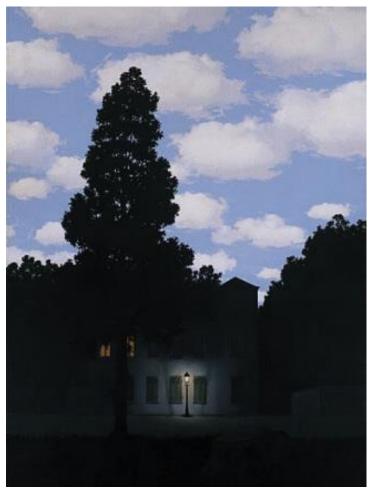
Light and Shading



Computer Vision

Derek Hoiem, University of Illinois

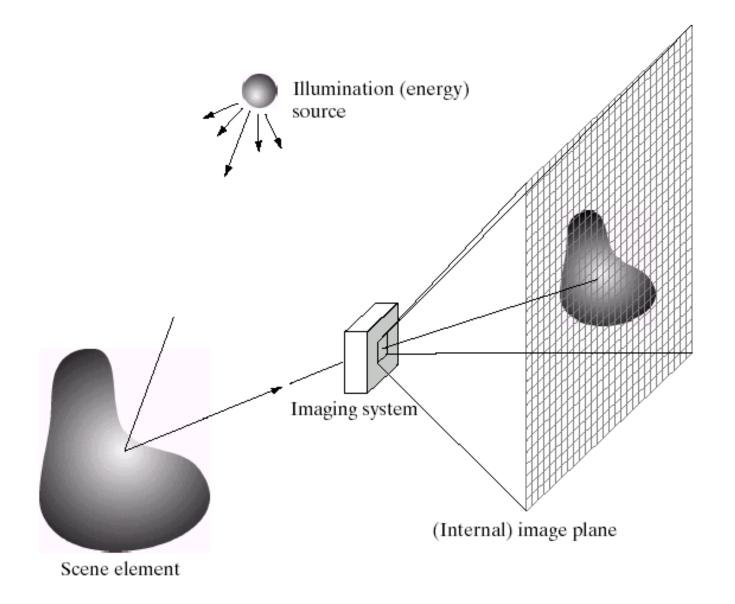
"Empire of Light", Magritte

FAQ

Final project: see guidelines here
 https://courses.engr.illinois.edu/cs543/Project Guidelines.pdf

- Waitlist
 - 80 students on the list so far
 - Sign up by Friday (link on course page)
- Note other highly related courses
 - ECE 418: Image & Video Processing
 - ECE 547: Topics in Image Processing

How light is recorded



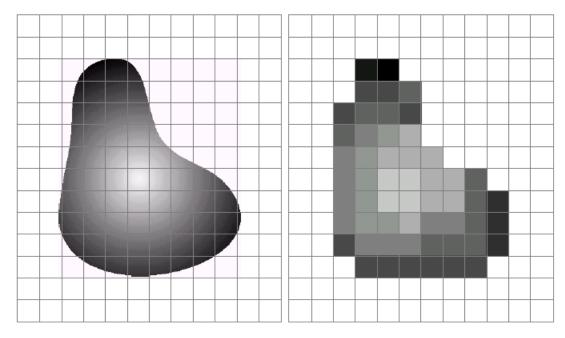
Digital camera

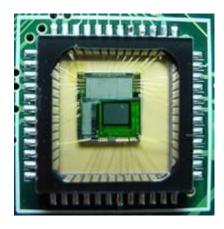


A digital camera replaces film with a sensor array

- Each cell in the array is light-sensitive diode that converts photons to electrons
- Two common types: Charge Coupled Device (CCD) and CMOS
- http://electronics.howstuffworks.com/digital-camera.htm

Sensor Array





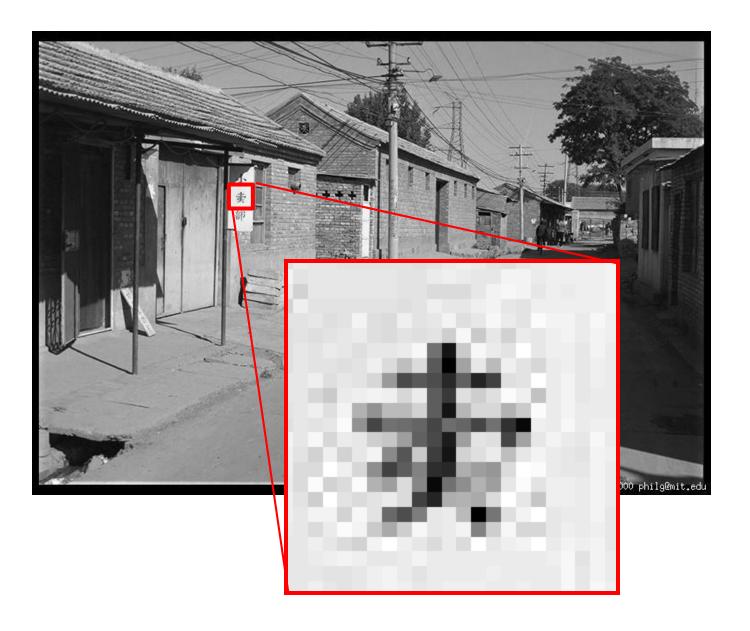
CMOS sensor

a b

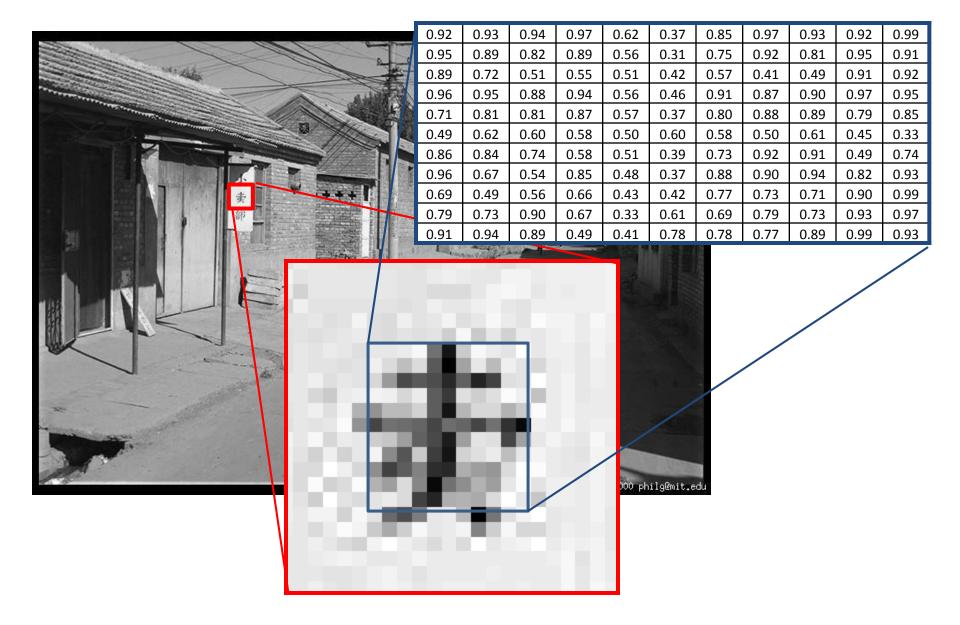
FIGURE 2.17 (a) Continuos image projected onto a sensor array. (b) Result of image sampling and quantization.

