

Sampling and Reconstruction



Overview

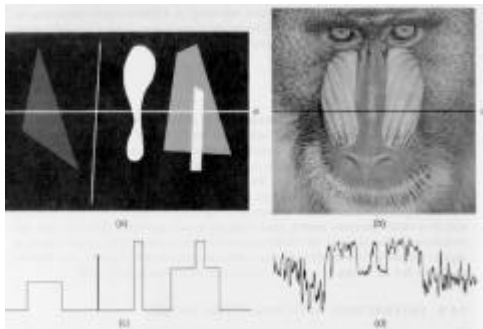
- Introduction
- Sampling Theory
 - Fourier Transform
 - Convolution & Convolution Theorem
- Reconstruction
 - Sampling Theorem
 - Reconstruction in theory and practice
- Interpolation - Zero Insertion

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Image Data

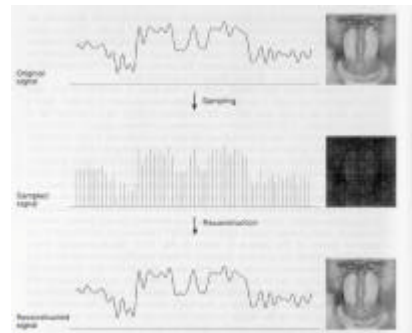


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Image Storage and Retrieval

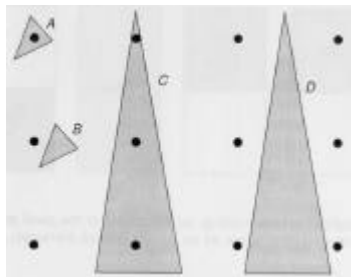


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Sampling Problems



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Sampling Theory

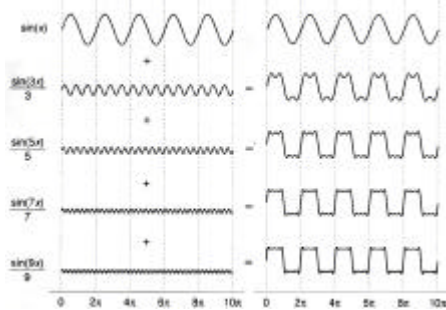
- Relationship between Signal and Samples
- View Image Data as Signals
- Signals can be plotted as intensity vs. time - spatial domain
- Signals can be represented as sum of sine waves - frequency domain

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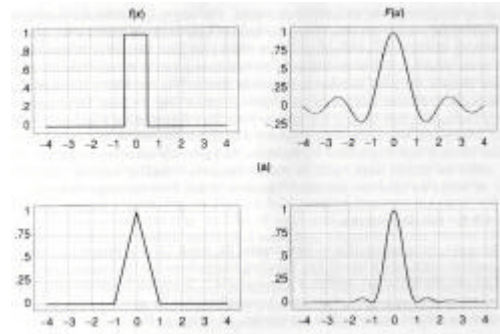
Square Wave Approximation



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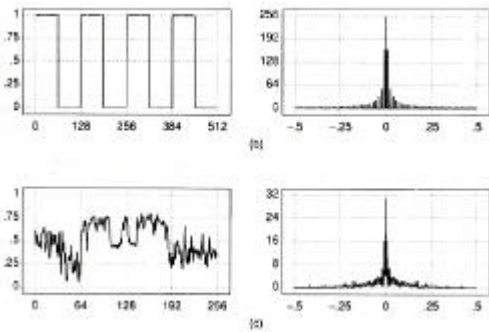
Box & Tent



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Square Wave & Scanline



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Fourier Transform

Link between spatial and frequency domain

$$f(x) = \int_{-\infty}^{\infty} F(\omega) e^{2\pi j \omega x} d\omega$$

$$F(\omega) = \int_{-\infty}^{\infty} f(x) e^{-2\pi j \omega x} dx$$

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Fourier Transform

- Yields complex functions for frequency domain
- Extends to higher dimensions
- Complex part is phase information - usually ignored

