



Sampling & Reconstruction

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Course News

Assignment 3 (project)

- Due April 1

Homework

- This week: H6 – texture mapping
- Next week: no new homework (focus on quiz prep)

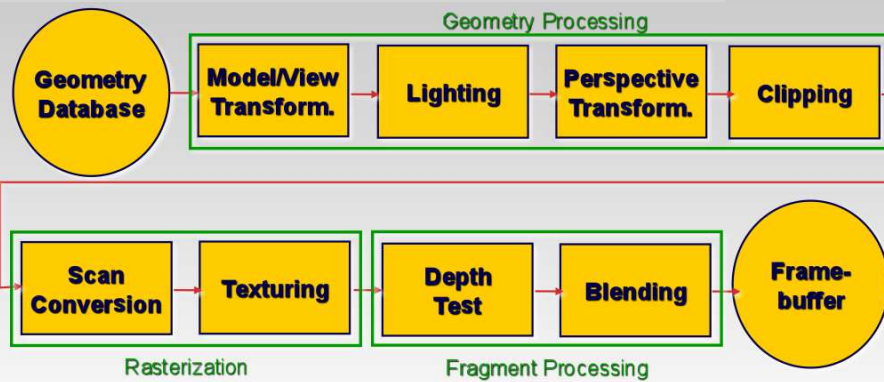
Quiz 2

- Friday, March 13
- Topics:
 - *Everything after transformations up to and including this lecture*
 - *Questions on rendering pipeline as a whole*

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The Rendering Pipeline



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Samples

- Most things in the real world are **continuous**
- 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 discrete function to a continuous one is called **reconstruction**
- The process of mapping a continuous variable to a discrete one is called **quantization**
- Rendering an image requires sampling and quantization
- Displaying an image involves reconstruction

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Images

An image is a 2D function $I(x, y)$

- Specifies intensity for each point (x, y)
- (we consider each color channel independently)

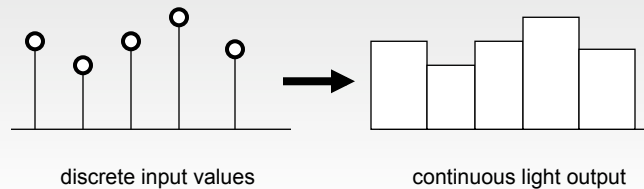


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Image Sampling and Reconstruction

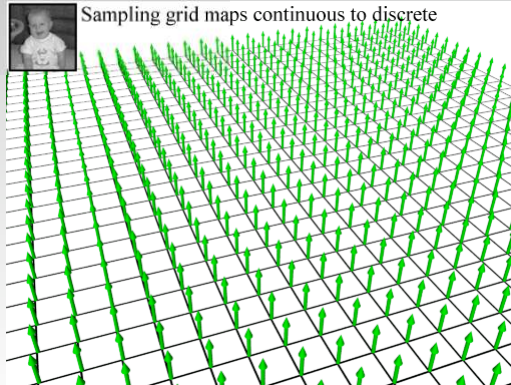
- Convert **continuous** image to **discrete** set of samples
- Display hardware **reconstructs** samples into continuous image
- *Finite sized source of light for each pixel*



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Point Sampling an Image

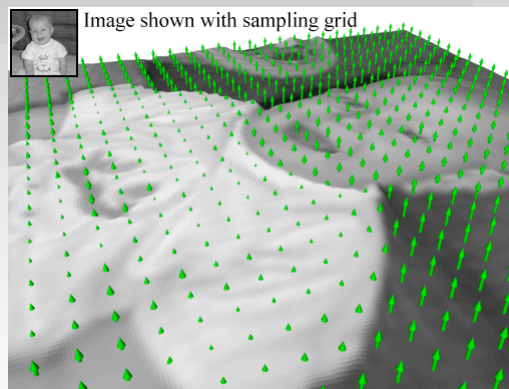
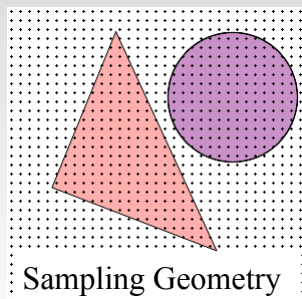
- Simplest sampling is on a grid
- Sample depends solely on value at grid points



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Point Sampling

Multiply sample grid by image intensity to obtain a discrete set of points, or samples.



