
Computer Graphics

- Anti-Aliasing & Super-Sampling -

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Organization

- **Registration HISPOS:**
 - deadline: **December 1st**
- **Final Exam**
 - date change: **February 19th**
- **Rendering Competition**
 - final deadline: **January 31st**
 - pre-deadline: January 24th

Overview

- **Last time**
 - Filtering
 - Signal processing
- **Today**
 - Sampling
 - Anti-aliasing & supersampling
- **Next lecture**
 - The Human Visual System

Discrete Fourier Transform

- **Equally-spaced function samples**
 - Function values known only at discrete points
 - Physical measurements
 - Pixel positions in an image !
- **Fourier Analysis**

$$a_k = 1/N \sum_i \sin(2\pi k i / N) f_i, \quad b_k = 1/N \sum_i \cos(2\pi k i / N) f_i$$

- yields amplitude and phase per frequency
- Sum over all measurement points N
- $k=0,1,2, \dots, ?$ Highest possible frequency ?
 \Rightarrow **Nyquist frequency**
 - Sampling rate N_i
 - 2 samples / period \Leftrightarrow 0.5 cycles per pixel
- $\Rightarrow k \leq N / 2$

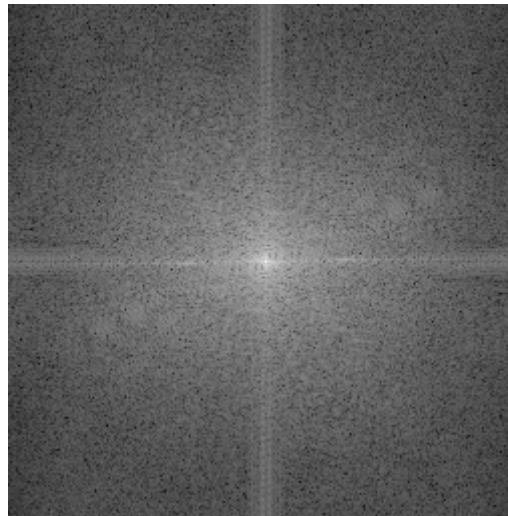
An Exam

Fourier transformed

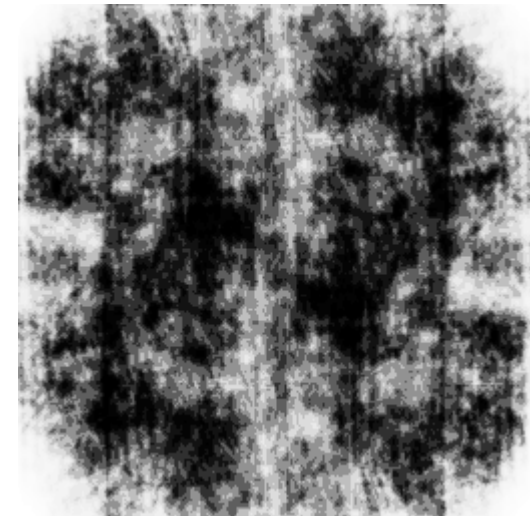
reconstructed



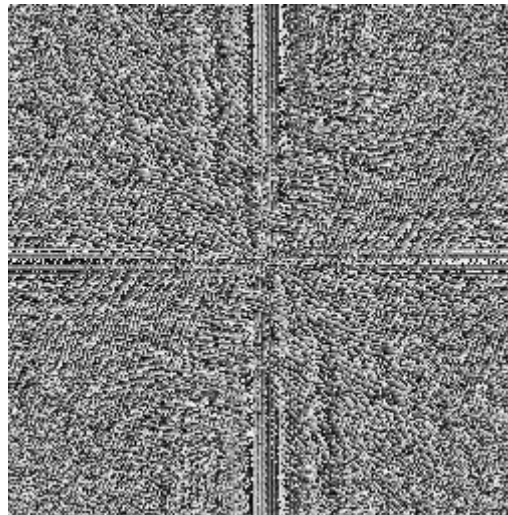
$f(x)$



Amplitude



ignoring Phase



Phase



using Phase+Amplitude

Spatial vs. Frequency Domain

- **Important basis functions**

- Box \leftrightarrow sinc

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

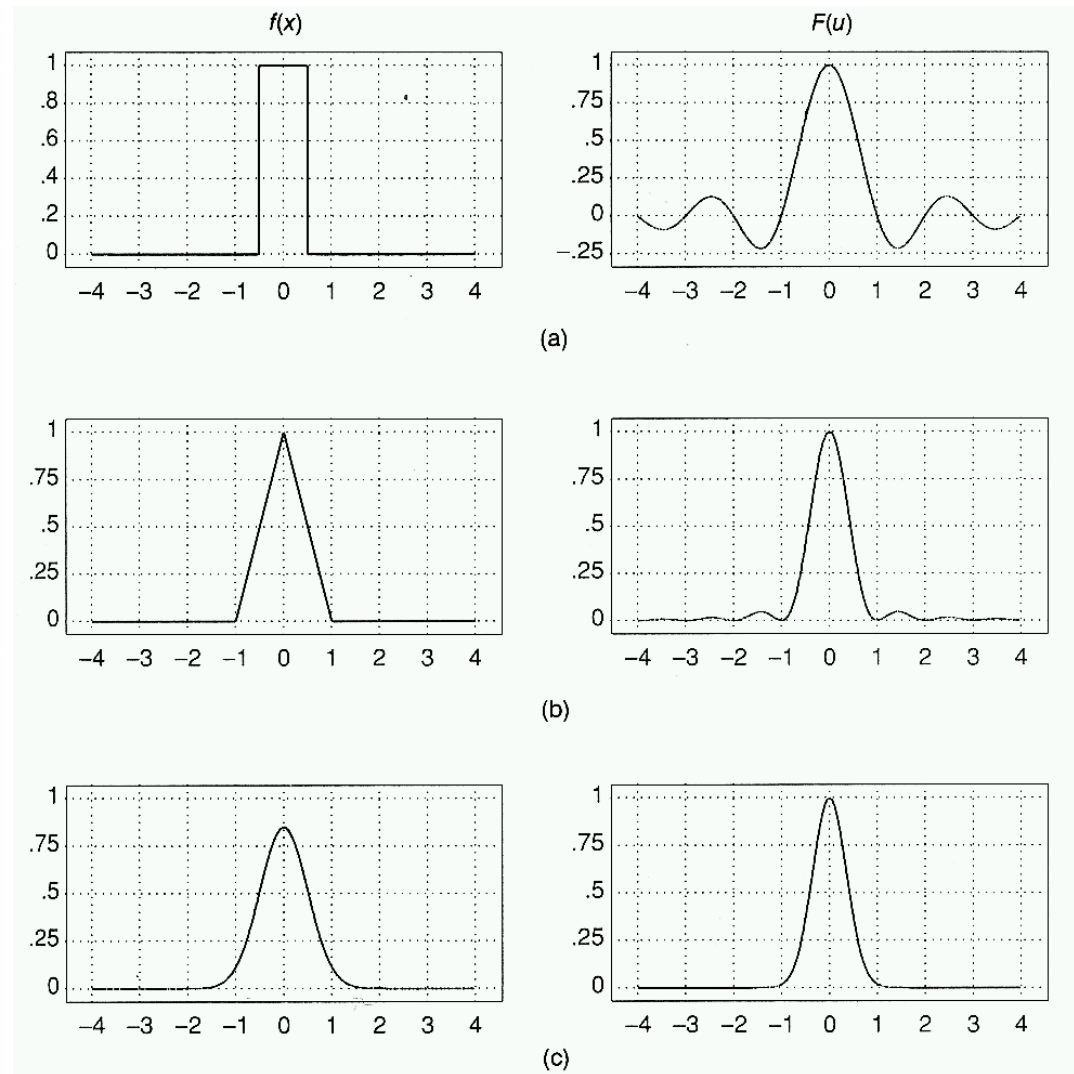
$$\text{sinc}(0) = 1$$

$$\int \text{sinc}(x) dx = 1$$

- Wide box \rightarrow small sinc
- Negative values
- Infinite support

- Triangle \leftrightarrow sinc^2

- Gauss \leftrightarrow Gauss

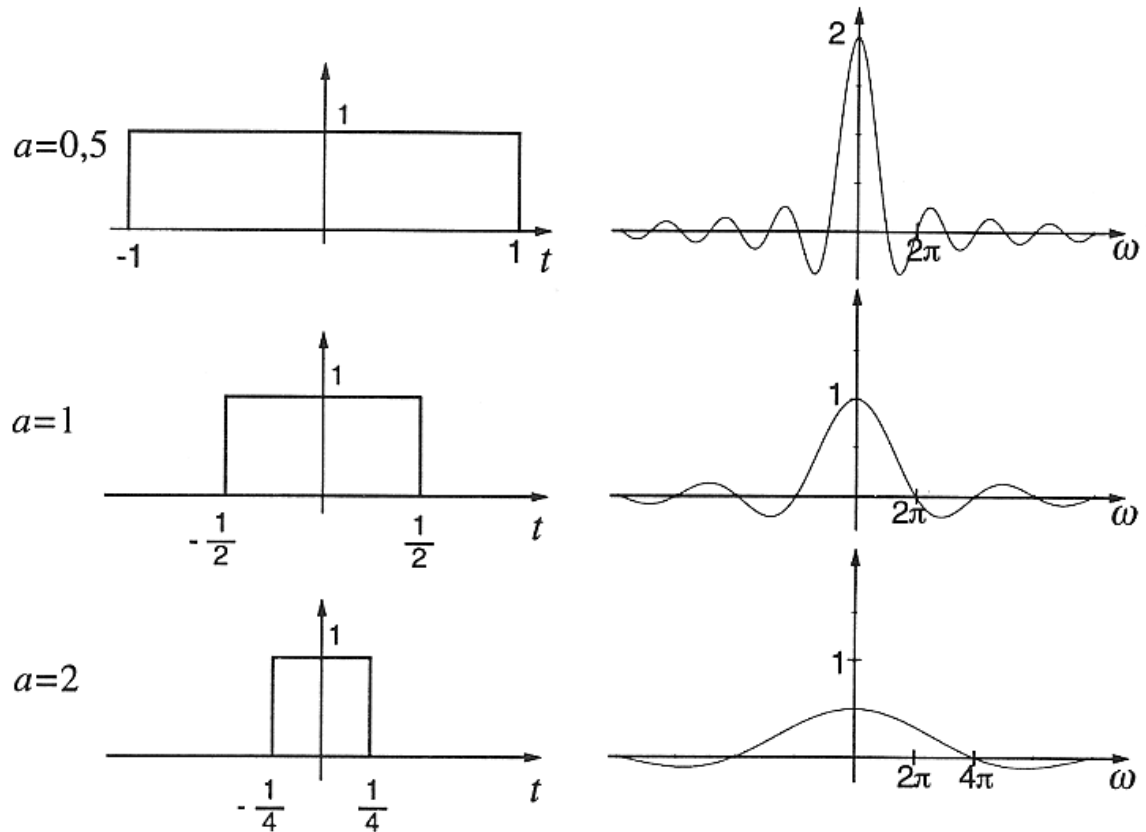


Spatial vs. Frequency Domain

- Transform behavior
- Example:
box function

$$\text{rect}(at) \quad \longleftrightarrow \quad \frac{1}{|a|} \text{si}\left(\frac{\omega}{2a}\right)$$

