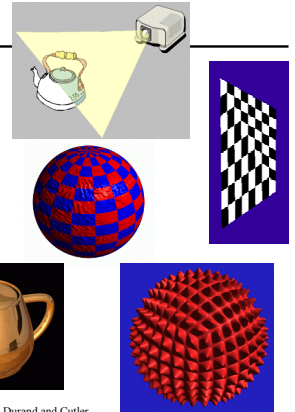


# Sampling, Aliasing, & Mipmaps

## Last Time?

- 2D Texture Mapping
- Perspective Correct Interpolation
- Common Texture Coordinate Projections
- Bump Mapping
- Displacement Mapping
- Environment Mapping



MIT EECS 6.837, Durand and Cutler

## Texture Maps for Illumination

- Also called "Light Maps"



Quake

MIT EECS 6.837, Durand and Cutler

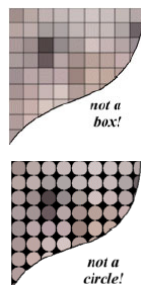
## Today

- **What is a Pixel?**
- Examples of Aliasing
- Signal Reconstruction
- Reconstruction Filters
- Anti-Aliasing for Texture Maps

MIT EECS 6.837, Durand and Cutler

## What is a Pixel?

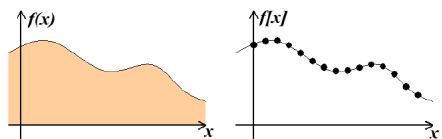
- A pixel is not:
  - a box
  - a disk
  - a teeny tiny little light
- A pixel is a point
  - it has no dimension
  - it occupies no area
  - it cannot be seen
  - it can have a coordinate
- A pixel is more than just a point, it is a sample!



MIT EECS 6.837, Durand and Cutler

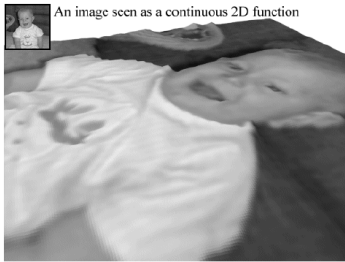
## More on Samples

- Most things in the real world are *continuous*, yet everything in a computer is *discrete*
- The process of mapping a continuous function to a discrete one is called *sampling*
- The process of mapping a continuous variable to a discrete one is called *quantization*
- To represent or render an image using a computer, we must both sample and quantize



## An Image is a 2D Function

- An *ideal image* is a function  $I(x,y)$  of intensities.
- It can be plotted as a height field.
- In general an image cannot be represented as a continuous, analytic function.
- Instead we represent images as tabulated functions.
- How do we fill this table?



MIT EECS 6.837, Durand and Cutler

## Sampling Grid

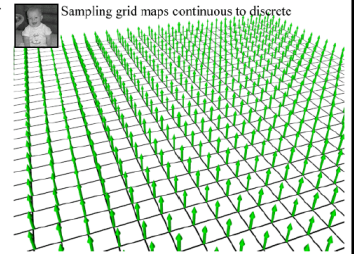
- We can generate the table values by multiplying the continuous image function by a sampling grid of Kronecker delta functions.

The definition of the 2-D Kronecker delta is:

$$\delta(x, y) = \begin{cases} 1, & (x, y) = (0, 0) \\ 0, & \text{otherwise} \end{cases}$$

And a 2-D sampling grid:

$$\sum_{i=0}^{h-1} \sum_{j=0}^{w-1} \delta(i - x, j - y)$$

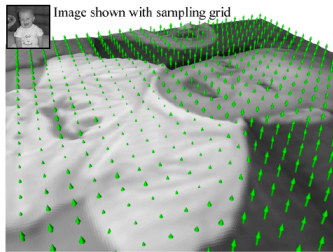
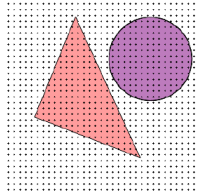


MIT EECS 6.837, Durand and Cutler

## Sampling an Image

- The result is a set of point samples, or pixels.

