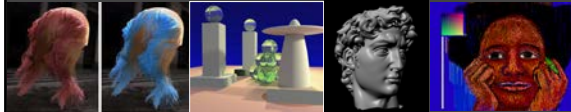


Advanced Computer Graphics

CSE 163 [Spring 2017], Lecture 2

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

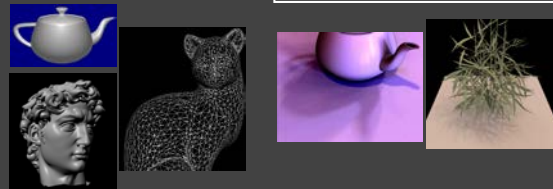


Course Outline

- 3D Graphics Pipeline

Modeling
(Creating 3D Geometry)

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



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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 28.
 - 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...



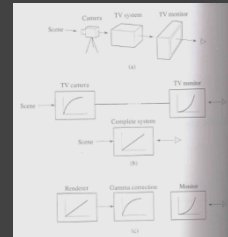
Gamma Correction

- 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

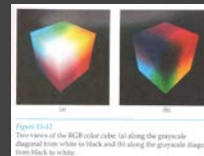
Example

- 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]



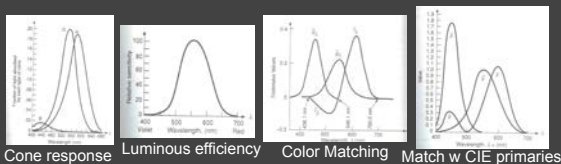
Color

- 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



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]



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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

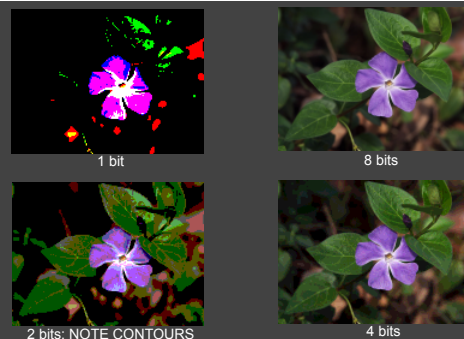
NTSC	640x480	8 bits	30 Hz
PC	1280x1024	24 bits RGB	75 Hz
Film	3000x2000	12 bits	24 Hz
Laser Printer	7000x2000	1 (on or off)	

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

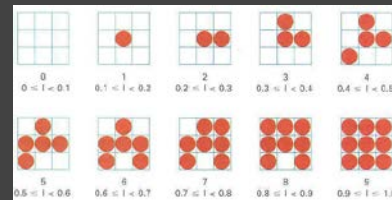
- Motivation: bilevel printing. Trade off spatial resolution for more intensity levels
- Dots of appropriate size to simulate grey levels
- Area of dots proportional intensity



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)

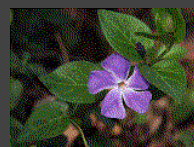
Dithering



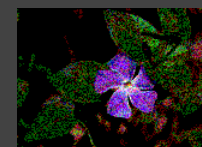
8 bits original



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

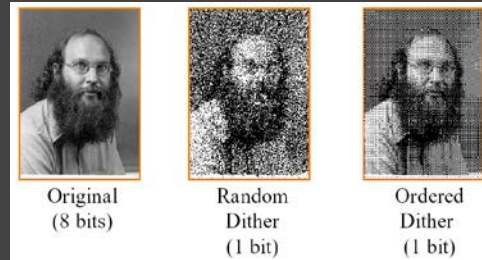
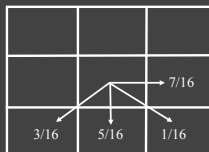


Image and example courtesy Tom Funkhouser

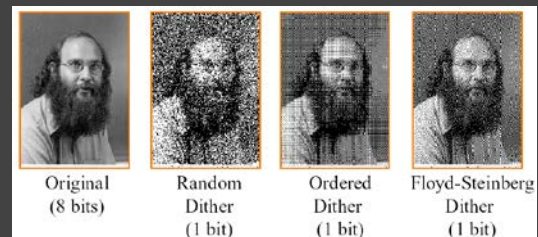
Error Diffusion

- 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) += α · Error;
 pixel(x-1,y+1) += β · Error;
 pixel(x,y+1) += γ · Error;
 pixel(x+1,y+1) += δ · 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