

Bioengineering 280A
Principles of Biomedical Imaging

Fall Quarter 2013
CT/Fourier Lecture 5

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Topics

- Sampling Requirements in CT
- Sampling Theory
- Aliasing

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Recap

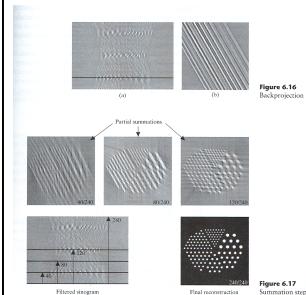


Figure 6.16
backprojection step.

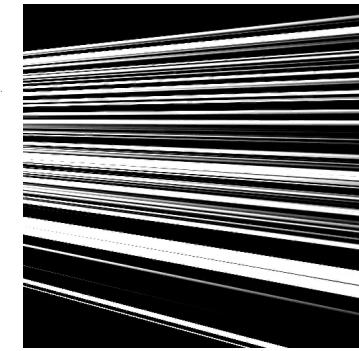


Figure 6.17
summation step.

Prince and Links 2005

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Seutens 2002

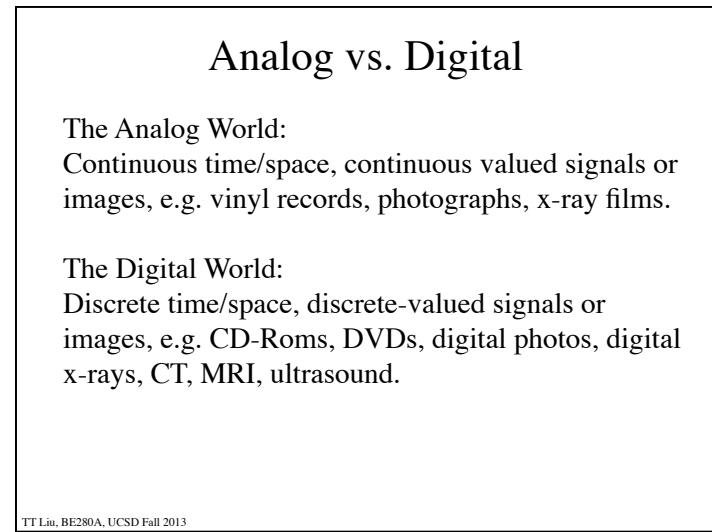
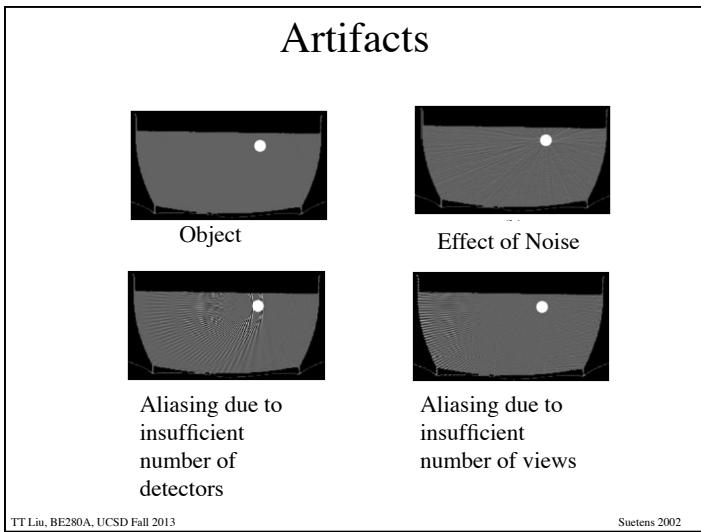
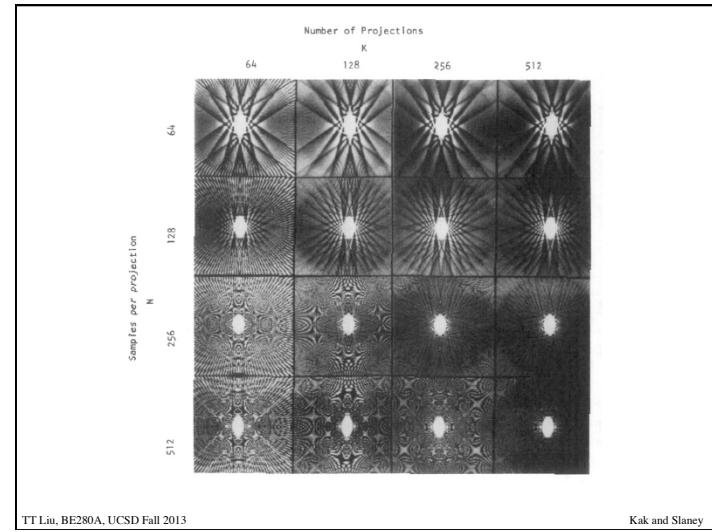
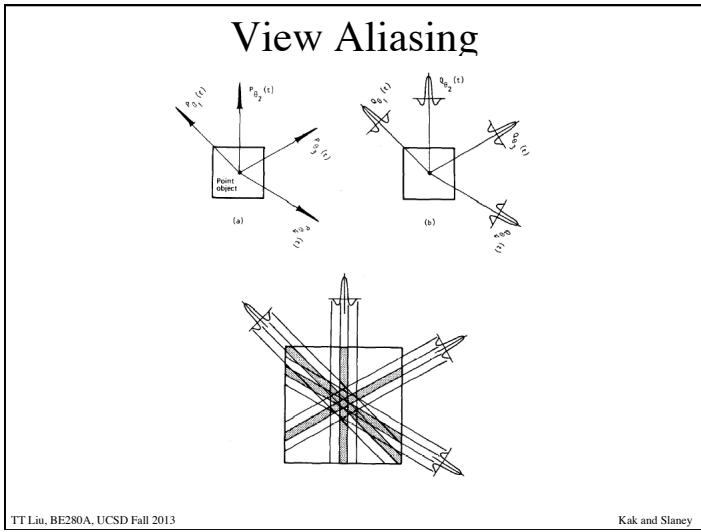
CT Sampling Requirements