The same analysis can be applied to geometric objects:



MIT EECS 6.837, Durand and Cutler

## Questions?

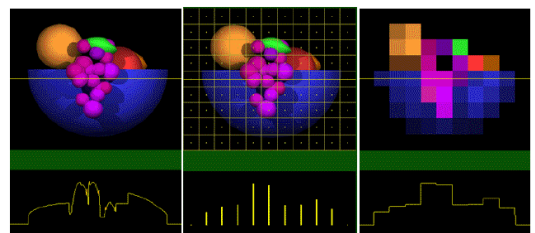
MIT EECS 6.837, Durand and Cutler

## Today

- What is a Pixel?
- **Examples of Aliasing**
- Signal Reconstruction
- Reconstruction Filters
- Anti-Aliasing for Texture Maps

MIT EECS 6.837, Durand and Cutler

## Examples of Aliasing



Original Image

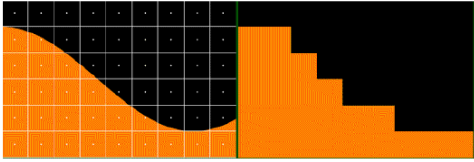
Samples

Reconstruction

MIT EECS 6.837, Durand and Cutler

## Examples of Aliasing

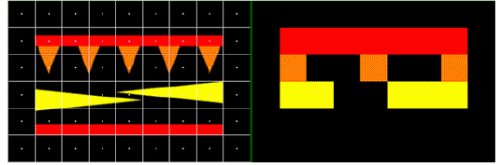
### Jagged boundaries



MIT EECS 6.837, Durand and Cutler

## Examples of Aliasing

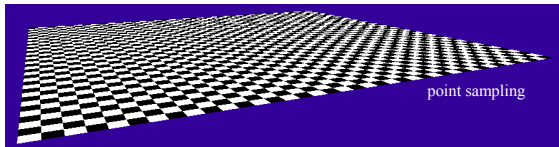
### Improperly rendered detail



MIT EECS 6.837, Durand and Cutler

## Examples of Aliasing

### Texture Errors



MIT EECS 6.837, Durand and Cutler

## Questions?

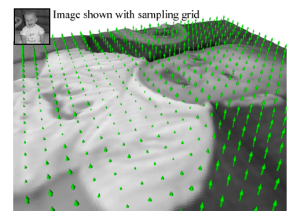
## Today

- What is a Pixel?
- Examples of Aliasing
- **Signal Reconstruction**
  - Sampling Density
  - Fourier Analysis & Convolution
- Reconstruction Filters
- Anti-Aliasing for Texture Maps

MIT EECS 6.837, Durand and Cutler

## Sampling Density

- How densely must we sample an image in order to capture its essence?
- If we under-sample the signal, we won't be able to accurately reconstruct it...

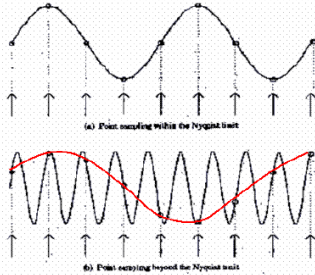


MIT EECS 6.837, Durand and Cutler

## Nyquist Limit / Shannon's Sampling Theorem

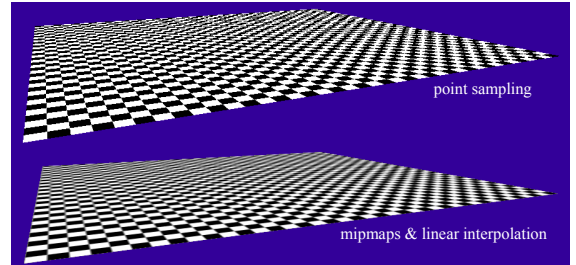
- If we insufficiently sample the signal, it may be mistaken for something simpler during reconstruction (that's aliasing!)

Image from Robert L. Cook, "Stochastic Sampling and Distributed Ray Tracing", An Introduction to Ray Tracing, Andrew Glassner, ed., Academic Press Limited, 1989.



