

Bioengineering 280A
Principles of Biomedical Imaging

Fall Quarter 2006
CT/Fourier Lecture 3

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Topics

- Sampling Requirements in CT
- Sampling Theory

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CT Sampling Requirements

What should the size of the detectors be?

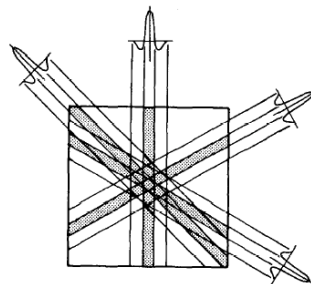
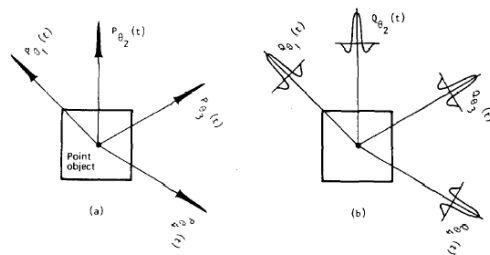
How many detectors do we need?

How many views do we need?

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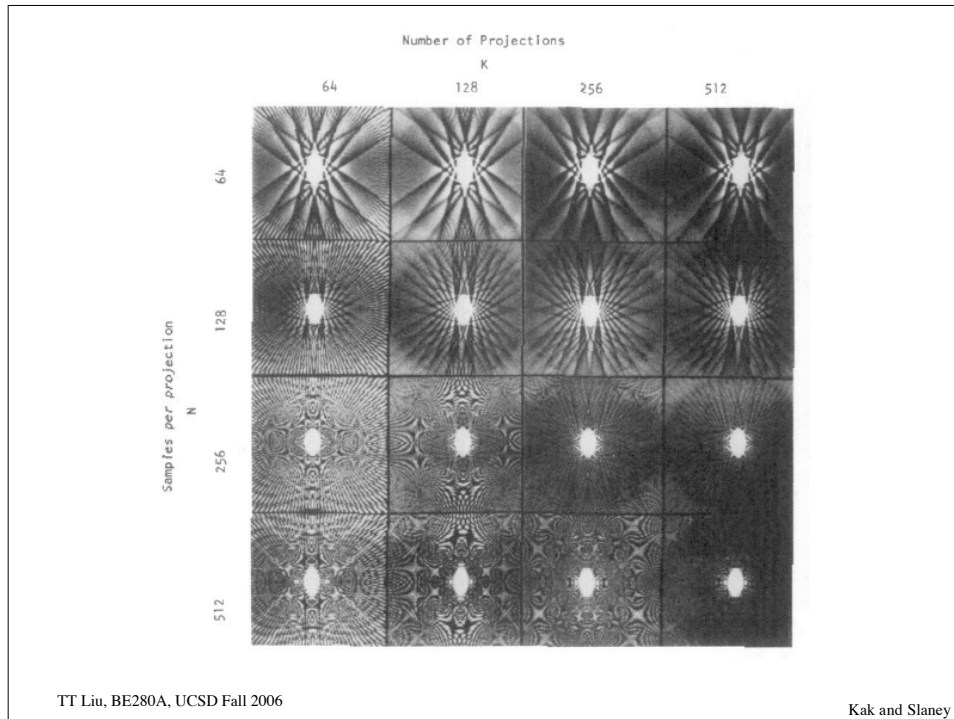
Suetens 2002

View Aliasing



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Kak and Slaney



Analog vs. Digital

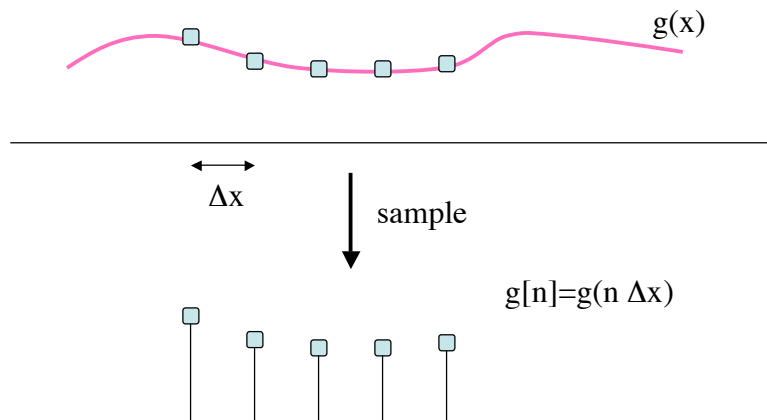
The Analog World:

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.