Each sensor cell records amount of light coming in at a small range of orientations

The raster image (pixel matrix)



The raster image (pixel matrix)

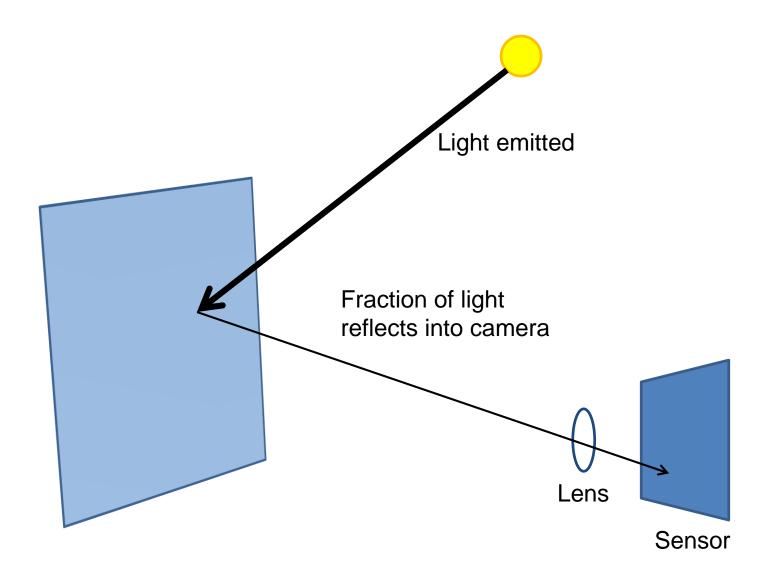


Today's class: Light and Shading



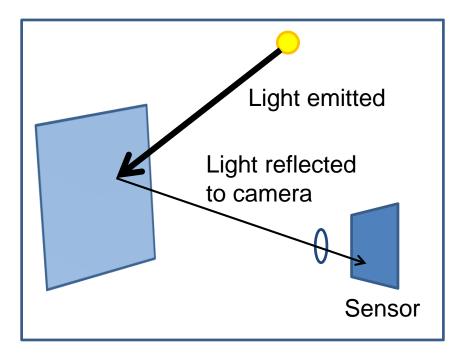
- What determines a pixel's intensity?
- What can we infer about the scene from pixel intensities?

How does a pixel get its value?



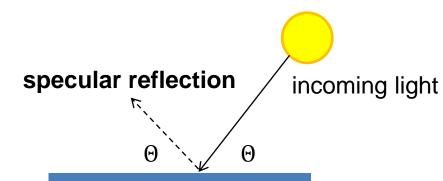
How does a pixel get its value?

- Major factors
 - Illumination strength
 and direction
 - Surface geometry
 - Surface material
 - Nearby surfaces
 - Camera gain/exposure

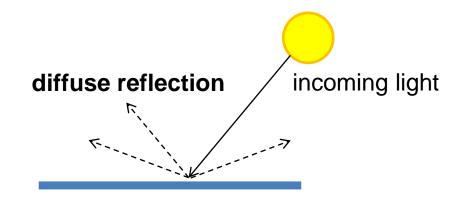


Basic models of reflection

- Specular: light bounces off at the incident angle
 - E.g., mirror

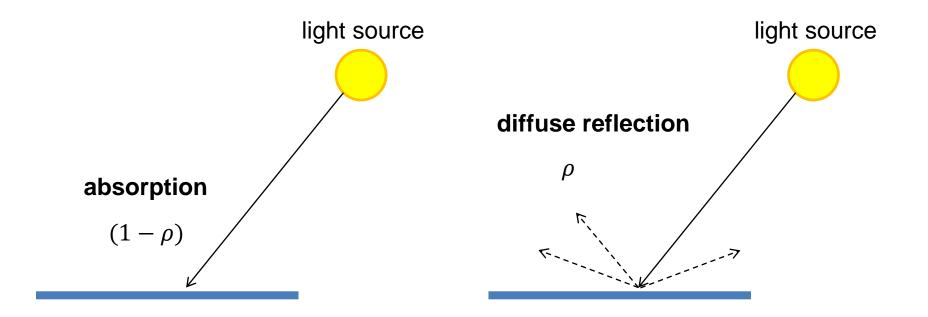


- Diffuse: light scatters in all directions
 - E.g., brick, cloth, rough wood



Lambertian reflectance model

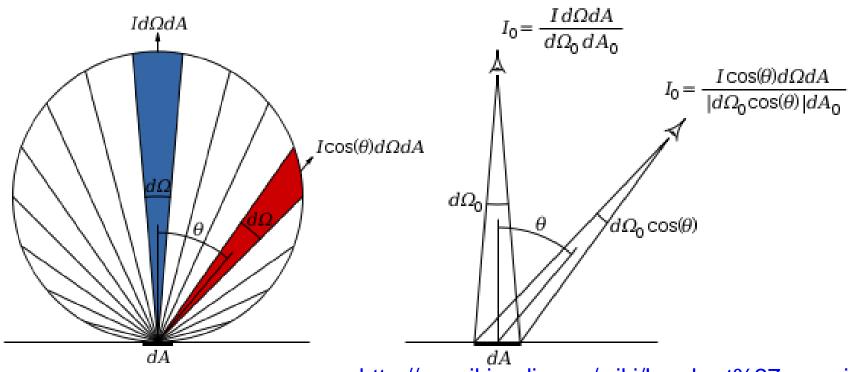
- Some light is absorbed (function of albedo ρ)
- Remaining light is scattered (diffuse reflection)
- Examples: soft cloth, concrete, matte paints



Diffuse reflection: Lambert's cosine law

Intensity does not depend on viewer angle.

- Amount of reflected light proportional to $cos(\theta)$
- Visible solid angle also proportional to $cos(\theta)$

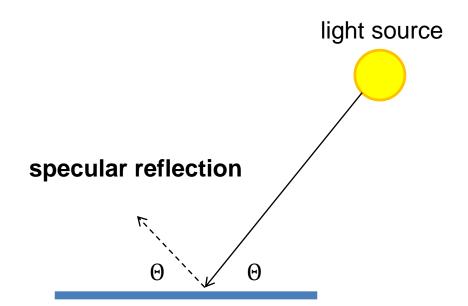


http://en.wikipedia.org/wiki/Lambert%27s_cosine_law

Specular Reflection

- Reflected direction depends on light orientation and surface normal
 - E.g., mirrors are fully specular
 - Most surfaces can be modeled with a mixture of diffuse and specular components







Flickr, by piratejohnny

Most surfaces have both specular and diffuse components

 Specularity = spot where specular reflection dominates (typically reflects light source)



Photo: northcountryhardwoodfloors.com



Typically, specular component is small

Intensity and Surface Orientation

Intensity depends on illumination angle because less light comes in at oblique angles.

