

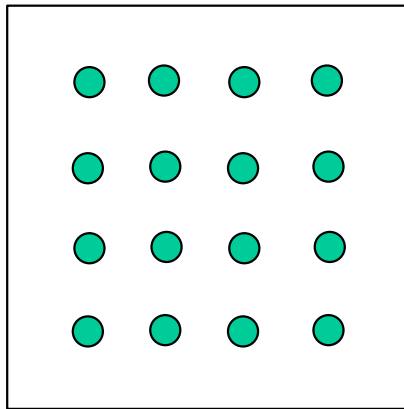


# **CS 563 Advanced Topics in Computer Graphics Sampling and Reconstruction III**

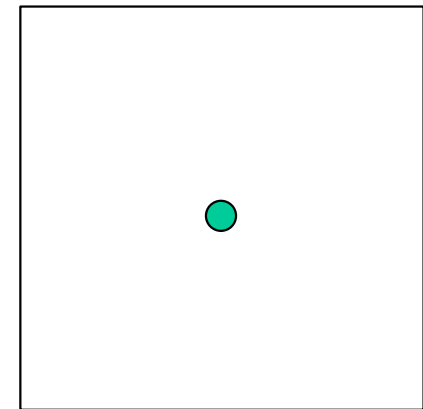
by Emmanuel Agu


# Uniform Supersampling

- Increasing the sampling rate:
  - Moves each spectra copy further apart
  - Potentially reducing the overlap and thus aliasing
- Resulting samples must be resampled (filtered) to image sampling rate



$$Pixel = \sum_k w_k \times Sample_k$$



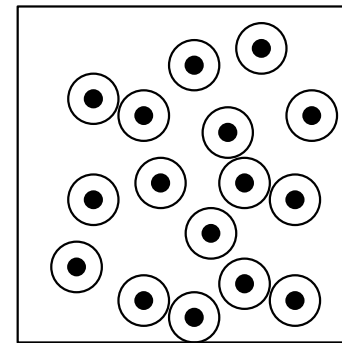
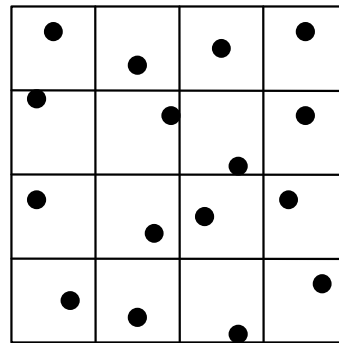
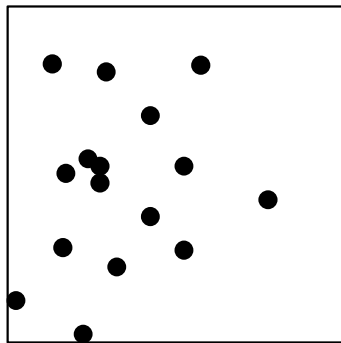


## Non-Uniform Sampling - Intuition

- Non-uniform sampling
  - Essentially, non-uniform sampling converts aliases into broadband noise
  - Less noticeable by eye
  - Noise is incoherent, and much less objectionable
  - Based on Yellot theory (1983)

# Non-Uniform Sampling - Patterns

- Poisson
  - Pick  $n$  random points in sample space
- Uniform Jitter
  - Subdivide sample space into  $n$  regions
- Poisson Disk
  - Pick  $n$  random points, but not too close

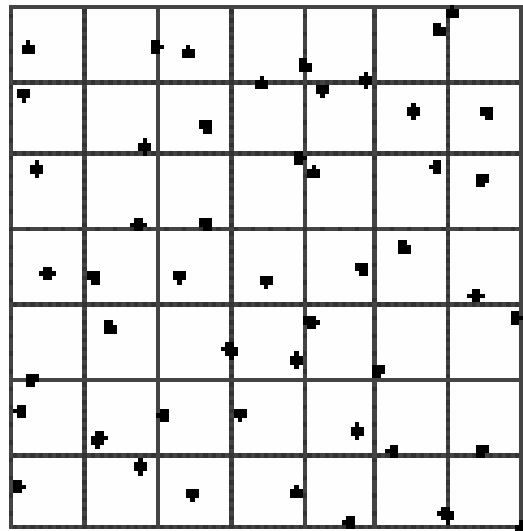




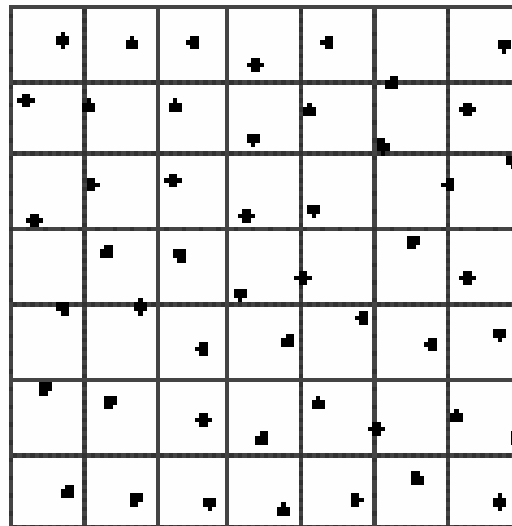
# Best-Candidate Sampling

- Jittered stratification
  - Randomness (inefficient)
  - Clustering problems
  - Undersampling (“holes”)
- Stratified, Low Discrepancy Sequences
  - Still (visibly) aliased
- “Ideal”: Poisson disk distribution
  - too computationally expensive
- Best candidate sampling - approximation to Poisson disk