## Examples of Aliasing

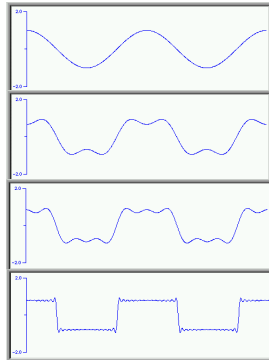
### Texture Errors



MIT EECS 6.837, Durand and Cutler

## Remember Fourier Analysis?

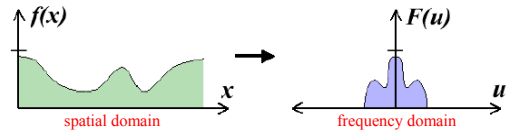
- All periodic signals can be represented as a summation of sinusoidal waves.



Images from <http://axion.physics.ubc.ca/341-02/fourier/fourier.html>

## Remember Fourier Analysis?

- Every periodic signal in the *spatial domain* has a dual in the *frequency domain*.



- This particular signal is *band-limited*, meaning it has no frequencies above some threshold

MIT EECS 6.837, Durand and Cutler

## Remember Fourier Analysis?

- We can transform from one domain to the other using the Fourier Transform.

frequency domain      spatial domain

Fourier Transform

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-i2\pi(ux+vy)} dx dy$$

Inverse Fourier Transform

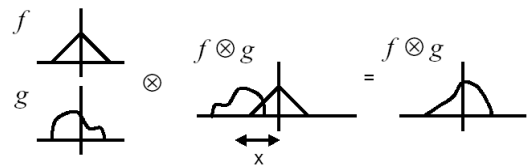
$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{i2\pi(ux+vy)} du dv$$

MIT EECS 6.837, Durand and Cutler

## Remember Convolution?

Convolution describes how a system with impulse response,  $h(x)$ , reacts to a signal,  $f(x)$ .

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



CS174 Fall 99 Lecture 7

Copyright © Mark Meyer

Images from Mark Meyer <http://www.gg.caltech.edu/~cs174ta/>

## Remember Convolution?

- Some operations that are difficult to compute in the spatial domain can be simplified by transforming to its dual representation in the frequency domain.
- For example, convolution in the spatial domain is the same as multiplication in the frequency domain.

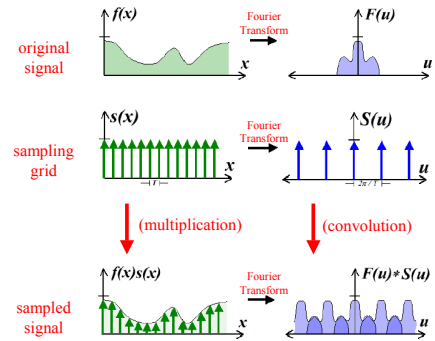
$$f(x) * h(x) \rightarrow F(u)H(u)$$

- And, convolution in the frequency domain is the same as multiplication in the spatial domain

$$F(u) * H(u) \rightarrow f(x)h(x)$$

MIT EECS 6.837, Durand and Cutler

## Sampling in the Frequency Domain

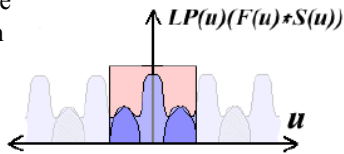


MIT EECS 6.837, Durand and Cutler

## Reconstruction

- If we can extract a copy of the original signal from the frequency domain of the sampled signal, we can reconstruct the original signal!

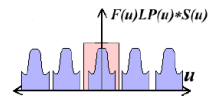
- But there may be overlap between the copies.



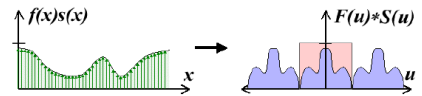
MIT EECS 6.837, Durand and Cutler

## Guaranteeing Proper Reconstruction

- Separate by removing high frequencies from the original signal (low pass pre-filtering)



- Separate by increasing the sampling density



- If we can't separate the copies, we will have overlapping frequency spectrum during reconstruction  $\rightarrow$  aliasing.