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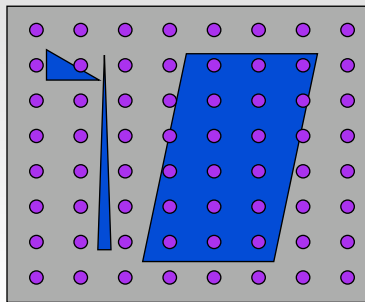


Sampling Errors

Some objects missed entirely, others poorly sampled

- Could try unweighted or weighted area sampling
- But how can we be sure we show everything?

Need to think about entire class of solutions!



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Image As Signal

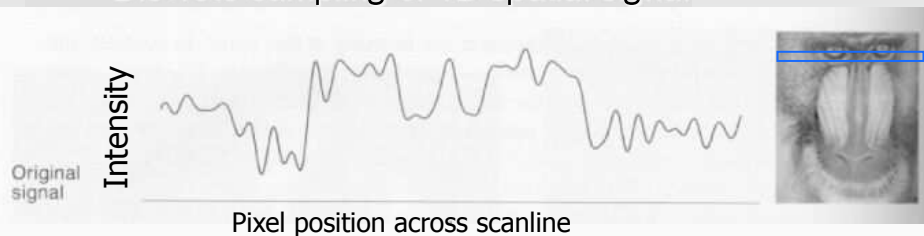
Image as spatial signal

2D raster image

- Discrete sampling of 2D spatial signal

1D slice of raster image

- Discrete sampling of 1D spatial signal



Examples from Foley, van Dam, Feiner, and Hughes

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Sampling Theory

How would we generate a signal like this out of simple building blocks?

Theorem

- Any signal can be represented as an (infinite) sum of sine waves at different frequencies

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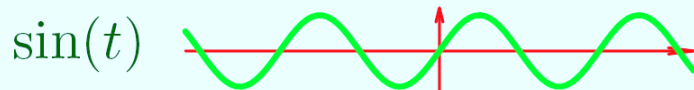
Sampling Theory in a Nutshell

Terminology

- Wavelength – length of repeated sequence on infinite signal
- Frequency – $1/\text{wavelength}$ (number of repeated sequences in unit length)

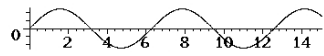
Example – sine wave

- Wavelength = 2π
- Frequency = $1/2\pi$

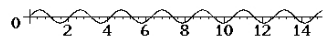


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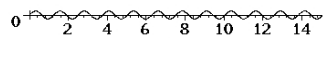
Fourier Transform – Summing Sinusoids



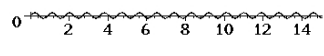
$\sin(x)$



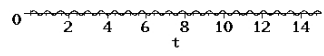
$+\sin(3x)/3$



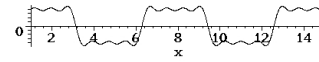
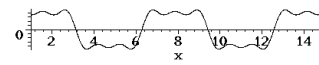
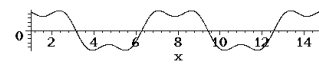
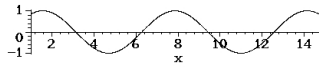
$+\sin(5x)/5$