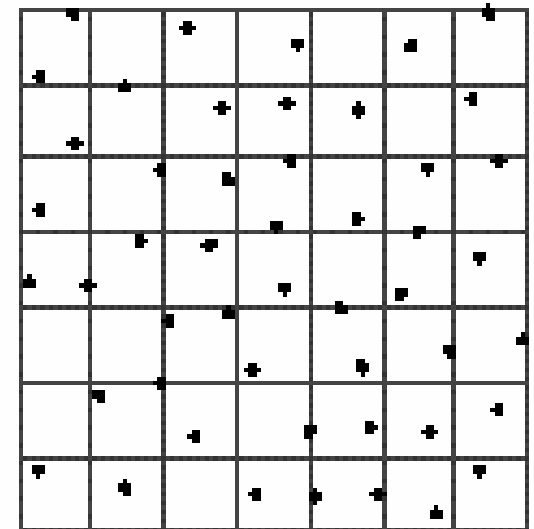
# Best-Candidate Sampling



Jittered



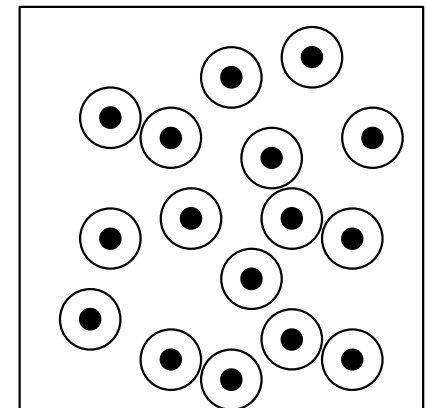
Poisson Disk



Best Candidate

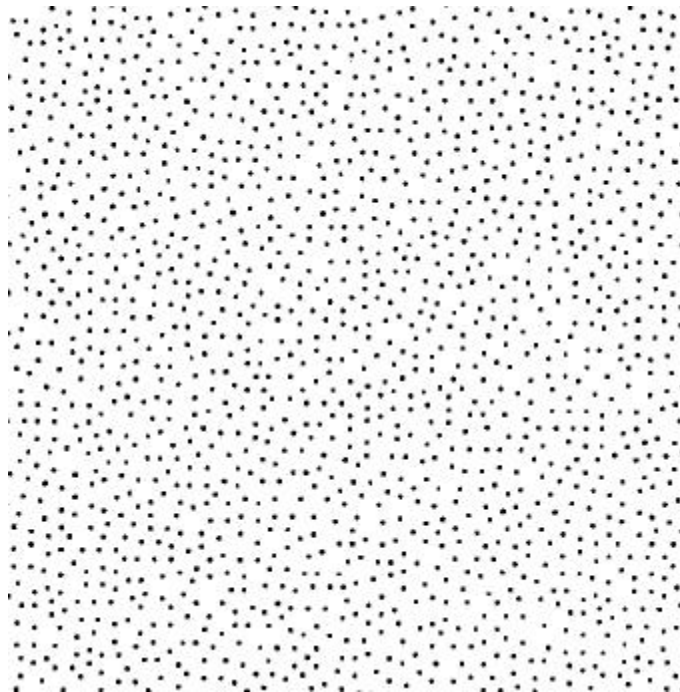
# Poisson Disk

- Comes from eye structure of – rods and cones
- Dart Throwing
- No two points are closer than a threshold
- Very expensive – time consuming
- Compromise – Best Candidate Sampling
  - Don Mitchell
  - Generates many **potential** candidates randomly, only insert **farthest one** to all previous samples.
  - Compute “tilable pattern” offline that is reused by tiling the image plane (translating and scaling).
  - Toroidal topology – paste on toroid
  - Affects distance between points on top to bottom

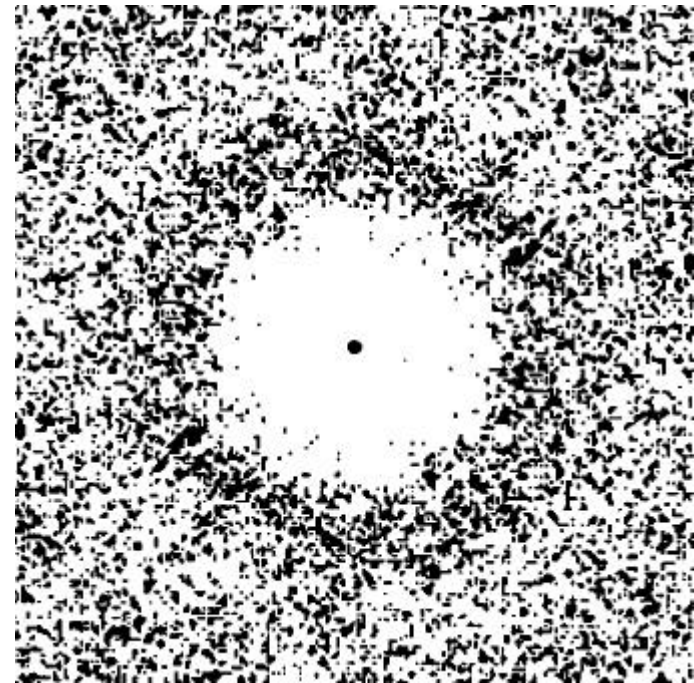


# Poisson Disk Sampling

- Spectral characteristics:
  - **Poisson:** completely uniform (white noise). High and low frequencies equally present
  - **Poisson disc:** Pulse at origin (DC component of image), surrounded by empty ring (no low frequencies), surrounded by white noise