What should the size of the detectors be?

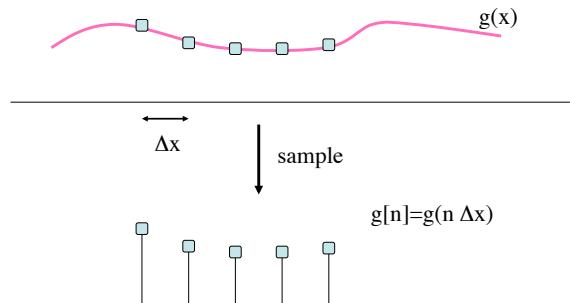
How many detectors do we need?

How many views do we need?

Suetens 2002



The Process of Sampling



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Questions

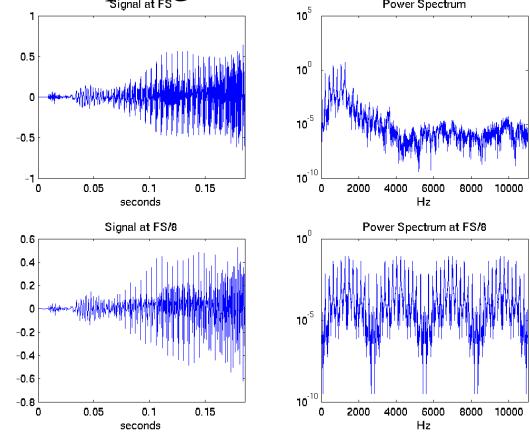
How finely do we need to sample?

What happens if we don't sample finely enough?

