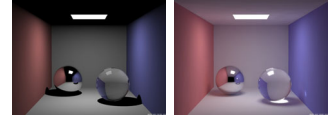


Sampling, Aliasing, & Mipmaps

Last Time?

- Global illumination
“physically accurate
light transport”



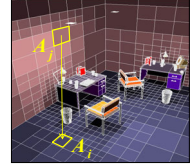
- The rendering equation

$$L(x', \omega') = E(x', \omega') + \int \rho_x(\omega, \omega') L(x, \omega) G(x, x') V(x, x') dA$$

- The discrete radiosity equation

$$B_i = E_i + \rho_i \sum_{j=1}^n F_{ij} B_j$$

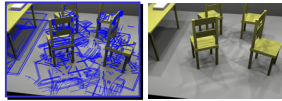
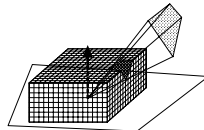
- Diffuse reflection only
- View independent



MIT EECS 6.837

Last Time?

- Form factors
 - F_{ij} = fraction of light energy leaving patch j that arrives at patch i
 - Hemicube algorithm
- Advanced techniques
 - Progressive radiosity
 - Adaptive subdivision
 - Discontinuity meshing
 - Hierarchical radiosity



MIT EECS 6.837

Today

- **What is a Pixel?**
- Examples of aliasing
- Sampling & Reconstruction
- Filters in Computer Graphics
- Anti-Aliasing for Texture Maps

MIT EECS 6.837

What is a Pixel?

- My research during for my PhD was on sampling & aliasing with point-sampled surfaces, i.e., 3D objects
- This lecture is about sampling images

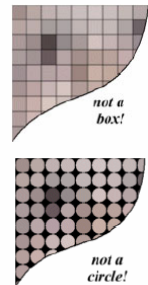


No triangles, just samples in 3D

MIT EECS 6.837

What is a Pixel?

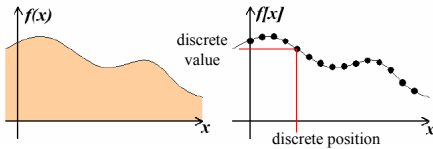
- A pixel is not:
 - a box
 - a disk
 - a teeny tiny little light
- A pixel “looks different” on different display devices
- A pixel is a sample
 - it has no dimension
 - it occupies no area
 - it cannot be seen
 - it has a coordinate
 - it has a value



MIT EECS 6.837

More on Samples

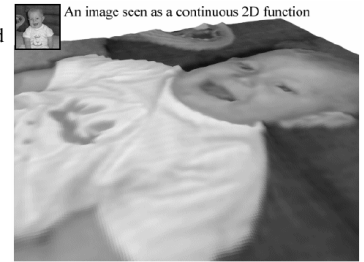
- Most things in the real world are *continuous*, yet everything in a computer is *discrete*
- The process of mapping a continuous function to a discrete one is called *sampling*
- The process of mapping a continuous variable to a discrete one is called *quantization*
- To represent or render an image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



MIT EECS 6.837

An Image is a 2D Function

- An *ideal image* is a continuous 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

Sampling Grid

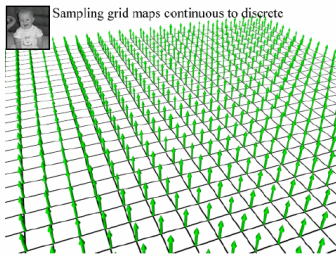
- We can generate the table values by multiplying the continuous image function by a sampling grid of Kronecker delta functions.

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:

$$\sum_{j=0}^{k-1} \sum_{i=0}^{w-1} \delta(u-i, v-j)$$

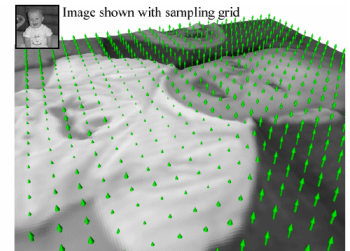
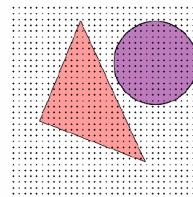


MIT EECS 6.837

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

Questions?

