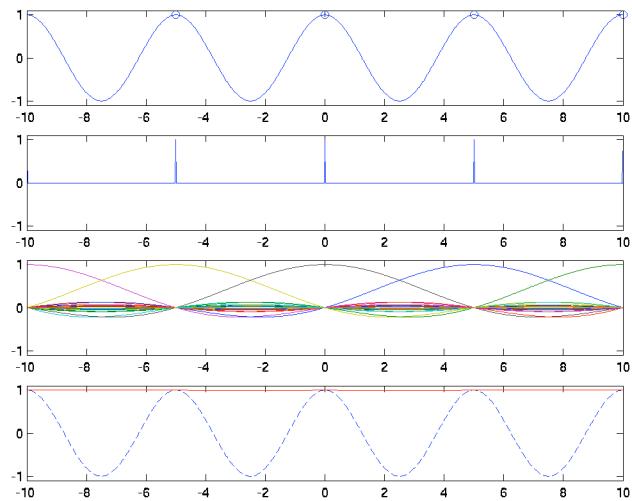
Aliasing



Aliasing occurs when the Nyquist condition is not satisfied.
This occurs for $K_s \leq 2B$

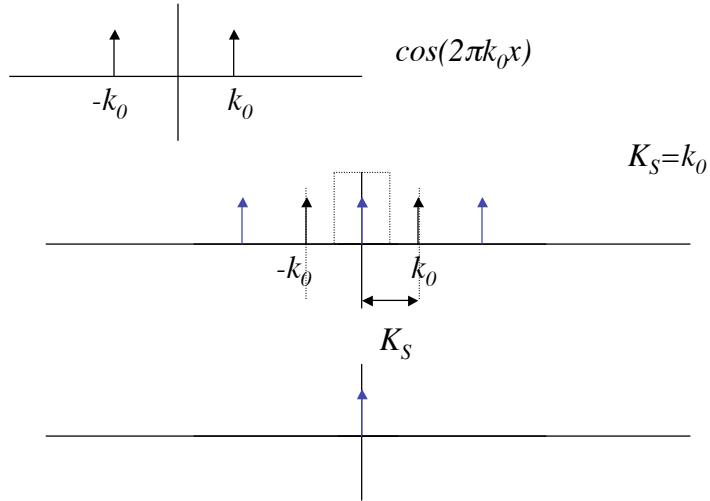
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Aliasing Example



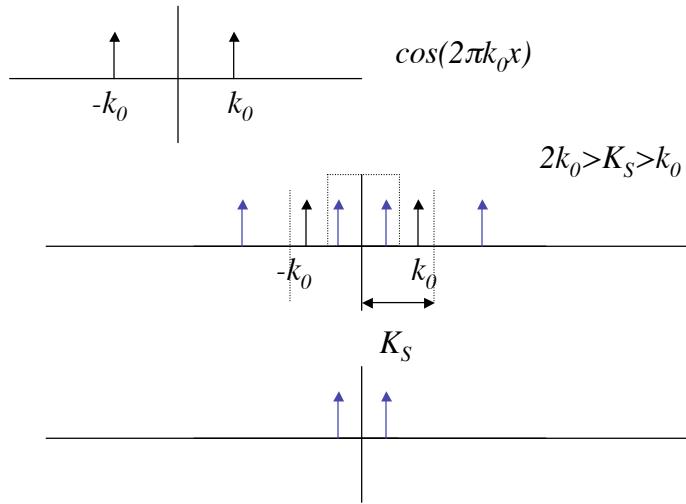
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Aliasing Example



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Aliasing Example



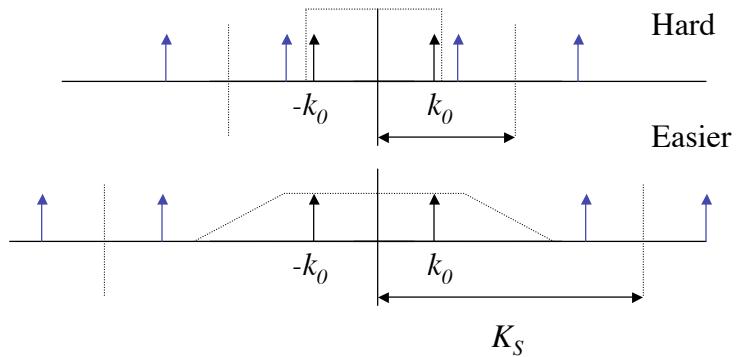
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Practical Considerations

Why sample higher than the Nyquist frequency?

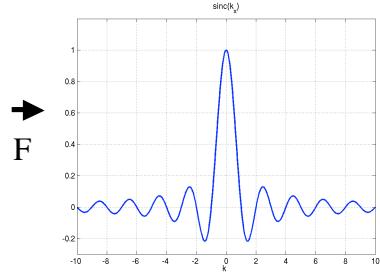
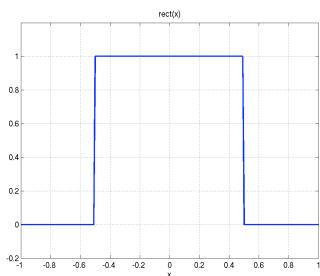
- true sinc interpolation is not practical since the sinc function goes from -infinity to infinity

- the requirements on the low-pass filter are reduced.