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

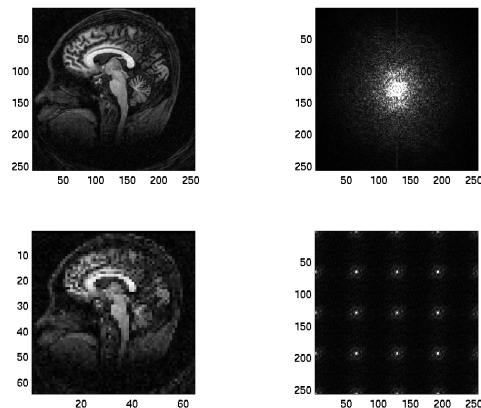
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Sampling in the Time Domain



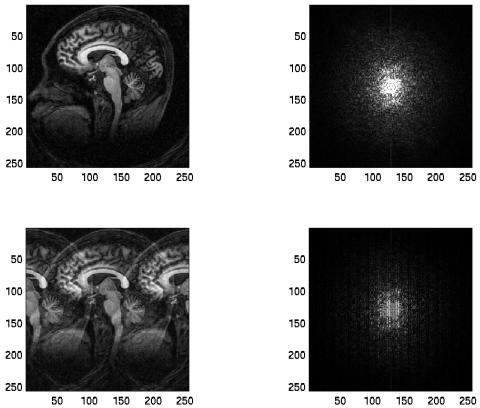
T

Sampling in Image Space



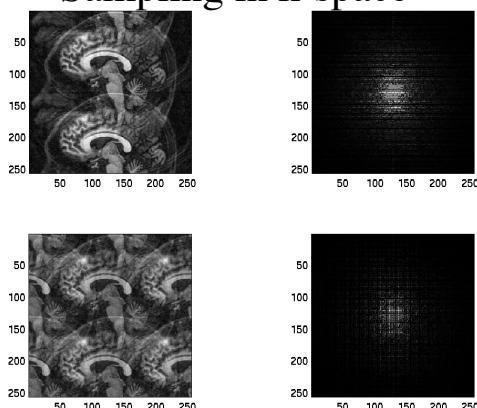
TT

Sampling in k-space



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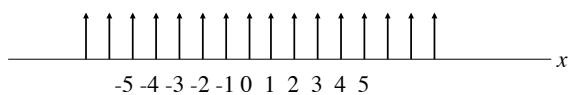
Sampling in k-space



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Comb Function

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

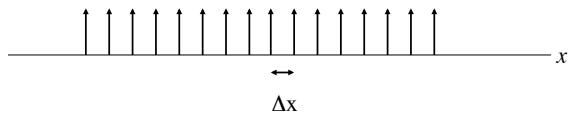


Other names: Impulse train, bed of nails, shah function.

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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

$$g_s(x) = g(x) \frac{1}{\Delta x} \text{comb}\left(\frac{x}{\Delta x}\right)$$

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

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

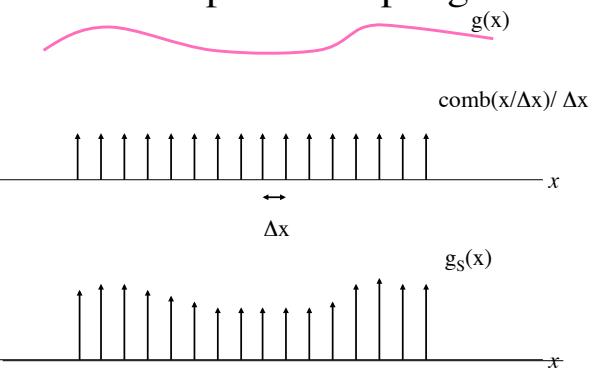
Recall the sifting property $\int_{-\infty}^{\infty} g(x)\delta(x-a) = g(a)$