The Process of Sampling



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Questions

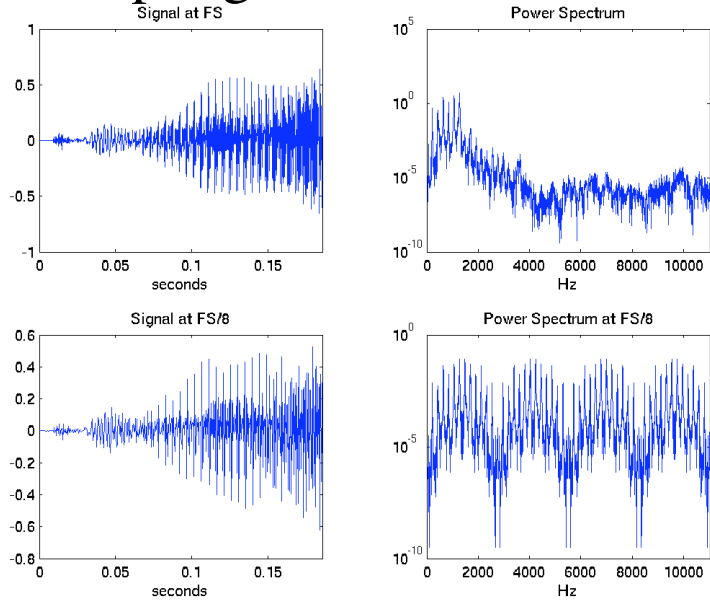
How finely do we need to sample?

What happens if we don't sample finely enough?

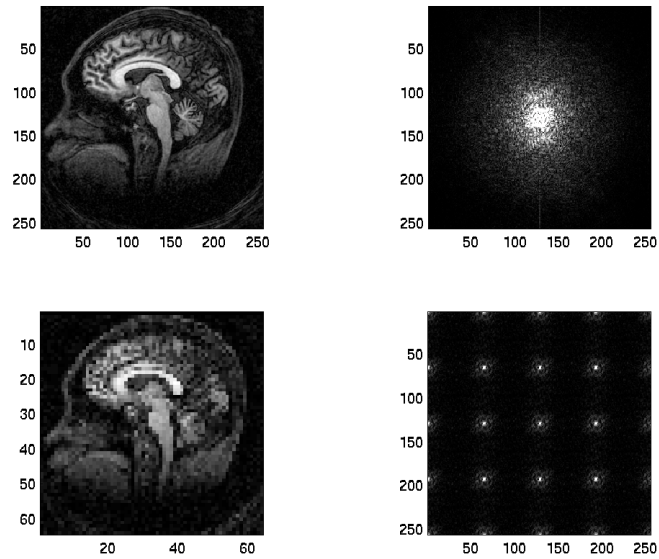
Can we reconstruct the original signal or image from its samples?

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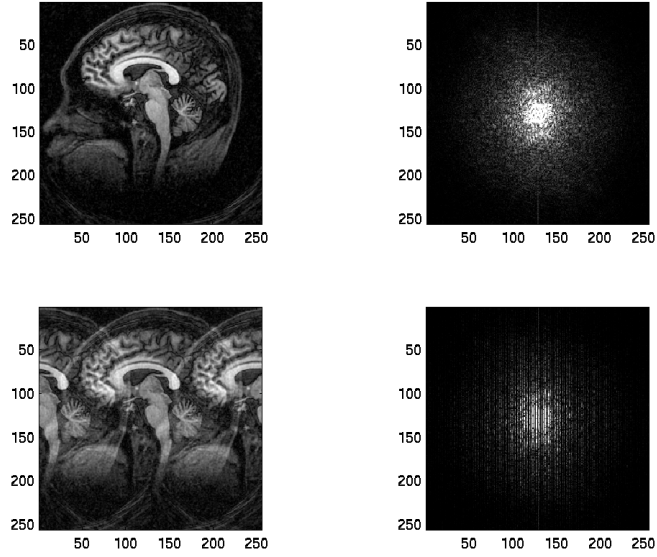
Sampling in the Time Domain