MIT EECS 6.837, Durand and Cutler

## Questions?

MIT EECS 6.837, Durand and Cutler

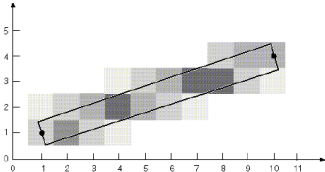
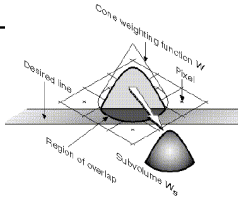
## Today

- What is a Pixel?
- Examples of Aliasing
- Signal Reconstruction
- **Reconstruction Filters**
  - Pre-Filtering, Post-Filtering
  - Ideal, Gaussian, Box, Bilinear, Bicubic
- Anti-Aliasing for Texture Maps

MIT EECS 6.837, Durand and Cutler

## Pre-Filtering

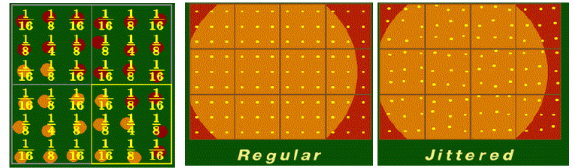
- Filter continuous primitives
- Treat a pixel as an area
- Compute weighted amount of object overlap
- What weighting function should we use?



Source: Foley, VanDam, Fetsner, Hughes - Computer Graphics, Second Edition, Addison Wesley

## Post-Filtering

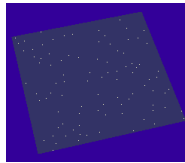
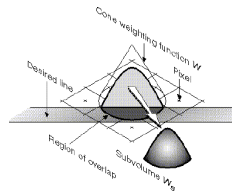
- Filter samples
- Compute the weighted average of many samples
- Regular or jittered sampling (better)



MIT EECS 6.837, Durand and Cutler

## Reconstruction Filters

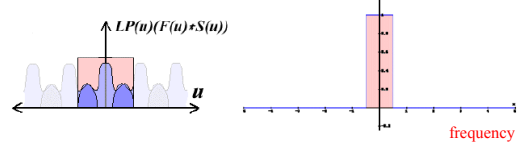
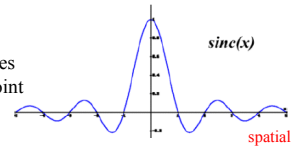
- Weighting function
- Area of influence often bigger than "pixel"
- Sum of weights = 1
  - Each pixel contributes the same total to image
  - Constant brightness as object moves across the screen.
- No negative weights/colors (optional)



MIT EECS 6.837, Durand and Cutler

## The Ideal Reconstruction Filter

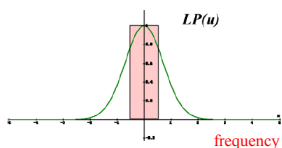
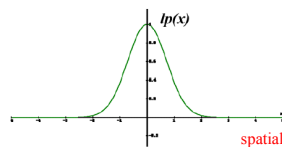
- Unfortunately it has *infinite* spatial extent
  - Every sample contributes to every interpolated point
- Expensive/impossible to compute



MIT EECS 6.837, Durand and Cutler

## Gaussian Reconstruction Filter

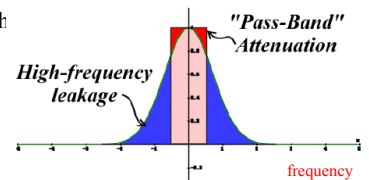
- This is what a CRT does for free!



MIT EECS 6.837, Durand and Cutler

## Problems with Reconstruction Filters

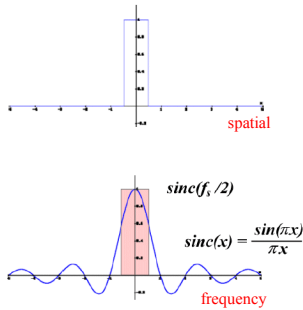
- Many visible artifacts in re-sampled images are caused by poor reconstruction filters
- Excessive pass-band attenuation results in blurry images
- Excessive high-frequency leakage causes "ringing" and can accentuate the sampling grid (anisotropy)



MIT EECS 6.837, Durand and Cutler

## Box Filter / Nearest Neighbor

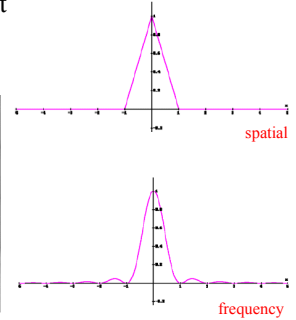
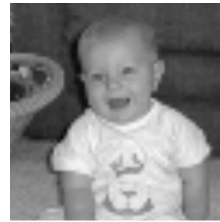
- Pretending pixels are little squares.