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)

$$F[\text{comb}(x)] = \text{comb}(k_x)$$

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

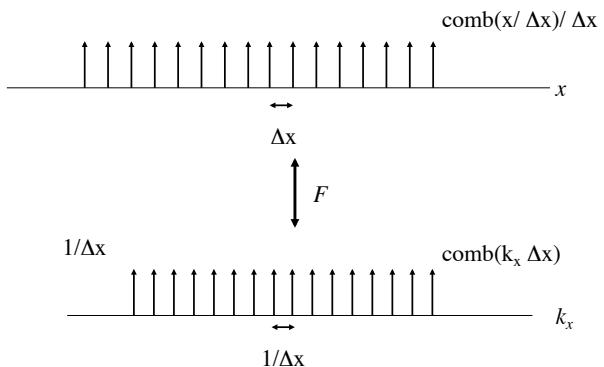
$$F\left[\frac{1}{\Delta x} \text{comb}\left(\frac{x}{\Delta x}\right)\right] = \frac{1}{\Delta x} \Delta x \text{comb}(k_x \Delta x)$$

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

$$= \frac{1}{\Delta x} \sum_{n=-\infty}^{\infty} \delta(k_x - \frac{n}{\Delta x})$$

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Fourier Transform of comb(x/ Δx)



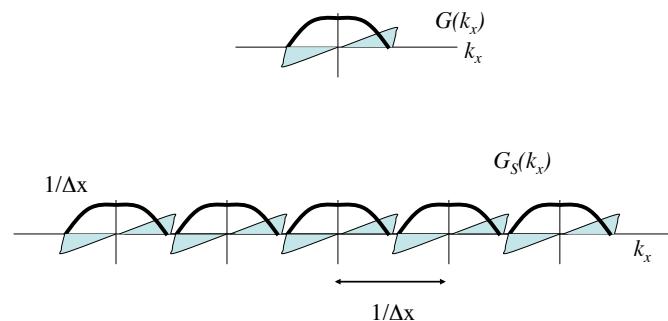
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Fourier Transform of $g_s(x)$

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

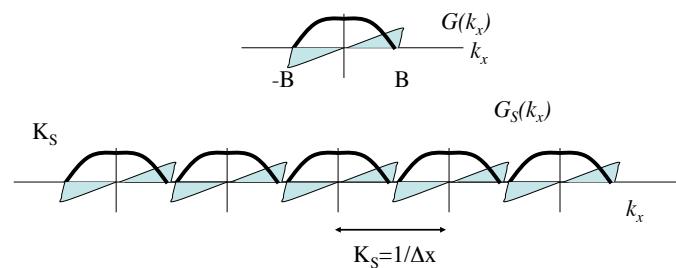
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Fourier Transform of $g_s(x)$



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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 $B = 2 \text{ cm}^{-1}$.

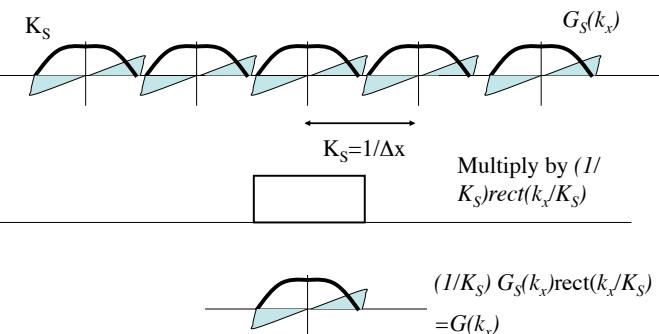
Thus, smallest spatial period is 0.5 cm .

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

This corresponds to 2 samples per spatial period.