Alternative: Hartley transform

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Discrete Fourier Transform

For discrete signals (i.e. sets of samples)

$$f(x) = \sum_{\omega=0}^{N-1} F(\omega) \cdot e^{2\pi j \omega x / N}$$

$$F(\omega) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) \cdot e^{-2\pi j \omega x / N}$$

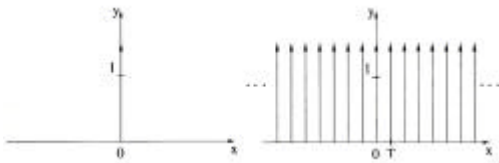
N samples: $O(N^2)$ complexity

Fast FT (FFT): $O(N \log N)$

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Base Functions



Dirac Pulse

Comb Function

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FT of Base Functions

- Impulse function: constant 1, i.e. equal energy at all frequencies
- Comb function: comb with reciprocal spacing

$$\text{comb}_T(x) \Leftrightarrow \text{comb}_{1/T}(\omega)$$

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Convolution

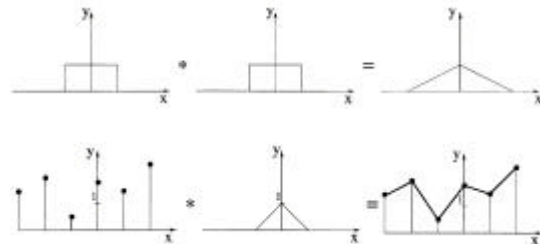
- Operation on two functions
- Produces a new function which is a sliding weighted average of a function. The second function provides the weights.

$$(f_1 * f_2)(x) = \int_{-\infty}^{\infty} f_1(x') f_2(x - x') dx'$$

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Convolution - Examples



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Convolution Theorem

The spectrum of the convolution of two functions is equivalent to the product of the transforms of both input signals, and vice versa.

$$f_1 * f_2 \equiv F_1 F_2$$

$$F_1 * F_2 \equiv f_1 f_2$$

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Example - Low-Pass

Low-pass filtering performed on Mandrill scanline

Spatial domain: convolution with sinc function

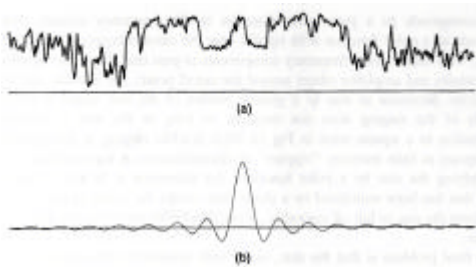
Frequency domain: cutoff of high frequencies - multiplication with box filter

Sinc function corresponds to box function and vice versa!

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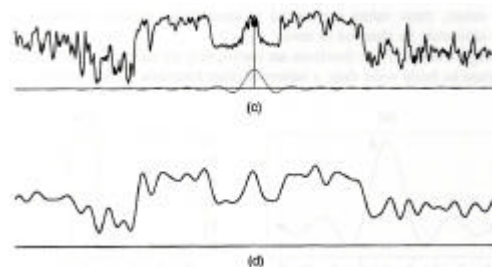
Low-Pass in Spatial Domain 1



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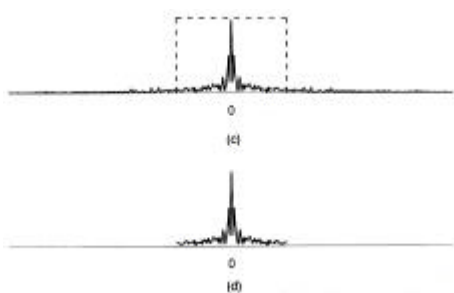
Low-Pass in Spatial Domain 2



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Low-Pass in Frequency Domain



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Sampling

The process of sampling is a multiplication of the signal with a comb function.