- Fourier transform:
sinc



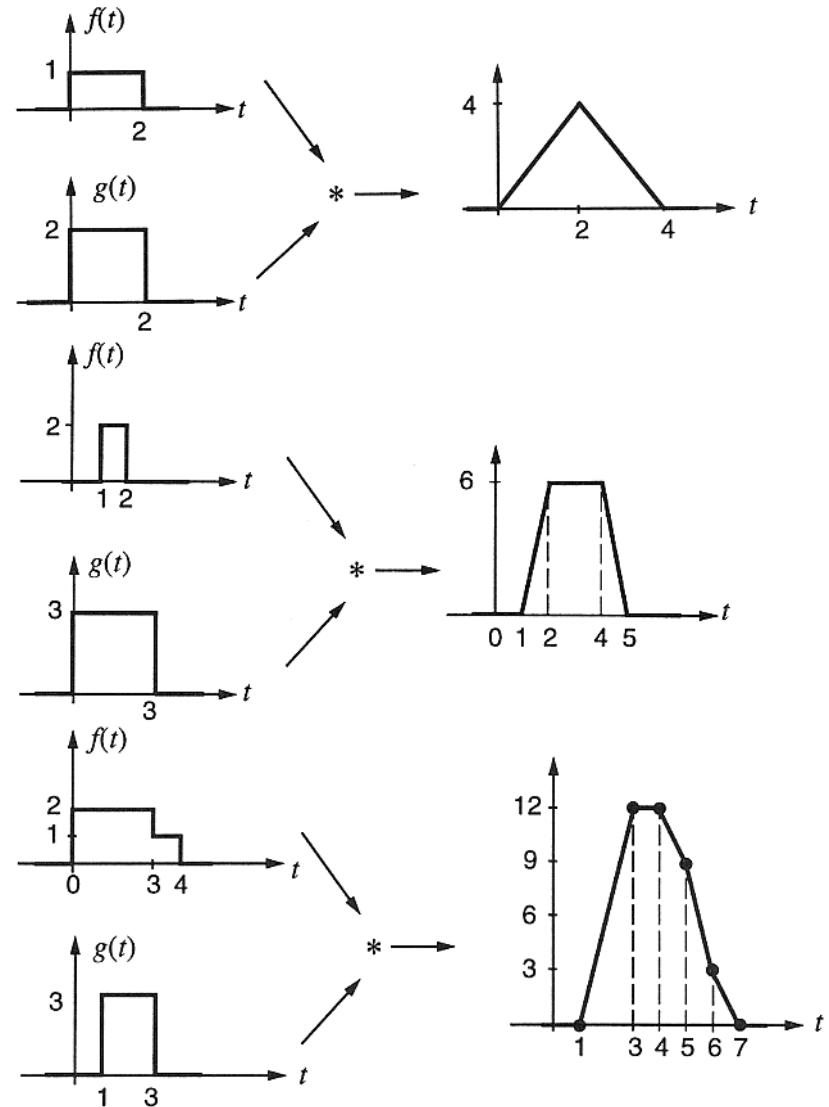
- Wide box:
narrow sinc

- Narrow box:
wide sinc

Convolution

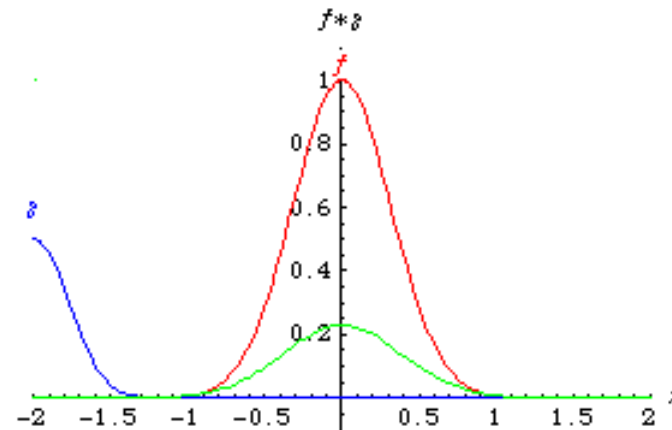
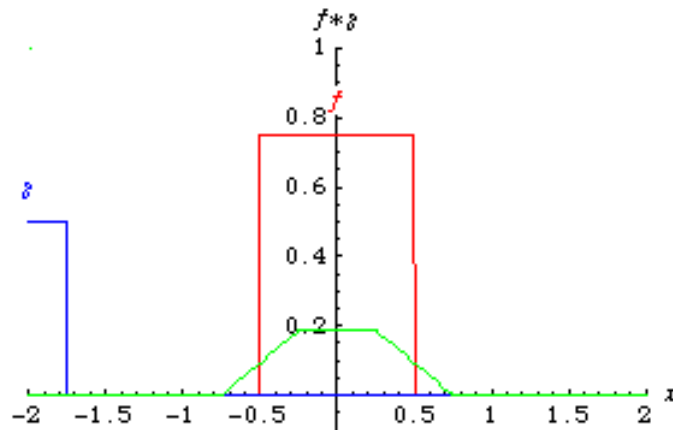
$$f(x) \otimes g(x) = \int_{-\infty}^{\infty} f(\tau)g(x - \tau)d\tau$$

- **Two functions f, g**
- **Shift one function against the other by x**
- **Multiply function values**
- **Integrate overlapping region**
- **Numerical convolution:**
 - For each x :
integrate over non-zero domain



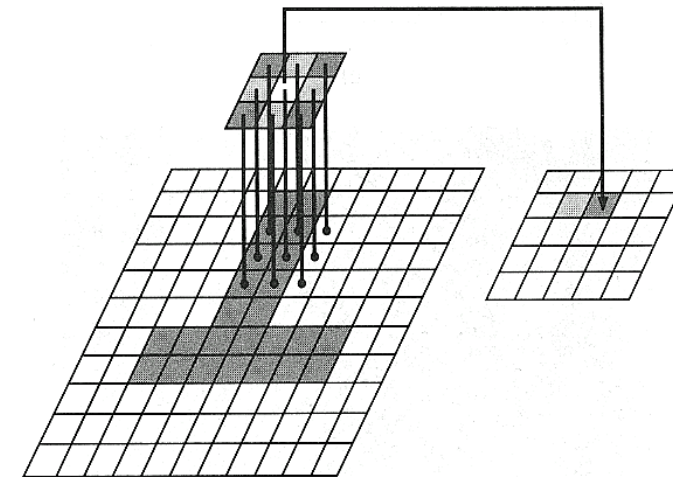
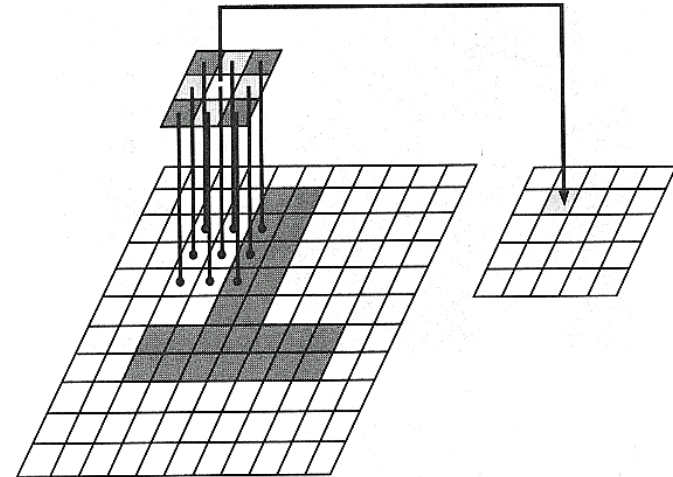
Convolution

- **Examples**
 - Box functions
 - Gauss functions



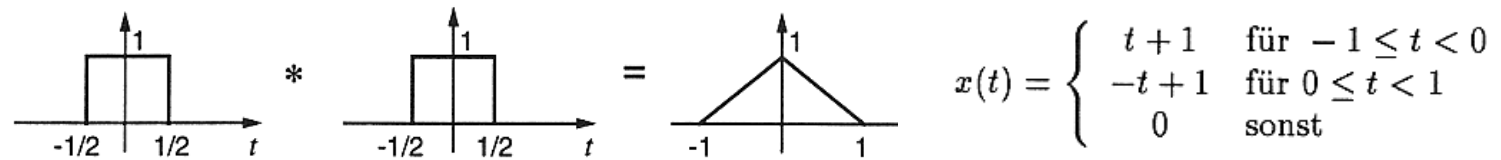
Convolution and Filtering

- **Technical Realization**
 - In image domain
 - Pixel mask with weights
 - OpenGL: Convolution extension
- **Problems (e.g. sinc)**
 - Large filter support
 - Large mask
 - A lot of computation
 - Negative weights
 - Negative light?



Convolution Theorem

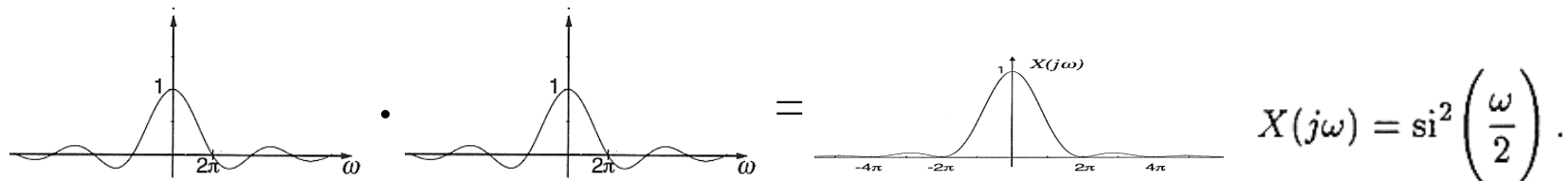
- Convolution in image domain:
multiplication in Fourier domain
- Convolution in Fourier domain:
multiplication in image domain
 - Multiplication much cheaper than convolution !



$$\text{rect}(t) * \text{rect}(t) = x(t)$$

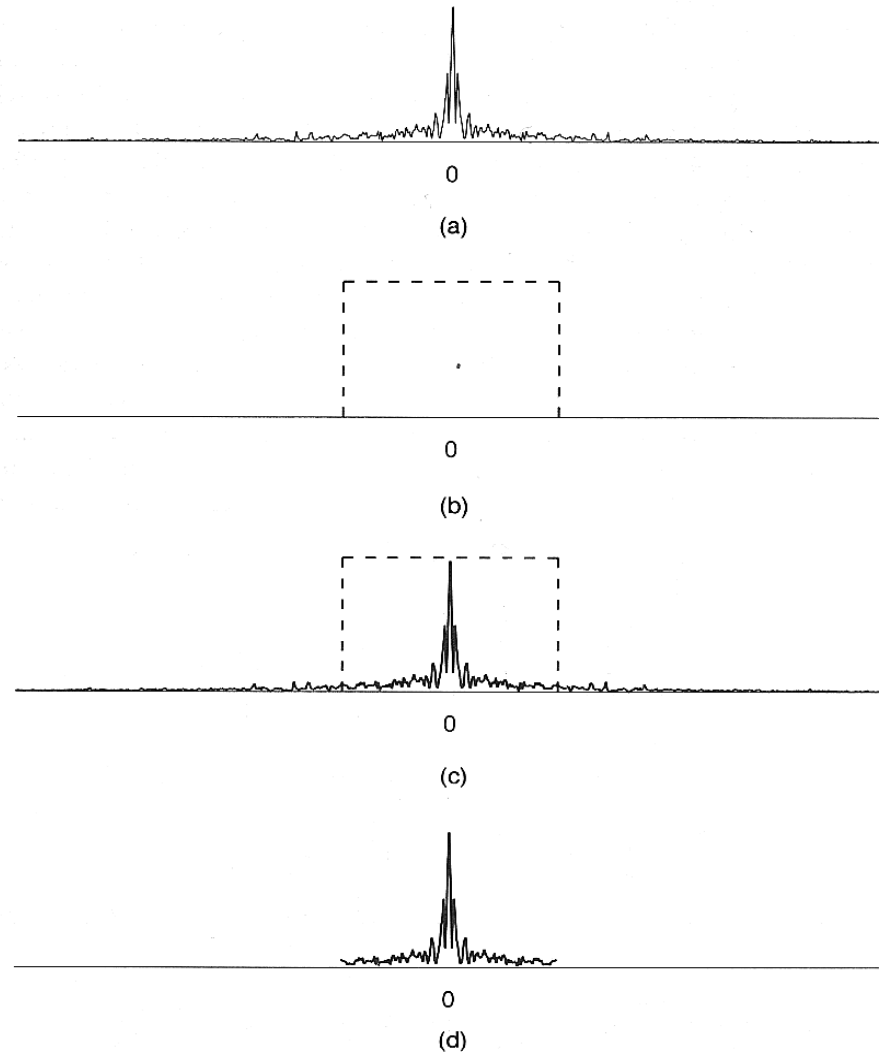


$$\text{si}\left(\frac{\omega}{2}\right) \cdot \text{si}\left(\frac{\omega}{2}\right) = X(j\omega)$$