Spatial Domain

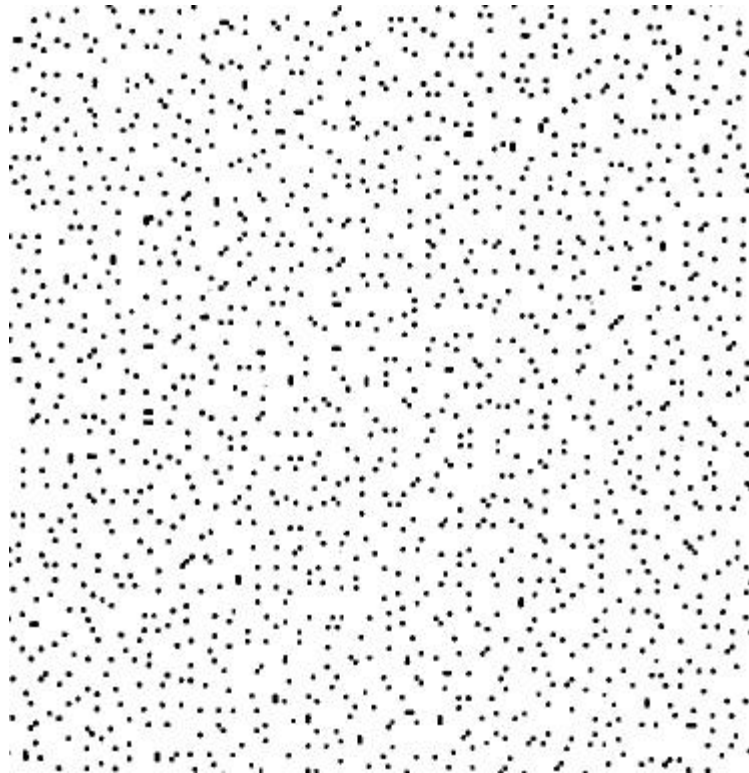


Fourier Domain

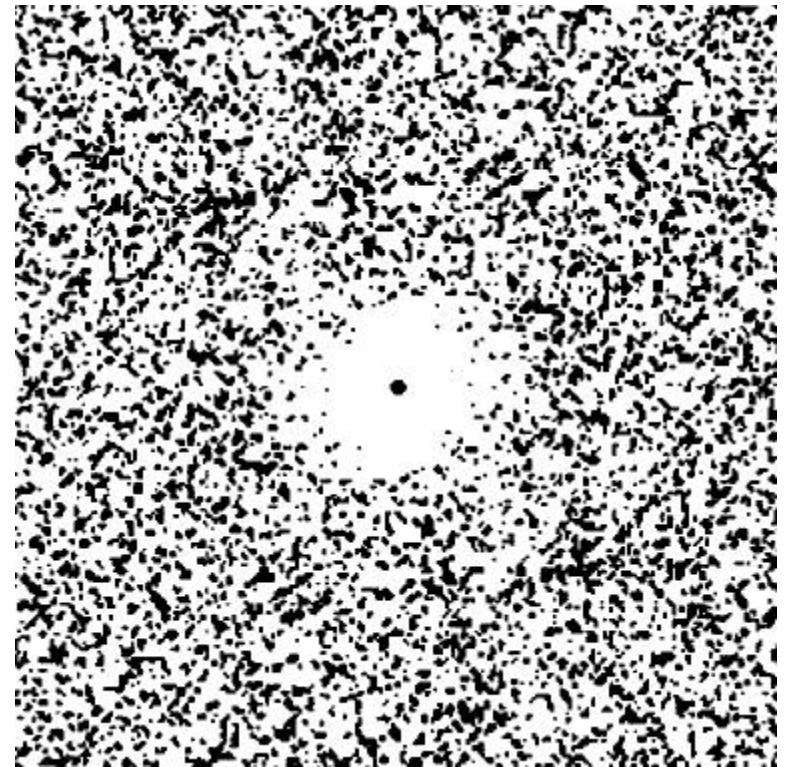


# Uniform Jittered Sampling

- Spectral characteristics:
  - Jitter: Approximates Poisson disc spectrum,
  - **But** with a smaller empty disc.