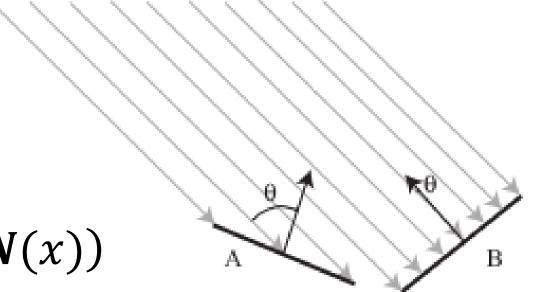
 ρ = albedo

S =directional source

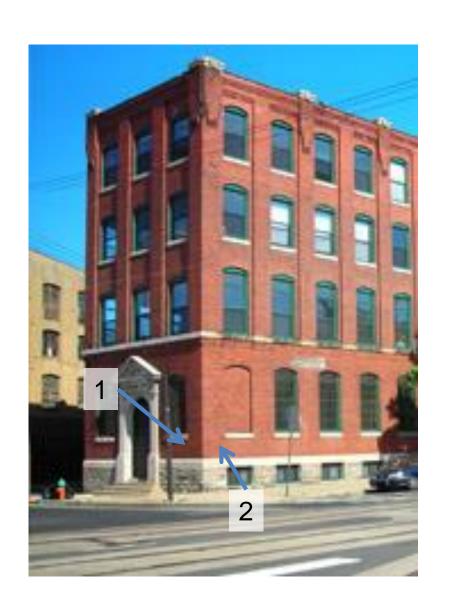
N =surface normal

I = reflected intensity

$$I(x) = \rho(x)(S \cdot N(x))$$



Slide: Forsyth

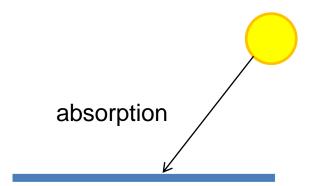


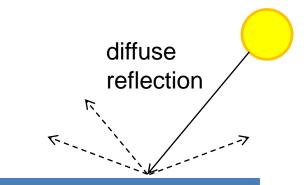
Recap

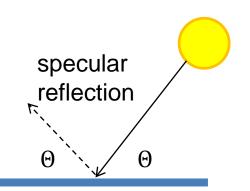
- When light hits a typical surface
 - Some light is absorbed (1- ρ)
 - More absorbed for low albedos

- Some light is reflected diffusely
 - Independent of viewing direction

- Some light is reflected specularly
 - Light bounces off (like a mirror), depends on viewing direction

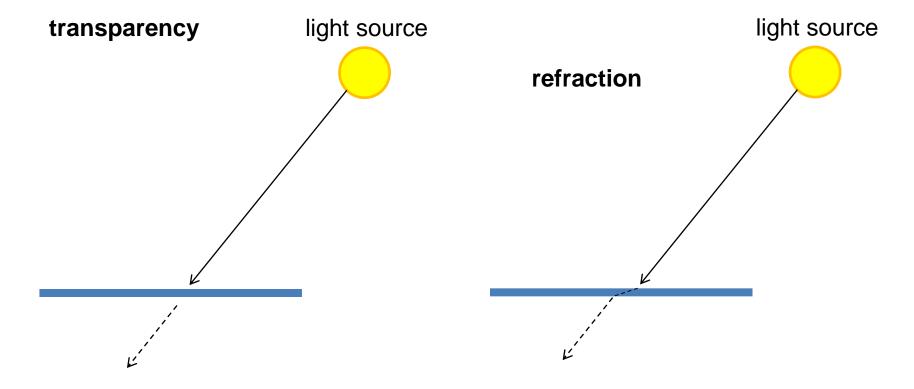




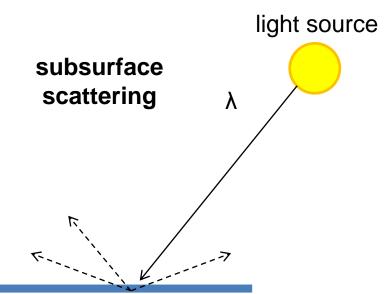


Other possible effects



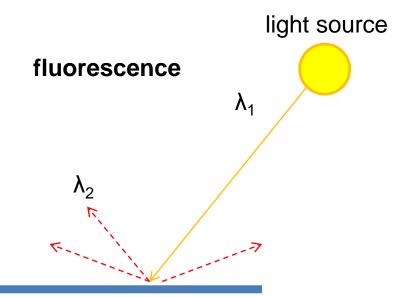


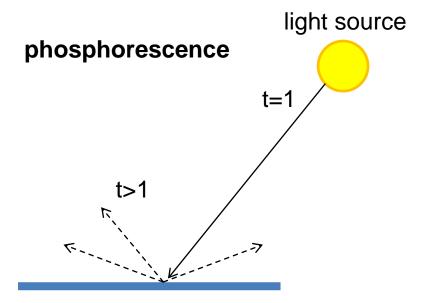






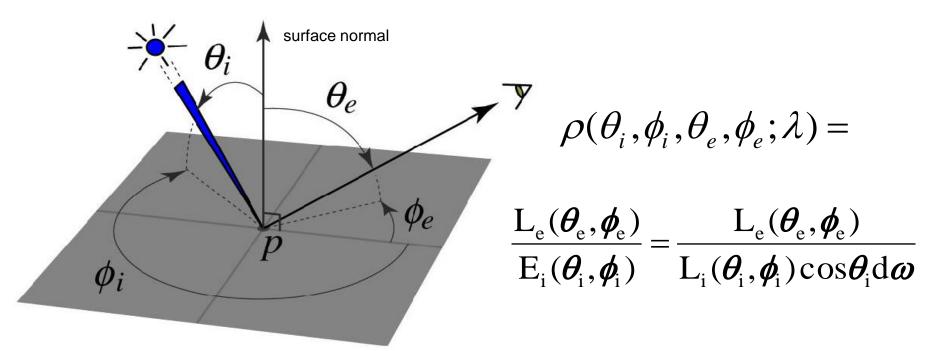






BRDF: Bidirectional Reflectance Distribution Function

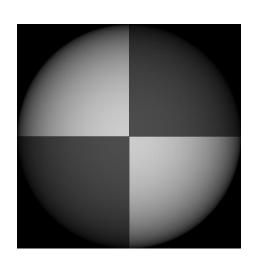
 Model of local reflection that tells how bright a surface appears when viewed from one direction when light falls on it from another

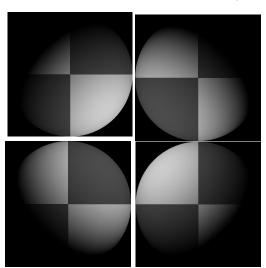


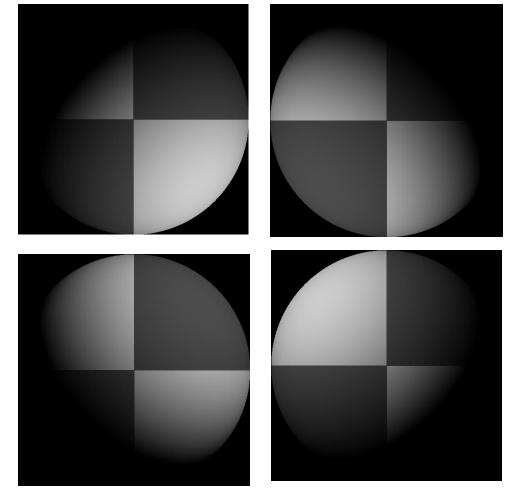
Application: photometric stereo

Assume:

- a set of point sources that are infinitely distant
- a set of pictures of an object, obtained in exactly the same camera/object configuration but using different sources
- a Lambertian object (or the specular component has been identified and removed)