Sampling in Image Space

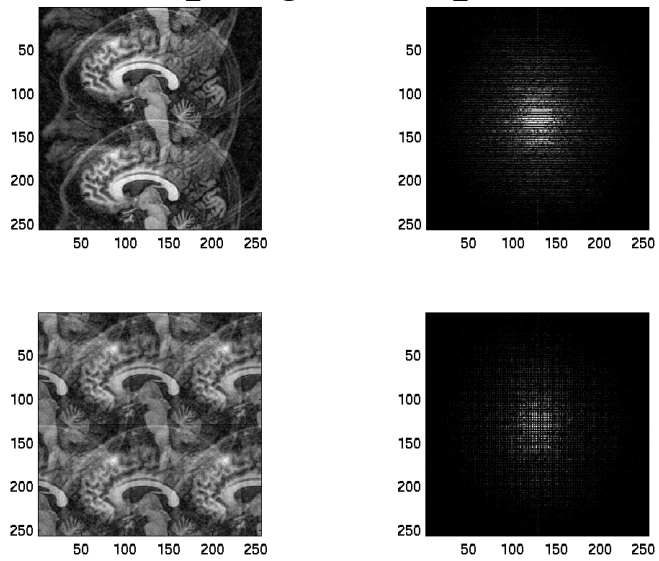


Sampling in k-space



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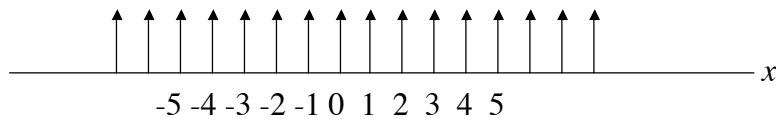
Sampling in k-space



1

Comb Function

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

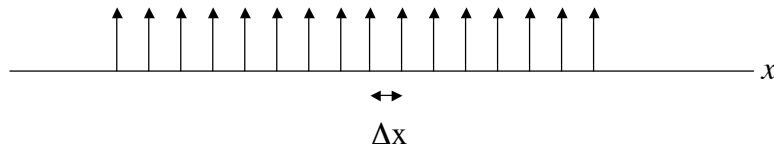


Other names: Impulse train, bed of nails, shah function.

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Scaled Comb Function

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



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1D spatial sampling

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

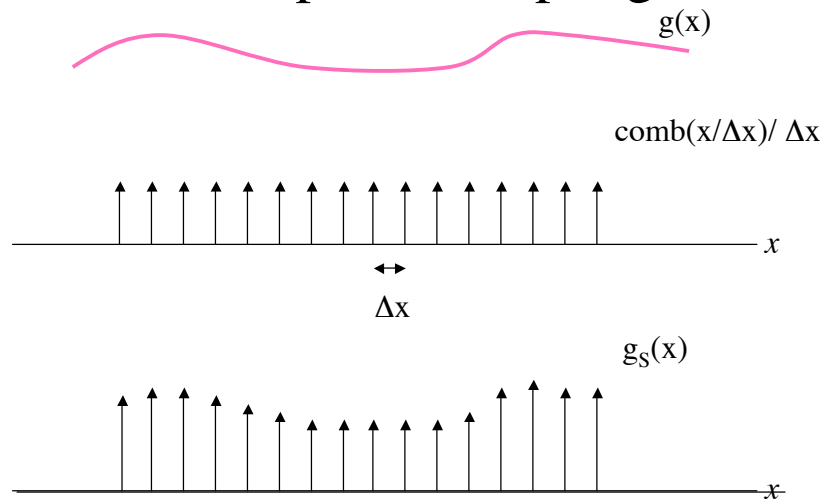
Recall the sifting property $\int_{-\infty}^{\infty} g(x) \delta(x - a) = g(a)$