Spatial Domain



Fourier Domain

# Poisson Disk algorithm

$i \leftarrow 0$

while  $i < N$

$x_i \leftarrow \text{unit}()$

*Throw a dart.*

$y_i \leftarrow \text{unit}()$

$\text{reject} \leftarrow \text{false}$

---

for  $k \leftarrow 0$  to  $i - 1$

*Check the distance to all other samples.*

$d \leftarrow (x_i - x_k)^2 + (y_i - y_k)^2$

---

if  $d < (2r_p)^2$  then

$\text{reject} \leftarrow \text{true}$

*This one is too close—forget it.*

break

endif

---

endfor

---

if not  $\text{reject}$  then

$i \leftarrow i + 1$

*Append this one to the pattern.*

endif

---

endwhile

---

# Texture

Jitter with 1 sample/pixel



Best Candidate with 1 sample/pixel



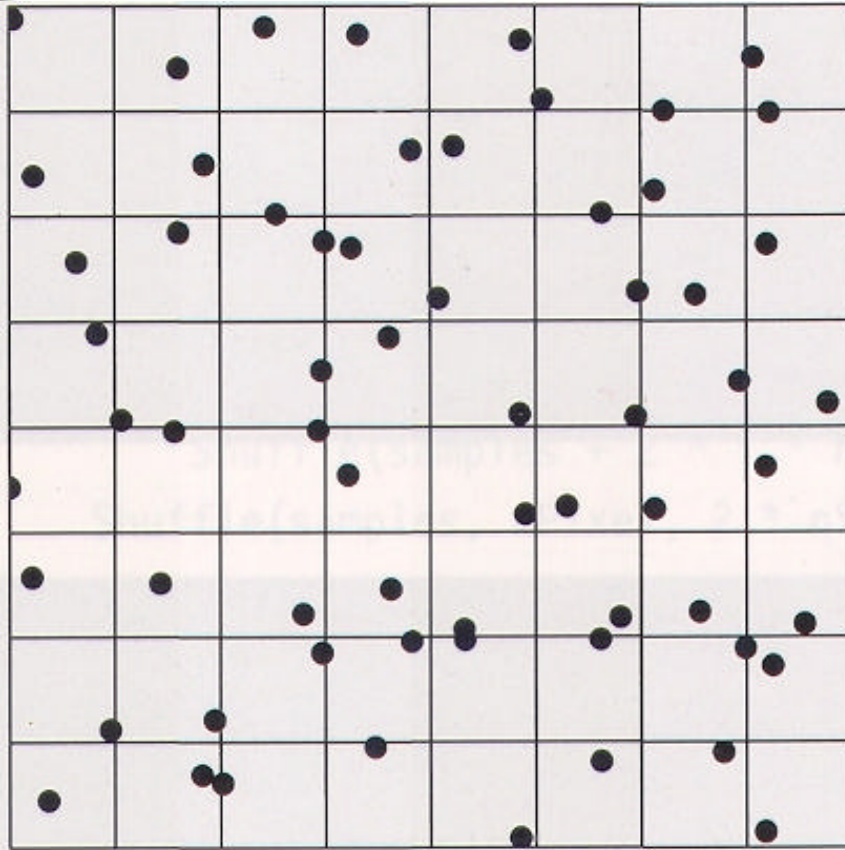
Jitter with 4 sample/pixel



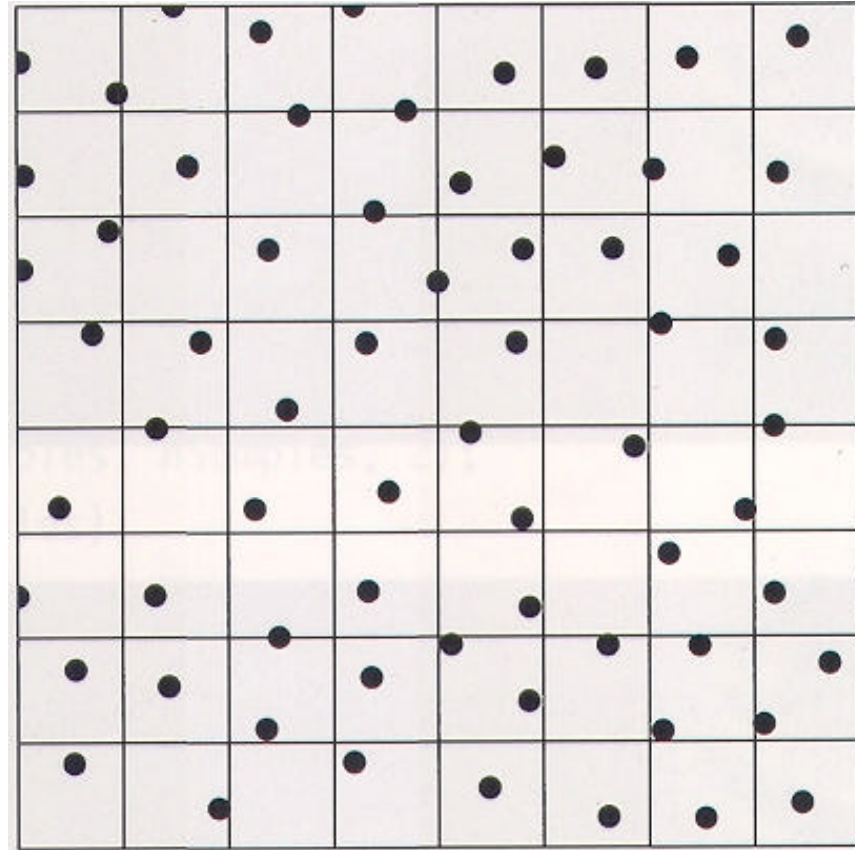
Best Candidate with 4 sample/pixel



# Best candidate sampling



*stratified jittered*



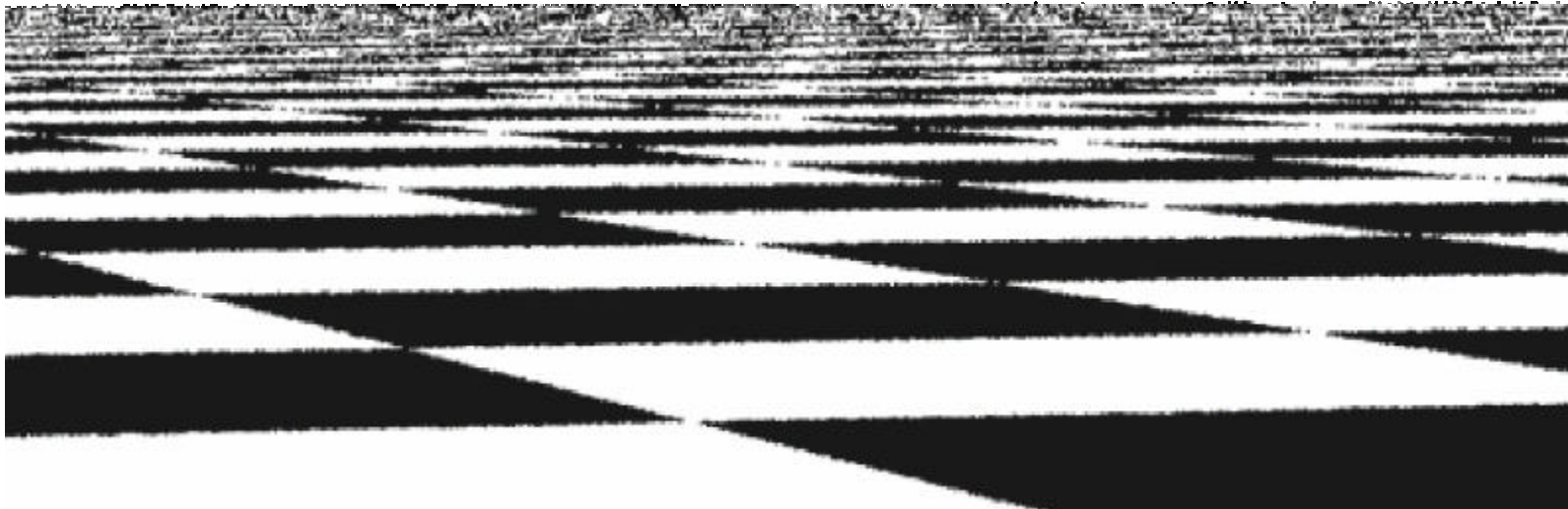
*best candidate*

*It avoids holes and clusters.*