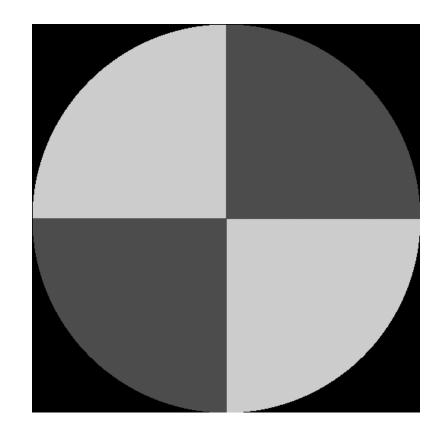


Intensity for pixel x Source i direction and strength

Each image is: $I_i(x) = \mathbf{S}_i \cdot (p(x)\mathbf{N}(x))$

So if we have enough images with known sources, we can solve for $\mathbf{B}(x) = p(x)\mathbf{N}(x)$

albedo times 3D normal vector

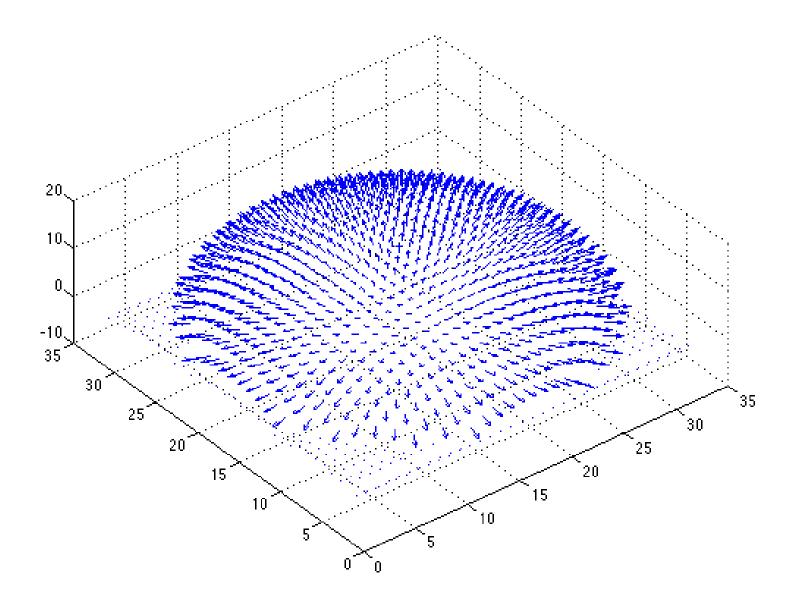


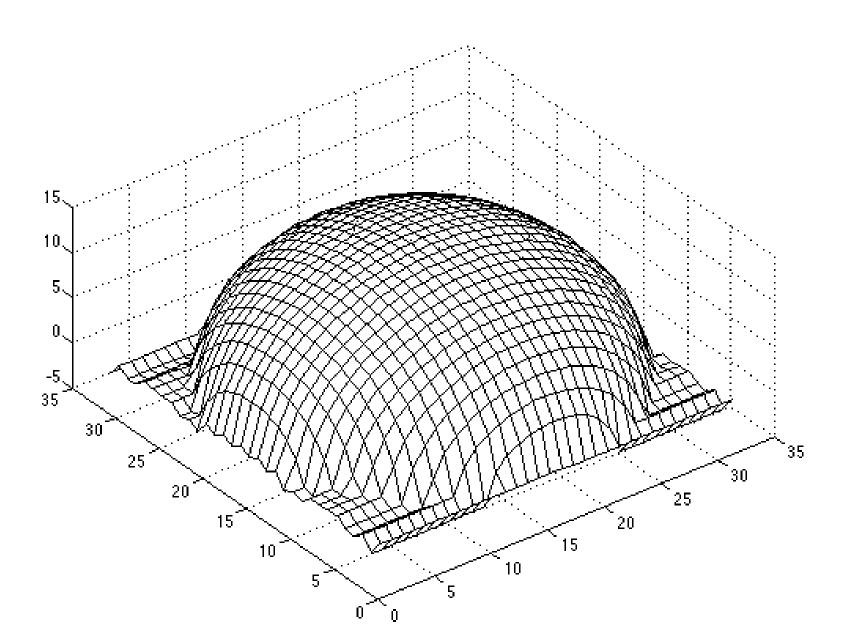
$$\mathbf{B}(x) = p(x)\mathbf{N}(x)$$

And the albedo (shown here) is given by:

$$p(x) = \sqrt{\boldsymbol{B}(x) \cdot \boldsymbol{B}(x)}$$

(the normal is a unit vector)

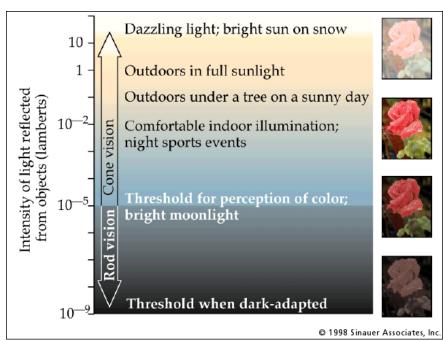


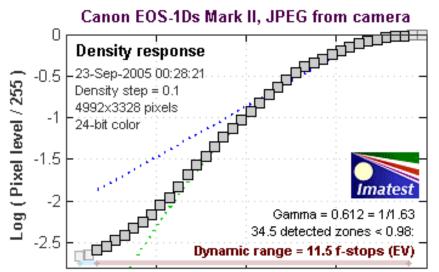


Dynamic range and camera response

 Typical scenes have a huge dynamic range

- Camera response is roughly linear in the mid range (15 to 240) but non-linear at the extremes
 - called saturation or undersaturation

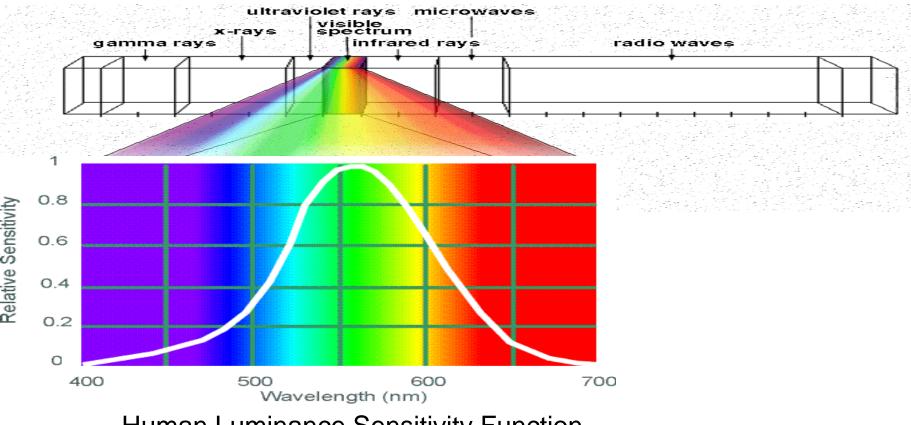




Log Exposure (Target density)

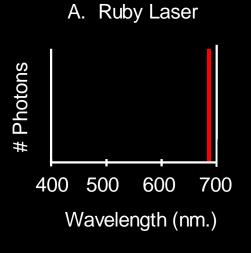
Color

Light is composed of a spectrum of wavelengths



Human Luminance Sensitivity Function

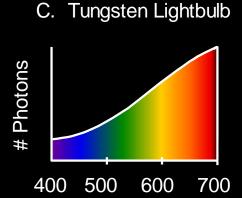
Some examples of the spectra of light sources