Filtering

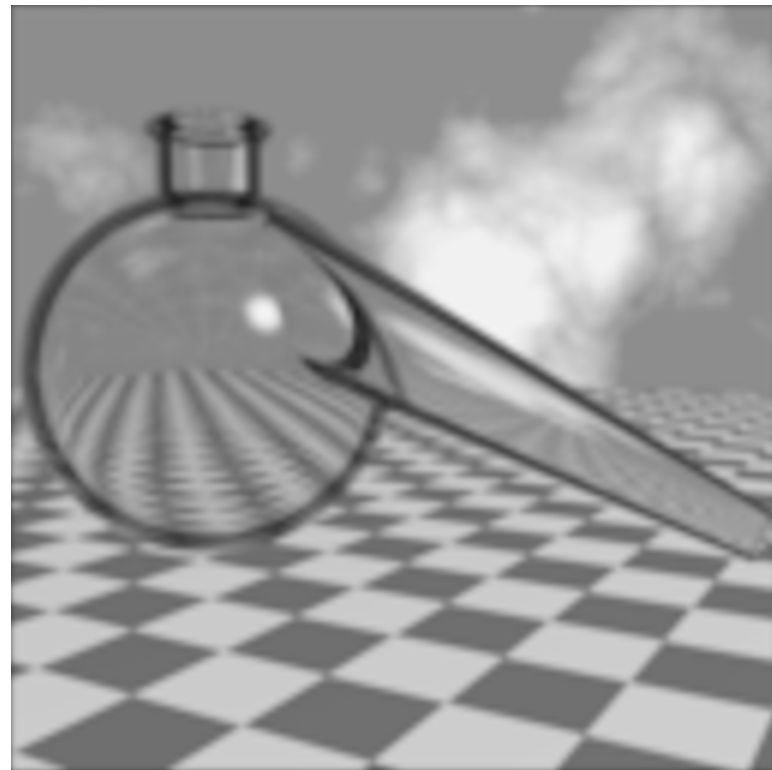
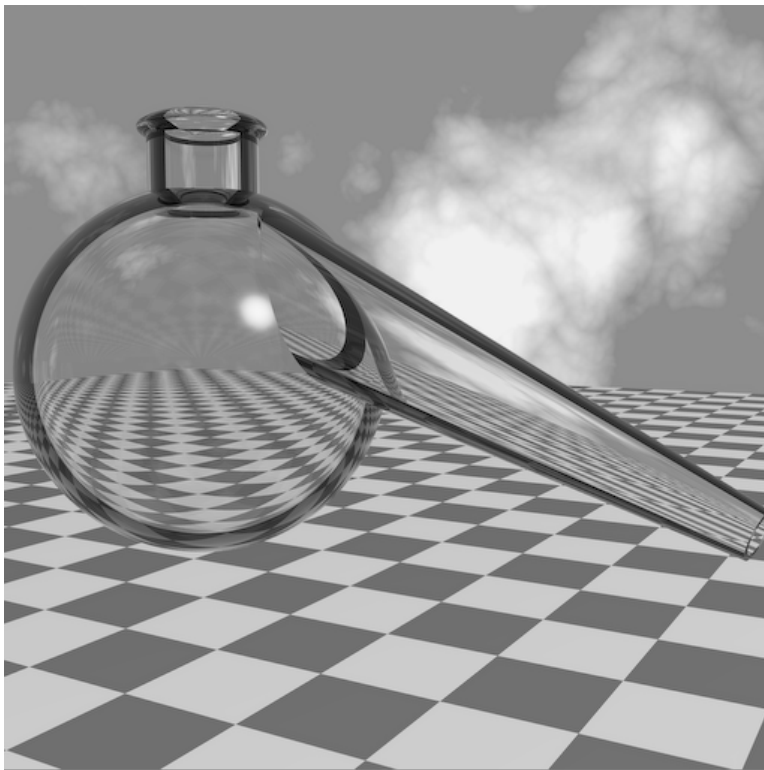
- **Low-pass filtering**
 - Convolution with sinc in spatial domain, or
 - Multiplication with box in frequency domain
- **High-pass filtering**
 - Only high frequencies
- **Band-pass filtering**
 - Only intermediate



Low-pass filtering in frequency domain: multiplication with box

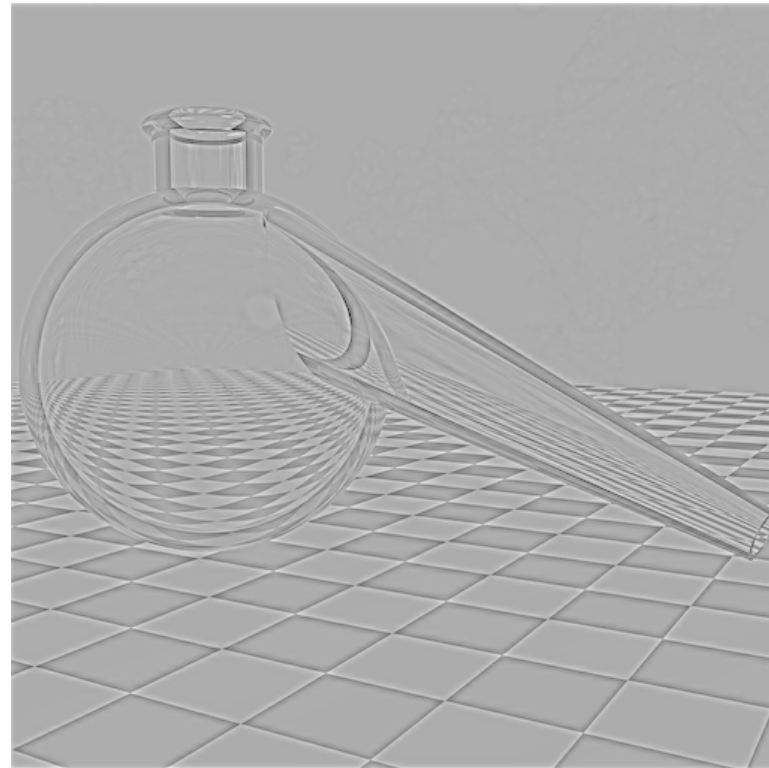
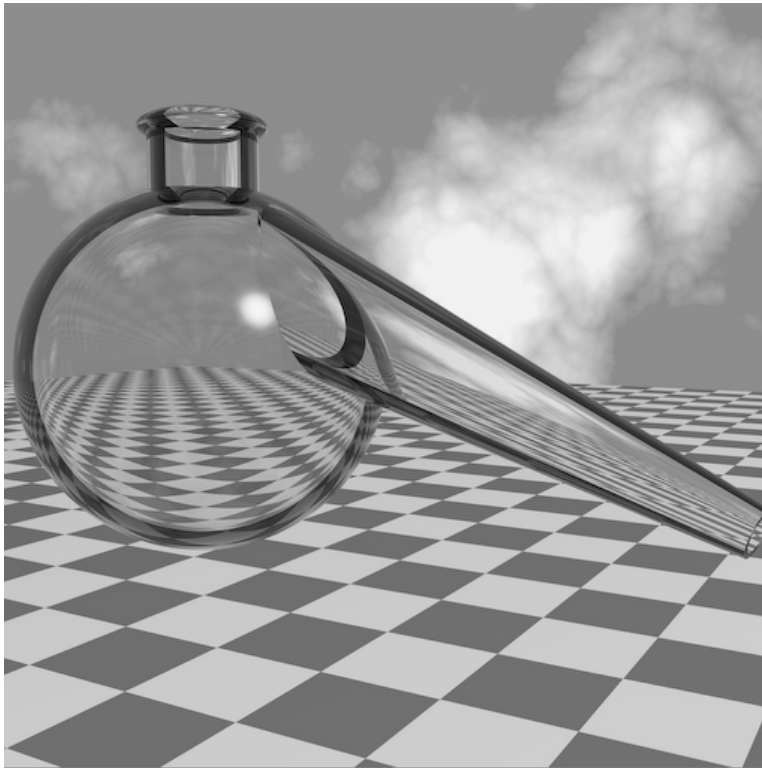
Low-Pass Filtering

- „Blurring“



High-Pass Filtering

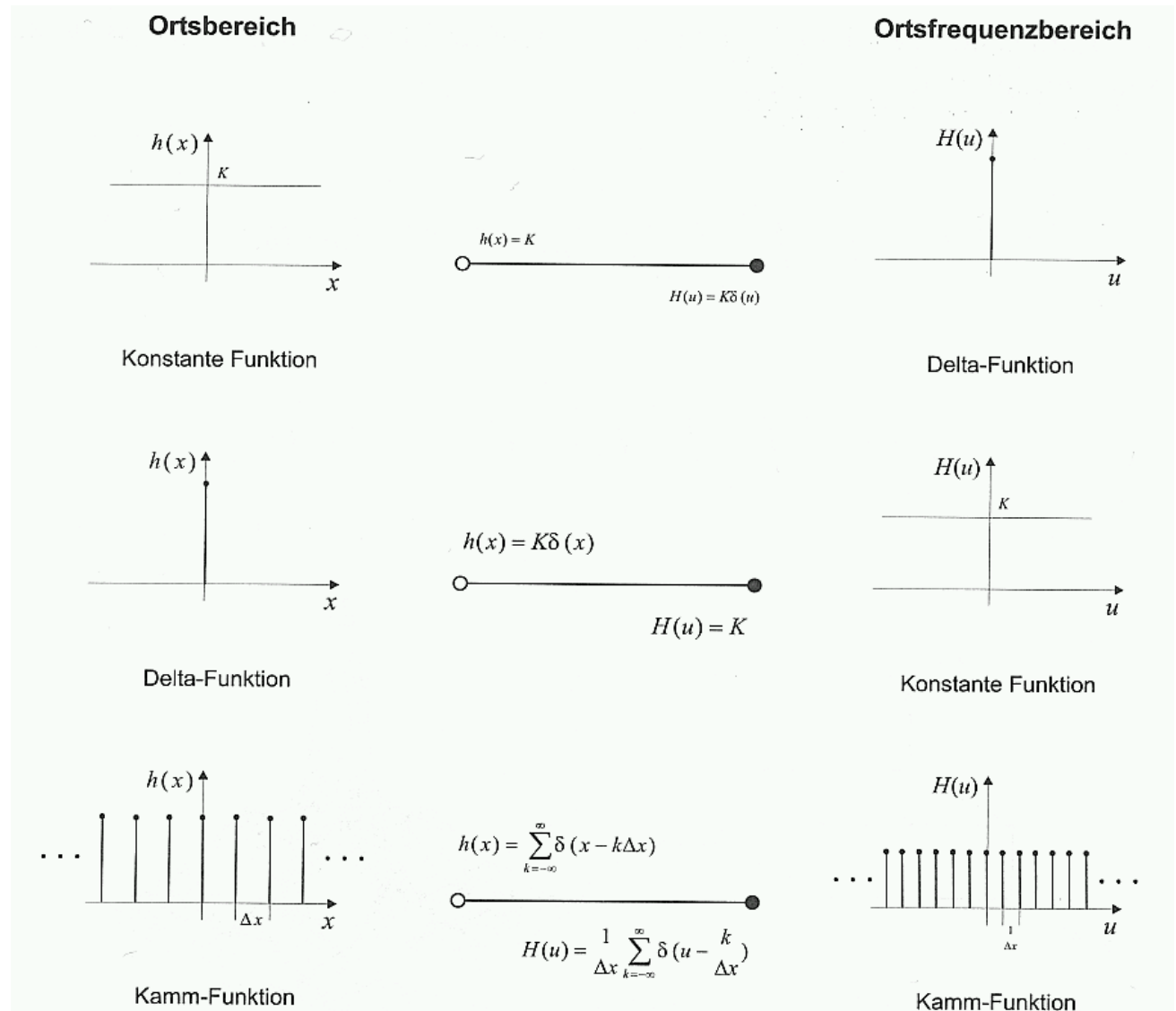
- **Enhances discontinuities in image**
 - Useful for edge detection



Sampling

Sampling

- **Constant & δ -Function**
– flash



- **Comb/Shah function**

Sampling

- **Constant & δ -Function**

- Duality

$$f(x) = K$$

$$F(\omega) = K\delta(\omega)$$

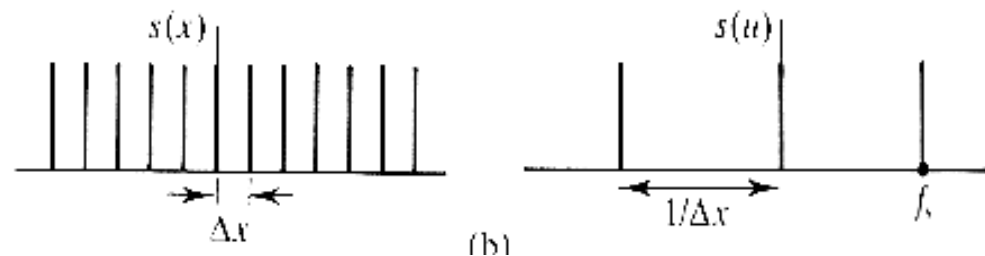
- And vice versa

- **Comb function**

- Duality: The dual of a comb function is again a comb function
 - Inverse wave length, amplitude scales with inverse wave length