# Best candidate sampling



*stratified jittered, 1 sample/pixel*



*best candidate, 1 sample/pixel*

# Best candidate sampling



*stratified jittered, 4 sample/pixel*



*best candidate, 4 sample/pixel*



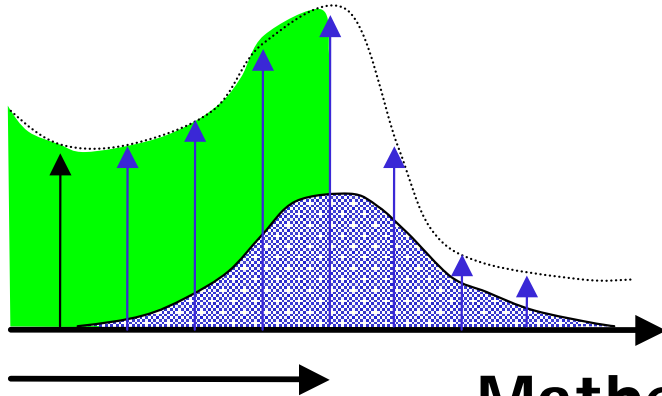


# Ideal Reconstruction

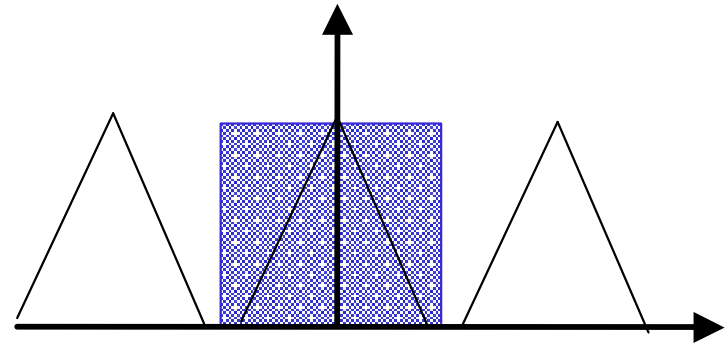
- Ideally, use perfect low-pass filter - the sinc function
  - to bandlimit the sampled signal
- Thus remove all copies of spectra introduced by sampling
- Unfortunately,
  - The sinc has infinite extent and we must use simpler filters with finite extents. Physical processes in particular do not reconstruct with sincs
  - The sinc may introduce ringing which are perceptually objectionable

# How? - Reconstruction

## Spatial Domain:



## Frequency Domain:



## Mathematically:

- Convolution:

$$f(x) * h(x)$$

$$\int_{-\infty}^{\infty} f(t) \times h(x - t) dt$$

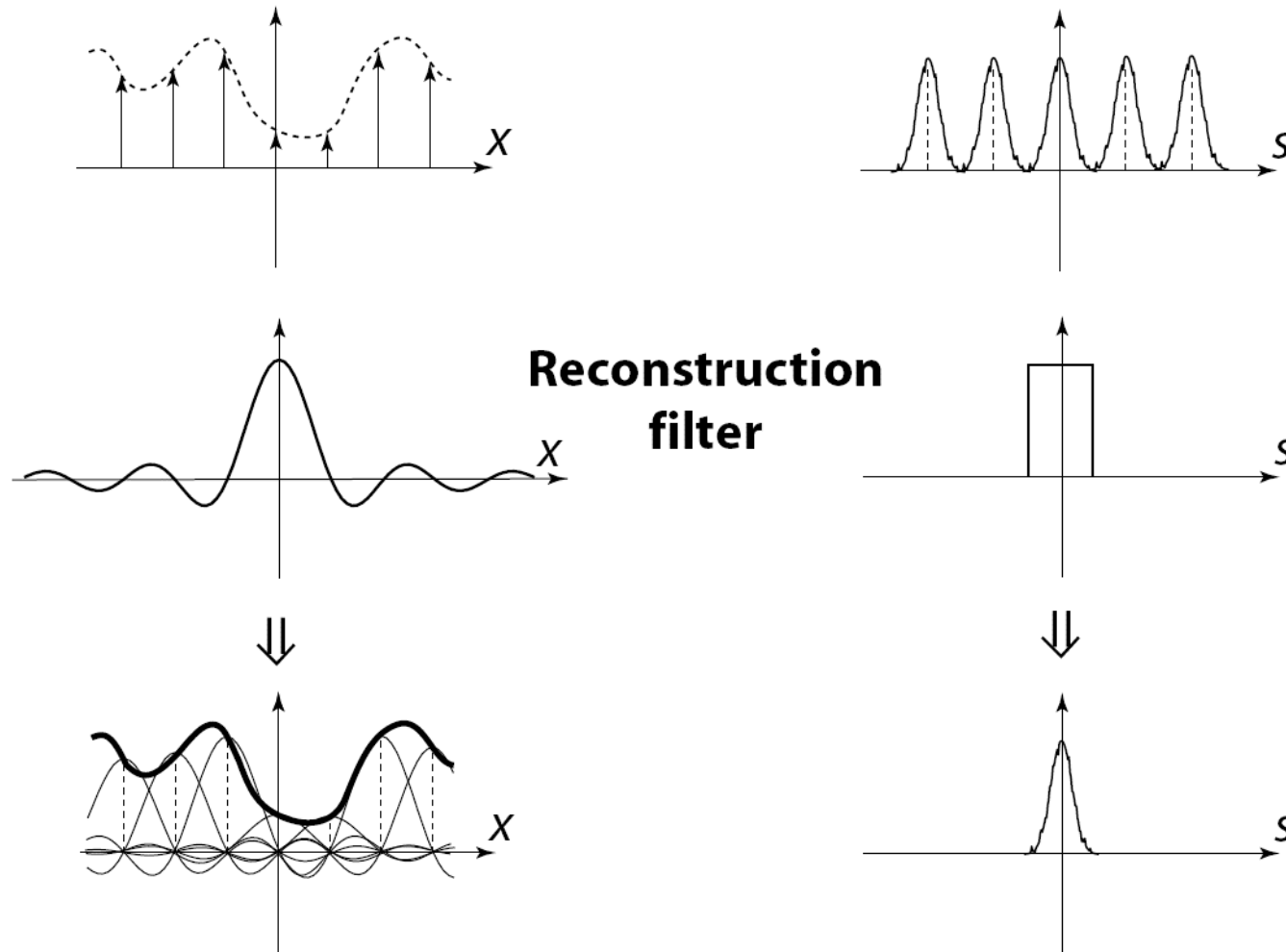
Evaluated at discrete points (sum)