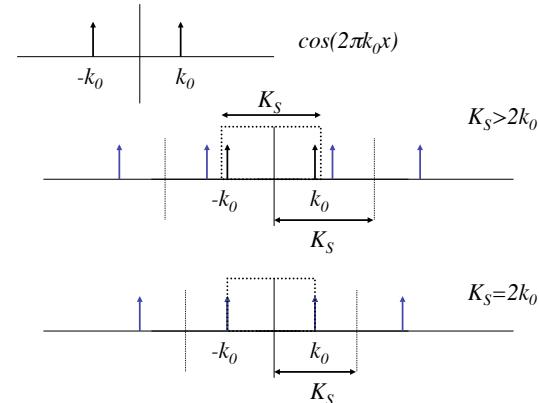
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Reconstruction from Samples



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Example Cosine Reconstruction



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Reconstruction from Samples

If the Nyquist condition is met, then

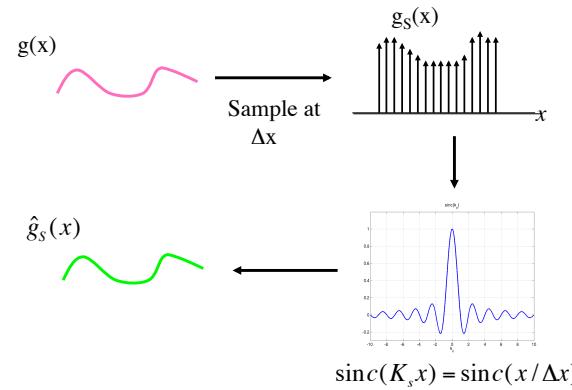
$$\hat{G}_s(k_x) = \frac{1}{K_s} G_s(k_x) \text{rect}(k_x/K_s) = G(k_x)$$

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

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

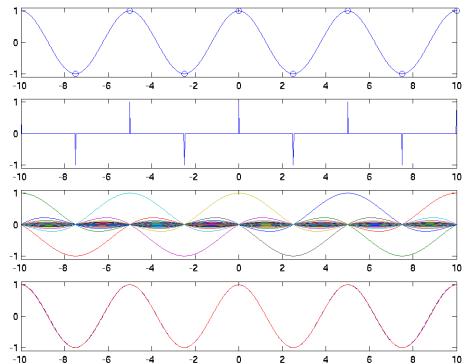
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Reconstruction from Samples



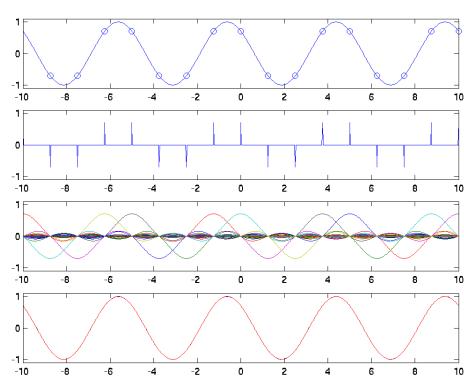
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Cosine Example with $K_s=2k_0$



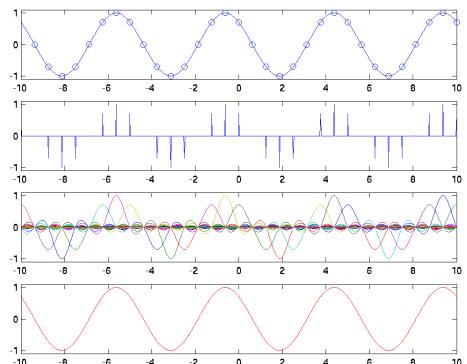
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Example with $K_s=4k_0$



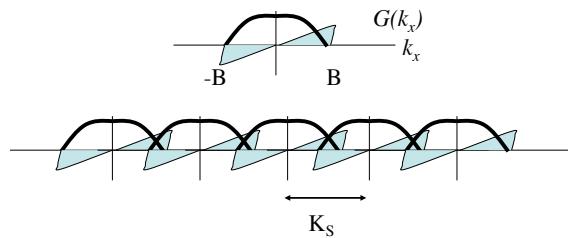
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Example with $K_s=8k_0$