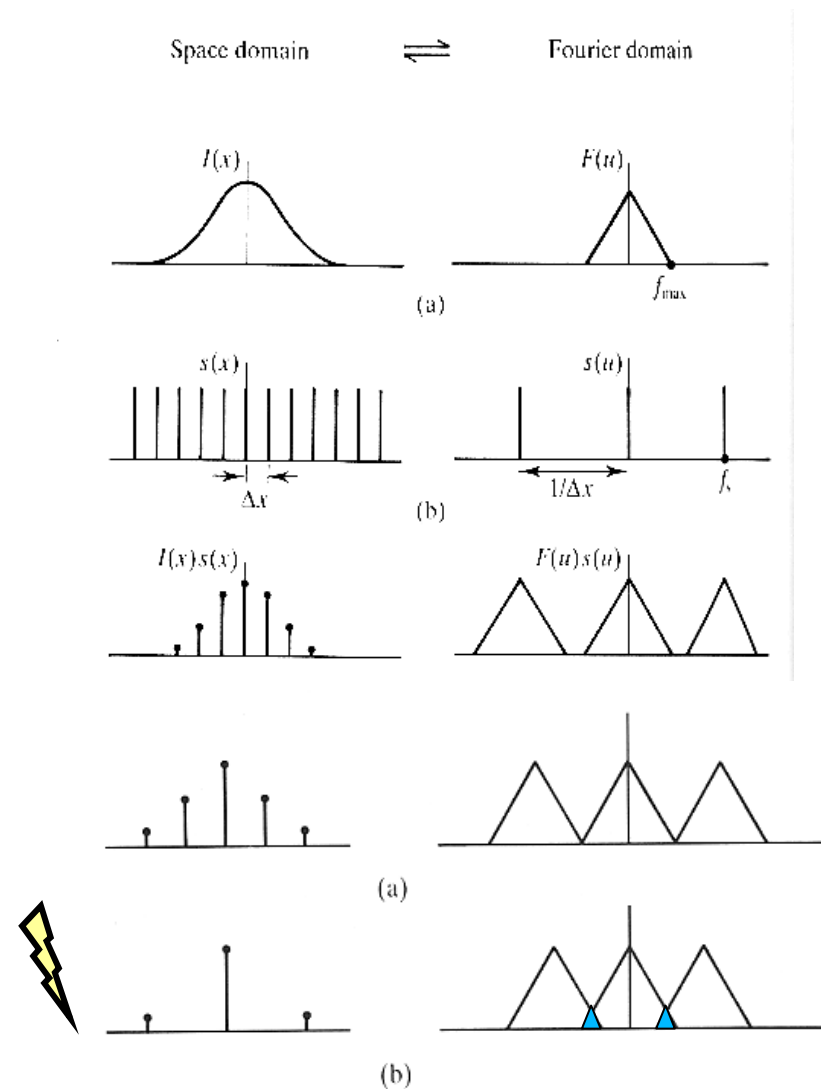
$$f(x) = \sum_{k=-\infty}^{\infty} \delta(x - k\Delta x)$$

$$F(\omega) = \frac{1}{\Delta x} \sum_{k=-\infty}^{\infty} \delta(\omega - k \frac{1}{\Delta x})$$



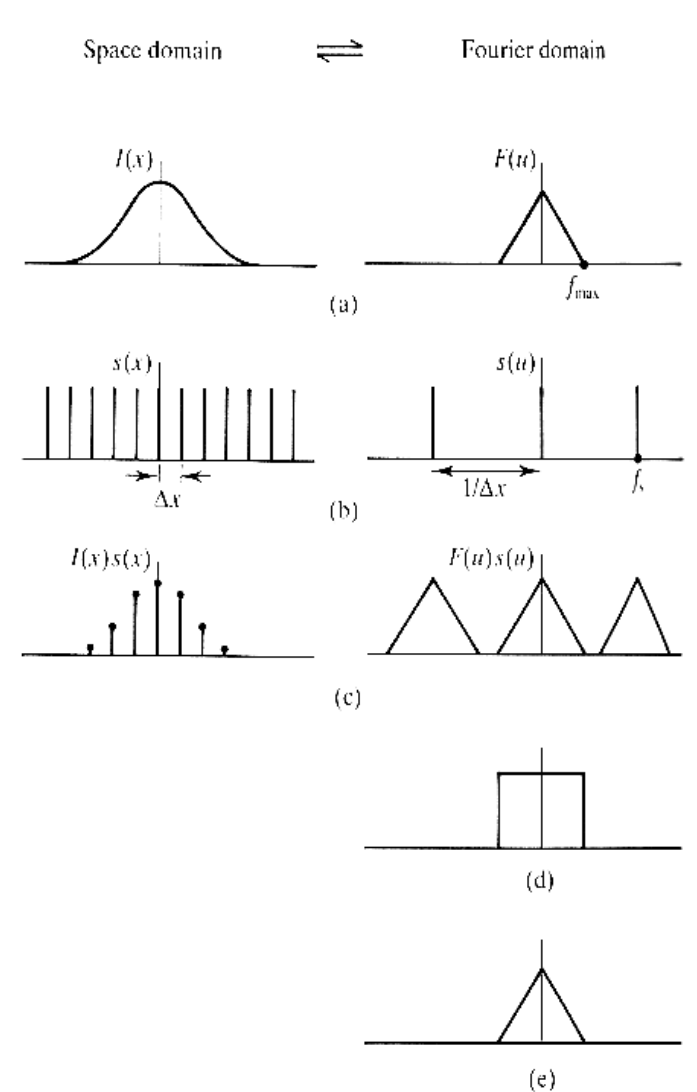
Sampling

- **Continuous function**
 - Band-limited Fourier transform
- **Sampled at discrete points**
 - Multiplication with Comb function in space domain
 - Corresponds to convolution in Fourier domain
 - Multiple copies of the original spectrum
- **Frequency bands overlap ?**
 - No: good
 - Yes: bad, *aliasing*



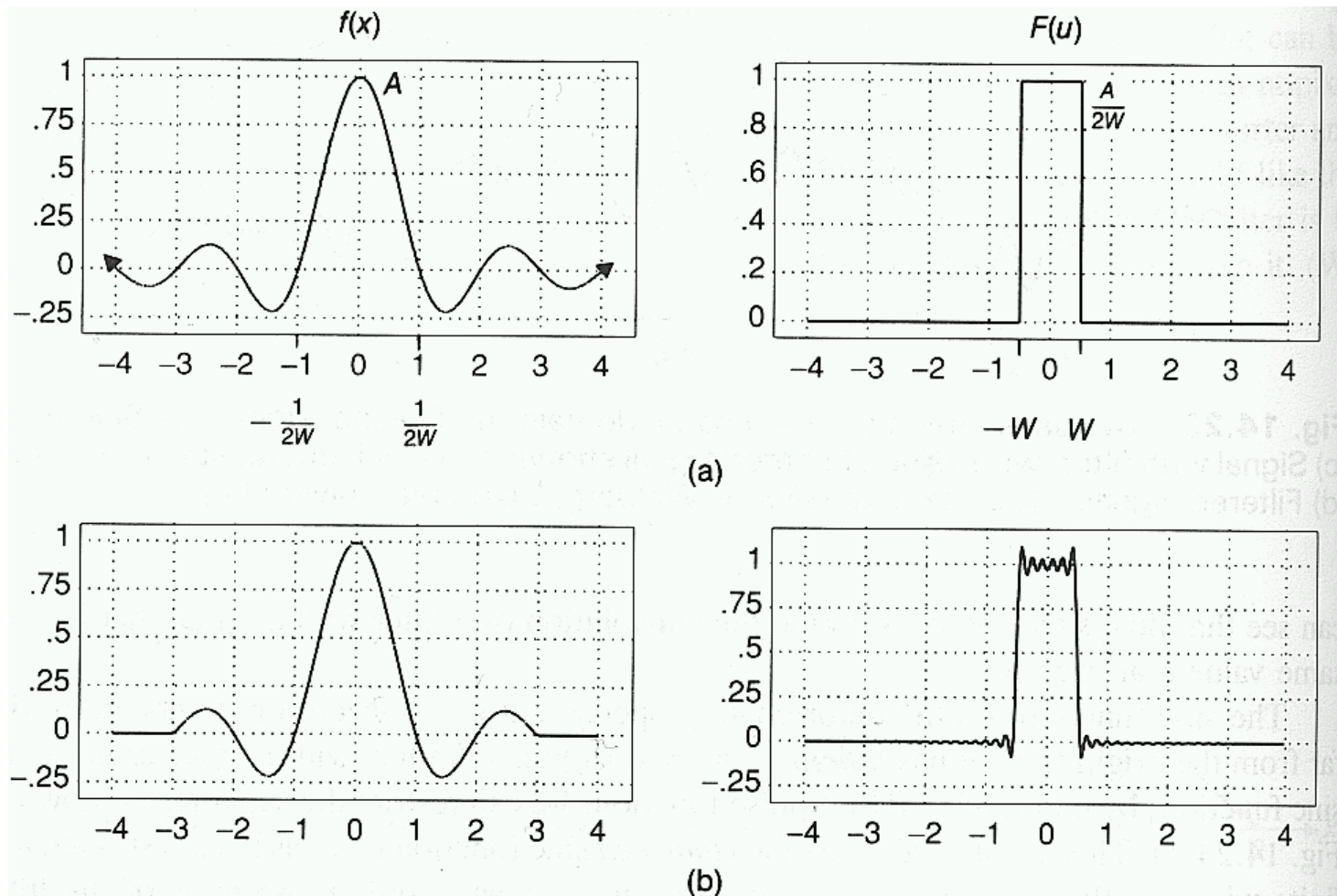
Reconstruction

- **Only original frequency band desired**
- **Filtering**
 - In Fourier domain: multiplication with windowing function around origin
 - In spatial domain: convolution with Fourier transform of windowing function
- **Optimal filtering function**
 - Box function in Fourier domain
 - Unlimited region of support
 - Corresponds to sinc in space domain
 - Spatial domain only allows approximations (due to finite support)



Reconstruction Filter

- Cutting off the support is *not* a good solution



Sampling and Reconstruction

Original function and its band-limited frequency spectrum

Signal sampling:

Mult./conv. with comb

Comb dense enough (sampling $\geq 2 \cdot \text{bandlimit}$)

Frequency spectrum is replicated

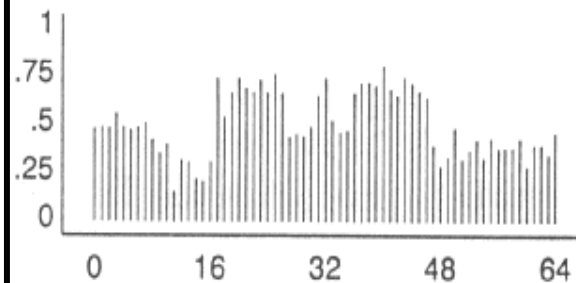
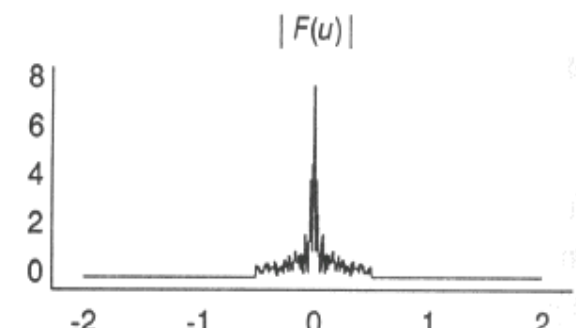
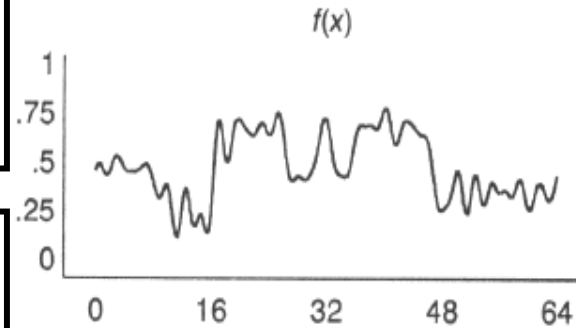
Bands do not overlap

Correct filtering

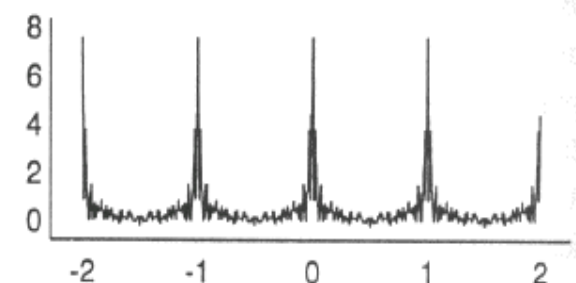
Fourier: Box (mult.)

Image: sinc (conv.)

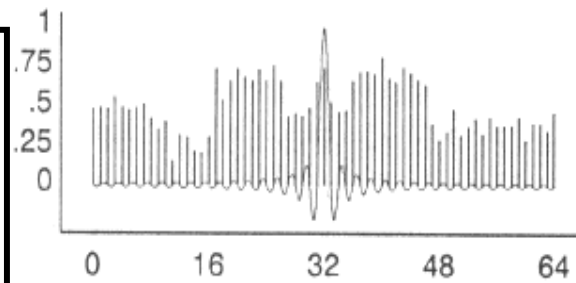
Only one copy



(a)



(b)

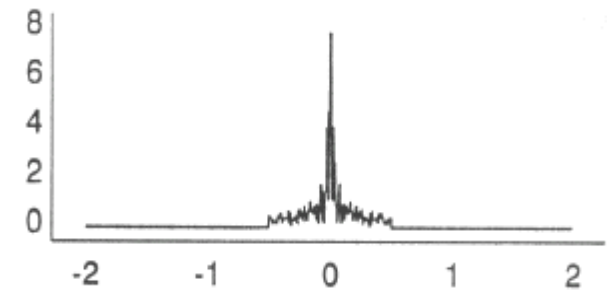
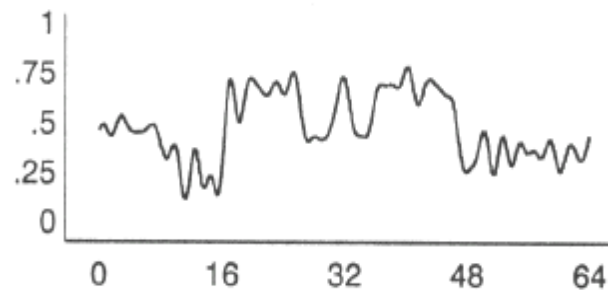


(c)

Sampling and Reconstruction

Reconstruction
with ideal sinc

Identical signal



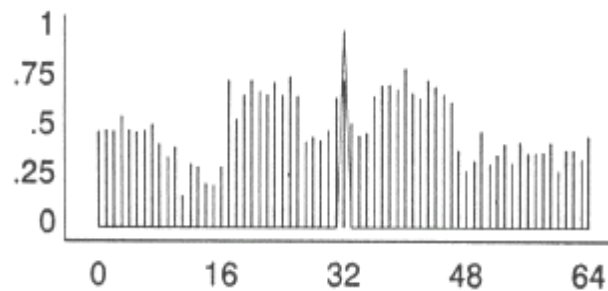
Approximate filtering

Space: tri (conv.)