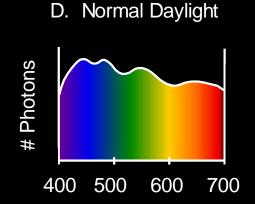


Bhotons # 400 500 600 700

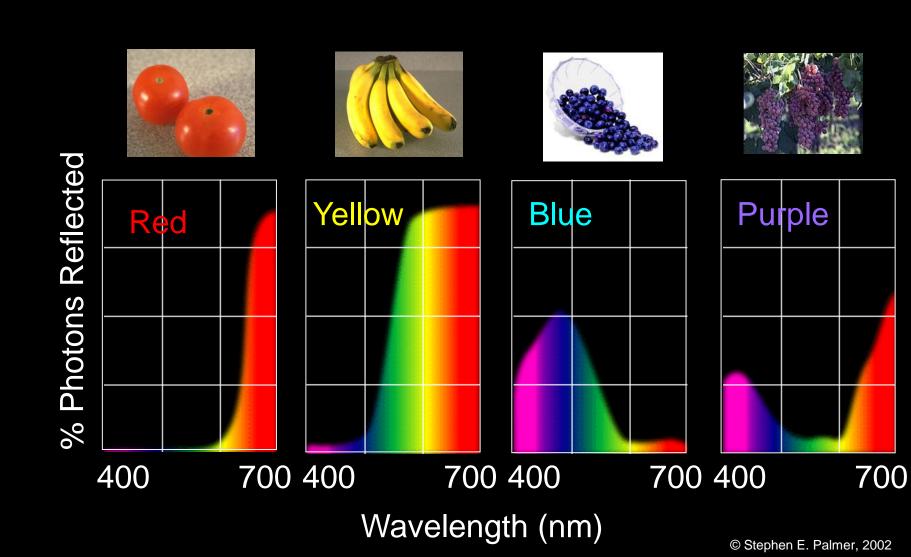
Wavelength (nm.)

B. Gallium Phosphide Crystal

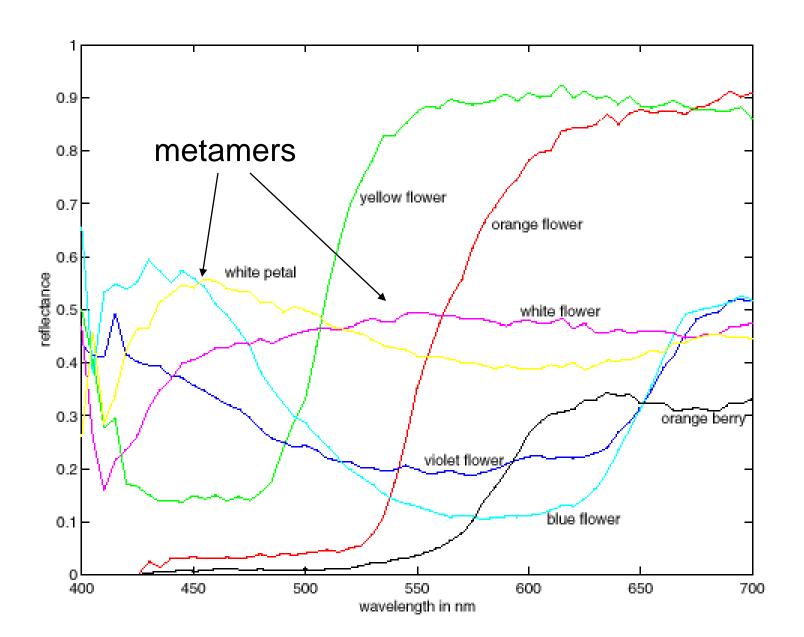




Some examples of the <u>reflectance</u> spectra of <u>surfaces</u>

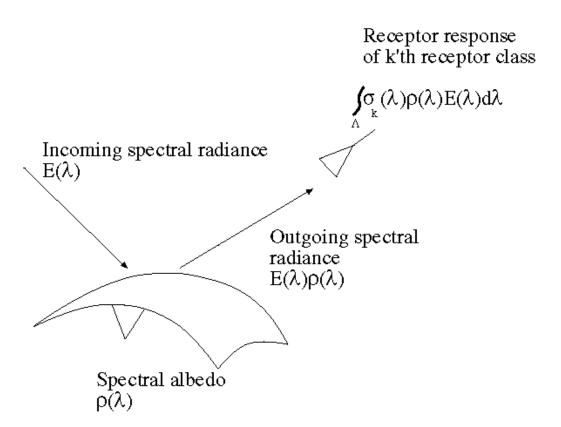


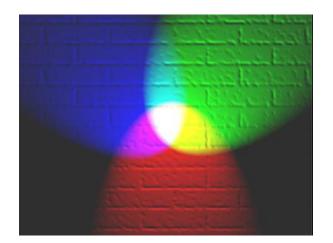
More spectra



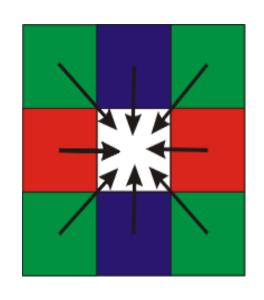
The color of objects

- Colored light arriving at the camera involves two effects
 - The color of the light source (illumination + inter-reflections)
 - The color of the surface

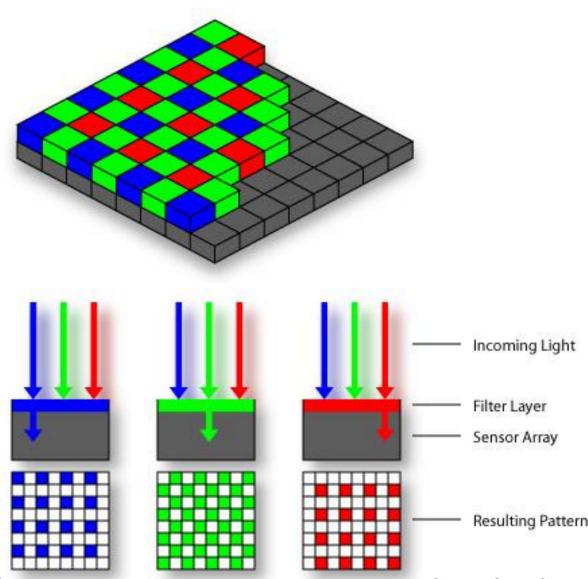




Color Sensing: Bayer Grid



Estimate RGB at each cell from neighboring values



Color Image

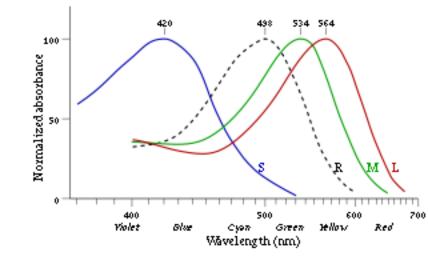


R

Why RGB?

If light is a spectrum, why are images RGB?