$+\sin(7x)/7$

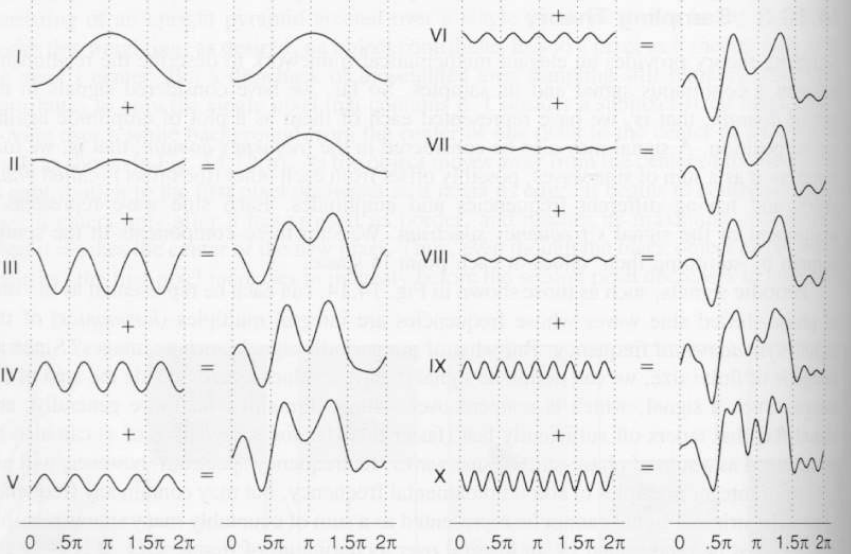


$+\sin(9x)/9$



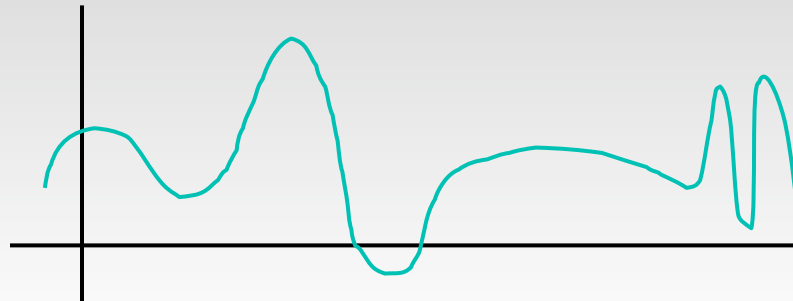
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Fourier Transform





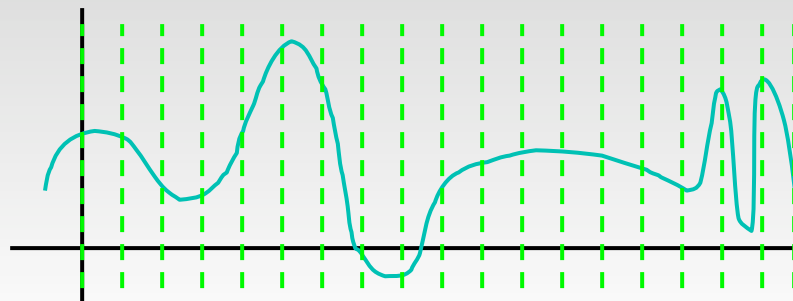
1D Sampling and Reconstruction



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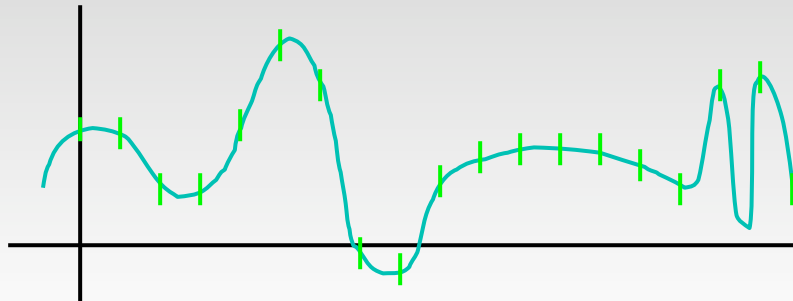
1D Sampling and Reconstruction



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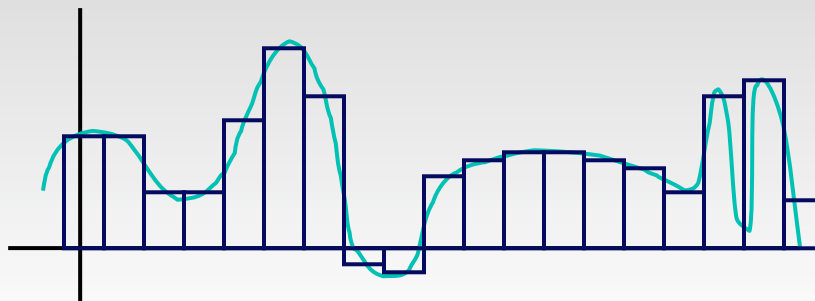
1D Sampling and Reconstruction



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1D Sampling and Reconstruction



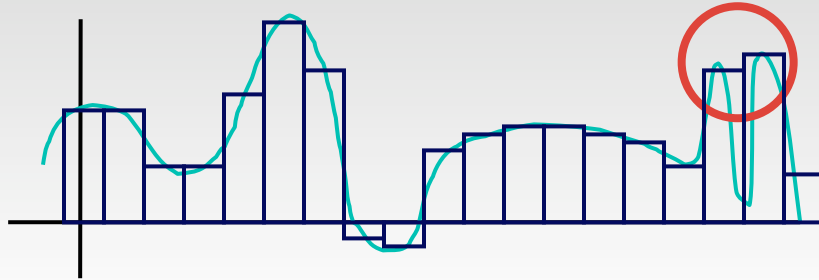
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1D Sampling and Reconstruction