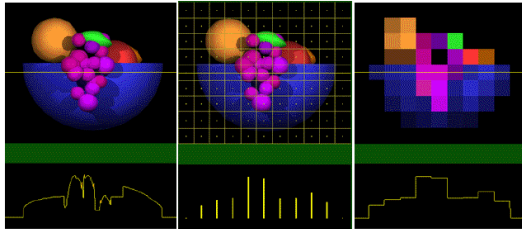
MIT EECS 6.837

Today

- What is a Pixel?
- **Examples of Aliasing**
- Sampling & Reconstruction
- Filters in Computer Graphics
- Anti-Aliasing for Texture Maps

MIT EECS 6.837

Examples of Aliasing



Original Image

Samples

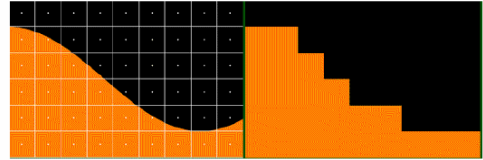
Reconstruction

- Aliasing occurs because of *sampling* and *reconstruction*

MIT EECS 6.837

Examples of Aliasing

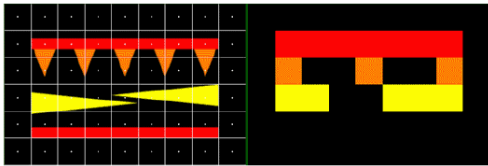
Jagged boundaries



MIT EECS 6.837

Examples of Aliasing

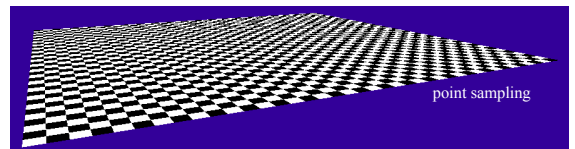
Improperly rendered detail



MIT EECS 6.837

Examples of Aliasing

Texture Errors



MIT EECS 6.837

Questions?

MIT EECS 6.837

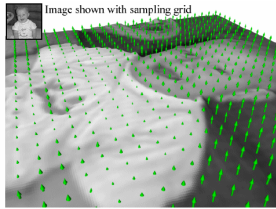
Today

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

MIT EECS 6.837

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...

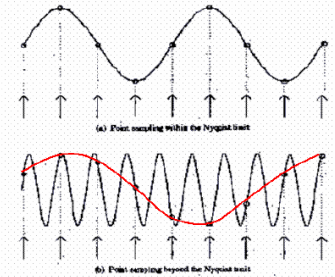


MIT EECS 6.837

Sampling Density

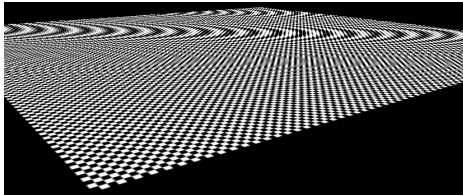
- If we insufficiently sample the signal, it may be mistaken for something simpler during reconstruction (that's aliasing!)

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



Sampling Density

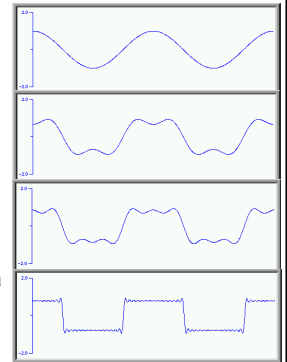
- Aliasing in 2D because of insufficient sampling density



MIT EECS 6.837

Remember Fourier Analysis?

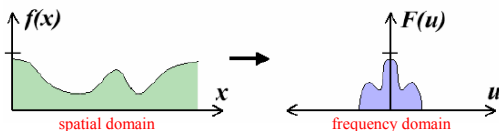
- All periodic signals can be represented as a summation of sinusoidal waves.



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

Remember Fourier Analysis?

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

$$\text{Fourier Transform} \quad \begin{matrix} \text{frequency domain} & \text{spatial domain} \\ \downarrow & \uparrow \\ F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-i2\pi(ux+vy)} dx dy \end{matrix}$$

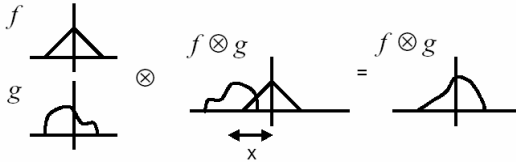
$$\text{Inverse Fourier Transform} \quad f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{i2\pi(ux+vy)} du dv$$

MIT EECS 6.837

Remember Convolution?

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

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



CS174F:6.837 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.