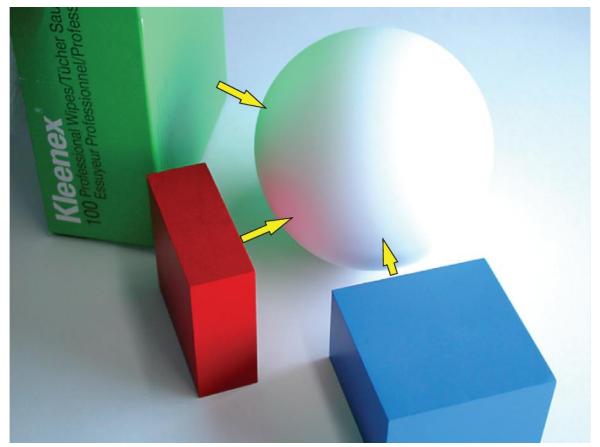
Human color receptors



- Long (red), Medium (green), and Short (blue) cones, plus intensity rods
- Fun facts
 - "M" and "L" on the X-chromosome
 - That's why men are more likely to be color blind (see what it's like: http://www.vischeck.com/vischeck/vischeckImage.php)
 - "L" has high variation, so some women are tetrachromatic
 - Some animals have 1 (night animals), 2 (e.g., dogs), 4 (fish, birds), 5 (pigeons, some reptiles/amphibians), or even 12 (mantis shrimp) types of cones

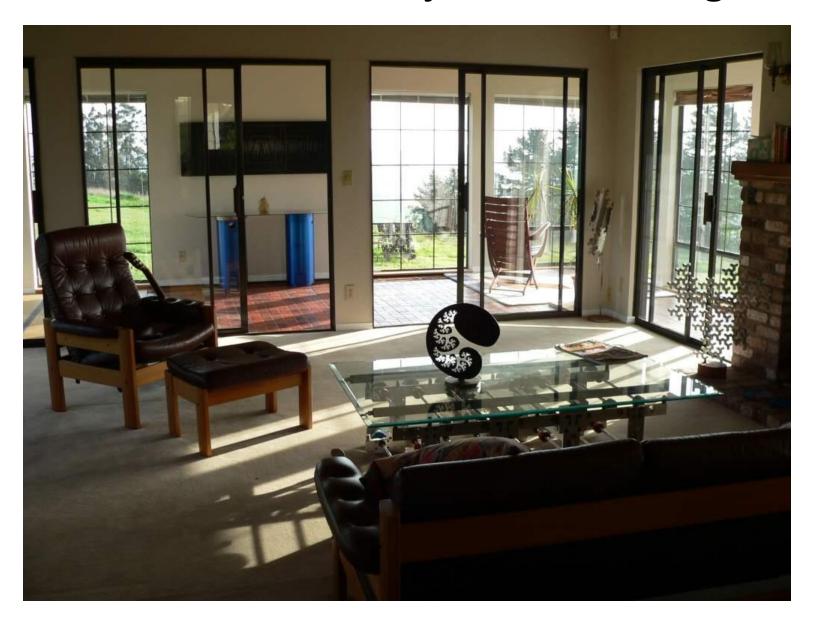
So far: light → surface → camera

- Called a local illumination model
- But much light comes from surrounding surfaces

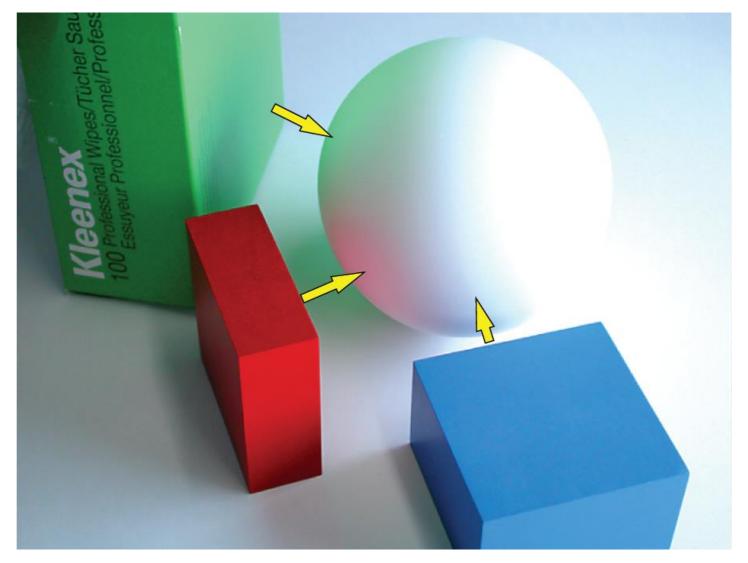


From Koenderink slides on image texture and the flow of light

Inter-reflection is a major source of light



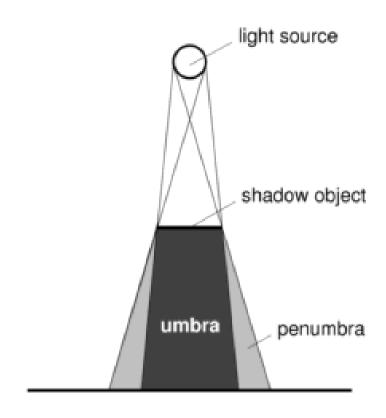
Inter-reflection affects the apparent color of objects



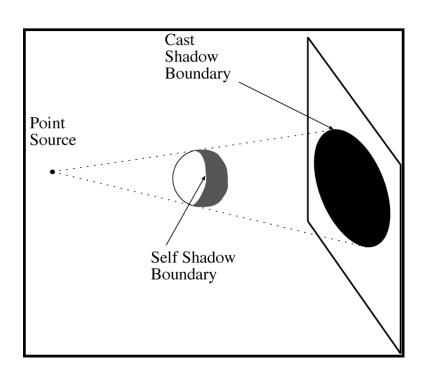
From Koenderink slides on image texture and the flow of light

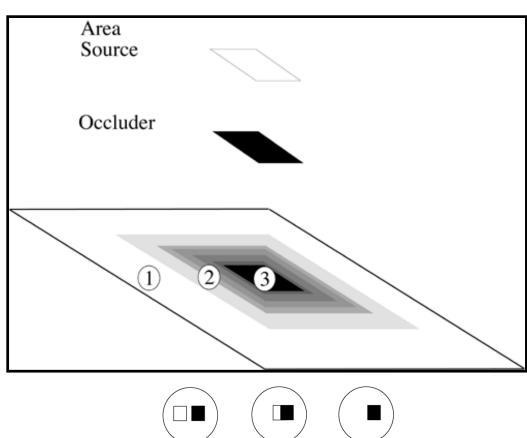
Scene surfaces also cause shadows

 Shadow: reduction in intensity due to a blocked source



Shadows





Models of light sources

- Distant point source
 - One illumination direction
 - E.g., sun
- Area source
 - E.g., white walls, diffuser lamps, sky
- Ambient light
 - Substitute for dealing with interreflections
- Global illumination model
 - Account for interreflections in modeled scene

Recap



Possible factors: albedo, shadows, texture, specularities, curvature, lighting direction

What does the intensity of a pixel tell us?