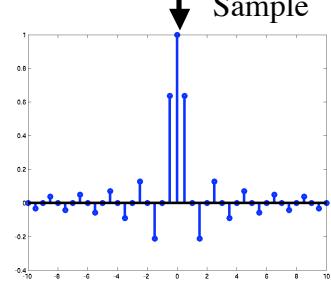
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Fourier Sampling



Instead of sampling the signal, we sample its Fourier Transform

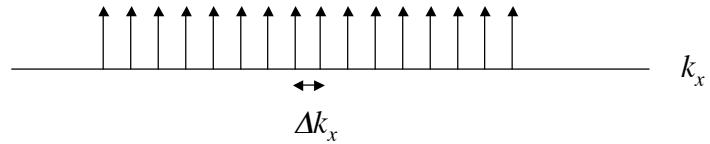
??? \leftarrow
F⁻¹



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Fourier Sampling

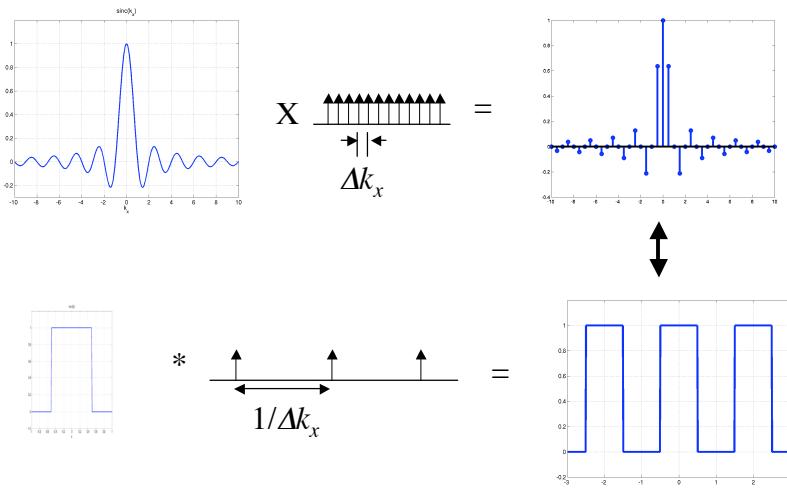
$$(I / \Delta k_x) \operatorname{comb}(k_x / \Delta k_x)$$



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

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Fourier Sampling -- Inverse Transform



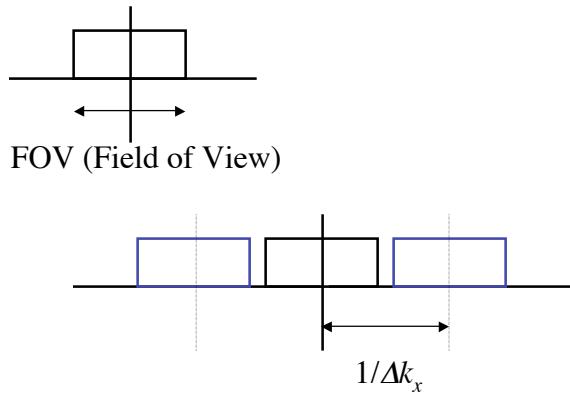
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Fourier Sampling -- Inverse Transform

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

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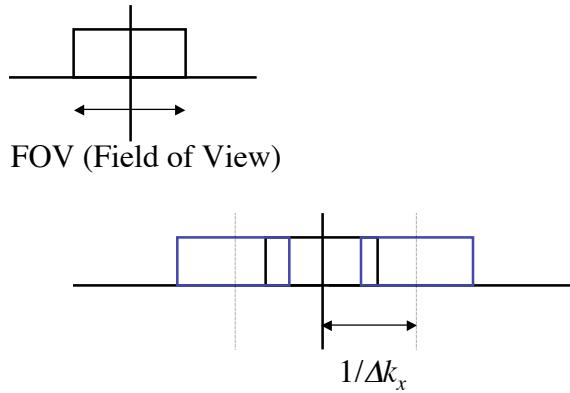
Nyquist Condition



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

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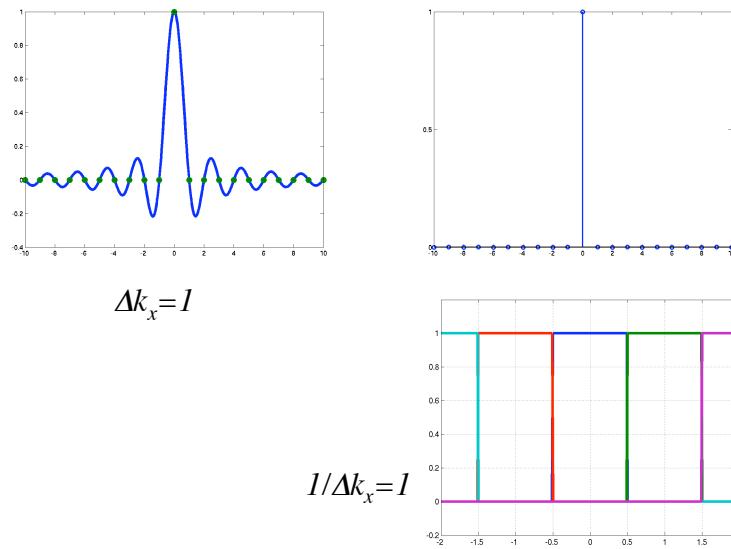
Aliasing



