## **Computer Graphics**

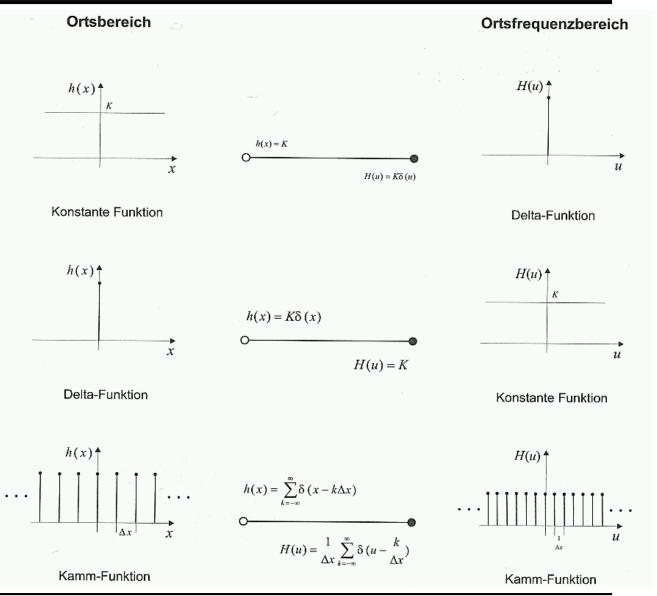
## Sampling Theory & Anti-Aliasing

**Philipp Slusallek** 

# Dirac Comb (1)

- Constant & δ-function
  - flash

Comb/Shah
function



# Dirac Comb (2)

### Constant & δ-Function

Duality

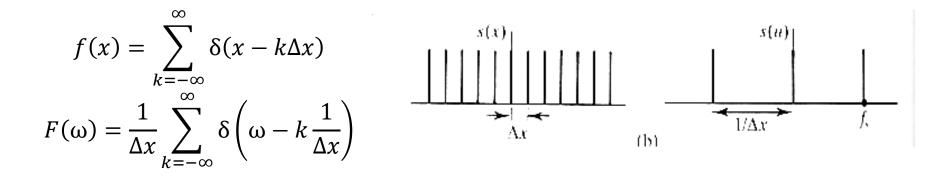
$$f(x) = K$$

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

And vice versa

## Comb function

- Duality: the dual of a comb function is again a comb function
  - Inverse wavelength
  - Amplitude scales with inverse wavelength



# Sampling

## Continuous function

- Assume band-limited
- Finite support of Fourier transform
  - Depicted here as triangle-shaped finite spectrum (not meant to be a tent function)

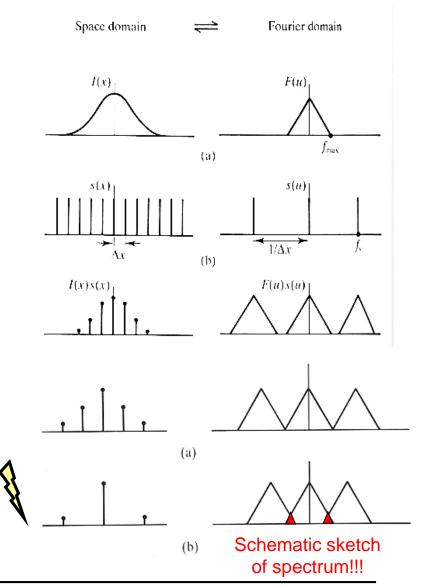
## Sampling at discrete points

- Multiplication with Comb function in spatial domain
- Corresponds to convolution in Fourier domain

 $\Rightarrow$  Multiple copies of the original spectrum (convolution theorem!)

### • Frequency bands overlap ?

- No : good
- Yes: aliasing artifacts

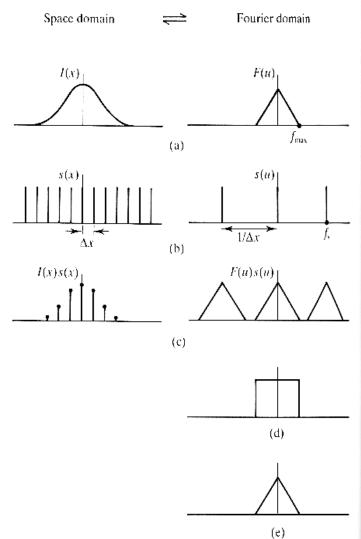


## Reconstruction

- Only original frequency band desired
- Filtering
  - In Fourier domain:
    - Multiplication with windowing function around origin
  - In spatial domain
    - Convolution with inverse Fourier transform of windowing function