MIT EECS 6.837, Durand and Cutler

## Tent Filter / Bi-Linear Interpolation

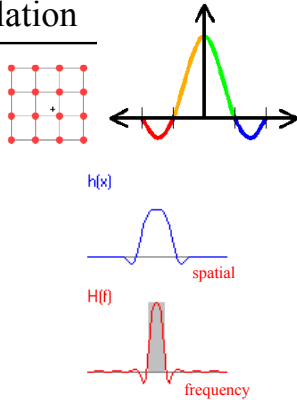
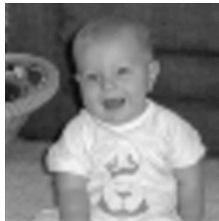
- Simple to implement
- Reasonably smooth



MIT EECS 6.837, Durand and Cutler

## Bi-Cubic Interpolation

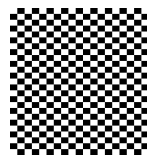
- Begins to approximate the ideal spatial filter, the sinc function



MIT EECS 6.837, Durand and Cutler

## Why is the Box filter bad?

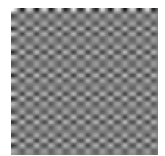
- (Why is it bad to think of pixels as squares)



Original high-resolution image



Down-sampled with a 5x5 box filter (uniform weights)



Down-sampled with a 5x5 Gaussian filter (non-uniform weights)

notice the ugly horizontal banding



MIT EECS 6.837, Durand and Cutler

## Questions?

MIT EECS 6.837, Durand and Cutler

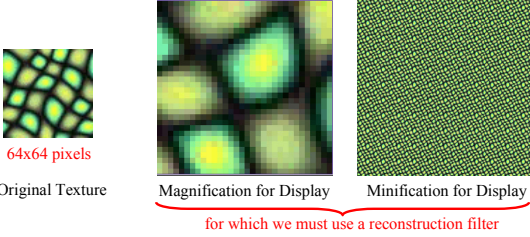
## Today

- What is a Pixel?
- Examples of Aliasing
- Signal Reconstruction
- Reconstruction Filters
- **Anti-Aliasing for Texture Maps**
  - Magnification & Minification
  - Mipmaps
  - Anisotropic Mipmaps

MIT EECS 6.837, Durand and Cutler

## Sampling Texture Maps

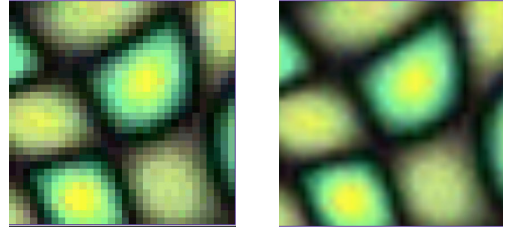
- When texture mapping it is rare that the screen-space sampling density matches the sampling density of the texture.



MIT EECS 6.837, Durand and Cutler

## Linear Interpolation

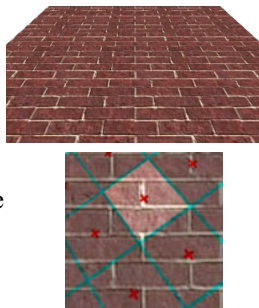
- Tell OpenGL to use a tent filter instead of a box filter.
- Magnification looks better, but blurry
  - (texture is under-sampled for this resolution)



MIT EECS 6.837, Durand and Cutler

## Spatial Filtering

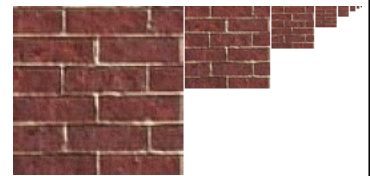
- Remove the high frequencies which cause artifacts in minification.
- Compute a spatial integration over the extent of the sample
- Expensive to do during rasterization, but it can be precomputed



