



Image Sampling and Reconstruction

Thomas Funkhouser
Princeton University
COS 426, Fall 1999



Overview

- Image sampling and reconstruction
 - Pixels are discrete samples of continuous function
 - Frequency analysis
- Image resampling
 - Aliasing
 - Filters and convolution

Image Processing

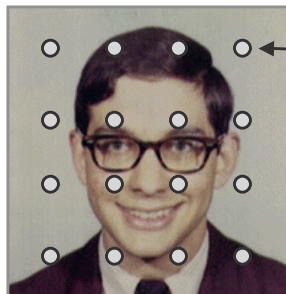


- Quantization
 - Uniform Quantization
 - Random dither
 - Ordered dither
 - Floyd-Steinberg dither
- Pixel operations
 - Add random noise
 - Add luminance
 - Add contrast
 - Add saturation
- Filtering
 - Blur
 - Detect edges
- Warping
 - Scale
 - Rotate
 - Warps
 - Morphs
- Combining
 - Composite

Image Sampling



- An image is a 2D rectilinear array of samples
 - Quantization due to limited intensity resolution
 - Sampling due to limited spatial and temporal resolution

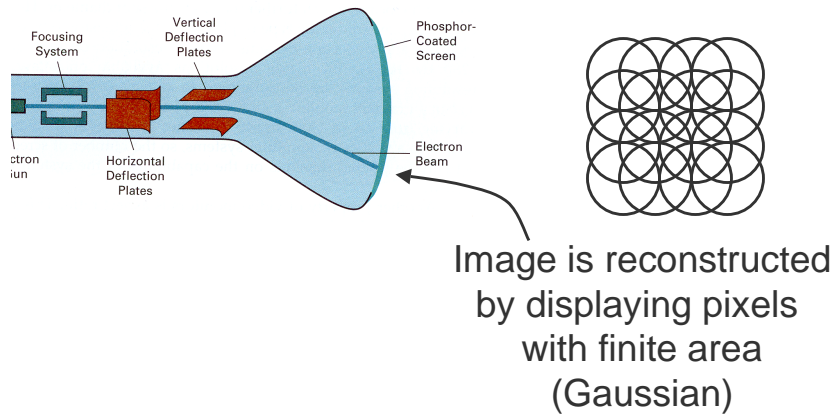


Pixels are
infinitely small
point samples

Image Reconstruction



- Re-create continuous image from samples
 - Example: cathode ray tube



Sampling and Reconstruction

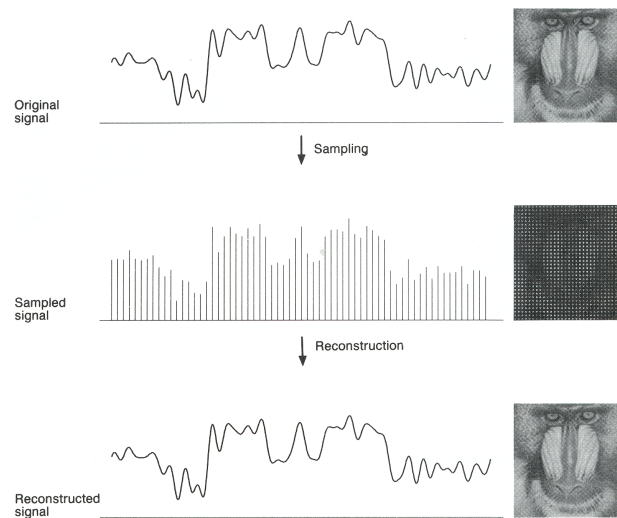
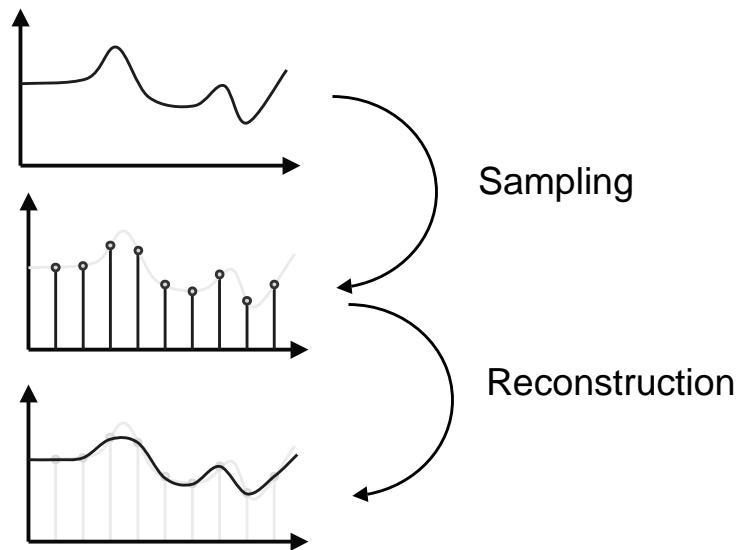


