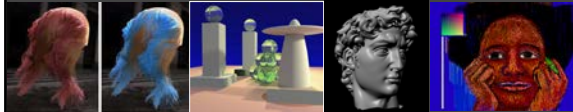


## Advanced Computer Graphics

CSE 163 [Spring 2018], Lecture 2

Ravi Ramamoorthi

<http://www.cs.ucsd.edu/~ravr>

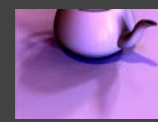
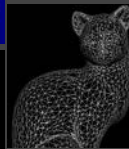


## Course Outline

- 3D Graphics Pipeline

**Modeling**  
(Creating 3D Geometry)

**Rendering**  
(Creating, shading images from  
geometry, lighting, materials)



## Course Outline

- 3D Graphics Pipeline

**Modeling**  
(Creating 3D Geometry)

**Rendering**  
(Creating, shading images from  
geometry, lighting, materials)

Unit 1: Foundations of Signal and Image Processing

Understanding the way 2D images are formed and displayed, the important concepts and algorithms, and to build an image processing utility like Photoshop  
Weeks 1 – 3. **Assignment 1**

## To Do

- Assignment 1, Due Apr 27.
  - Anyone need help finding partners?
  - Should already have downloaded code, skimmed assn
  - After today, enough to finish 3.2, 3.3 (first half)
  - Should START EARLY (this week) on assn
  - Second half next week.
- Class participation, discussion important
  - If you have to miss a class, see podcast if available
- Please sign up for Piazza

## Outline

- Intensity and Color (briefly)**
  - Basic operations (3.2 in assignment [10 points])
- Quantization, Halftoning and Dithering
  - (3.3 in assignment [10 points])
- Next week: Sampling and Reconstruction
  - Including signal processing and fourier analysis
  - Implementation of simple digital filters, resizing
  - Second half of assignment
- Lectures main source; will also try handout

## Intensities: Human Perception

- Human eye can perceive wide range of intensities
  - Dimly lit darkened room to bright sunlight
  - Radiance ratio in these cases is a million to one or more
- How does it work? [image only 256 gray levels]
  - Nonlinear human response  $S = I^p$   $p \approx .33$ 
    - Care about **ratio** of intensities (log scale). So jump from 0.1 to 0.11 as important as 0.50 to 0.55 (*not* .5 to .51)
    - E.g.: cycle through 50W, 100W, 150W (step from 50 to 100 much greater than from 100 to 150)
- Technically, equispaced intensities multiplicative
  - 0.02, 0.0203, 0.0206, ... 0.9848, 1.000 [for 100 values]
- Area of CG known as tonemapping (we ignore)

## Gamma Correction

- Website: <http://graphics.stanford.edu/gamma.html>
- Practical problem: Images look too dark/bright...



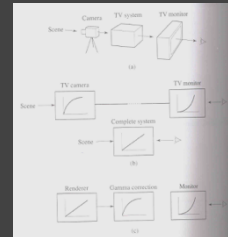
## Gamma Correction

- Monitors were CRT displays with nonlinear resp.

$$I = aV^\gamma \quad V = \left(\frac{I}{a}\right)^{1/\gamma}$$

$\gamma=2.5+$

- NTSC, use 2.2 (camera pre-corrected)
- Rendering linear (physical space) **Gamma Correct**



Watt Page 440

## Example

- Say RGB is something like (1, 0.5, 0)
- Values of 1 and 0 don't change (black, white, primary colors unaffected by gamma correction)
- Value of .5 becomes .707 (power of  $\frac{1}{2}$ , gamma = 2)
- Final color is (1, 0.707, 0) [brighter, less saturated]

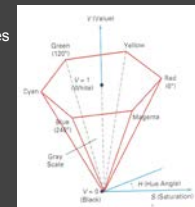


## Color

- Already seen: RGB model (color cube)
- Today: A very brief overview of real story
- Intuitive specify: Hue, Saturation, Lightness
  - Hexacone
  - Can convert HSV to RGB
  - Many other fancy, perceptual spaces

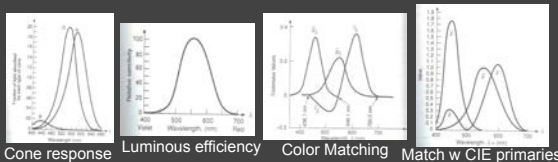


Figure 15.4.2  
Two views of the RGB color cube: (a) along the grayscale diagonal from white to black and (b) along the grayscale diagonal from black to white.



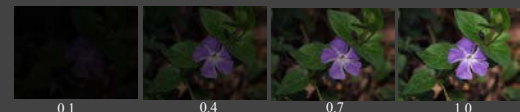
## Color: Tristimulus Theory

- Perception: Tri-stimulus theory
  - 3 types of cones: basis for RGB
  - Cone response functions
  - Luminous efficiency ( $G > R > B$ )
  - Color matching: Note "negative colors"
  - CIE overview



## Basic Image Processing (HW 1: 3.2)

- Brightness: Simply scale pixel RGB values (1 leaves image intact, 0 makes it black)



- Gamma Correction
- Crop (integer coords) to focus on important aspects

## Basic Image Processing (HW 1: 3.2)

- Contrast [0 is constant grey image, 1 is original]
  - Find constant grey image by averaging
  - Interpolate between this and original



## Basic Image Processing (HW 1: 3.2)

- Saturation [0 is greyscale, 1 is original colors]
  - Interpolate between grayscale (but not const) and orig.
  - Negative values correspond to inverting hues [negative]