## Optimal filtering function

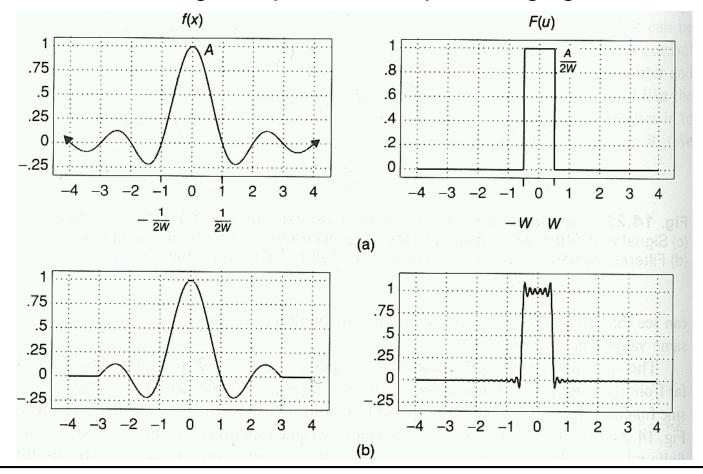
- Box function in Fourier domain
- Corresponds to sinc in space domain
  - Unlimited region of support
  - Spatial domain only allows approximations due to finite support.