Figure 19.9 FvDFH

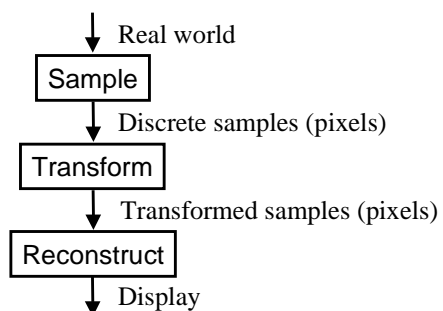
Sampling and Reconstruction



Sample Processing



- Apply function to samples
 - Linear functions
 - Linear reconstructions



Adjusting Brightness



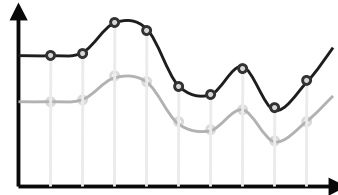
- Simply scale pixel components
 - Must clamp to range (e.g., 0 to 255)



Original



Brighter



Adjusting Contrast



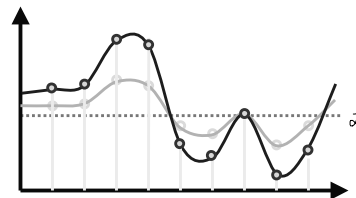
- Compute mean luminance \bar{L} for all pixels
 - luminance = $0.30*r + 0.59*g + 0.11*b$
- Scale deviation from \bar{L} for each pixel component
 - Must clamp to range (e.g., 0 to 255)



Original



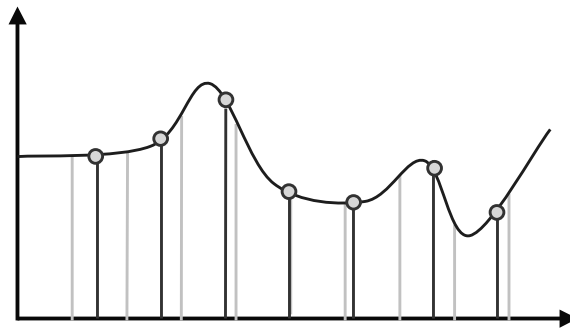
More Contrast



Resampling



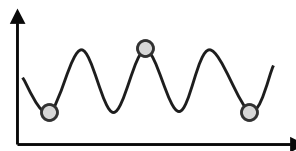
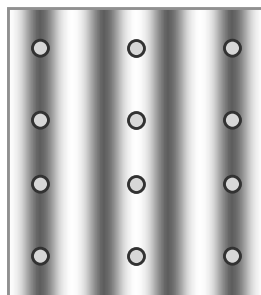
- What if samples are at new locations?
 - Image scaling, warping, etc.



Sampling Theory



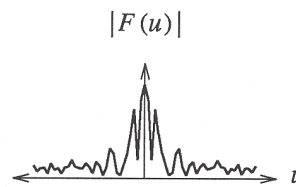
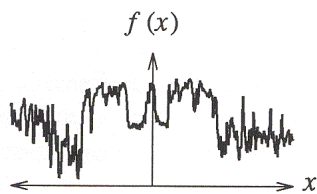
- How many samples are required to represent a given signal without loss of information?
- What signals can be reconstructed without loss for a given sampling rate?