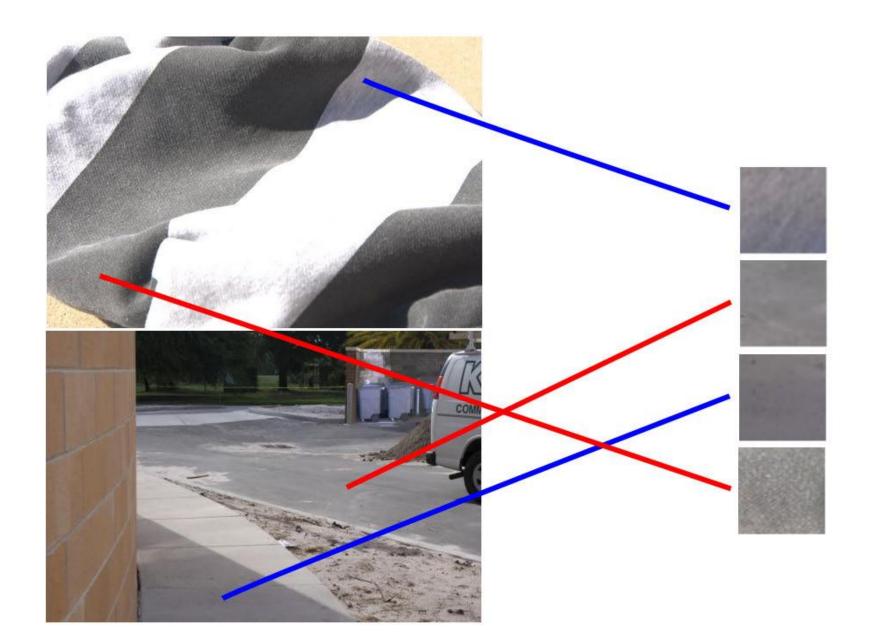
im(234, 452) = 0.58

0.92	0.93	0.94	0.97	0.62	0.37	0.85	0.97	0.93	0.92	0.99
0.95	0.89	0.82	0.89	0.56	0.31	0.75	0.92	0.81	0.95	0.91
0.89	0.72	0.51	0.55	0.51	0.42	0.57	0.41	0.49	0.91	0.92
0.96	0.95	0.88	0.94	0.56	0.46	0.91	0.87	0.90	0.97	0.95
0.71	0.81	0.81	0.87	0.57	0.37	0.80	0.88	0.89	0.79	0.85
0.49	0.62	0.60	0.58	0.50	0.60	0.58	0.50	0.61	0.45	0.33
0.86	0.84	0.74	0.58	0.51	0.39	0.73	0.92	0.91	0.49	0.74
0.96	0.67	0.54	0.85	0.48	0.37	0.88	0.90	0.94	0.82	0.93
0.69	0.49	0.56	0.66	0.43	0.42	0.77	0.73	0.71	0.90	0.99
0.79	0.73	0.90	0.67	0.33	0.61	0.69	0.79	0.73	0.93	0.97
0.91	0.94	0.89	0.49	0.41	0.78	0.78	0.77	0.89	0.99	0.93

The plight of the poor pixel

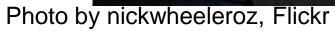
- A pixel's brightness is determined by
 - Light source (strength, direction, color)
 - Surface orientation
 - Surface material and albedo
 - Reflected light and shadows from surrounding surfaces
 - Gain on the sensor

A pixel's brightness tells us nothing by itself











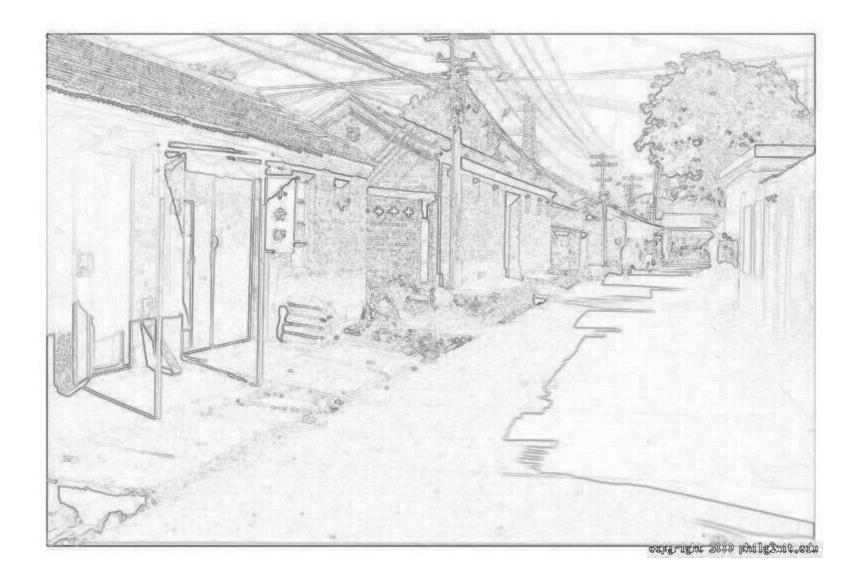
Slide: Forsyth

And yet we can interpret images...



- Key idea: for nearby scene points, most factors do not change much
- The information is mainly contained in *local* differences of brightness

Darkness = Large Difference in Neighboring Pixels



What is this?





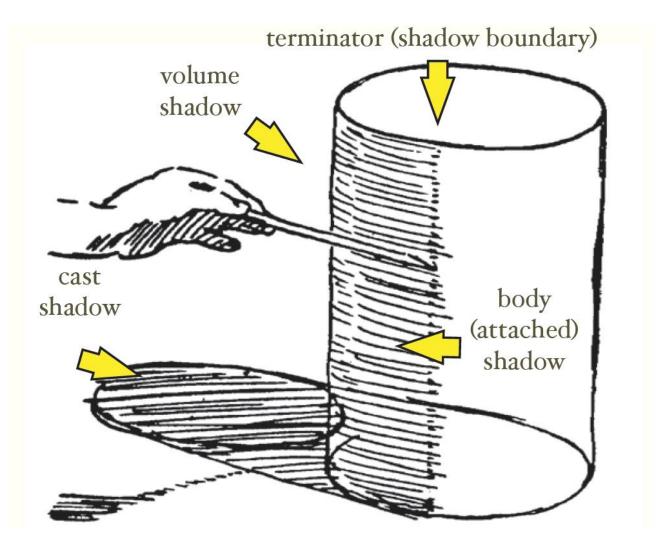
What differences in intensity tell us about shape

- Changes in surface normal
- Texture
- Proximity
- Indents and bumps
- Grooves and creases