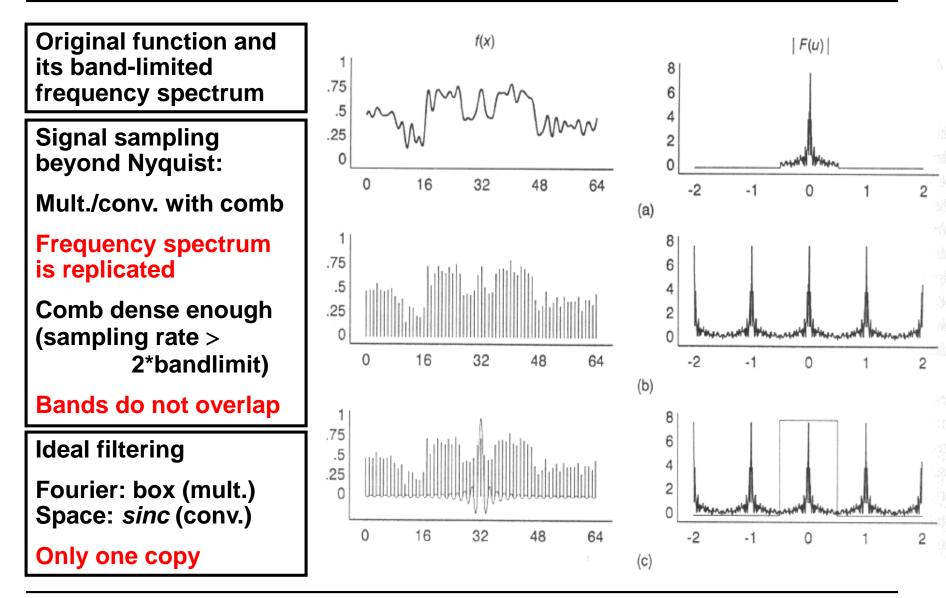
## **Reconstruction Filter**

 Cutting off the spatial support of the sinc function is NOT a good solution

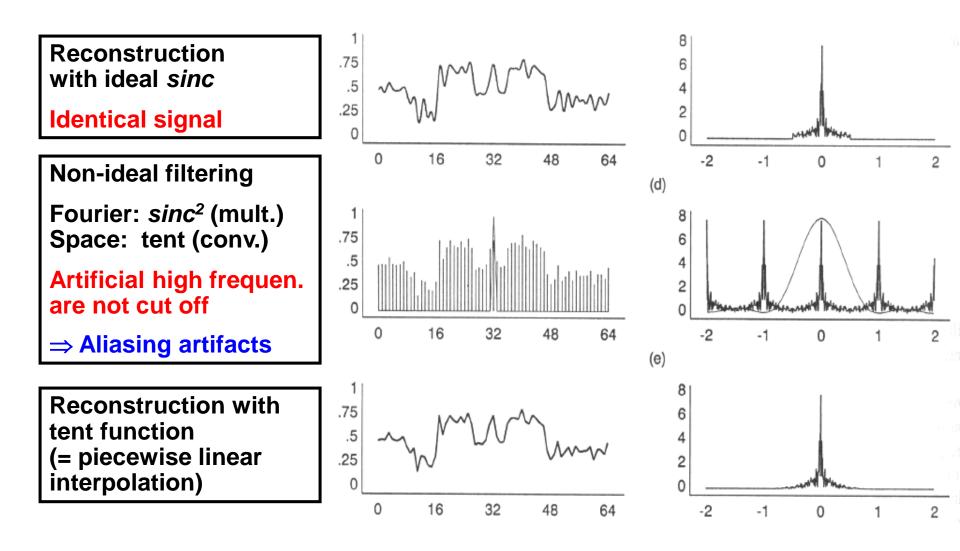
- Re-introduces high-frequencies  $\Rightarrow$  spatial ringing