Spectral Analysis



- Spatial domain:
 - Function: $f(x)$
 - Filtering: convolution
- Frequency domain:
 - Function: $F(u)$
 - Filtering: multiplication



Any signal can be written as a sum of periodic functions.

Fourier Transform

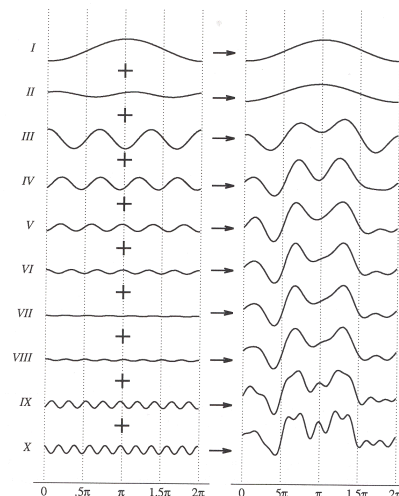
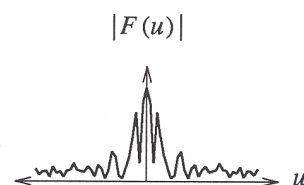
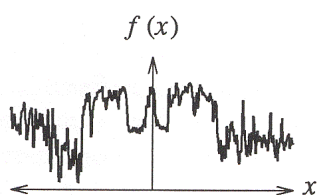


Figure 2.6 Wolberg

Fourier Transform

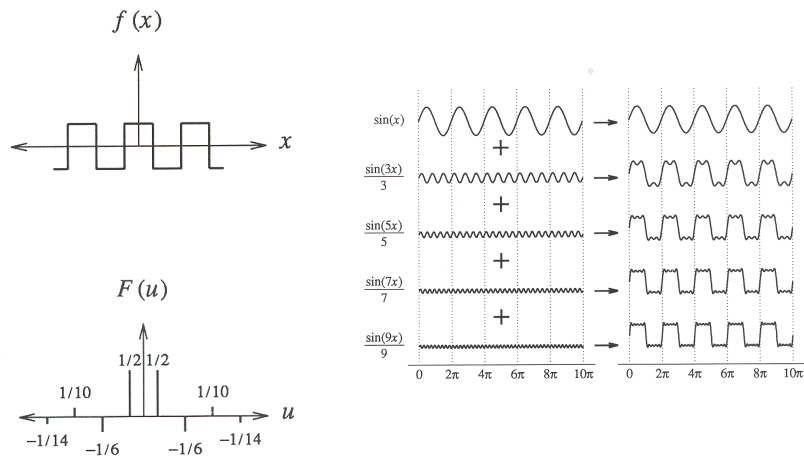


Figure 2.5 Wolberg

Fourier Transform



- Fourier transform:

$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-i2\pi x} dx$$

- Inverse Fourier transform:

$$f(x) = \int_{-\infty}^{\infty} F(u) e^{+i2\pi u} du$$

Aliasing



- Artifacts due to undersampling

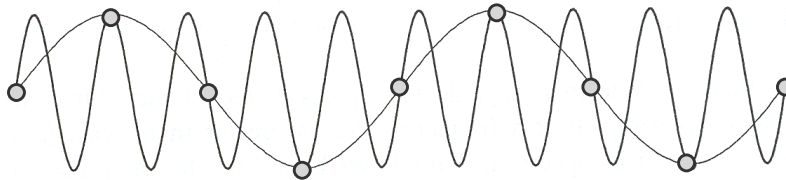
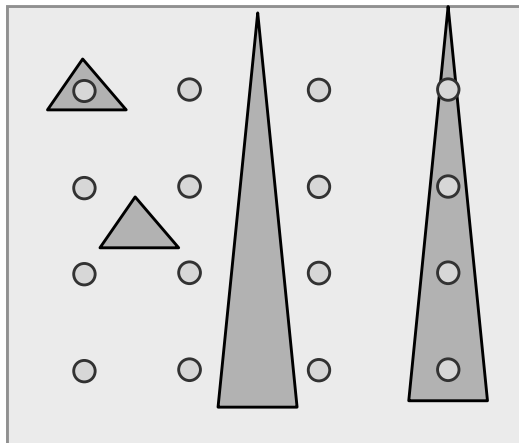