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

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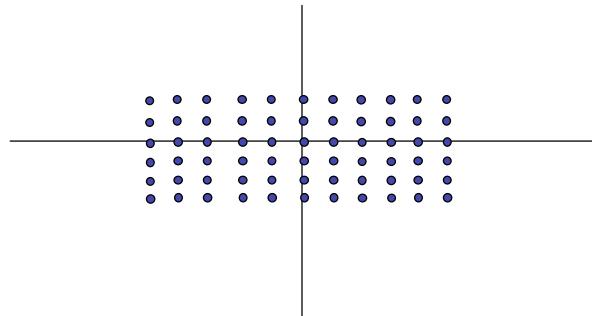
Aliasing Example



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2D Comb Function

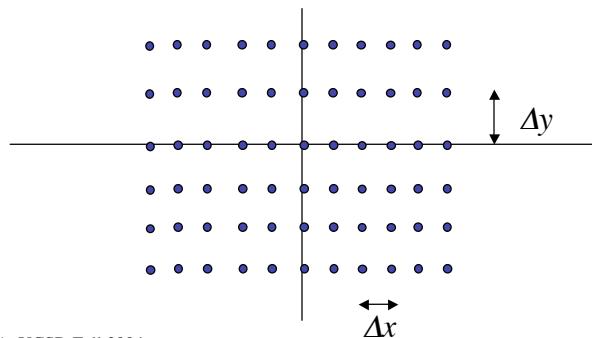
$$\begin{aligned} \text{comb}(x, y) &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x - m, y - n) \\ &= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} \delta(x - m) \delta(y - n) \\ &= \text{comb}(x) \text{comb}(y) \end{aligned}$$



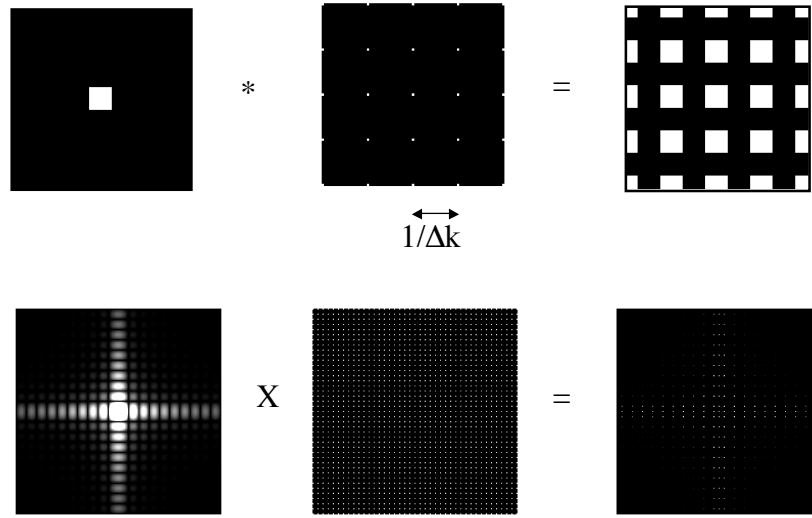
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Scaled 2D Comb Function

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



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2D k-space sampling

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

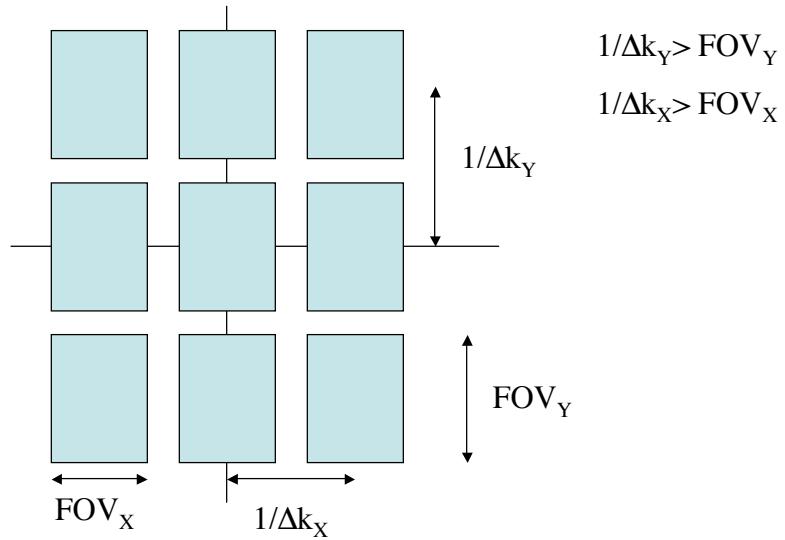
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2D k-space sampling

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

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Nyquist Conditions



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Aliasing