Problems

- Jaggies – abrupt changes
- Lose data



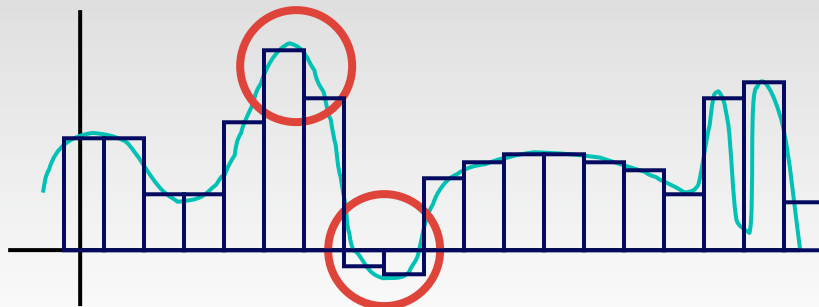
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1D Sampling and Reconstruction

Problems

- Jaggies – abrupt changes



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Sampling Theorem

- Continuous signal can be completely recovered from its samples

Iff

- Sampling rate greater than twice highest frequency present in signal

- Claude Shannon

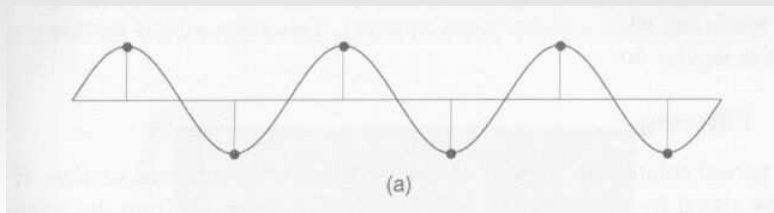
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Nyquist Rate

Lower bound on sampling rate

- Twice the highest frequency component in the image's spectrum



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Falling Below Nyquist Rate

When sampling below Nyquist Rate, resulting signal looks like a lower-frequency one

- This is **aliasing!**

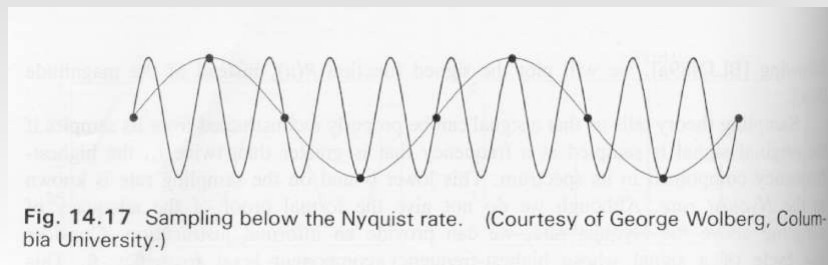
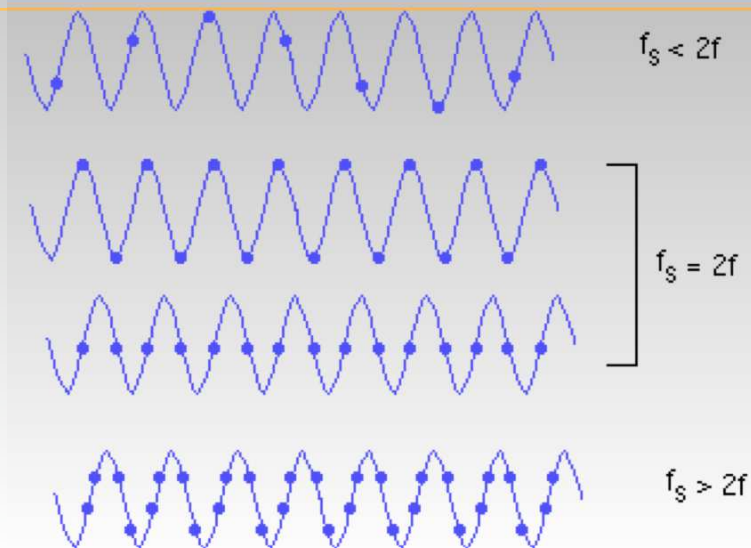


Fig. 14.17 Sampling below the Nyquist rate. (Courtesy of George Wolberg, Columbia University.)

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Nyquist Rate



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Aliasing

Incorrect appearance of high frequencies as low frequencies

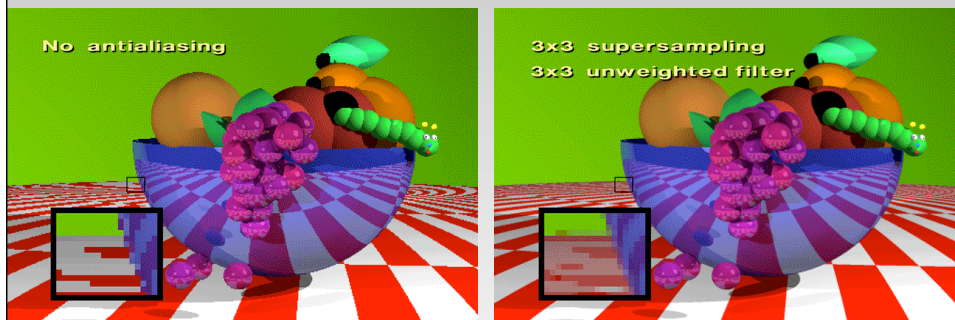
To avoid: anti-aliasing

- Supersample
 - Sample at higher frequency
- Low pass filtering
 - Remove high frequency function parts
 - Aka prefiltering, band-limiting