$$f_s(x) = f(x) \cdot \text{comb}_T(x)$$

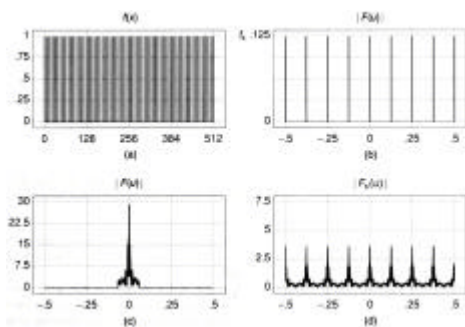
The frequency response is convolved with a transformed comb function.

$$F_s(\omega) = F(\omega) * \text{comb}_{1/T}(\omega)$$

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Spectrum of a Sampled Signal



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Reconstruction

Recovering the original function from a set of samples

- Sampling theorem
- Ideal reconstruction
 - Sinc function
- Reconstruction in practice

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Definitions

- A function is called *band-limited* if it contains no frequencies outside the interval $[-u, u]$. u is called the *bandwidth* of the function
- The *Nyquist frequency* of a function is twice its bandwidth, i.e. $w = 2u$

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Sampling Theorem

A function $f(x)$ that is

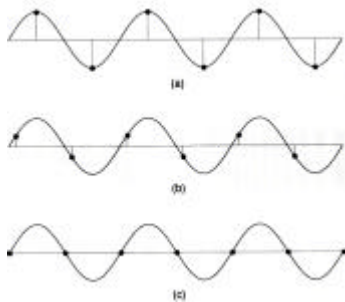
- band-limited and
 - sampled above the Nyquist frequency
- is completely determined by its samples.

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Sampling at Nyquist Frequency



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Sampling Below Nyquist f



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Ideal Reconstruction

- Replicas in frequency domain must not overlap
- Multiplying the frequency response with a box filter of the width of the original bandwidth restores original
- Amounts to convolution with Sinc function

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Sinc function

- Infinite in extent
- Ideal reconstruction filter
- FT of box function

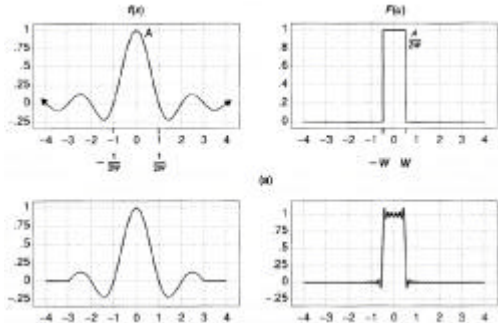
$$\text{sinc}(x) = \begin{cases} \frac{\sin px}{px} & \text{if } x \neq 0 \\ 1 & \text{if } x = 0 \end{cases}$$

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Sinc & Truncated Sinc



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Reconstruction: Examples

Sampling and reconstruction of the Mandrill image scanline signal

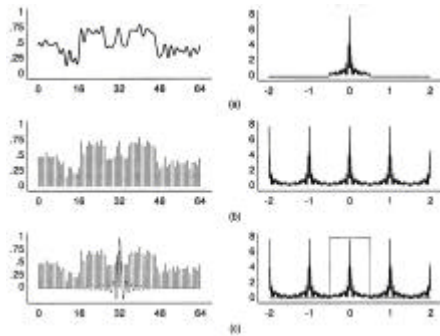
- with adequate sampling rate
- with inadequate sampling rate
- demonstration of band-limiting

With Sinc and tent reconstruction kernels

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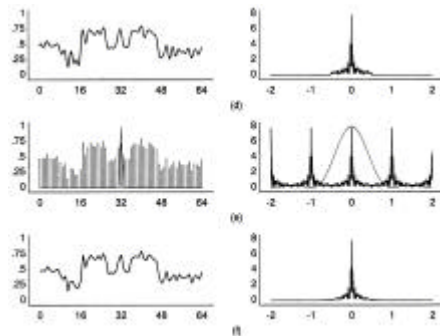
Adequate Sampling Rate