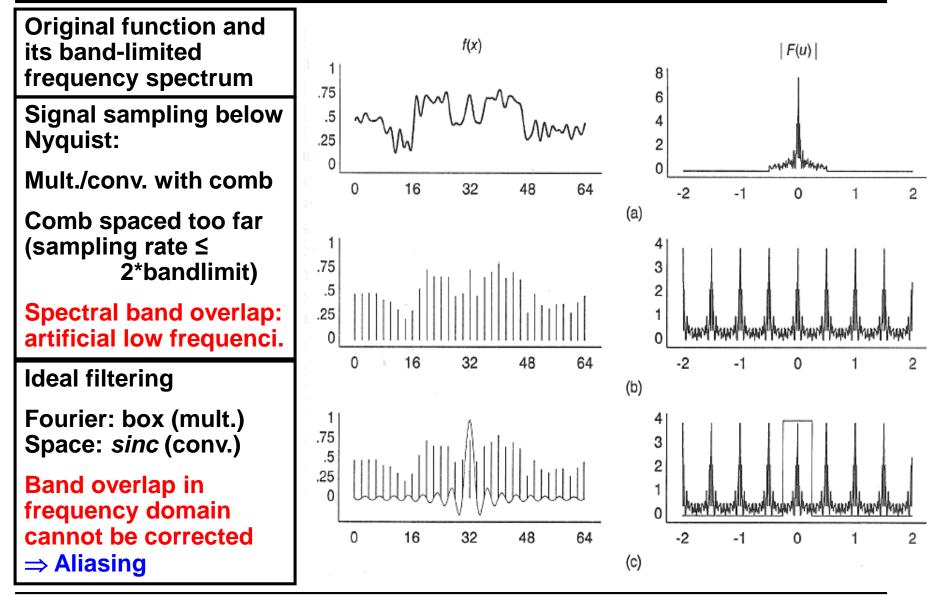
# Sampling and Reconstruction



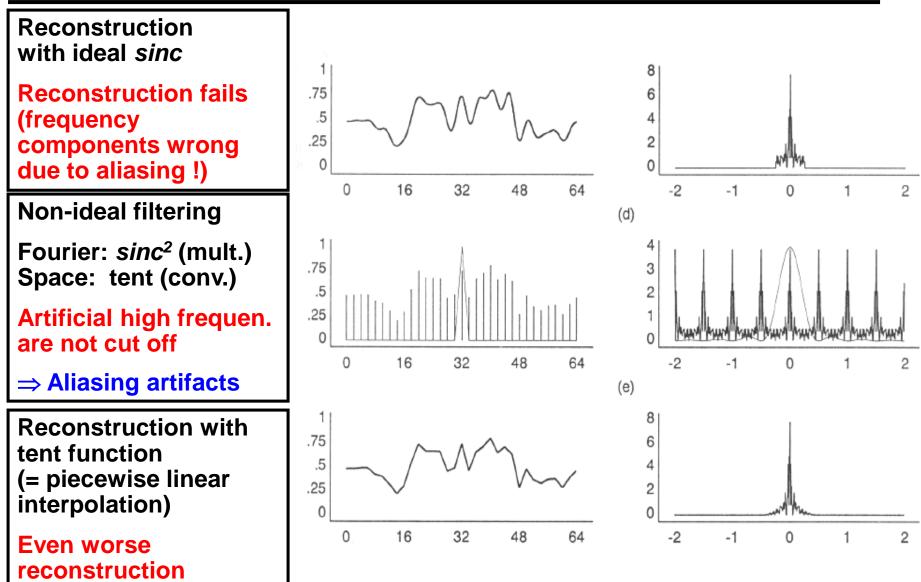
# Sampling and Reconstruction



# Sampling at Too Low Frequency

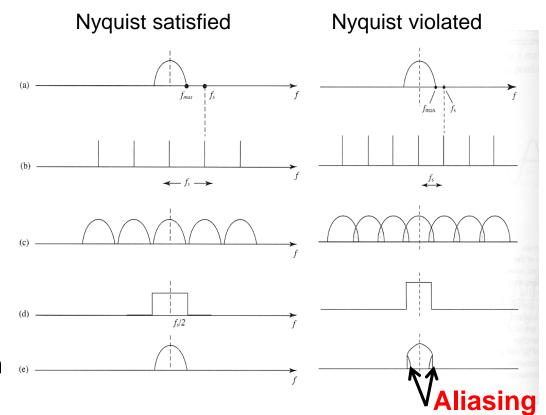


# Sampling at Too Low Frequency



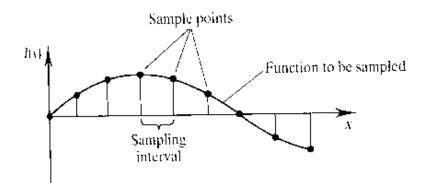
# Aliasing

- High frequency components from the replicated copies are treated like low frequencies during the reconstruction process
- In Fourier space:
  - Original spectrum
  - Sampling comb
  - Resulting spectrum
  - Reconstruction filter
  - Reconstructed spectrum (e)

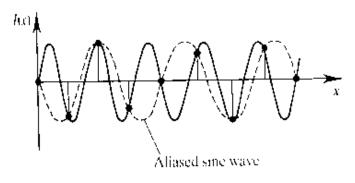


Different signals become "aliases" when sampled

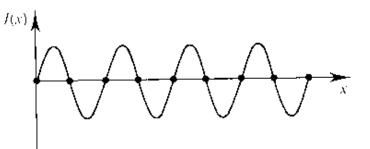
# Aliasing in 1D



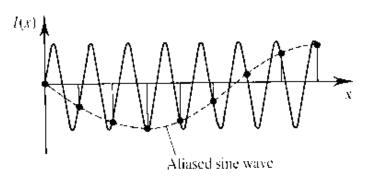
**Spatial frequency < Nyquist** 



**Spatial frequency > Nyquist** 

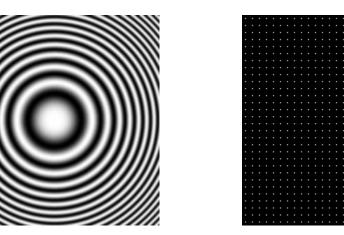


Spatial frequency = Nyquist 2 samples / period

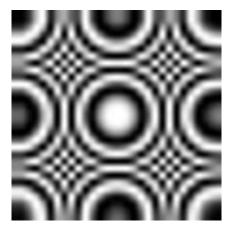


Spatial frequency >> Nyquist

# Aliasing in 2D



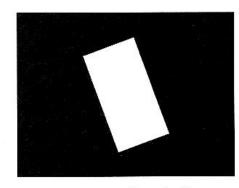




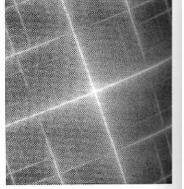
This original image sampled at these locations yields this reconstruction.

## Aliasing in 2D

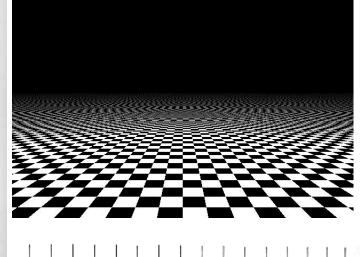
- Spatial sampling ⇒ repeated frequency spectrum
- Spatial conv. with box filter  $\Rightarrow$  spectral mult. with sinc

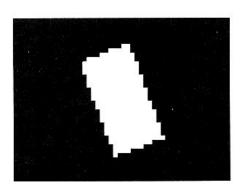


(a) Simulation of a perfect line

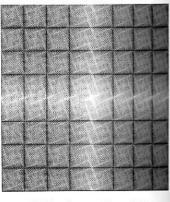


(b) Fourier transform of (a)