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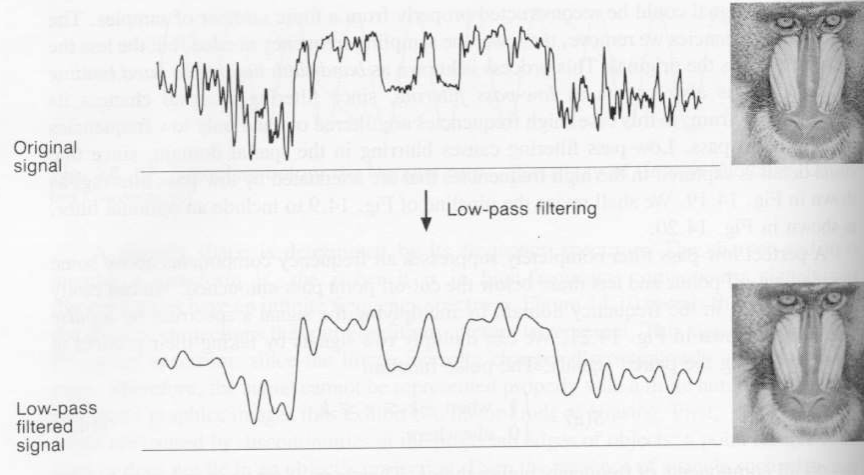
Supersampling



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Low-Pass Filtering



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Low-Pass Filtering

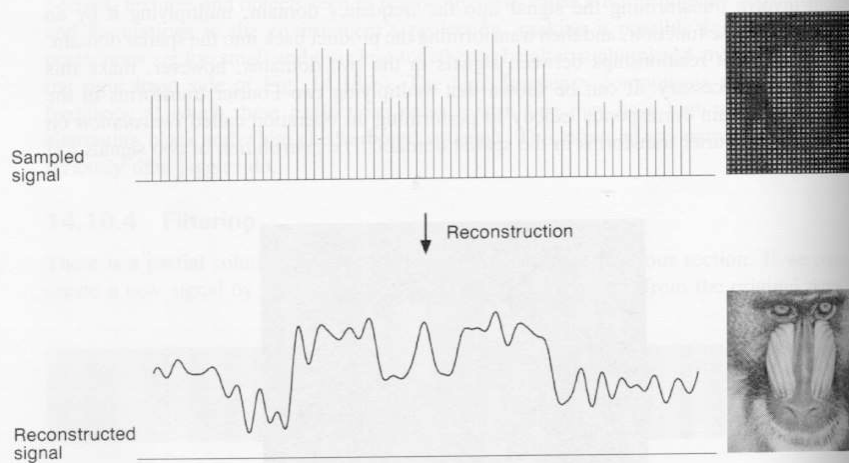


Fig. 14.20 The sampling pipeline with filtering. (Courtesy of George Wolberg, Columbia University.)

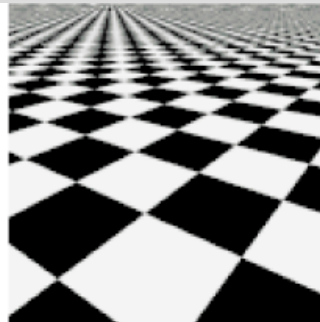


Previous Antialiasing Example

Texture mipmapping: low pass filter



(a)



(b)

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Discussion

Sampling & Reconstruction

- Fundamental issue in graphics, vision, and many other areas of computer science
 - *Whenever continuous signals need to be represented in a computer*
- Aliasing refers to the problem of reconstruction errors due to frequencies above the Nyquist limit
 - *These frequencies show up as erroneous low frequency content*

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Discussion

Anti-Aliasing Approaches

- Low-pass filtering (**before** sampling!)
 - *Avoids aliasing*
 - *May not be practical in all settings*
 - *For images: artifacts around edges?!*
- Supersampling
 - *General algorithmic approach*
 - *Hoever: even the higher resolution image has a Nyquist limit!*
 - *Slow*

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Coming Up:

Monday

- Programmable GPU architectures (Gordon Wetzstein)

Wednesday

- Shadows

Friday

- Quiz 2

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