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



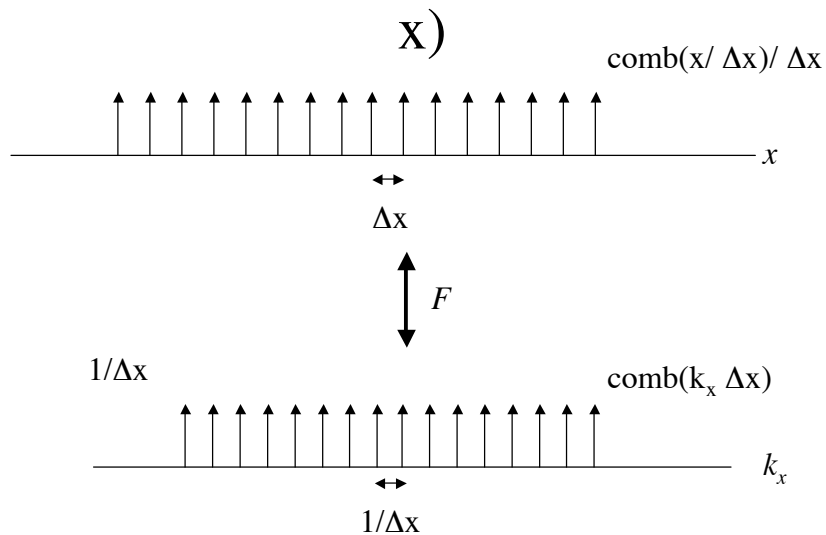
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Fourier Transform of comb(x)

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

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Fourier Transform of comb(x/ Δ



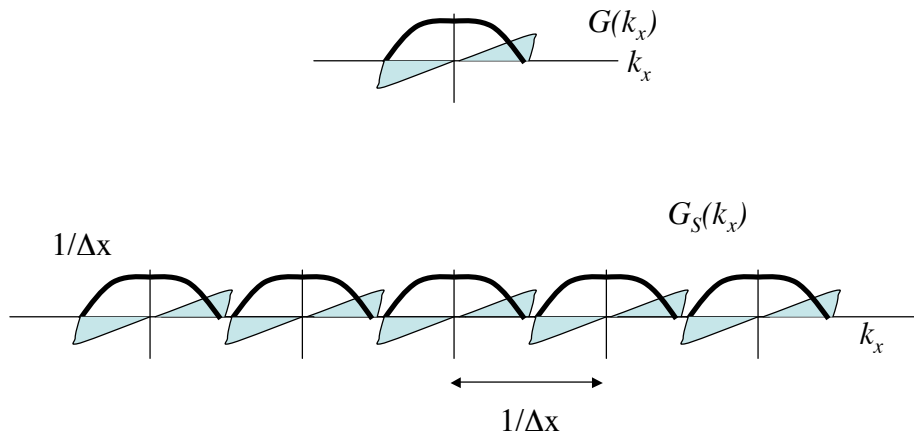
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Fourier Transform of $g_S(x)$

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

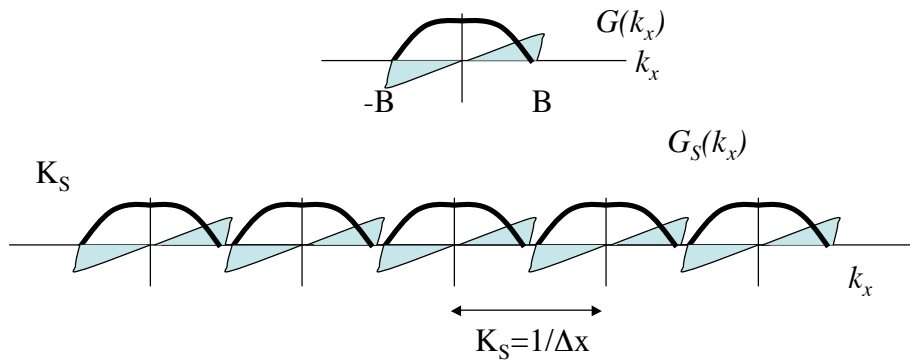
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Fourier Transform of $g_S(x)$



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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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Example

Assume that the highest spatial frequency in an object is $B = 2 \text{ cm}^{-1}$.