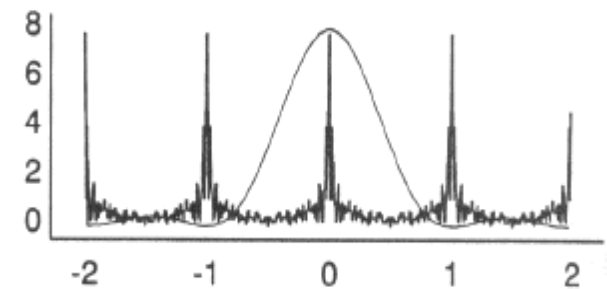
Fourier: sinc² (mult.)

High frequencies are
not ignored

→ Aliasing

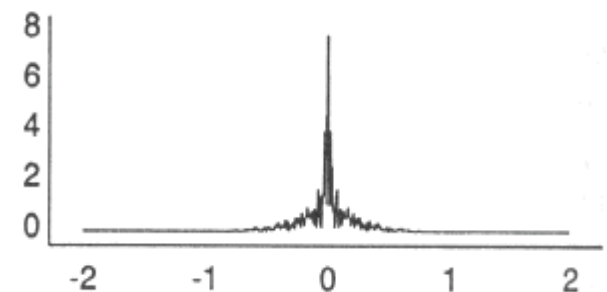
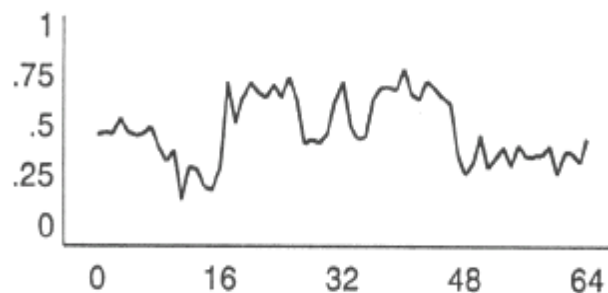


(d)



(e)

Reconstruction
with tri function
(= piecewise linear
interpolation)



Sampling with Too Low Frequency

Original function

Sampling below Nyquist:

Comb spaced to far
(sampling $< 2 \cdot \text{bandlimit}$)

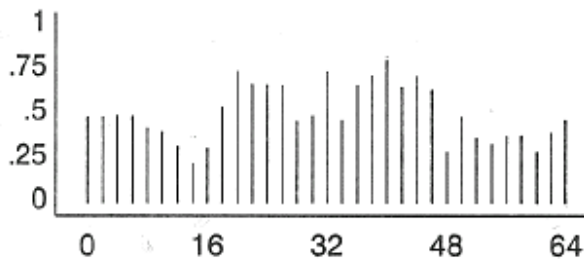
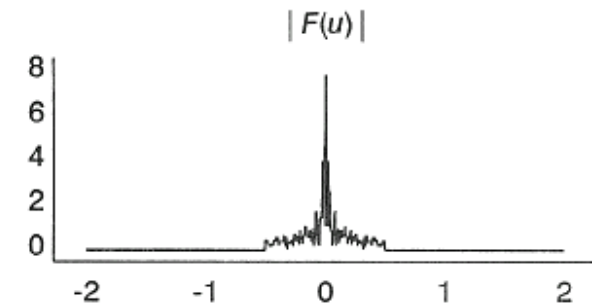
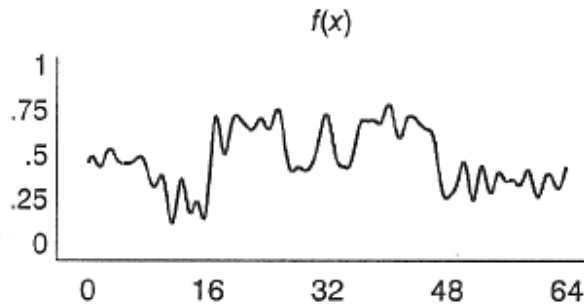
Frequency bands
overlap

Correct filtering

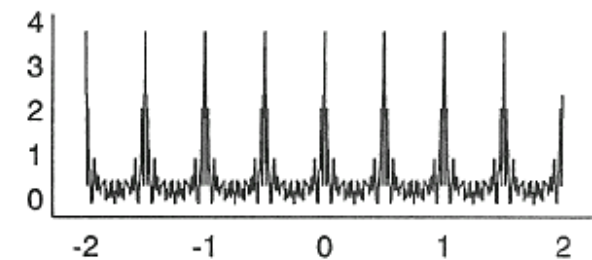
Image: sinc (conv.)

Fourier: box (mult.)

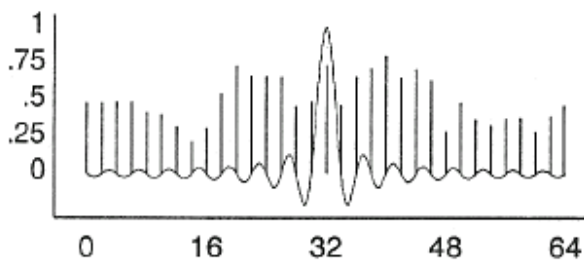
Band overlap in
frequency domain
cannot be corrected -
aliasing



(a)



(b)

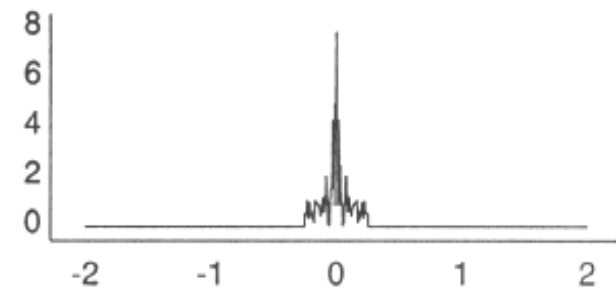
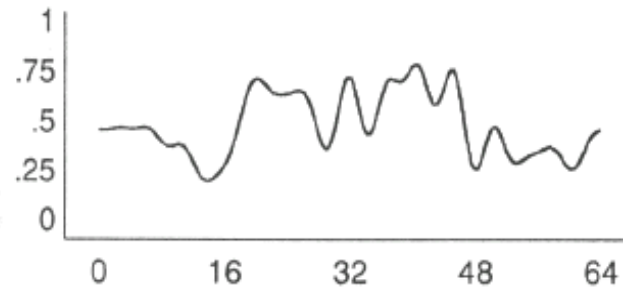


(c)

Sampling with Too Low Frequency

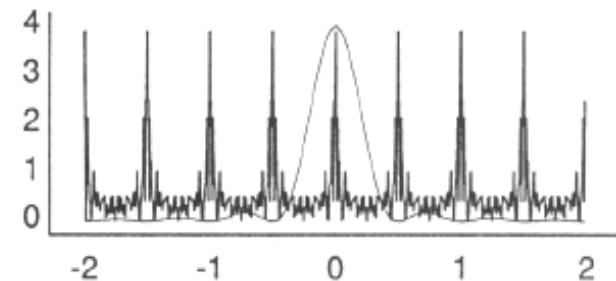
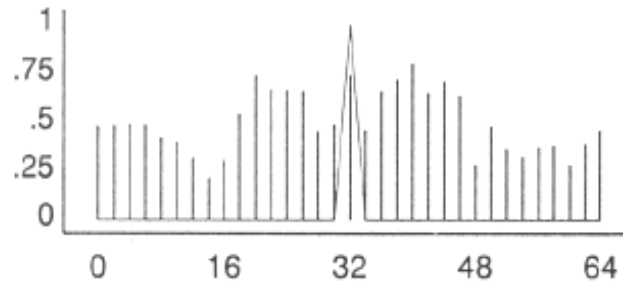
Reconstruction
with ideal sinc

Reconstruction
fails (frequency
components
wrong due to
aliasing !)



(d)

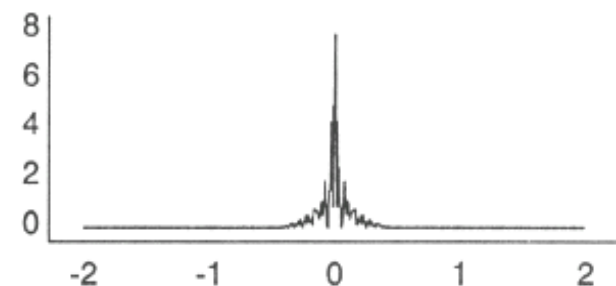
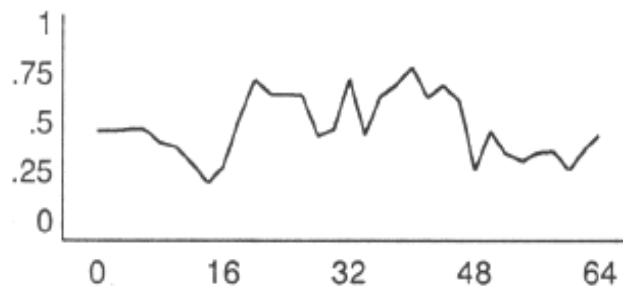
Filtering with sinc^2
function



(e)

Reconstruction
with tri function
(= piecewise linear
interpolation)

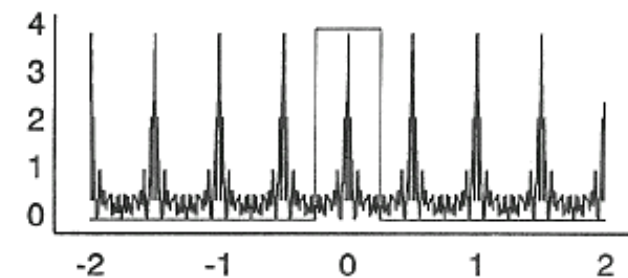
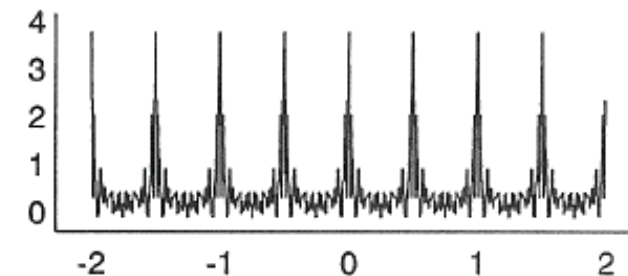
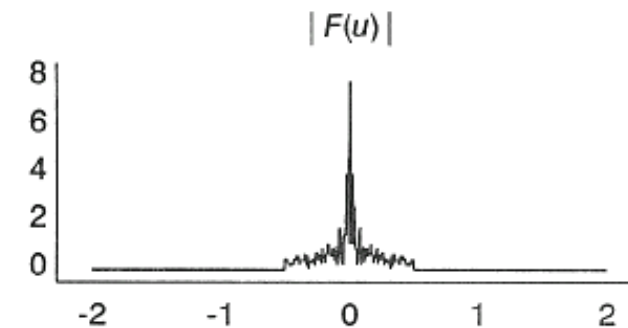
Even worse
reconstruction



Aliasing

- **Overlap between replicated copies in frequency spectrum**

High frequency components from the replicated copies are treated like low frequencies during the reconstruction process