Figure 14.17 FvDFH

Spatial Aliasing



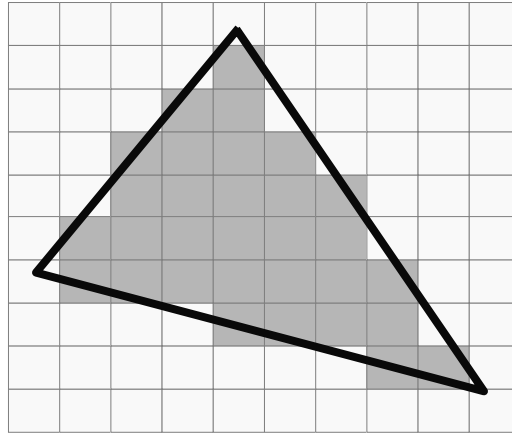
- Point sampling problems



Spatial Aliasing



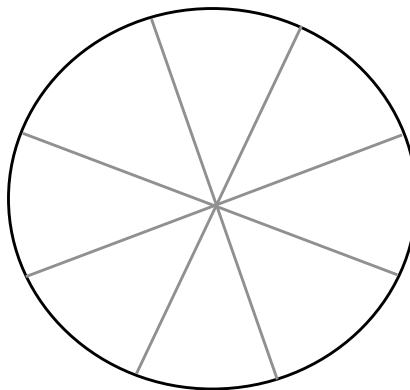
- Jaggies



Temporal Aliasing



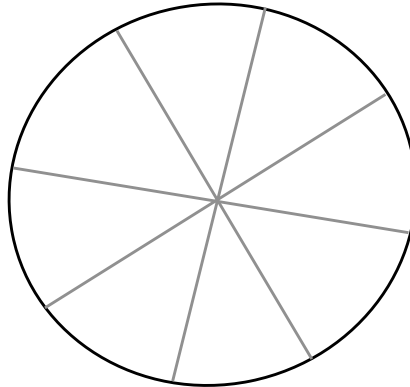
- Artifacts due to limited temporal resolution
 - Strobbing
 - Flickering



Temporal Aliasing



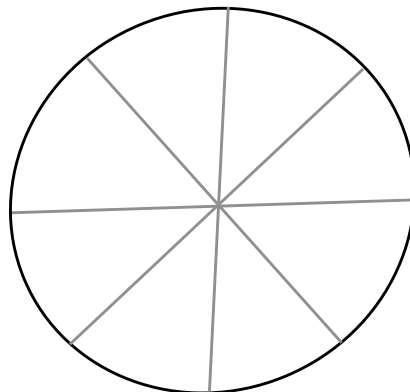
- Artifacts due to limited temporal resolution
 - Strobbing
 - Flickering



Temporal Aliasing



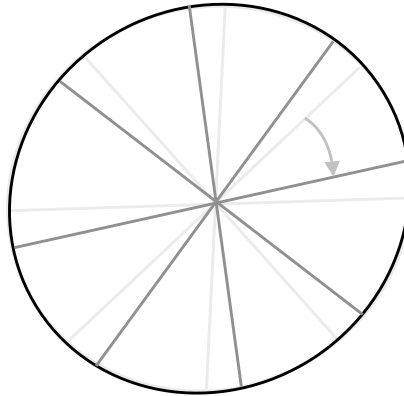
- Artifacts due to limited temporal resolution
 - Strobbing
 - Flickering



Temporal Aliasing



- Artifacts due to limited temporal resolution
 - Strobing
 - Flickering



Sampling Theorem



- A signal can be reconstructed from its samples, if the original signal has no frequencies above $1/2$ the sampling frequency - Shannon
- The minimum sampling rate for bandlimited function is called “Nyquist rate”

A signal is bandlimited if its highest frequency is bounded. The frequency is called the bandwidth.