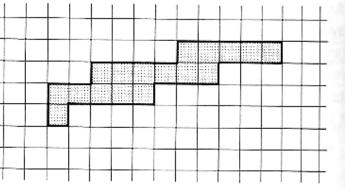




(c) Simulation of a jagged line



(d) Fourier transform of (c)



## **Causes for Aliasing**

- It all comes from sampling at discrete points
  - Multiplied with comb function
  - Comb function: repeats frequency spectrum
- Issue when using non-band-limited primitives
  - Hard edges  $\rightarrow$  infinitely high frequencies
- In reality, integration over finite region necessary
  - E.g., finite CCD pixel size, integrates in the analog domain

### 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
- Important:
  - Distinction between sampling errors and reconstruction errors

# **Sampling Artifacts**

## Spatial aliasing

- Stair cases, Moiré patterns (interference), etc...

## Solutions

- Increasing the sampling rate
  - OK, but infinite frequencies at sharp edges
- Post-filtering (after reconstruction)
  - Too late, does not work only leads to blurred stair cases
- Pre-filtering (blurring) of sharp geometry features
  - Slowly make geometry "fade out" at the edges?
  - Correct solution in principle, but blurred images might not be useful
  - Analytic low-pass filtering hard to implement
- Super-sampling (see later)
  - On the fly re-sampling: densely sample, filter, down sample

# Sampling Artifacts in 4D

- Temporal aliasing
  - Video of cart wheel, ...

## Solutions

- Increasing the frame rate
  - OK
- Post-filtering (averaging several frames)
  - Does not work only multiple details
- Pre-filtering (motion blur)
  - Should be done on the original analog signal
  - Possible for simple geometry (e.g., cartoons)
  - Problems with texture, etc...
- Super-sampling (see later)



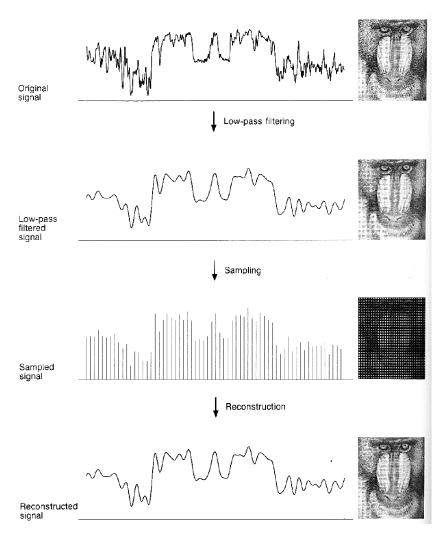
# Antialiasing by Pre-Filtering

## Filtering before sampling

- Analog/analytic original signal
- Band-limiting the signal
- Reduce Nyquist frequency for chosen sampling-rate
- Ideal reconstruction
  - Convolution with sinc

## Practical reconstruction

- Convolution with
  - Box filter, Bartlett (tent)
  - $\rightarrow$  Reconstruction error



# **Sources of High Frequencies**

#### Geometry

- Edges, vertices, sharp boundaries
- Silhouettes (view dependent)

- ...

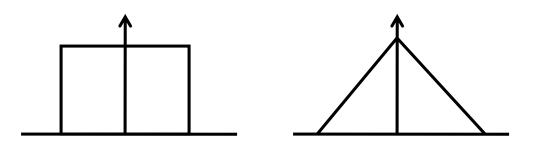
- Texture
  - E.g., checkerboard pattern, other discontinuities, ...

## Illumination

- Shadows, lighting effects, projections, ...

### Analytic filtering almost impossible

- Even with the most simple filters



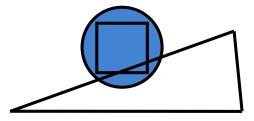
# Comparison

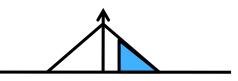
## Analytic low-pass filtering (pixel/triangle overlap)

- Ideally eliminates aliasing completely
- Hard to implement
  - · Weighted or unweighted area evaluation
  - 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
- But it helps: ...