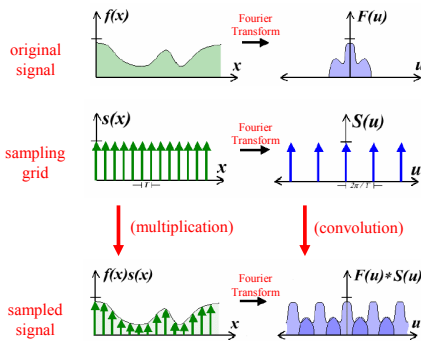
$$f(x) * h(x) \rightarrow F(u)H(u)$$

- 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

Sampling in the Frequency Domain

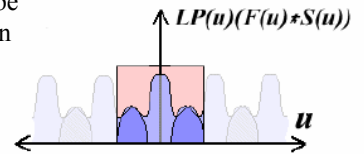


MIT EECS 6.837

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!

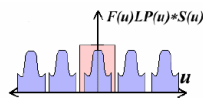
- But there may be overlap between the copies.



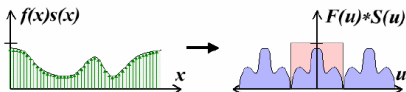
MIT EECS 6.837

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 → *aliasing*.

MIT EECS 6.837

Sampling Theorem

- When sampling a signal at discrete intervals, the sampling frequency must be *greater than twice* the highest frequency of the input signal in order to be able to reconstruct the original perfectly from the sampled version (Shannon, Nyquist)

MIT EECS 6.837

Questions?

MIT EECS 6.837

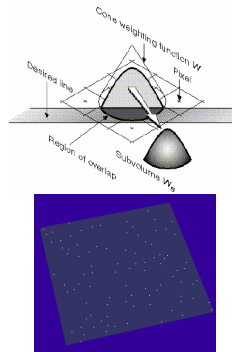
Today

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

MIT EECS 6.837

Filters

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



MIT EECS 6.837

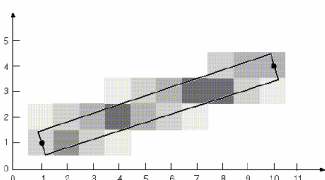
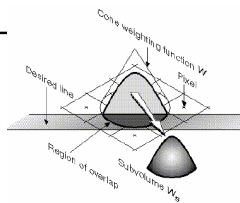
Filters

- Filters are used to
 - reconstruct a continuous signal from a sampled signal (reconstruction filters)
 - band-limit continuous signals to avoid aliasing during sampling (low-pass filters)
- Desired frequency domain properties are the same for both types of filters
- Often, the same filters are used as reconstruction and low-pass filters

MIT EECS 6.837

Pre-Filtering

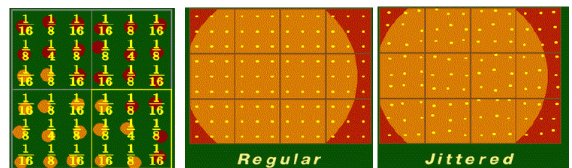
- Filter continuous primitives
- Treat a pixel as an area
- Compute weighted amount of object overlap
- What weighting function should we use?



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

Post-Filtering

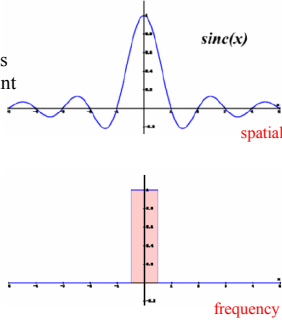
- Filter samples
- Compute the weighted average of many samples
- Regular or jittered sampling (better)



MIT EECS 6.837

The Ideal Filter

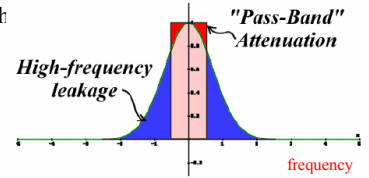
- Unfortunately it has *infinite* spatial extent
 - Every sample contributes to every interpolated point
- Expensive/impossible to compute



MIT EECS 6.837

Problems with Practical Filters

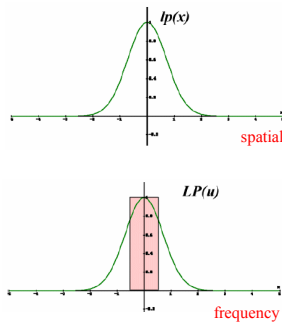
- Many visible artifacts in re-sampled images are caused by poor reconstruction filters
- Excessive pass-band attenuation results in blurry images
- Excessive high-frequency leakage causes "ringing" and can accentuate the sampling grid (anisotropy)



MIT EECS 6.837

Gaussian Filter

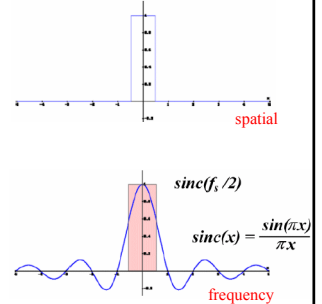
- This is what a CRT does for free!