Antialiasing



- Sample at higher rate
 - Not always possible
- Pre-filter to form bandlimited signal
 - Form bandlimited function (low-pass filter)

Unfortunately,
real image signals
are rarely bandlimited

Convolution



- Convolution of two functions (= filtering):

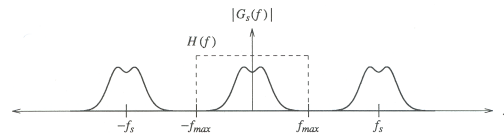
$$g(x) = f(x) \otimes h(x) = \int_{-\infty}^{\infty} f(\lambda) h(x - \lambda) d\lambda$$

- Convolution theorem
 - Convolution in frequency domain is same as multiplication in spatial domain, and vice-versa

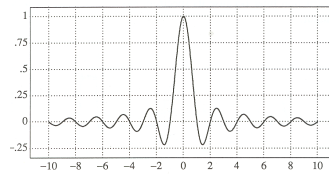
Ideal Low-Pass Filter



- Frequency domain



- Spatial domain



$$\text{Sinc}(x) = \frac{\sin \pi x}{\pi x}$$

Figure 4.5 Wolberg

Image Processing

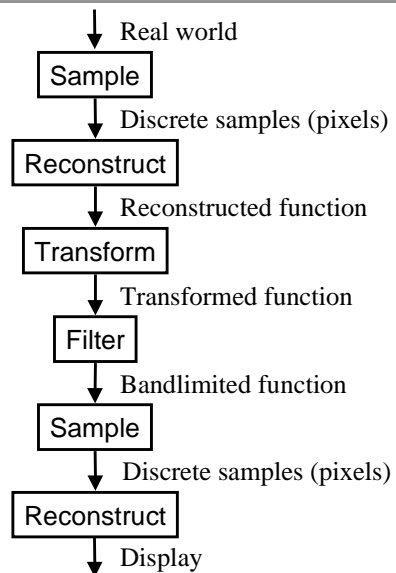


Image Processing

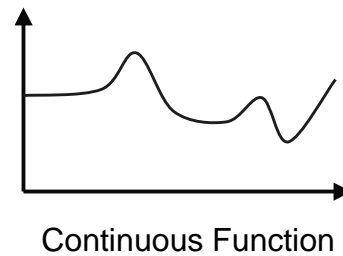
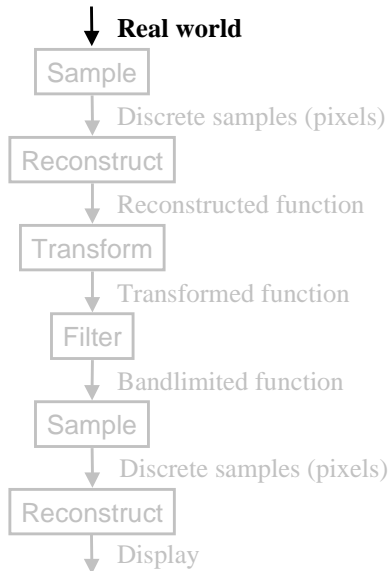


Image Processing

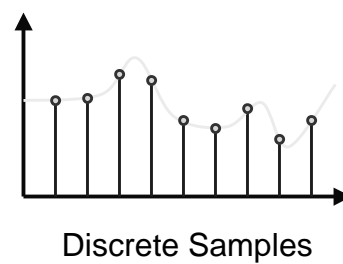
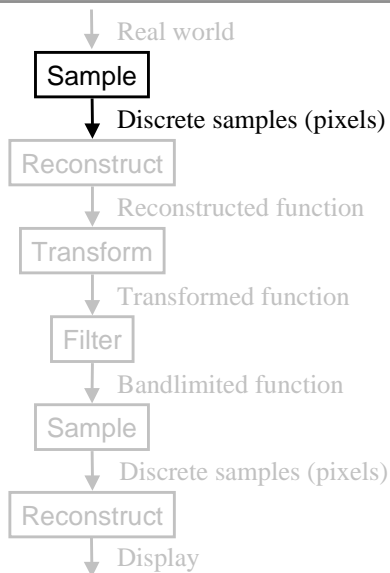