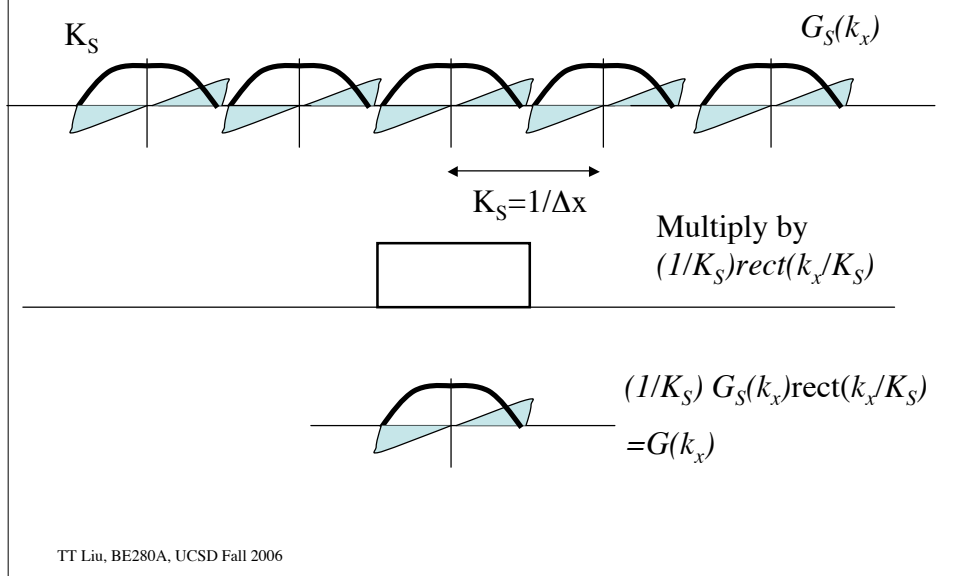
Thus, smallest spatial period is 0.5 cm.

Nyquist theorem says we need to sample with $\Delta x < 1/2B = 0.25 \text{ cm}$

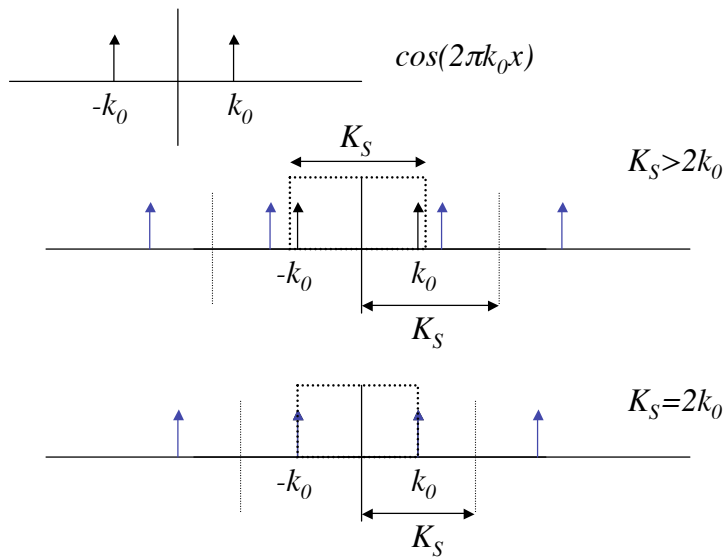
This corresponds to 2 samples per spatial period.