- Multiplication:

$$F(\omega) \times H(\omega)$$



# Reconstruction

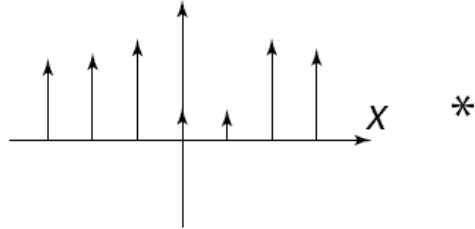


*The reconstructed function is obtained by interpolating among the samples in some manner*

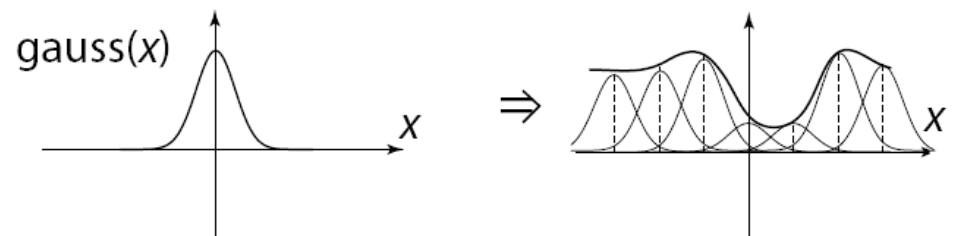
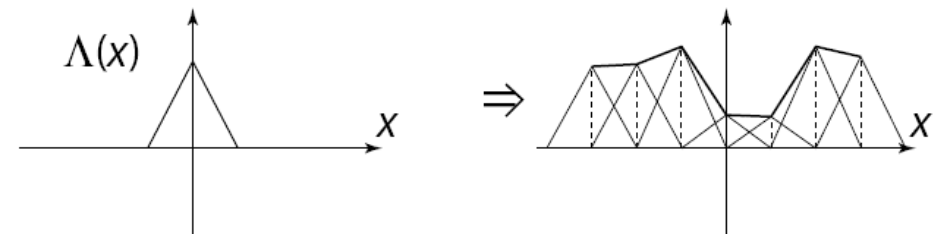
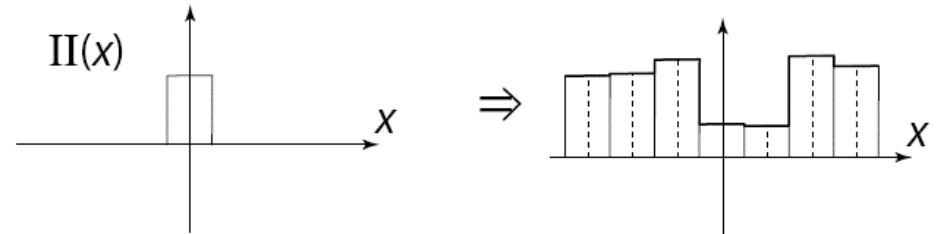
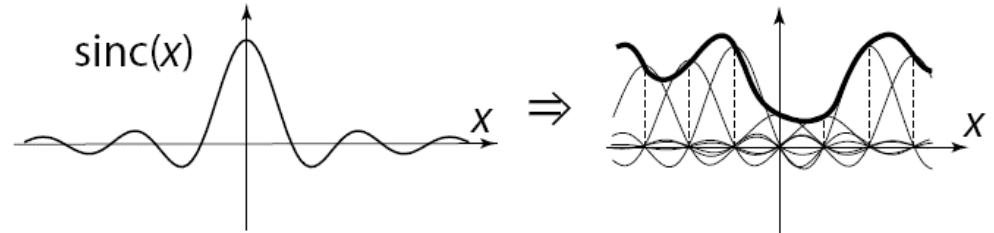
# Reconstruction filters

The sinc filter, while ideal, has two drawbacks:

- It has large support (slow to compute)
- It introduces ringing in practice



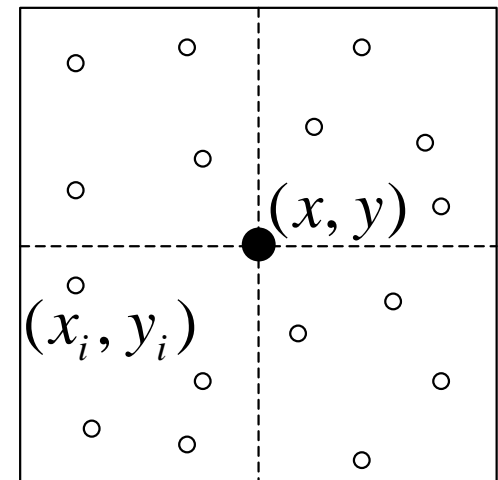
The box filter is bad because its Fourier transform is a sinc filter which includes high frequency contribution from the infinite series of other copies.



## Reconstruction filters

- Given image samples, we can do the following to compute pixel values.
  1. reconstruct continuous function  $L'$  from samples
  2. prefilter  $L'$  to remove frequency higher than Nyquist limit
  3. sample  $L'$  at pixel locations
- Instead, we consider an interpolation problem

$$I(x, y) = \frac{\sum_i f(x - x_i, y - y_i) L(x_i, y_i)}{\sum_i f(x - x_i, y - y_i)}$$



- provides an interface to  $f(x,y)$
- **Film** stores a pointer to a filter and use it to filter the output before writing it to disk.