Image Processing

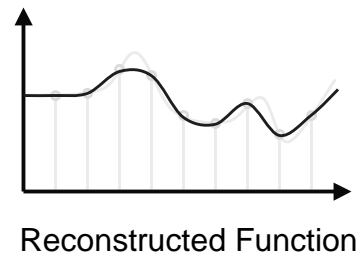
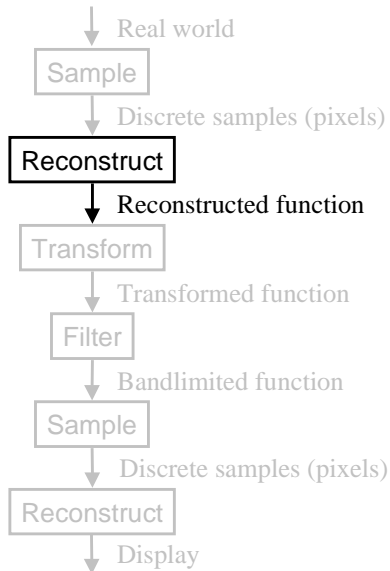


Image Processing

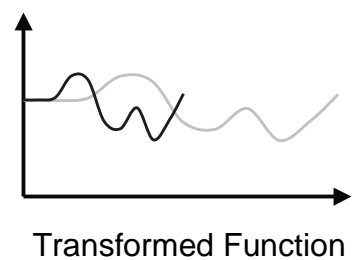
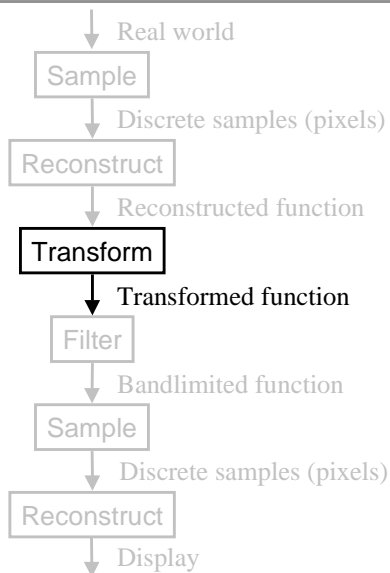


Image Processing

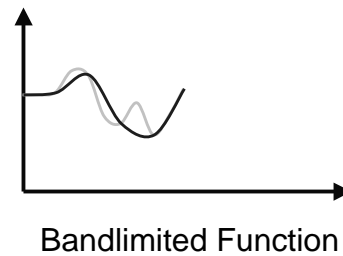
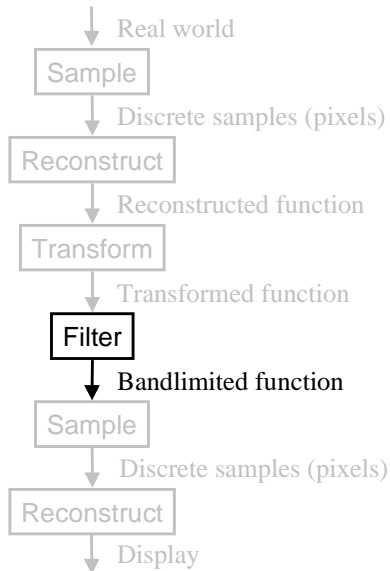