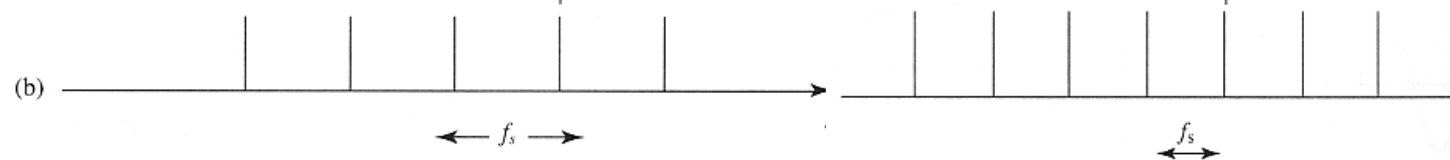
Aliasing

- In Fourier space

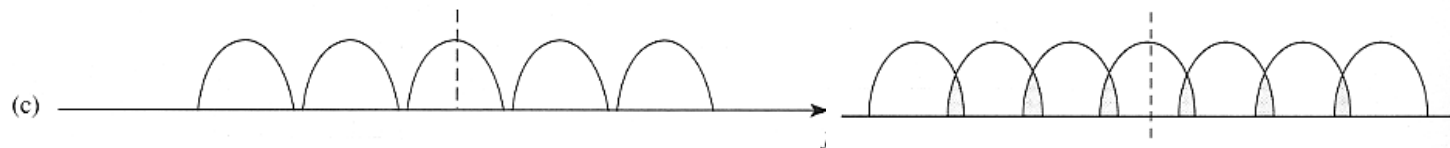
- Original spectrum



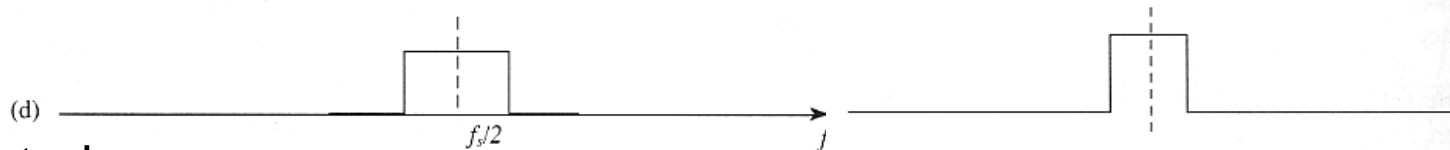
- Sampling comb



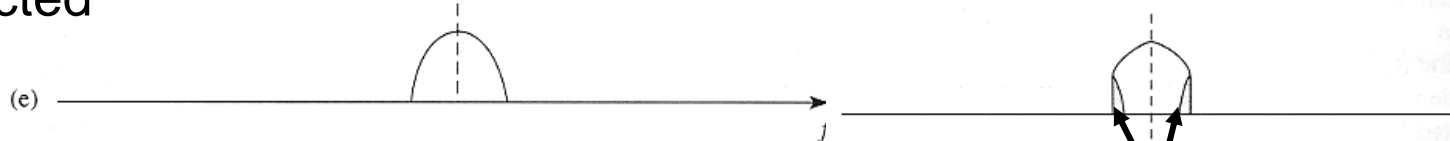
- Resulting spectrum



- Reconstruction Filter

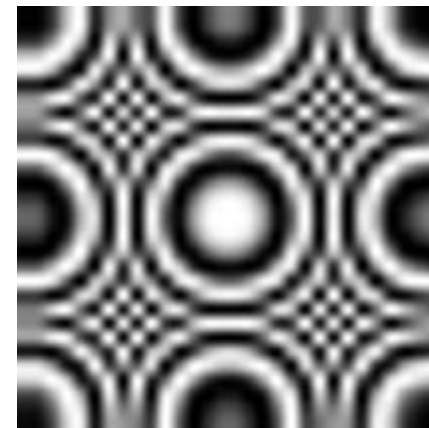
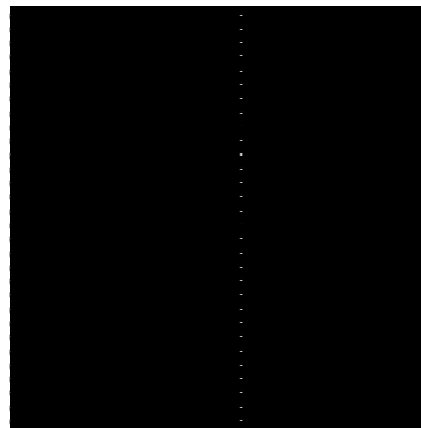
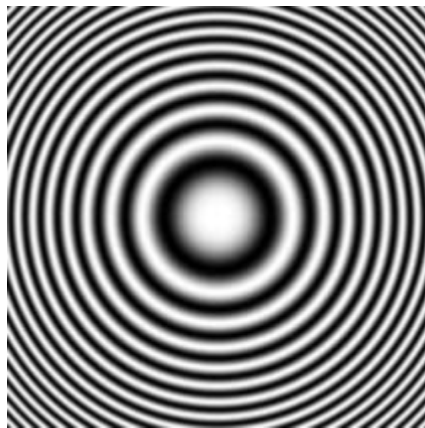


- Reconstructed spectrum



Aliasing

Aliasing 2D



[wikipedia]

original image sampled at these location yields this reconstruction.

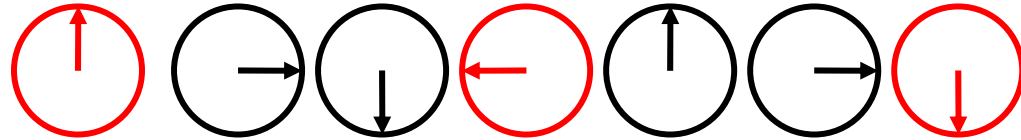
Sampling Artifacts

- **Spatial aliasing:**
 - Stair cases, Moiré patterns, etc.
- **Solutions:**
 - Increasing the sampling rate
 - Ok, but infinite frequencies at sharp edges
 - Post-filtering (after reconstruction)
 - Does not work - only leads to blurred stair cases
 - Pre-filtering (Blurring) of sharp geometry features
 - Slowly make geometry transparent at the edges
 - Correct solution in principal
 - Analytic low-pass filtering hard to implement
 - Super-sampling

Sampling Artifacts

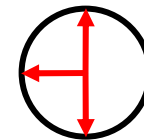
- **Temporal Aliasing**

- Cart wheels, ...



- **Solutions**

- Increasing the frame rate
 - OK
- Pre-filtering (Motion Blur)
 - Yes, possible for simple geometry (e.g., Cartoons)
 - Problems with texture, etc.
- Post-filtering (Averaging several frames)
 - Does not work – only multiple detail



- **Important**

- Distinction between **aliasing errors** and **reconstruction errors**

Aliasing

- **It all comes from sampling at discrete points**
 - Multiplied with comb function, no smoothly weighted filters
 - Comb function: repeats frequency spectrum
- **Or, from using non band limited primitives**
 - Hard edges \Rightarrow infinitely high frequencies
- **In reality, integration over finite region necessary**
 - E.g., finite CCD pixel size
- **Computer: Analytic integration often not possible**
 - No analytic description of radiance or visible geometry available
- **Only way: numerical integration**
 - Estimate integral by taking multiple point samples, average
 - Leads to aliasing
 - Computationally expensive
 - Approximate

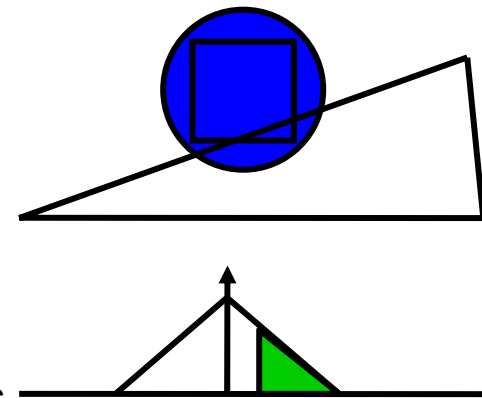
Sources of High Frequencies

- **Geometry**
 - Edges, vertices, sharp boundaries
 - Silhouettes (view dependent)
 - ...
 - **Texture**
 - E.g., checkerboard pattern, other discontinuities, ...
 - **Illumination**
 - Shadows, lighting effects, projections, ...
- ➔ **Analytic filtering almost impossible**
- Even with the most simple filters



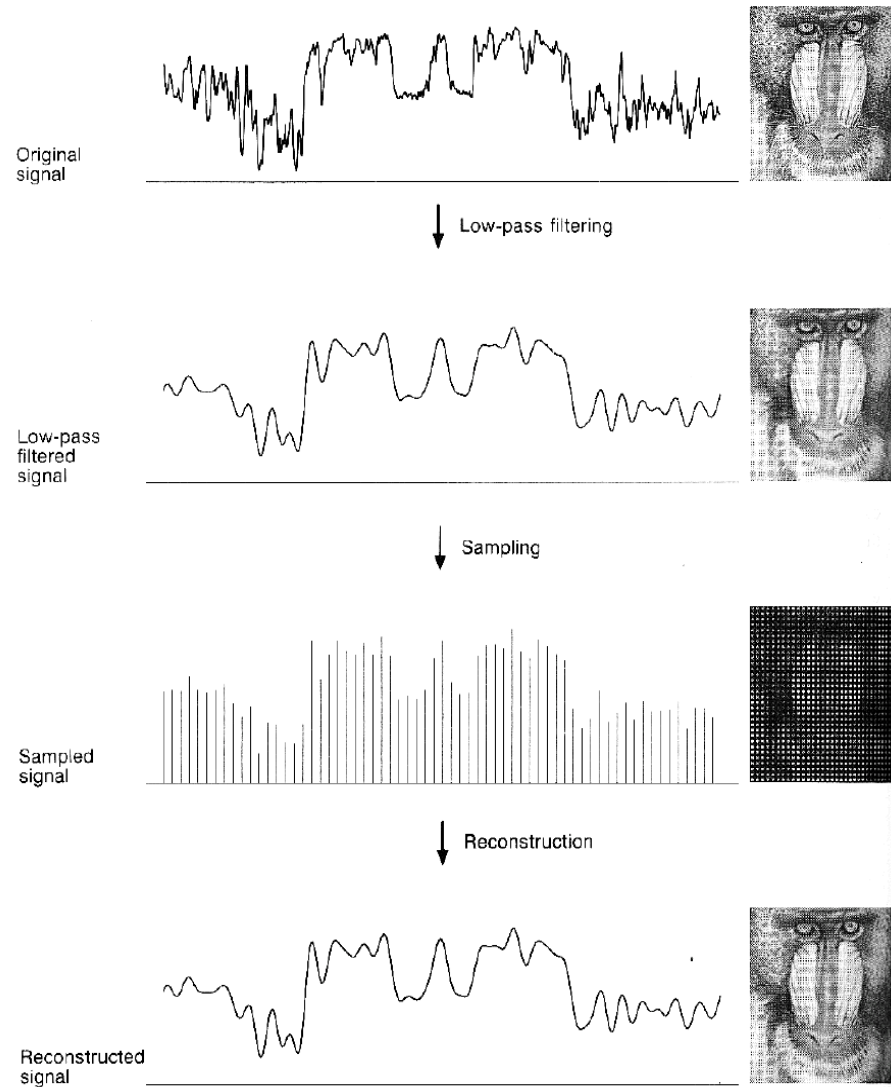
Comparison

- **Analytic low-pass filtering**
 - Ideally eliminates aliasing completely
 - Hard to implement
 - Weighted or unweighted area sampling
 - Compute distance from pixel to a line
 - Filter values can be stored in look-up tables
 - Possibly taking into account slope
 - Distance correction
 - Non rotationally symmetric filters
 - Does not work at corners
- **Over-/Super-sampling**
 - Very easy to implement
 - Does not eliminate aliasing completely
 - Sharp edges contain *infinitely* high frequencies



Antialiasing by Pre-Filtering

- **Filtering before sampling**
 - Band-limiting signal
 - Analog/analytic or
 - Reduce Nyquist frequency for chosen sampling-rate
- **Ideal reconstruction**
 - Convolution with sinc
- **Practical reconstruction**
 - Convolution with
 - Box filter, Bartlett (Tent)
 - Reconstruction error



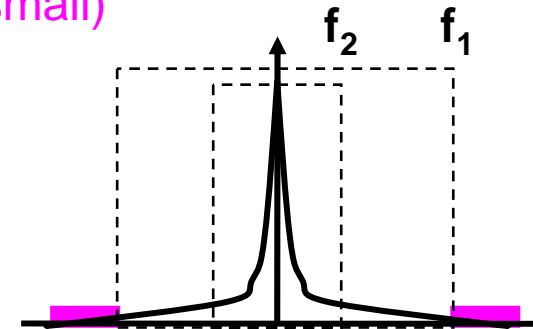
Re-Sampling Pipeline

- **Assumption**

- Energy in high frequencies decreases quickly
- Reduced aliasing by intermediate sampling with higher frequencies