MIT EECS 6.837, Durand and Cutler

## MIP Mapping

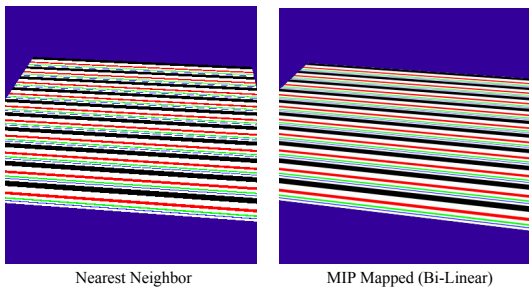
- Construct a pyramid of images that are pre-filtered and re-sampled at 1/2, 1/4, 1/8, etc., of the original image's sampling
- During rasterization we compute the index of the decimated image that is sampled at a rate closest to the density of our desired sampling rate
- MIP stands for *multum in parvo* which means *many in a small place*



MIT EECS 6.837, Durand and Cutler

## MIP Mapping Example

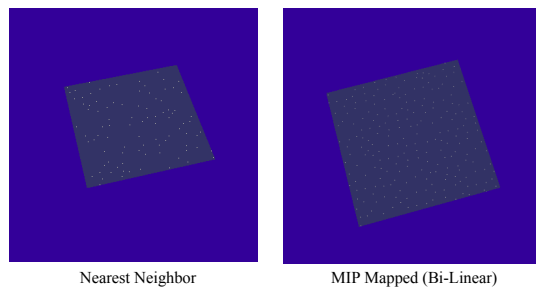
- Thin lines may become disconnected / disappear



MIT EECS 6.837, Durand and Cutler

## MIP Mapping Example

- Small details may "pop" in and out of view

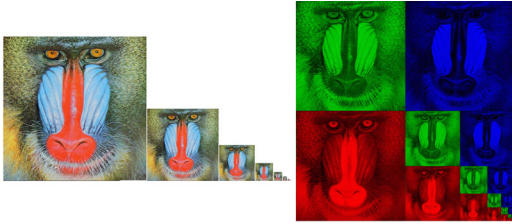


MIT EECS 6.837, Durand and Cutler



## Storing MIP Maps

- Can be stored compactly
- Illustrates the 1/3 overhead of maintaining the MIP map

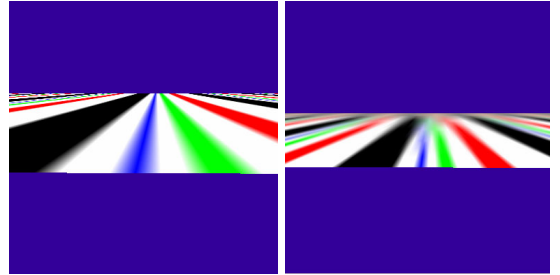


10-level mip map

Memory format of a mip map

## Anisotropic MIP-Mapping

- What happens when the surface is tilted?



Nearest Neighbor

MIP Mapped (Bi-Linear)

MIT EECS 6.837, Durand and Cutler

## Anisotropic MIP-Mapping

- We can use different mipmaps for the 2 directions
- Additional extensions can handle non axis-aligned views

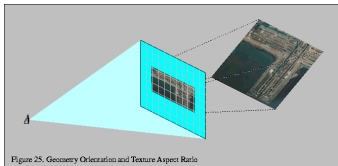


Figure 28. Geometry Orientation and Texture Aspect Ratio

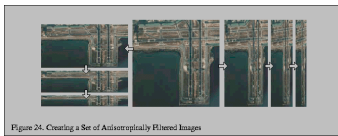


Figure 24. Creating a Set of Anisotropically Filtered Images

Images from <http://www.sgi.com/software/OpenGL/advanced98/notes/node37.html>

## Questions?

MIT EECS 6.837, Durand and Cutler

Next Time: Last Class!

Wrap Up &  
Final Project Review

MIT EECS 6.837, Durand and Cutler