Image Processing

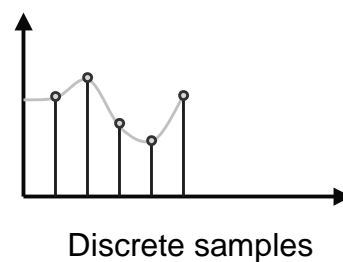
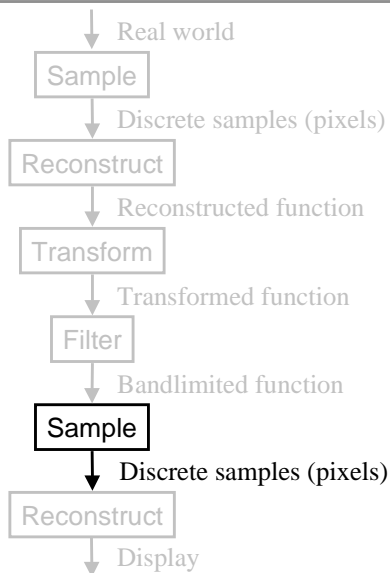
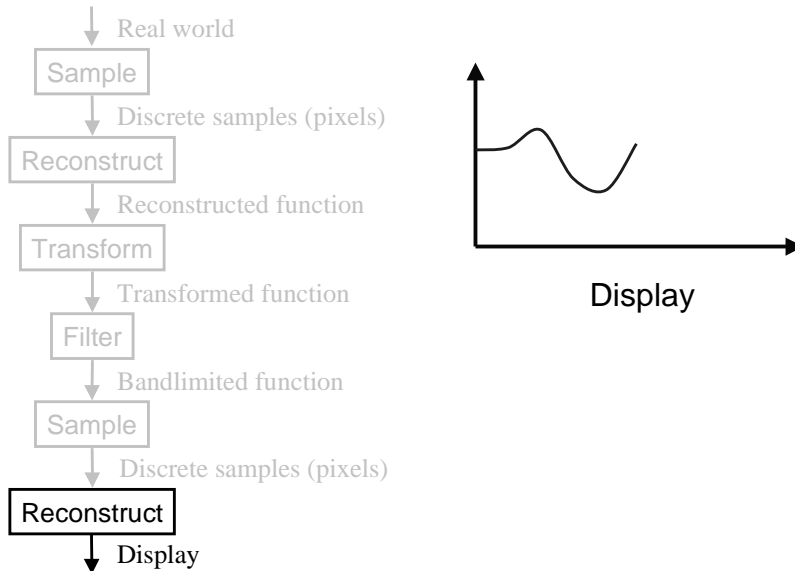


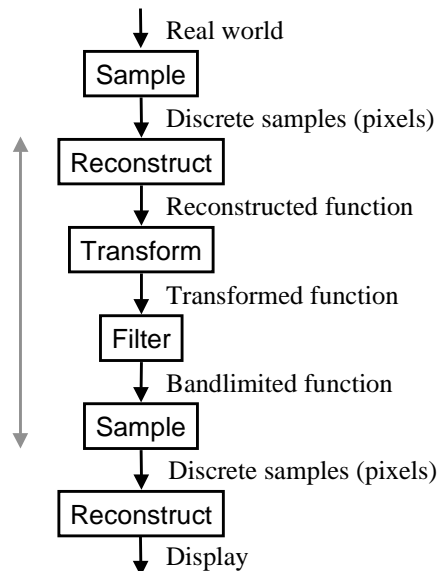
Image Processing



Practical Image Processing



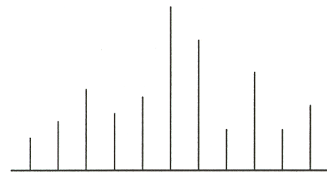
- Finite low-pass filters
 - Point sampling (bad)
 - Triangle filter
 - Gaussian filter



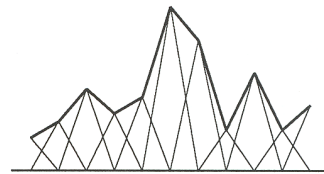
Triangle Filter



- Convolution with triangle filter



Input



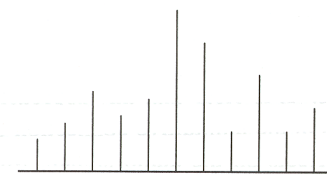
Output

Figure 2.4 Wolberg

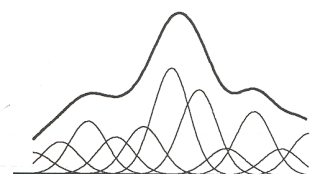
Gaussian Filter



- Convolution with Gaussian filter



Input



Output

Figure 2.4 Wolberg

Adjust Blurriness



- Convolve with filter whose support covers more than one pixel



Original



Blur

Edge Detection



- Convolve with a filter that finds differences between neighbor pixels



Original



Detect edges

Summary



- Image processing is a resampling problem
 - Aliasing
 - Filtering