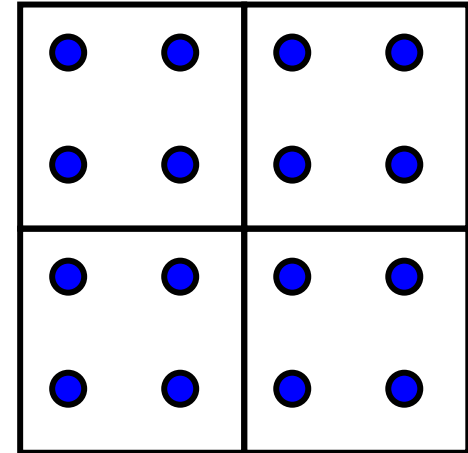
- **Algorithm**

- Super-sampling
 - Sample continuous signal with high frequency f_1
 - Aliasing with energy beyond f_1 (assumed to be small)
- Reconstruction of signal
 - Filtering with $g_1(x)$: e.g. convolution with sinc_{f_1}
 - Exact representation with sampled values !!
- Analytic low-pass filtering of signal
 - Filtering with filter $g_2(x)$ with $f_2 \ll f_1$
 - Signal is now band limited w.r.t. f_2
- Re-sampling with a sampling frequency that is compatible with f_2
 - No additional aliasing
- Filters $g_1(x)$ and $g_2(x)$ can be combined



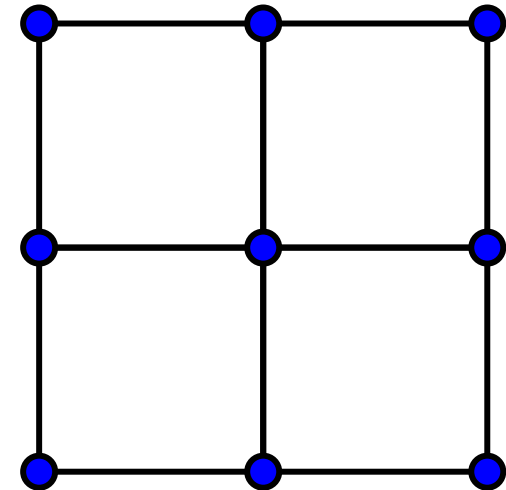
Super-Sampling in Practice

- **Regular super-sampling**
 - Averaging of N samples per pixel on a grid
 - N:
 - 4 quite good
 - 16 almost always sufficient
 - Samples
 - Rays, z-buffer, reflection, motion, ...
 - Filter Weights
 - Box filter
 - Others: B-spline, Pyramid (Bartlett), Hexagonal, ...
 - Regular super-sampling
 - Nyquist frequency for aliasing only shifted
 - Irregular sampling patterns



Super-Sampling Caveats

- **Popular mistake**
 - Sampling at the corners of every pixel
 - Pixel color by averaging
 - Free super-sampling ???
- **Problem**
 - Wrong reconstruction filter !!!
 - Same sampling frequency, but post-filtering with a hat function
 - Blurring: Loss of information
- **Post-Reconstruction Blur**



1x1 Sampling, 3x3 Blur



1x1 Sampling, 7x7 Blur

→ „Super-sampling“ does not come for free

Adaptive Super-Sampling

- **Adaptive super-sampling**
 - Idea: locally adapt sampling density
 - Slowly varying signal: low sampling rate
 - Strong changes: high sampling rate
 - Decide sampling density locally
 - Decision criterion needed
 - Differences of pixel values
 - Contrast (relative difference)
 - $|A-B| / |A|+|B|$

Adaptive Super-Sampling

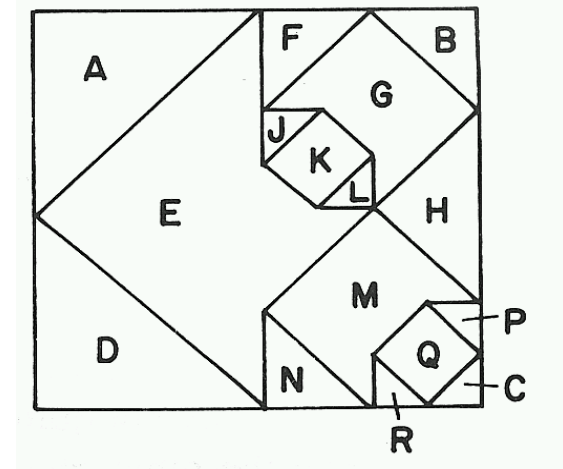
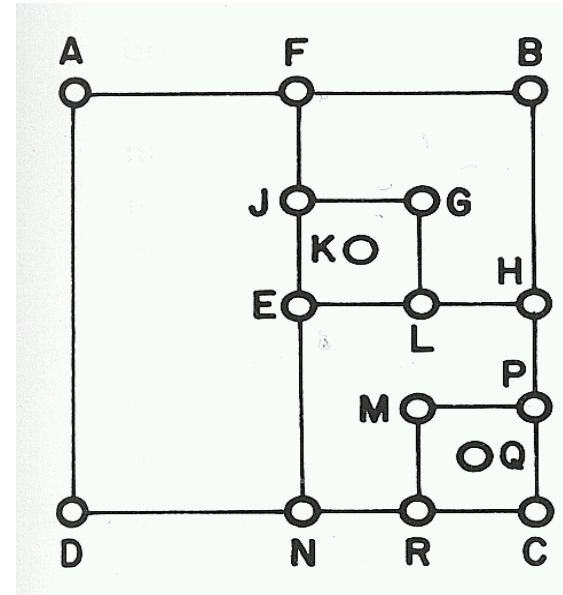
- **Algorithm**

- Sampling at corners and mid points
- Recursive subdivision of each quadrant
- Decision criterion
 - Differences, contrast, object-IDs, ray trees, ...
- Filtering with weighted averaging
 - $\frac{1}{4}$ from each quadrant
 - Quadrant: $\frac{1}{2}$ (midpoint + corner)
 - Recursion

$$\frac{1}{16} \left(\frac{A+E}{2} + \frac{D+E}{2} + \frac{1}{4} \left[\frac{F+G}{2} + \frac{B+G}{2} + \frac{H+G}{2} + \frac{1}{4} \left\{ \frac{J+K}{2} + \frac{G+K}{2} + \frac{L+K}{2} + \frac{E+K}{2} \right\} \right] \right. \\ \left. + \frac{1}{4} \left[\frac{E+M}{2} + \frac{H+M}{2} + \frac{N+M}{2} + \frac{1}{4} \left\{ \frac{M+Q}{2} + \frac{P+Q}{2} + \frac{C+Q}{2} + \frac{R+Q}{2} \right\} \right] \right)$$

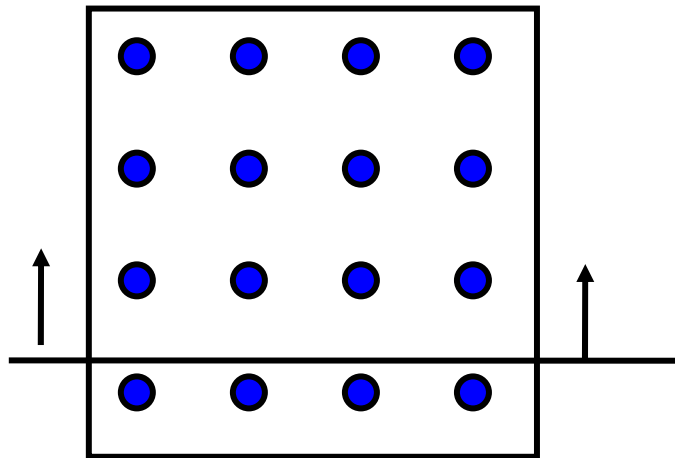
- **Extension**

- Jittering of sample points



Super-Sampling in Practice

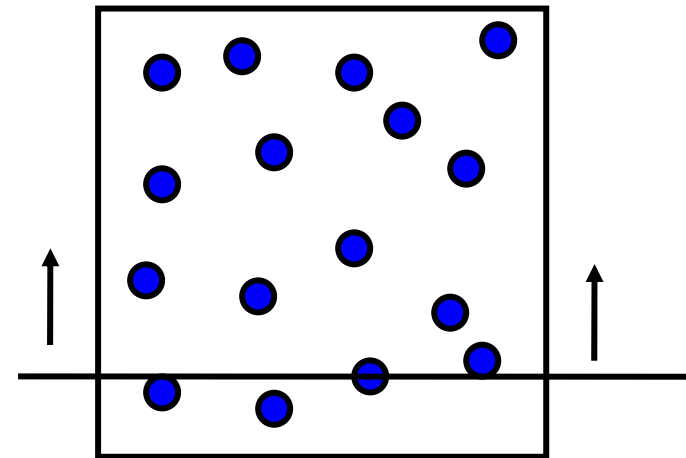
- **Problems with regular super-sampling**
 - Expensive: 4-fold to 16-fold effort
 - Non-adaptive: Same effort everywhere
 - Too regular: Apparent reduction of number of levels
- **Introduce irregular sampling pattern**



0 → 4/16 → 8/16 → 12/16 → 16/16

→ **Stochastic super-sampling**

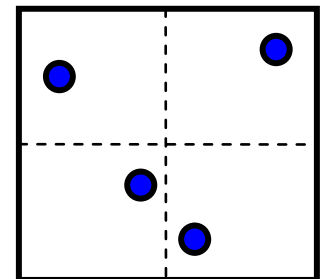
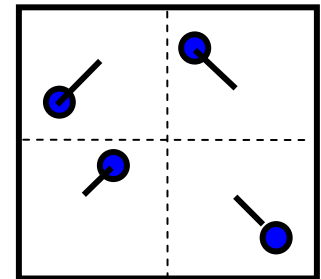
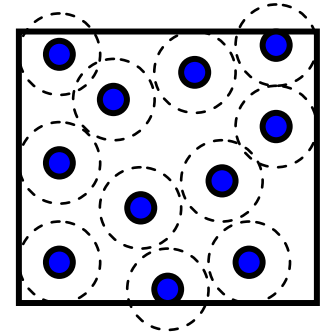
- Or analytic computation of pixel coverage and pixel mask



Better, but noisy

Stochastic Sampling

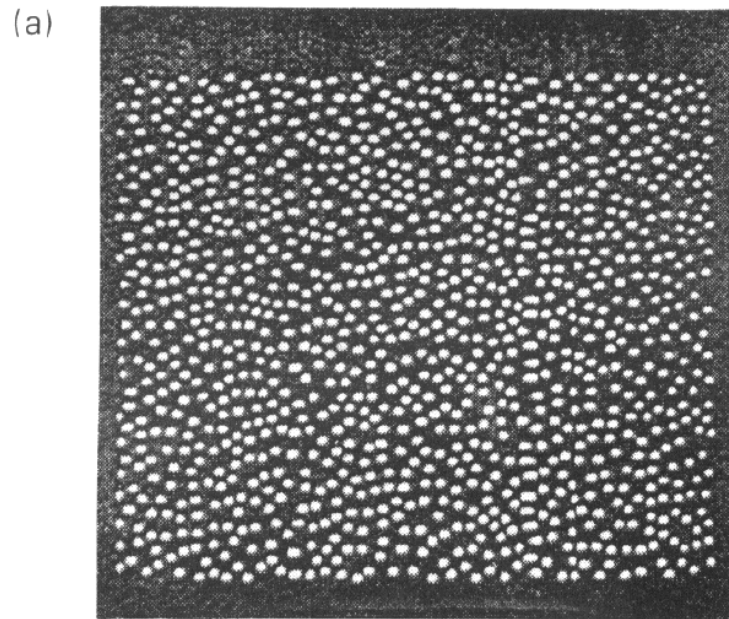
- **Requirements**
 - Even distribution
 - Little correlation between samples
 - Incremental generation
- **Generation of samples**
 - Poisson-disk sampling
 - Fixes a minimum distance between samples
 - Random generation of samples
 - Rejection, if too close to other samples
 - Jittered sampling
 - Random perturbation from regular positions
 - Stratified Sampling
 - Subdivision into areas with one random sample each
 - Improves even distribution
 - Quasi-random numbers (Quasi-Monte Carlo)
 - E.g. Halton Sequence
 - Advanced feature



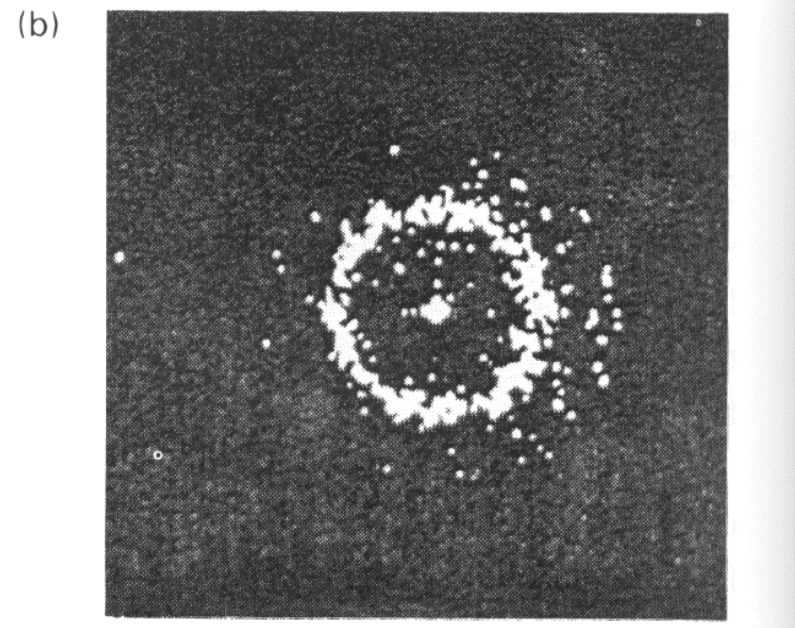
Poisson-Disk Sample Distribution

- **Motivation**

- Distribution of the optical receptors on the retina (here: ape)