Shadows as cues

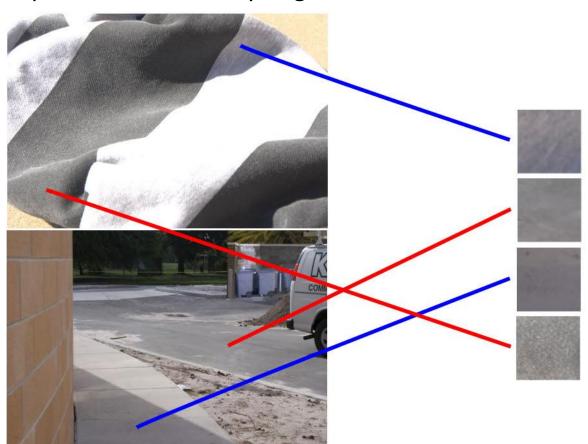


From Koenderink slides on image texture and the flow of light

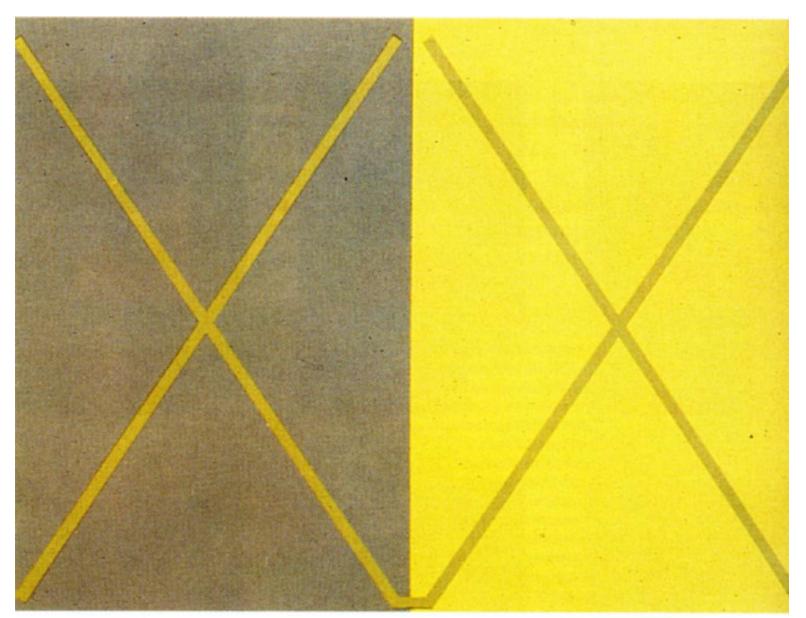
Slide: Forsyth

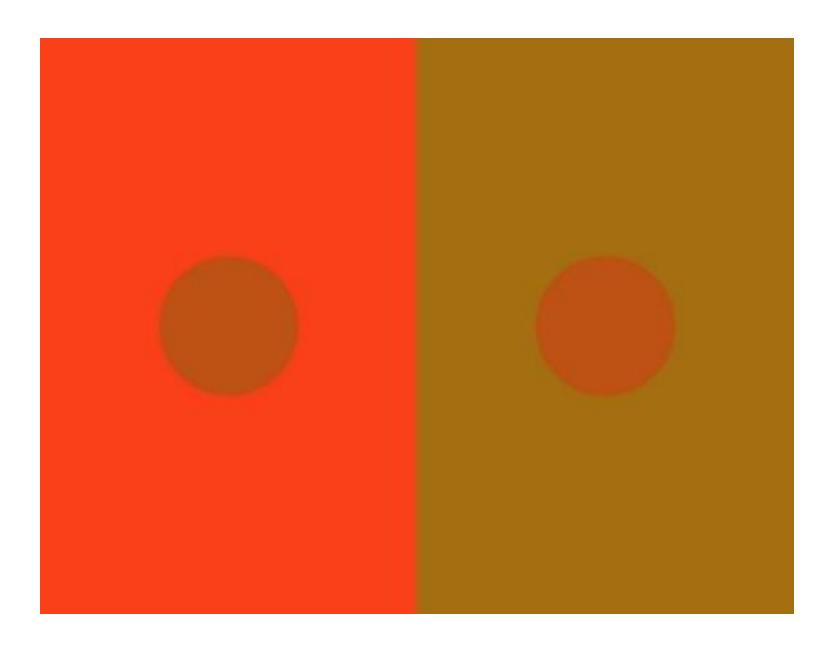
Color constancy

- Interpret surface in terms of albedo or "true color", rather than observed intensity
 - Humans are good at it
 - Computers are not nearly as good

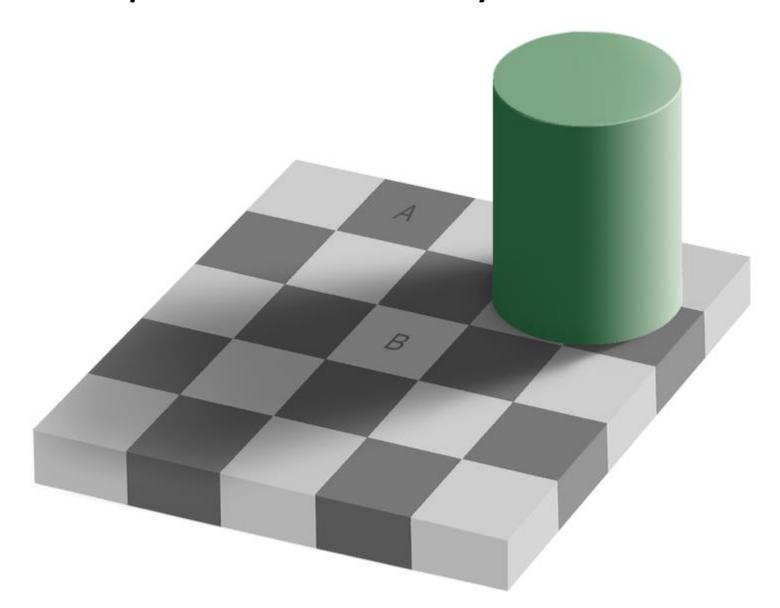


One source of constancy: local comparisons

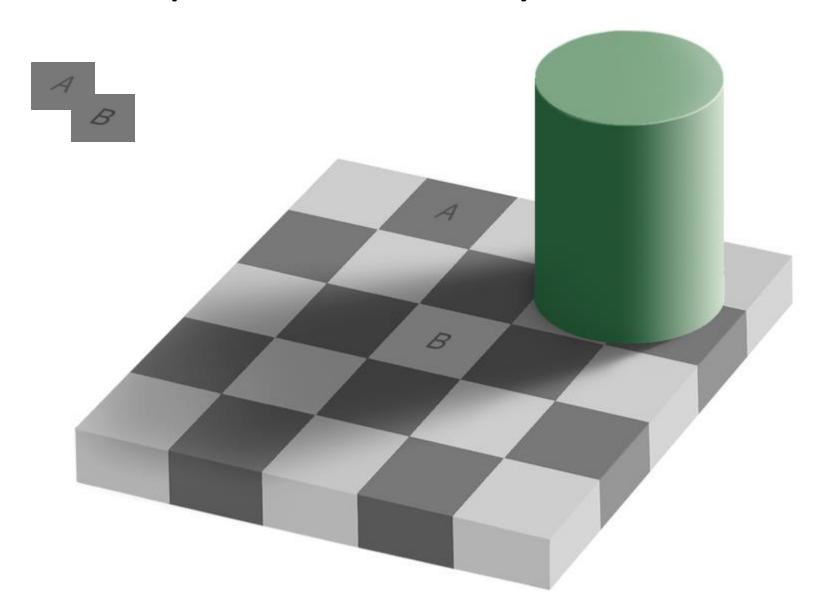




Perception of Intensity



Perception of Intensity



Color Correction

Simple idea: multiply R, G, and B values by separate constants

$$\begin{bmatrix} \tilde{r} \\ \tilde{g} \\ \tilde{b} \end{bmatrix} = \begin{bmatrix} \alpha_r & 0 & 0 \\ 0 & \alpha_g & 0 \\ 0 & 0 & \alpha_b \end{bmatrix} \begin{bmatrix} r \\ g \\ b \end{bmatrix}$$

- How to choose the constants?
 - "White world" assumption: brightest pixel is white
 - Divide by largest value
 - "Gray world" assumption: average value should be gray
 - E.g., multiply r channel by avg(r) /avg((r+g+b)/3)
 - White balancing: choose a reference as the white or gray color

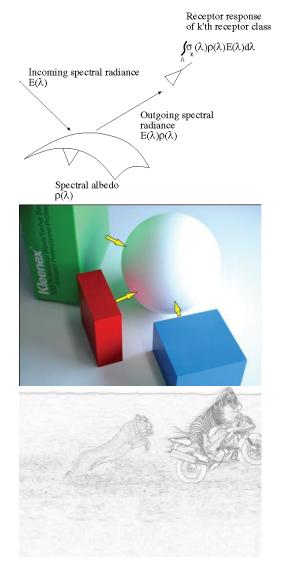
Things to remember

 Important terms: diffuse/specular reflectance, albedo, umbra/penumbra

 Observed intensity depends on light sources, geometry/material of reflecting surface, surrounding objects, camera settings

Objects cast light and shadows on each other

Differences in intensity are primary cues for shape



Thank you

Next class: Image Filters