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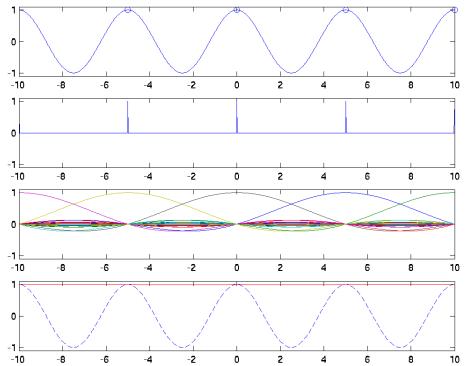
Aliasing



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

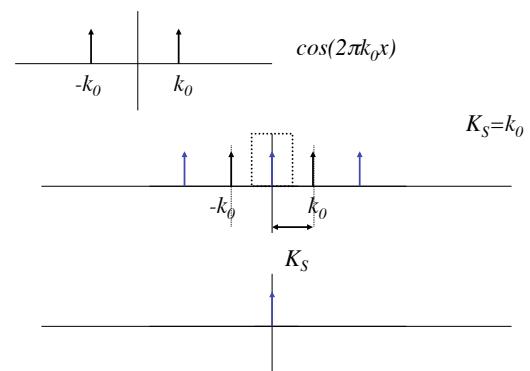
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Aliasing Example



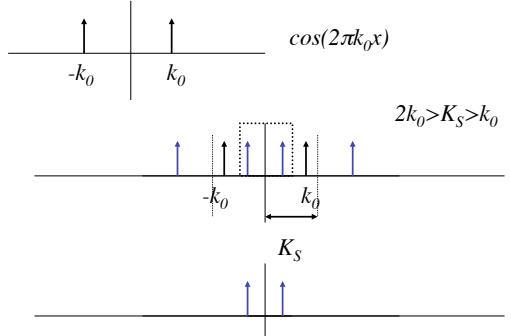
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Aliasing Example

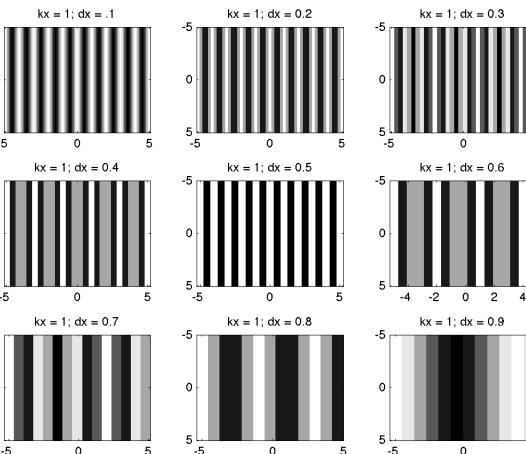


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Aliasing Example



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Example

1. Consider the function $g(x) = \cos^2(2\pi k_0 x)$. Sketch this function. You sample this signal in the spatial domain with a sampling rate $K_s = 1/\Delta x$ (e.g. samples spaced at intervals of Δx). What is the minimum sampling rate that you can use without aliasing? Give an intuitive explanation for your answer.

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Example

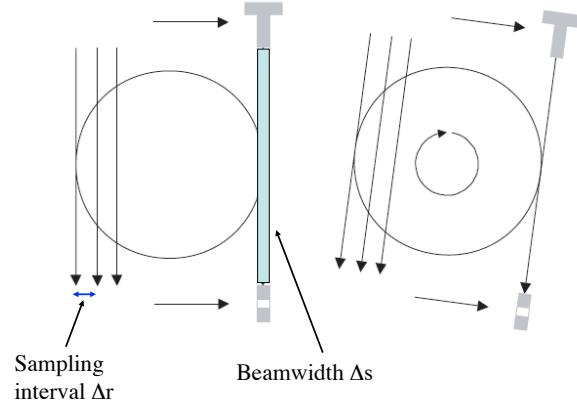
Assume that the Nyquist sampling periods of $f(x)$ and $g(x)$ are Δf and Δg , respectively. Determine the Nyquist sampling periods for