Distribution of the receptors



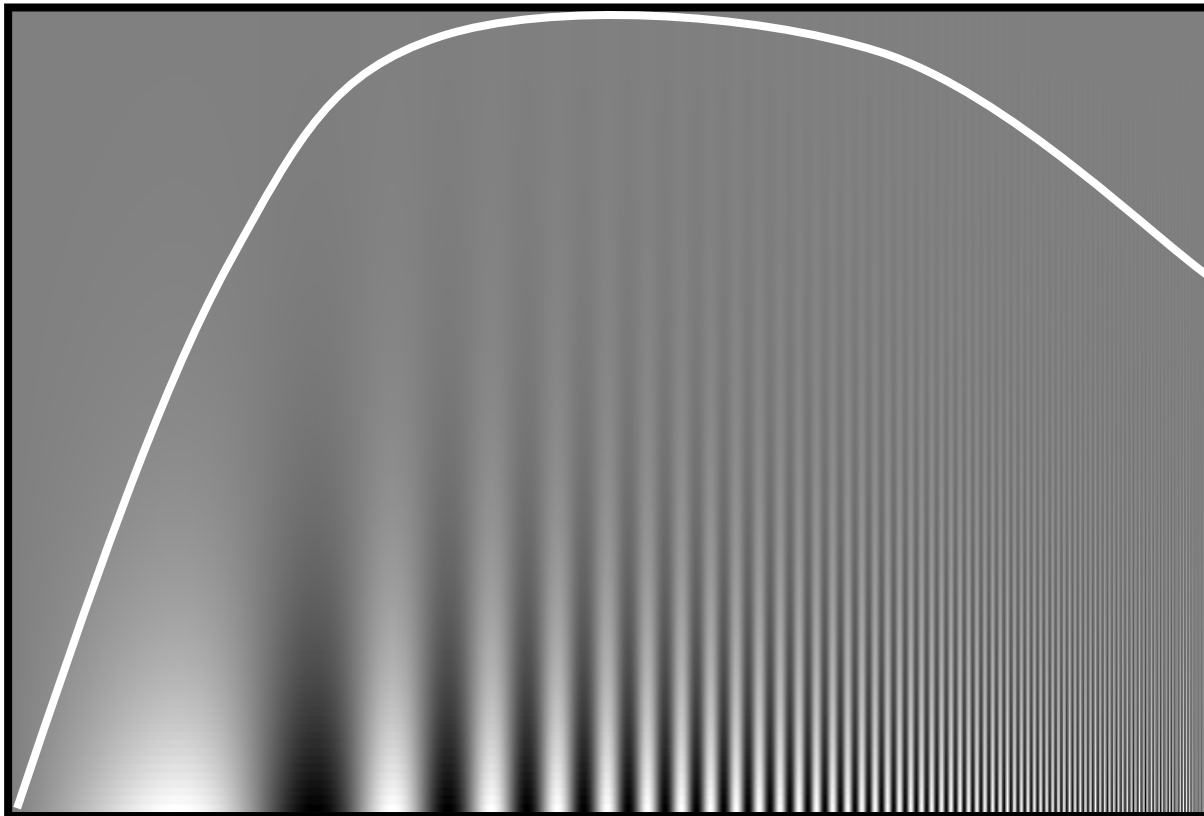
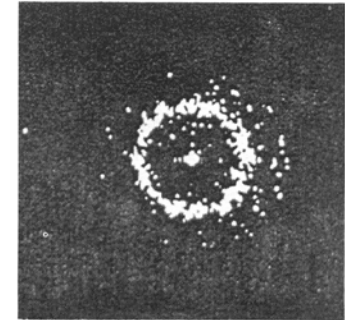
Fourier analysis

© Andrew Glassner, Intro to Raytracing

HVS: Poisson Disk Experiment

- **Human Perception**

- Very sensitive to regular structures
- Insensitive against (high frequency) noise

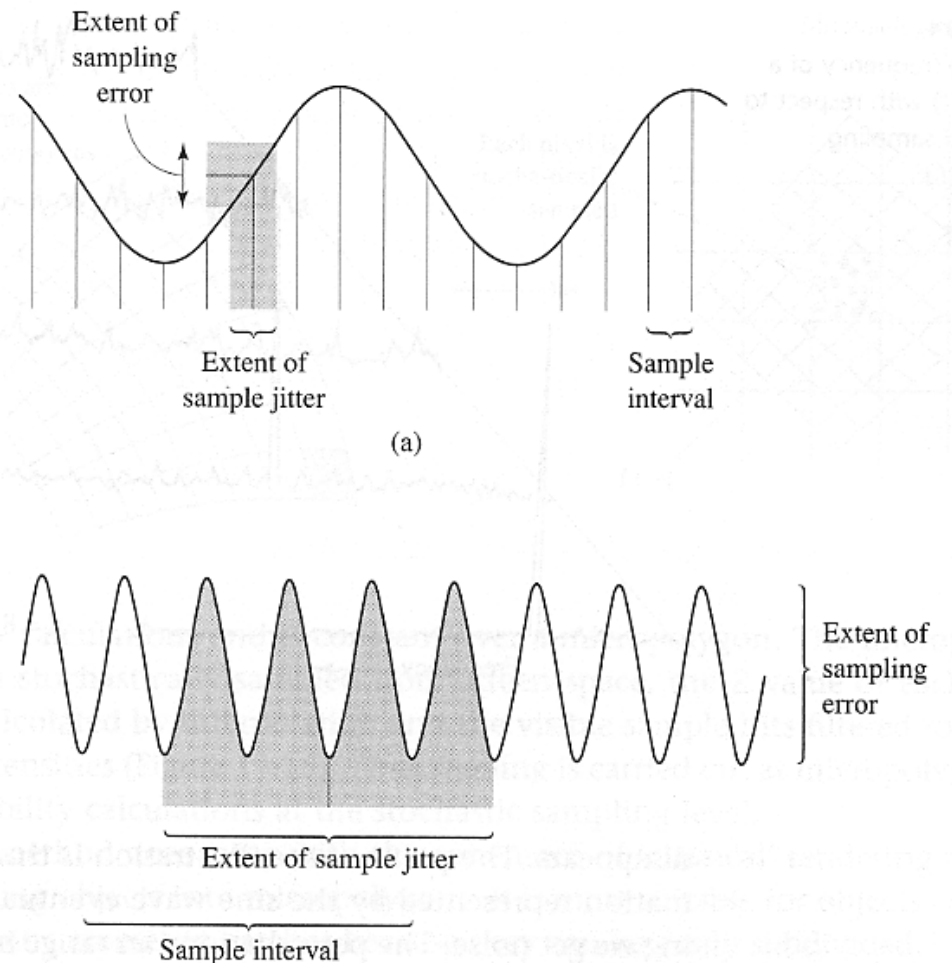


Campbell-Robson contrast sensitivity chart

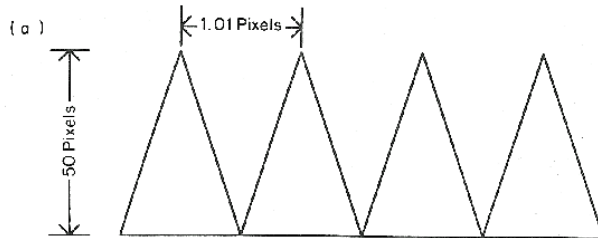
Stochastic Sampling

- **Stochastic Sampling**

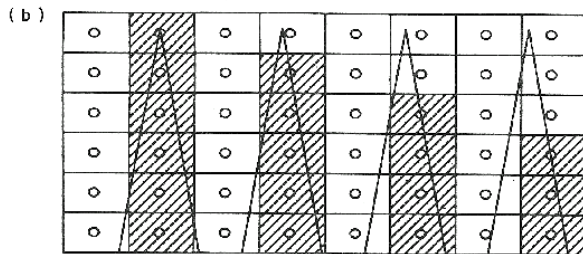
- Transforms energy in high frequency bands into noise
- Low variation in sample domain
 - Closely reconstructs target value
- High variation
 - Reconstructs average value



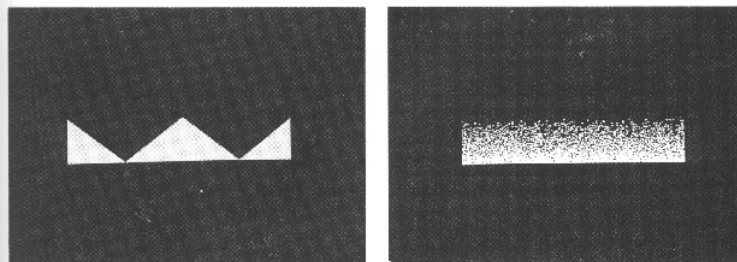
Examples



Triangle comb:
(Width: 1.01 pix, Height: 50 pix):
 1 sample, no jittering
 1 sample, jittering
 16 samples, no jittering
 16 samples, jittering

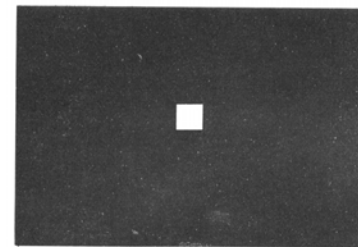


Motion Blur:
 1 sample, no jittering
 1 sample, jittering
 16 samples, no jittering
 16 samples, jittering

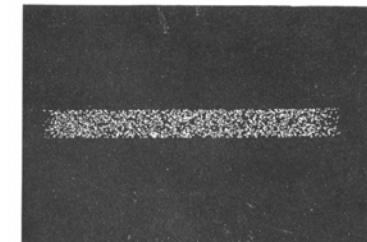


(c)

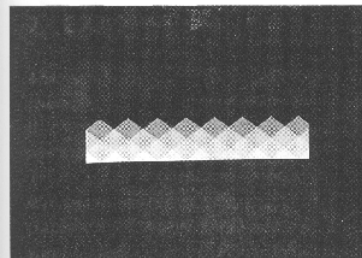
(d)



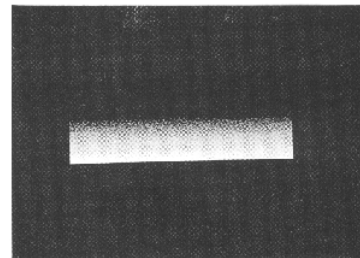
(a)



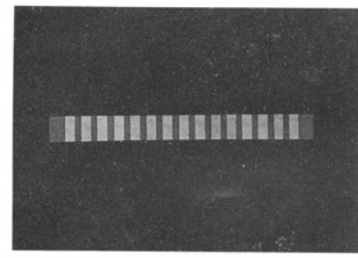
(b)



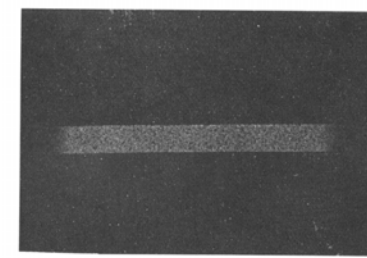
(e)



(f)



(c)



(d)

Comparison

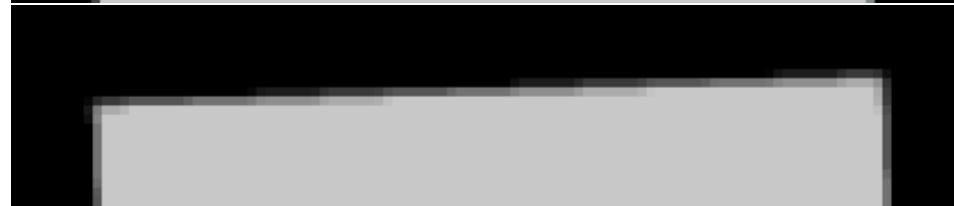
Regular, 1x1



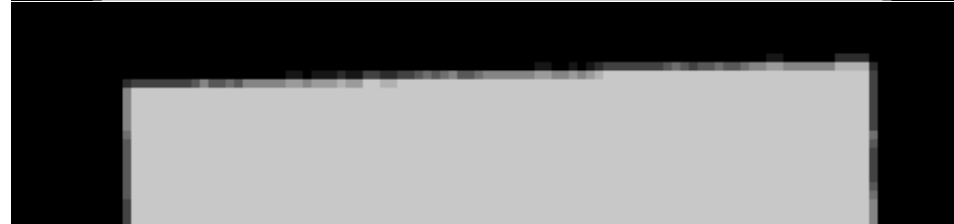
Regular 3x3



Regular, 7x7



Jittered, 3x3



Jittered, 7x7