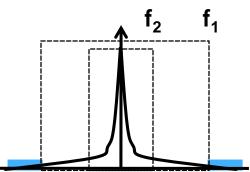
# **Re-Sampling Pipeline**

### Assumption

- Energy in higher frequencies typically decreases quickly
- Reduced aliasing by intermediate sampling at higher frequency

## Algorithm

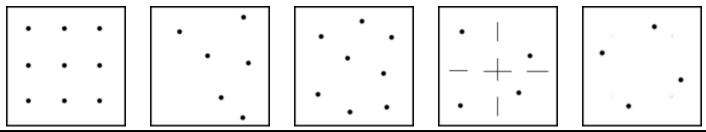
- Super-sampling
  - Sample continuous signal with high frequency  $f_1$
  - Aliasing with energy beyond  $f_1$  (assumed to be small)
- Reconstruction of signal
  - Filtering with  $g_1(x)$ : e.g. convolution with  $sinc_{f1}$
  - Exact representation with sampled values !!
- Analytic low-pass filtering of signal
  - Filtering with filter  $g_2(x)$  where  $f_2 \ll f_1$
  - 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

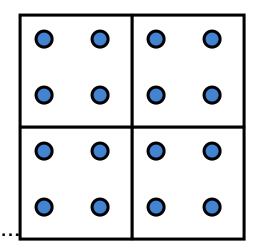


# **Super-Sampling in Practice**

### Regular super-sampling

- Averaging of N samples per pixel
- N: 4 (quite good), 16 (often sufficient)
- Samples: rays, z-buffer, motion, reflection, ...
- Filter weights
  - Box filter
  - Others: B-spline, pyramid (Bartlett), hexagonal,
- Sampling Patterns (left to right)
  - Regular: aliasing likely
  - Random: often clumps, incomplete coverage
  - Poisson Disc: close to perfect, but costly
  - Jittered: randomized regular sampling
  - Most often: rotated grid pattern





# Super-Sampling Caveats

### Popular mistake

- Sampling at the corners of every pixel
- Pixel color by averaging from corners
- Free super-sampling ???

### Problem

- Wrong reconstruction filter !!!
- Same sampling frequency, but post-filtering with a tent function
- Blurring: loss of information

### Post-reconstruction blur

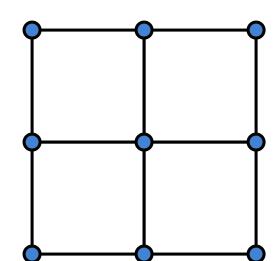


1x1 Sampling, 3x3 Blur



1x1 Sampling, 7x7 Blur

There is no "free" 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**

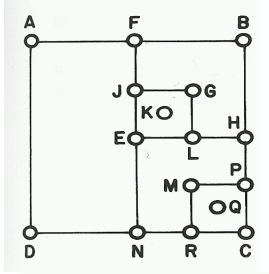
#### Recursive algorithm

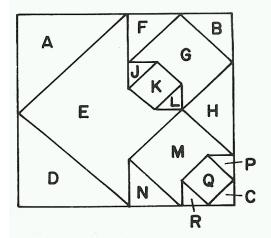
- Sampling at pixel corners and center
- Decision criterion for corner-center pairs
  - Differences, contrast, object-IDs, ray trees, ...
- Subdivide quadrant by adding 3 diag. points
- Filtering with weighted averaging
  - Tile: ¼ from each quadrant
  - Leaf quadrant: 1/2 (center + corner)
- Box filter with final weight proport. to area  $\rightarrow$

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

#### Extension

- Jittering of sample points

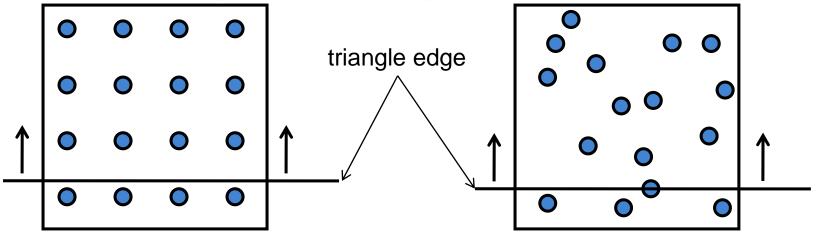




# **Stochastic Super-Sampling**

#### Problems with regular super-sampling

- Nyquist frequency for aliasing only shifted
- Expensive: 4-fold to 16-fold effort
- Non-adaptive: same effort everywhere
- Too regular: reduction of effective number of axis-aligned levels
- Introduce irregular sampling pattern



 $0 \rightarrow 4/16 \rightarrow 8/16 \rightarrow 12/16 \rightarrow 16/16$ : 5 levels