- $f(x - x_0)$
- $f(x) + g(x)$
- $f(x) * g(x)$

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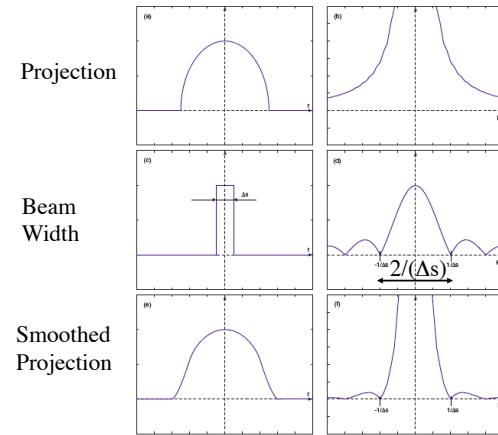
from Prince and Links 2006

Detector Sampling Requirements



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Smoothing of Projection



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Suetens 2002

Smoothing of Projection

$$g_s(l, \theta) = \text{rect}(l/\Delta s) * g(l, \theta)$$

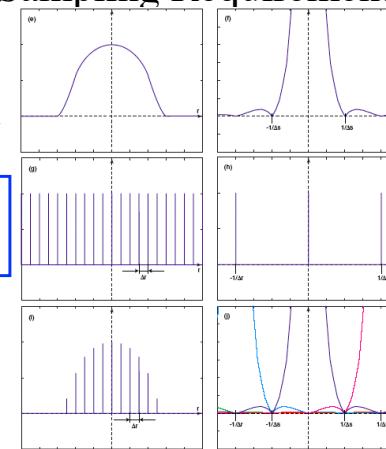
$$G_s(k_x, \theta) = \Delta s \sin c(k_x \Delta s) G(k_x, \theta)$$

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Suetens 2002

Sampling Requirements

Smoothed
Projection



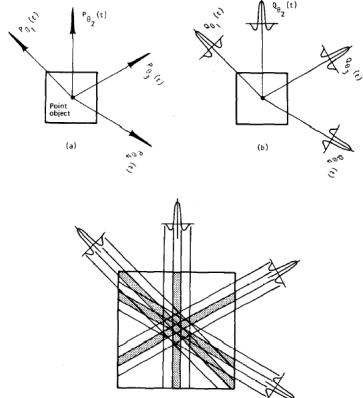
Detectors
 $\Delta r \leq \Delta s/2$

Sampled
Smooth
Projection

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View Aliasing



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Kak and Slaney

View Sampling Requirements

View Sampling -- how many views?

Basic idea is that to make the maximum angular sampling the same as the projection sampling.

$$\frac{\pi FOV}{N_{views}} = \Delta r$$

$$N_{views,360} = \frac{\pi FOV}{\Delta r} = \pi N_{proj} \quad (\text{for } 360 \text{ degrees})$$

$$N_{views,180} = \frac{\pi N_{proj}}{2} \quad (\text{for } 180 \text{ degrees})$$

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Suetens 2002

Example

beamwidth $\Delta s = 1 \text{ mm}$

Field of View (FOV) = 50 cm

$\Delta r = \Delta s/2 = 0.5 \text{ mm}$

500 mm / 0.5 mm = N = 1000 detector samples

$\pi * N = 3146$ views per 360 degrees

≈ 1500 views per 180 degrees

CT "Rule of Thumb"

$$N_{\text{view}} = N_{\text{detectors}} = N_{\text{pixels}}$$

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Suetens 2002

Example

Consider a rectangular object of width 20mm and height 40mm centered at (-10mm, -10mm). The attenuation coefficient of the object is 1 mm^{-1} . The object is imaged with a 1st generation CT scanner with a beamwidth of 1mm. The desired FOV is 100 mm.

Determine the appropriate detector size Δr and the number of radial samples needed to span the FOV. Assume that the middle two samples are acquired at coordinates of $-\Delta r/2$ and $\Delta r/2$.

Determine the number of angular samples required.

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