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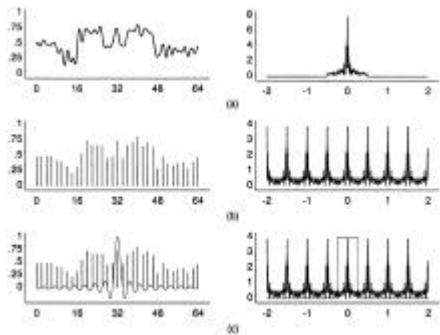
Adequate Sampling Rate



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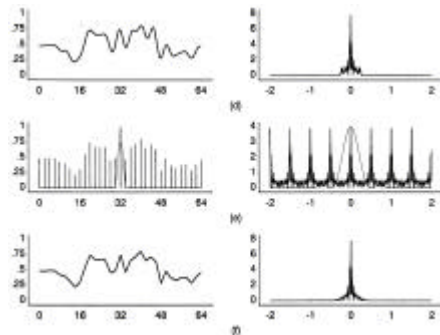
Inadequate Sampling Rate



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Inadequate Sampling Rate



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Band-Limiting a Signal

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Band-Limiting a Signal

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Reconstruction in Practice

Problem: which reconstruction kernel should be used?

- Genuine Sinc function unusable in practice
- Truncated Sinc often sub-optimal
- Various approximations exist; none is optimal for all purposes

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Tasks of Reconstruction Filters

- Remove the extraneous replicas of the frequency response
- Retain the original undistorted frequency response

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Used Reconstruction Filters

- Nearest neighbour
- Linear interpolation
- Symmetric cubic filters
- Windowed Sinc

More sophisticated ways of truncating the Sinc function

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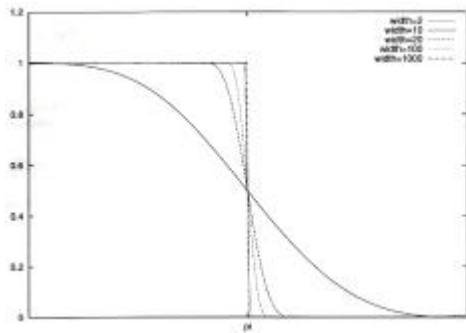
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Box & Tent Responses

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Windowed Sinc Responses



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Sampling & Reconstruction Errors

- Aliasing: due to overlap of original frequency response with replicas - information loss
- Truncation Error: due to use of a finite reconstruction filter instead of the infinite Sinc filter
- Non-Sinc error: due to use of a reconstruction filter that has a shape different from the Sinc filter

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Interpolation - Zero Insertion

Operates on series of n samples
Takes advantage of DFT properties

- Perform DFT on series
- Append zeros to the sequence
- Perform the inverse DFT

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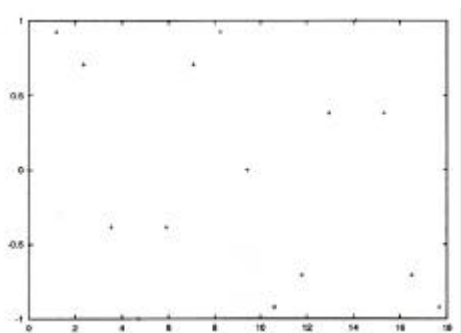
Zero Insertion - Properties

- Preserves frequency spectrum
- Original signal has to be sampled above Nyquist frequency
- Values can only be interpolated at evenly spaced locations
- The whole series must be accessible, and it is always completely processed

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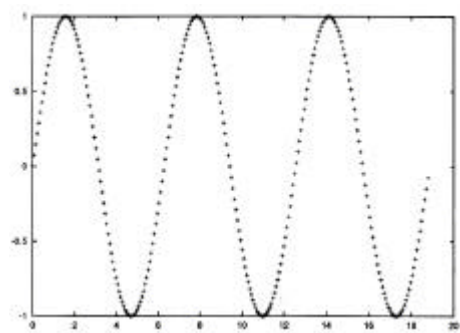
Zero Insertion - Original Series



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Zero Insertion - Interpolation



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- What we need around here is more aliasing, Jim Blinn, IEEE Computer Graphics and Applications, January 1989
- Return of the Jaggy, Jim Blinn, IEEE Computer Graphics and Applications, March 1989