## Outline

- Intensity and Color (briefly)
  - Basic operations (3.2 in assignment [10 points])
- Quantization, Halftoning and Dithering
  - (3.3 in assignment [10 points])
- Next week: Sampling and Reconstruction
  - Including signal processing and fourier analysis
  - Implementation of simple digital filters, resizing
  - Second half of assignment

## Images and Resolution

- Image is a 2D rectilinear discrete array of samples
- There are resolution issues:
  - Intensity resolution: Each pixel has only Depth bits
  - Spatial resolution: Image is only width\*height pixels
  - Temporal resolution: Monitor refreshes only at some rate

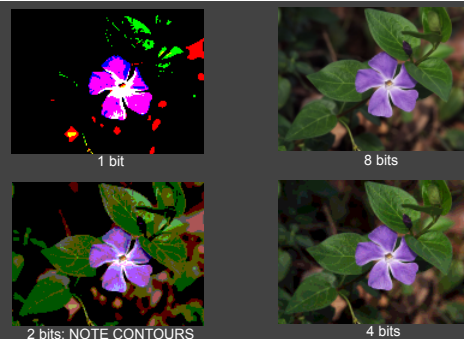
NTSC	640x480	8 bits	30 Hz
PC	1280x1024	24 bits RGB	75 Hz
Film	3000x2000	12 bits	24 Hz
Laser Printer	7000x2000	1 (on or off)	

Some material for slides courtesy Greg Humphreys and Tom Funkhouser

## Sources of Error or Artifacts

- Quantization: Not enough intensity resolution (bits)
  - Halftoning/dithering: Reduce visual artifacts due to quantization*
- Spatial and Temporal Aliasing: not enough resolution
  - Sampling and reconstruction to reduce visual artifacts due to aliasing (next week)

## Uniform Quantization



## Uniform Quantization



2 bits: NOTE CONTOURS

## Reducing Quantization

- Halftoning
- Dithering
  - Random Dither
  - Error Diffusion (Floyd-Steinberg)

## Halftoning

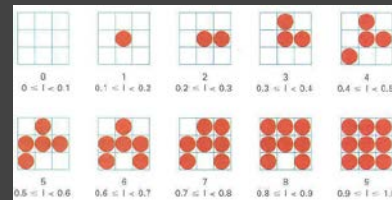
- Motivation: bilevel printing. Trade off spatial resolution for more intensity levels
- Dots of appropriate size to simulate grey levels
- Area of dots proportional intensity



Figure 14-34  
An enlarged section of a photograph reproduced with a halftoning method, showing how tones are represented with varying size dots.

## Halftone Patterns

- Cluster of dots (pixels) to represent intensity (trading spatial resolution for increased intensity resolution)
- Exploits spatial integration in eye



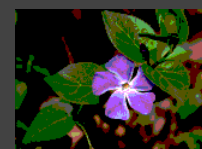
## Reducing Quantization

- Halftoning
- Dithering (*distribute errors among pixels*)
  - Random Dither
  - Error Diffusion (Floyd-Steinberg)

## Dithering



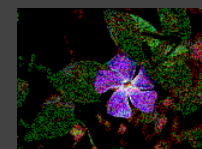
8 bits original



2 bits quantize: Note Contours



2 bits FLOYD STEINBERG



2 bits random dither: noise not contours

## Random Dither

- Randomize quantization errors [see assignment for exact details on adding random noise]
- Seems silly (add random noise), but eye more tolerant of high-frequency noise than contours or aliasing
- More complex algorithms (not considered here) are ordered dither with patterns of thresholds rather than completely random noise

## Random Dither

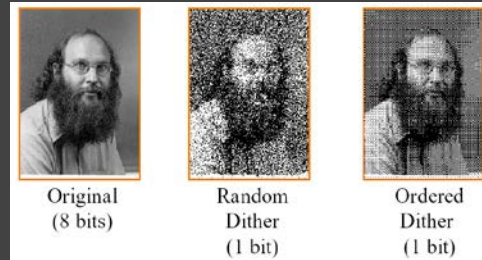
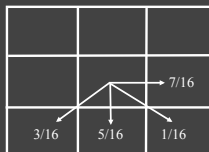


Image and example courtesy Tom Funkhouser

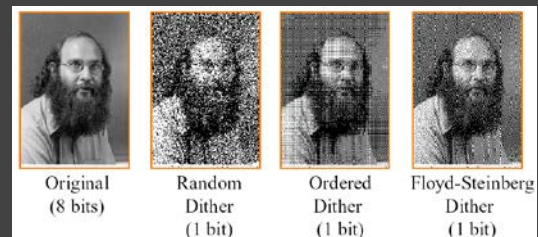
## Error Diffusion

- Spread quantization error to neighboring pixels to the right and below (later in the process)
- Reduces net error, gives best results



Error = pixel(x,y) - quantize(x,y);  
 pixel(x+1,y) +=  $\alpha$  · Error;  
 pixel(x-1,y+1) +=  $\beta$  · Error;  
 pixel(x,y+1) +=  $\gamma$  · Error;  
 pixel(x+1,y+1) +=  $\delta$  · Error;

## Floyd Steinberg Results



## Quantization (Sec 3.3 Ass 1)

- Simple quantization (should be straightforward)
- Random Dither (just add noise, pretty simple)
- Floyd-Steinberg (trickiest)
  - Must implement a diffusion of error to other pixels (simply add in appropriate error to them)
  - Uses fractions, so must use floating point
  - And possibly negative numbers since error can be minus
  - Boundary conditions (what if no right etc.) toroidal [may not be relevant in this case] or change weights appropriately, but don't darken boundaries