17 levels: better, but noisy

- Stochastic super-sampling
  - Or analytic computation of pixel coverage and pixel mask

## 27

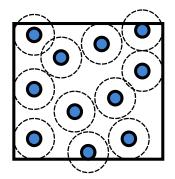
# **Stochastic Sampling**

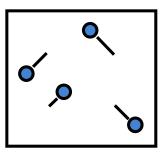
#### Requirements

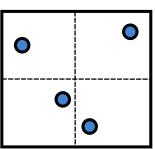
- Even sample distribution: no clustering
- Little correlation between positions: no alignment
- Incremental generation: on demand as needed

## Generation of samples

- Poisson-disk sampling
  - Random generation of samples
  - Rejection if closer than min distance to other samples
- Jittered sampling
  - Random perturbation from regular positions
- Stratified sampling
  - Subdivision into areas with one random sample in each
  - Improves even distribution
- Quasi-random numbers (Quasi-Monte Carlo)
  - E.g. Halton sequence
  - Advanced feature: see RIS course for more details

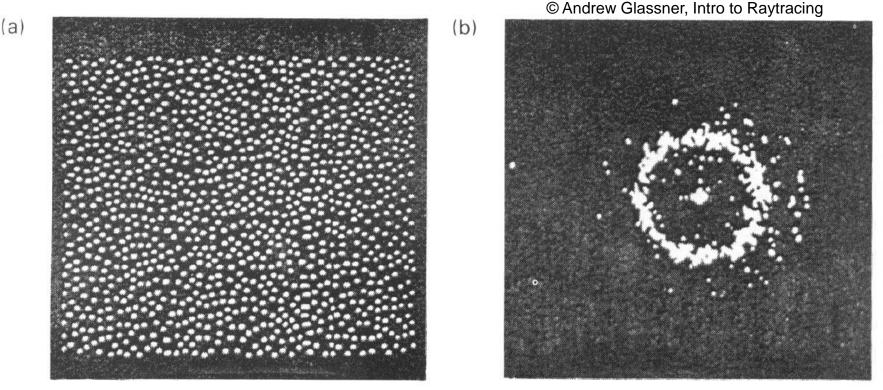






## Poisson-Disk Sample Distribut.

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



Distribution of the photo-receptors

Fourier analysis

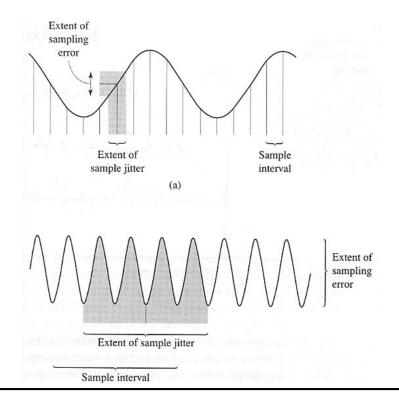
## **Stochastic Sampling**

### Slowly varying function in sample domain

Closely reconstructs target value with few samples

## Quickly varying function in sample domain

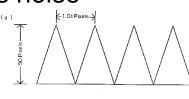
- Transforms energy in high-frequency bands into noise
- Reconstructs average value as sample count increases

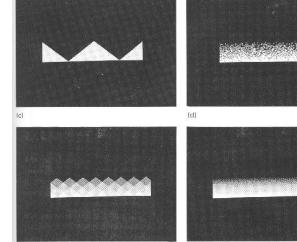


# Examples

## Spatial sampling: triangle comb

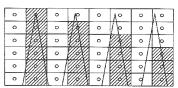
- (c) 1 sample/pixel, no jittering: aliasing
- (d) 1 spp, jittering: noise
- (e) 16 spp, no jittering: less aliasing
- (f) 16 spp, jittering: less noise

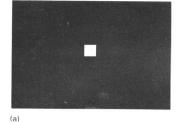


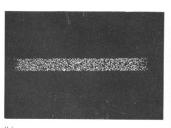


(f)

(d)







### Temporal sampling: motion blur

- (a) 1 time sample, no jittering: aliasing
- (b) 1 time sample, jittering/pixel: noise
- (c) 16 samples, no jittering: less aliasing
- (d) 16 samples, jittering/pixel: less noise



(c)

# Comparison

• Regular, 1x1

- Regular, 3x3
- Regular, 7x7

- Jittered, 3x3
- Jittered, 7x7