MIT EECS 6.837

Box Filter / Nearest Neighbor

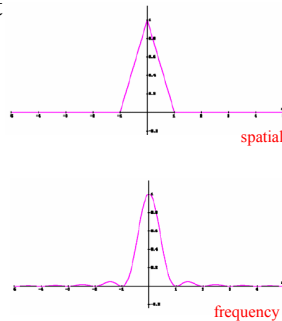
- Pretending pixels are little squares.



MIT EECS 6.837

Tent Filter / Bi-Linear Interpolation

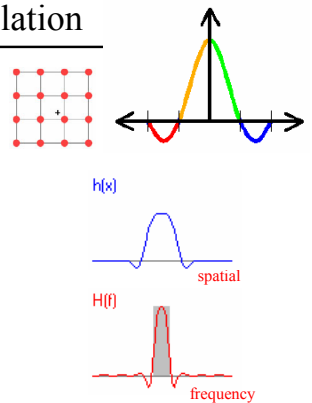
- Simple to implement
- Reasonably smooth



MIT EECS 6.837

Bi-Cubic Interpolation

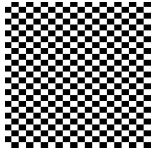
- Begins to approximate the ideal spatial filter, the sinc function



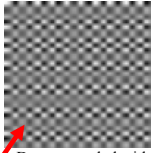
MIT EECS 6.837

Why is the Box filter bad?

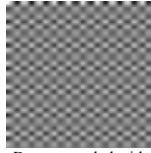
- (Why is it bad to think of pixels as squares)



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

1/16	1/8	1/16
1/8	1/4	1/8
1/16	1/8	1/16

MIT EECS 6.837

Questions?

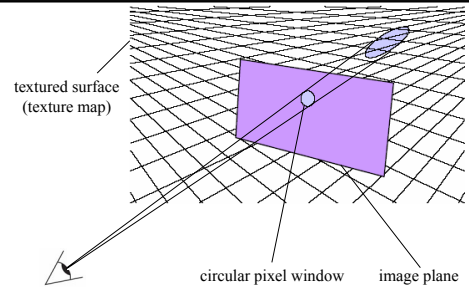
MIT EECS 6.837

Today

- What is a Pixel?
- Examples of Aliasing
- Sampling & Reconstruction
- Filters in Computer Graphics
- Anti-Aliasing for Texture Maps
 - Magnification & Minification
 - Mipmaps
 - Anisotropic Mipmaps

MIT EECS 6.837

Sampling Texture Maps



- How to map the texture area seen through the pixel window to a single pixel value?

MIT EECS 6.837

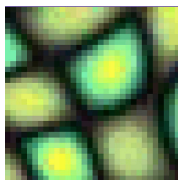
Sampling Texture Maps

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

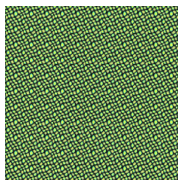


64x64 pixels

Original Texture



Magnification for Display



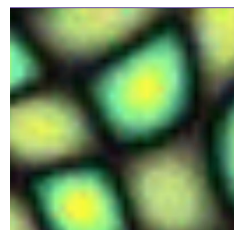
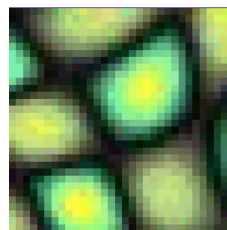
Minification for Display

for which we must use a reconstruction filter

MIT EECS 6.837

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

Spatial Filtering

- Remove the high frequencies which cause artifacts in texture minification.
- Compute a spatial integration over the extent of the pixel
- This is equivalent to convolving the texture with a filter kernel centered at the sample (i.e., pixel center)!
- Expensive to do during rasterization, but an approximation it can be precomputed



projected texture in image plane

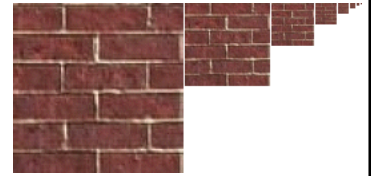


box filter in texture plane

MIT EECS 6.837

MIP Mapping

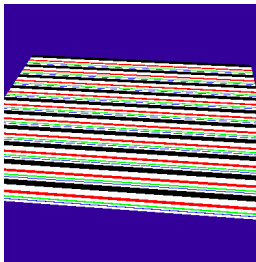
- Construct a pyramid of images that are pre-filtered and re-sampled at $1/2$, $1/4$, $1/8$, etc., of the original image's sampling
- During rasterization we compute the index of the decimated image that is sampled at a rate closest to the density of our desired sampling rate
- MIP stands for *multum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