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Reconstruction from Samples



Example Cosine Reconstruction



Reconstruction from Samples

If the Nyquist condition is met, then

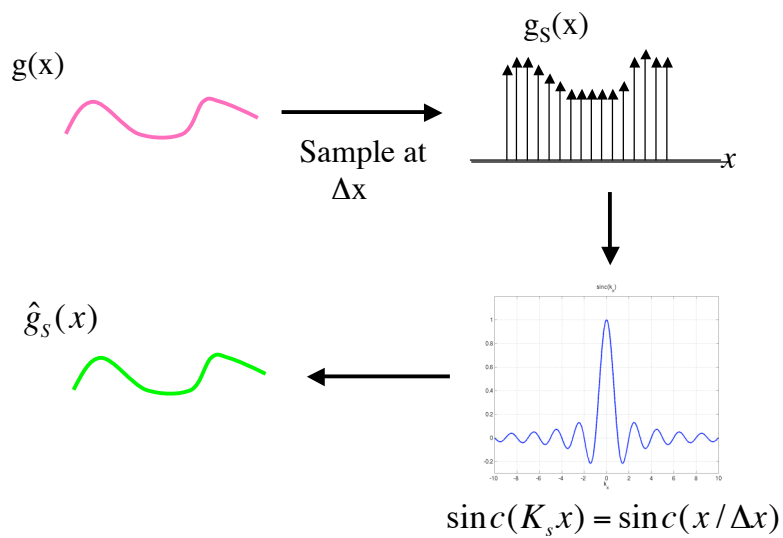
$$\hat{G}_S(k_x) = \frac{1}{K_S} G_S(k_x) \text{rect}(k_x / K_S) = G(k_x)$$

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

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

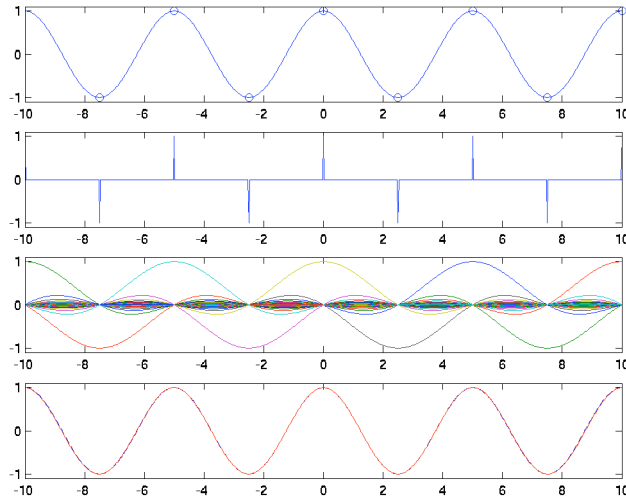
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Reconstruction from Samples



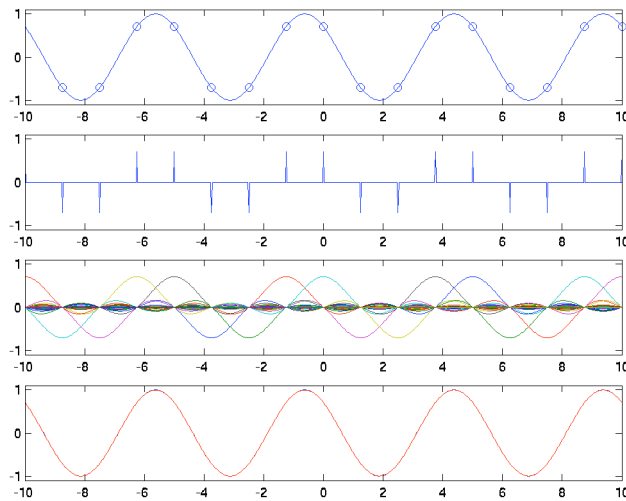
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Cosine Example with $K_s=2k_0$



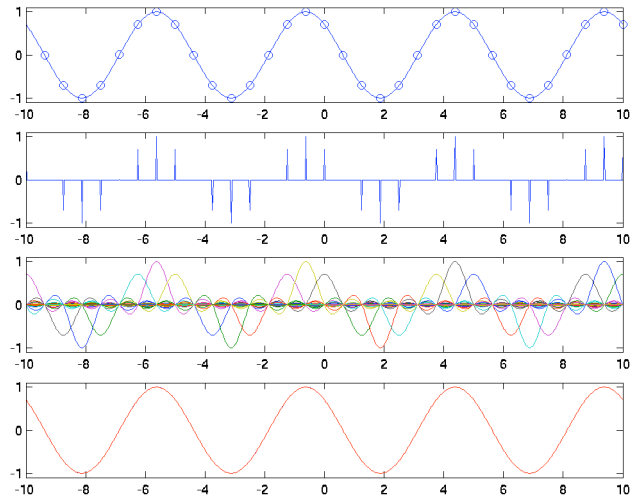
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Example with $K_s=4k_0$



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Example with $K_s = 8k_0$



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