width, half of support

`Filter::Filter(float xw, float yw)`

`Float Evaluate(float x, float y);`

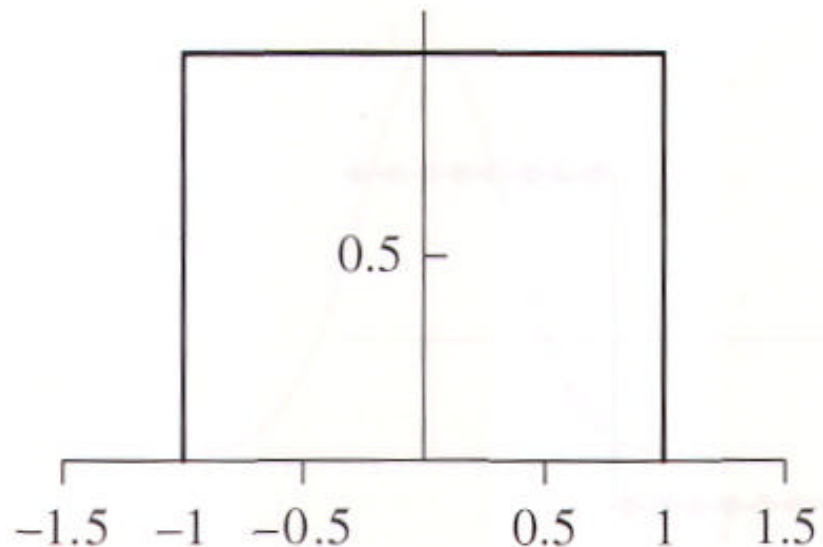
- `filters/*`

## Box filter

- Most commonly used in graphics. It's just about the worst filter possible, incurring postaliasing by high-frequency leakage.

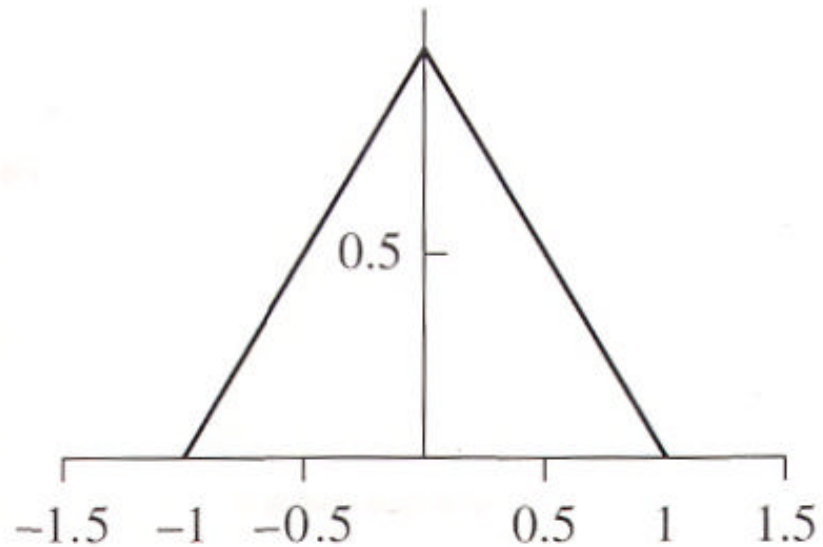
```
Float BoxFilter::Evaluate(float x, float  
y)  
{  
    return 1.;  
}
```

- **Note:** input always in Range **-1 to 1**



# Triangle filter

```
Float TriangleFilter::Evaluate(float x,  
    float y)  
{  
    return max(0.f, xWidth-fabsf(x)) *  
           max(0.f, yWidth-fabsf(y));  
}
```



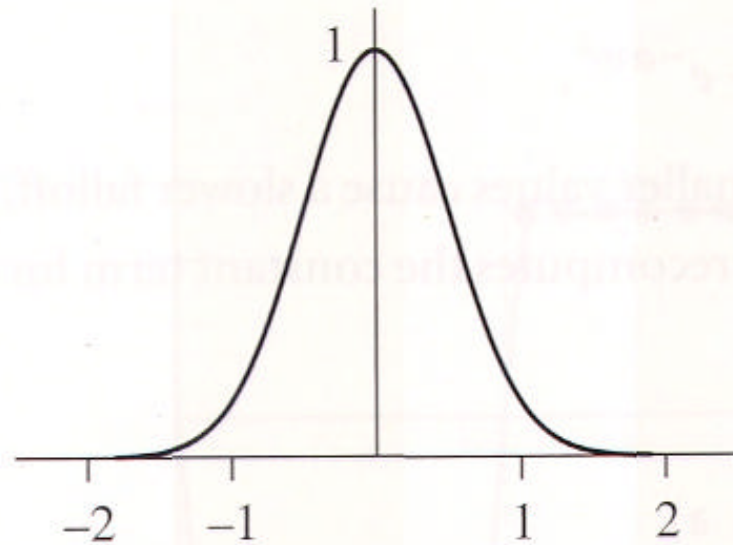
# Gaussian filter

- Gives reasonably good results in practice

```
Float GaussianFilter::Evaluate(float x,  
    float y)  
{  
    return Gaussian(x, expX)*Gaussian(y, expY);  
}
```

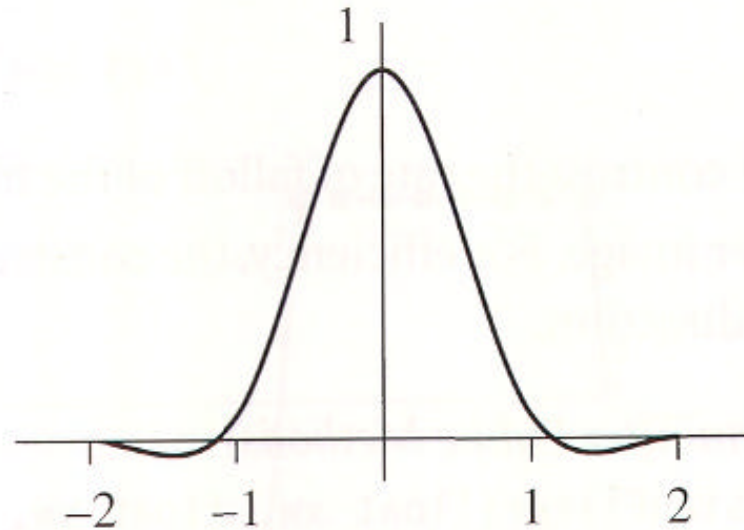
1D Gaussian filter:

$$f(x) = e^{-ax^2} - e^{-aw}$$



## Mitchell filter

- parametric filters, tradeoff between ringing and blurring
- Negative lobes

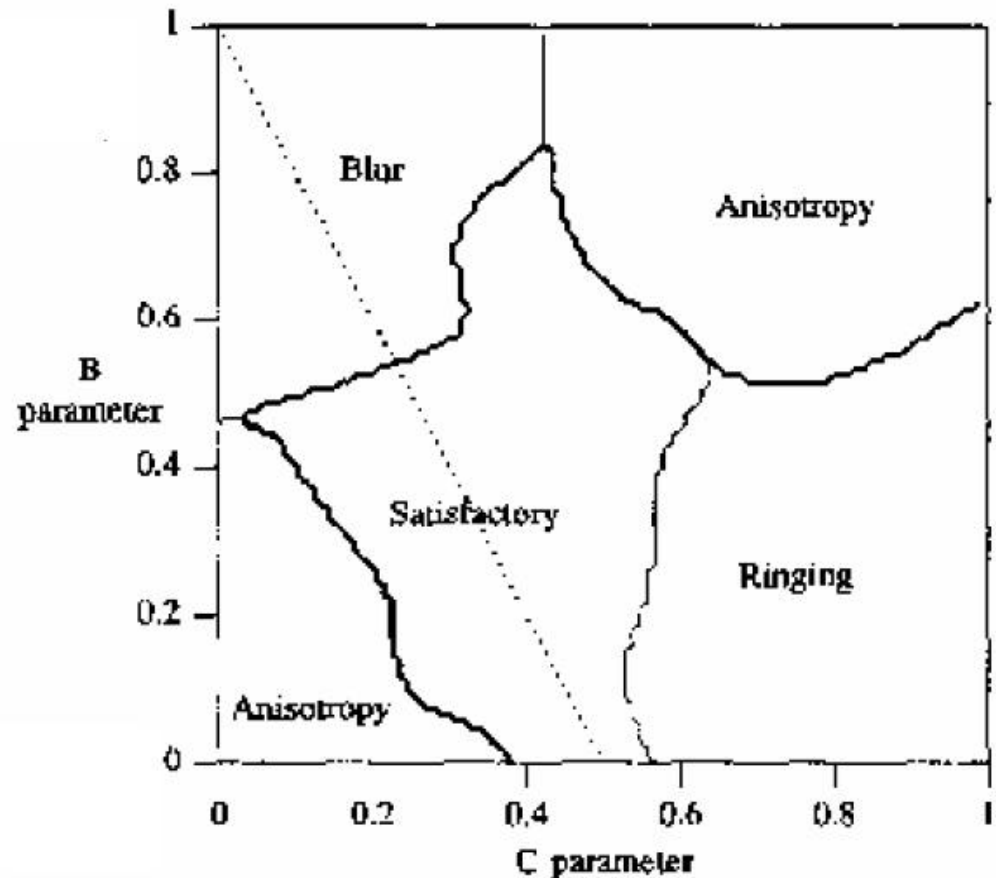




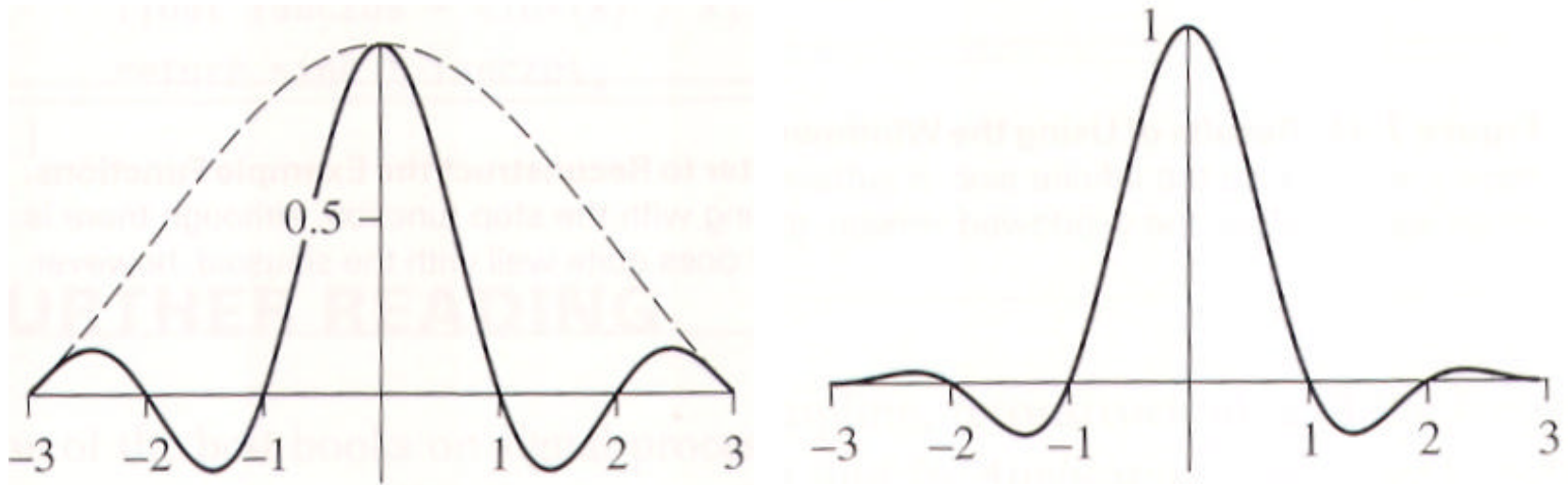
# Mitchell filter

$$h(x) = \frac{1}{6} \begin{cases} (12 - 9B - 6C)x^3 + (-18 + 12B + 6C)x^2 + (6 - 2B) & |x| < 1 \\ (-B - 6C)x^3 + (6B + 30C)x^2 + (-12B - 48C)x + (8B + 24C) & 1 < |x| < 2 \\ 0 & \text{otherwise} \end{cases}$$

- Separable filter
- Two parameters, B and C,  $B + 2C = 1$  suggested



# Windowed sinc filter



$$w(x) = \frac{\sin px / t}{px / t}$$



## References

- Yung-Yu Chuang, Image Synthesis, class slides, National Taiwan University, Fall 2005
- Rick Parent, 782: Advanced 3D Image Generation
- Pat Hanrahan, CS 348B, Spring 2005 class slides