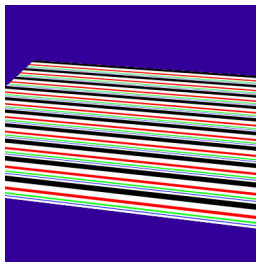
MIT EECS 6.837

MIP Mapping Example

- Thin lines may become disconnected / disappear



Nearest Neighbor

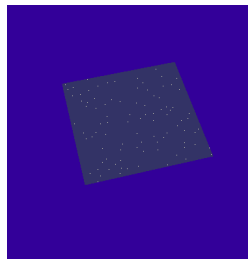


MIP Mapped (Bi-Linear)

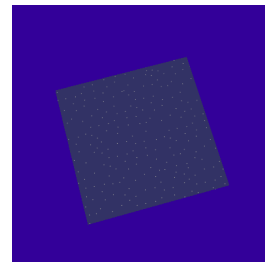
MIT EECS 6.837

MIP Mapping Example

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



Nearest Neighbor

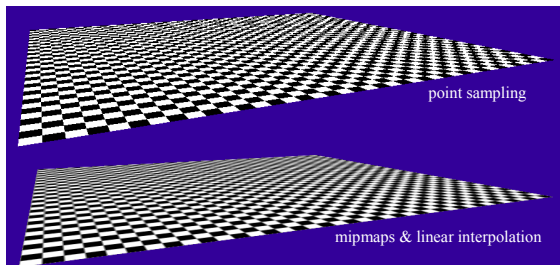


MIP Mapped (Bi-Linear)

MIT EECS 6.837

Examples of Aliasing

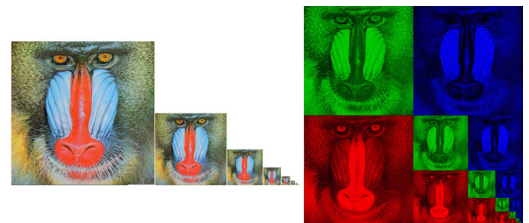
Texture Errors



MIT EECS 6.837

Storing MIP Maps

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

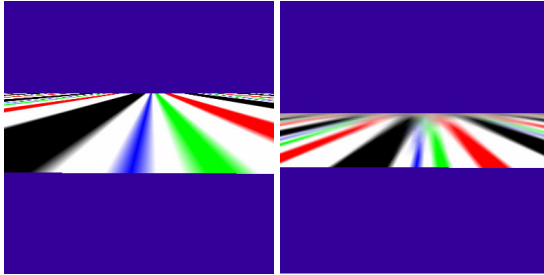


10-level mip map

Memory format of a mip map

Anisotropic MIP-Mapping

- What happens when the surface is tilted?

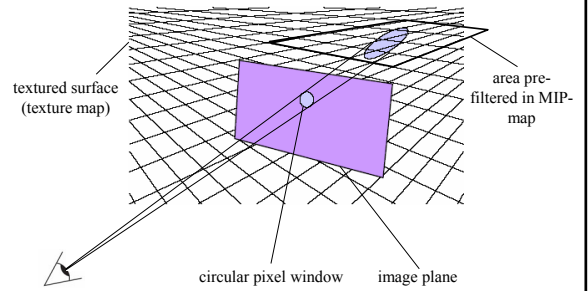


Nearest Neighbor

MIP Mapped (Bi-Linear)

MIT EECS 6.837

Anisotropic MIP-Mapping



- Square MIP-map area is a bad approximation

MIT EECS 6.837

Anisotropic MIP-Mapping

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

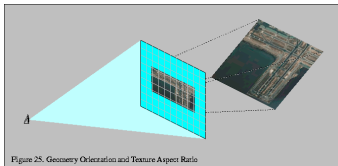


Figure 25. Geometry Orientation and Texture Aspect Ratio

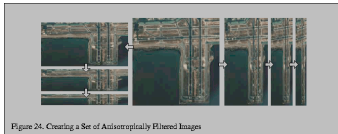


Figure 24. Creating a Set of Anisotropically Filtered Images

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

MIT EECS 6.837

Questions?

Next Time:

Supersampling & Basic Monte Carlo Techniques

MIT EECS 6.837