

Introduction to

Computer Vision

Understanding Variability

Why so different?







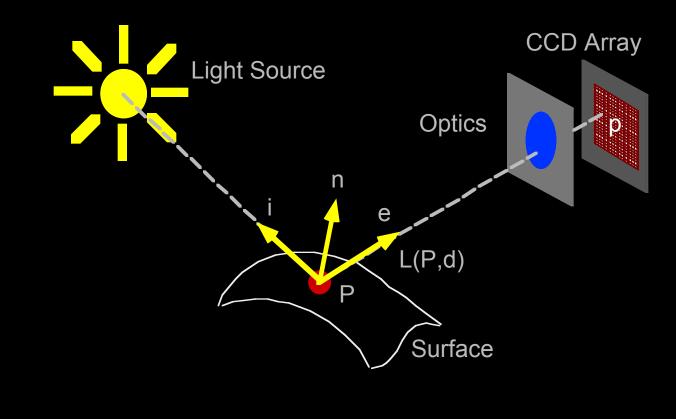
Light and Optics

- Pinhole camera model
- Perspective projection
- Thin lens model
- Fundamental equation
- Distortion: spherical & chromatic aberration, radial distortion
- Reflection and Illumination: color, Lambertian and specular surfaces, Phong, BRDF
- Sensing Light
- Conversion to Digital Images
- Sampling Theorem
- Other Sensors: frequency, type,



Basic Radiometry

Radiometry is the part of image formation concerned with the relation among the amounts of light energy emitted from light sources, reflected from surfaces, and registered by sensors.





Lecture Assumptions

Typical imaging scenario:

- visible light
- ideal lenses
- standard sensor (e.g. TV camera)
- opaque objects

Goal

To create 'digital' images which can be processed to recover some of the characteristics of the 3D world which was imaged.







World	reality
Optics	focus {light} from world on sensor
Sensor	converts {light} to {electrical energy}
Signal	representation of incident light as continuous electrical energy
Digitizer	converts continuous signal to discrete signal
Digital Rep.	final representation of reality in computer memory



Factors in Image Formation

Geometry

concerned with the relationship between points in the three-dimensional world and their images

Radiometry

 concerned with the relationship between the amount of light radiating from a surface and the amount incident at its image

Photometry

concerned with ways of measuring the intensity of light

Digitization

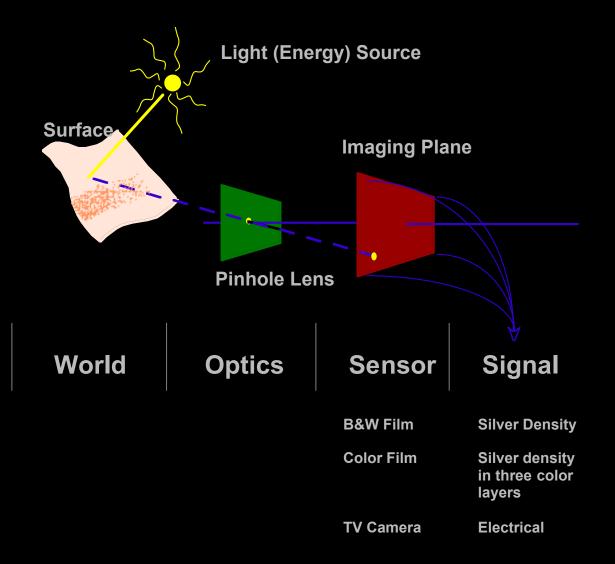
concerned with ways of converting continuous signals (in both space and time) to digital approximations



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Image Formation







Geometry describes the projection of:

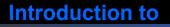
three-dimensional (3D) world



two-dimensional (2D) image plane.

Typical Assumptions

- Light travels in a straight line
- Optical Axis: the perpendicular from the image plane through the pinhole (also called the central projection ray)
- Each point in the image corresponds to a particular direction defined by a ray from that point through the pinhole.
- Various kinds of projections:
 - - perspective oblique
 - orthographic isometric
 - - spherical



Basic Optics

Two models are commonly used:

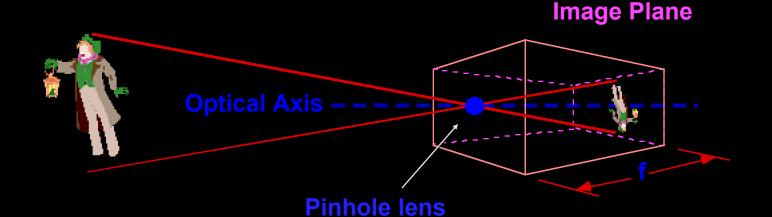
- Pin-hole camera
- Optical system composed of lenses
- Pin-hole is the basis for most graphics and vision
 - Derived from physical construction of early cameras
 - Mathematics is very straightforward
 - Thin lens model is first of the lens models
 - Mathematical model for a physical lens
 - Lens gathers light over area and focuses on image plane.



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Pinhole Camera Model



- World projected to 2D Image
 - Image inverted
 - Size reduced
 - Image is dim
 - No direct depth information
- f called the focal length of the lens
- Known as perspective projection



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Pinhole camera image

Amsterdam



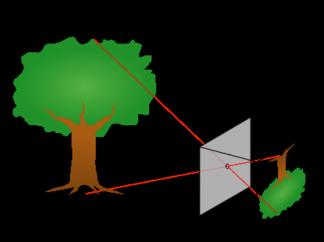
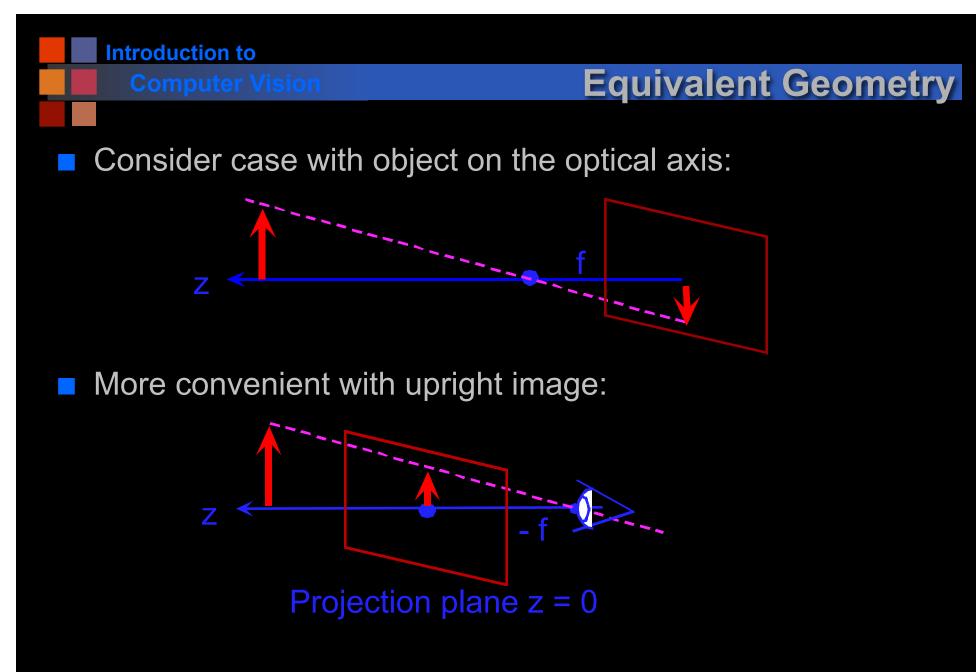


Photo by Robert Kosara, robert@kosara.net http://www.kosara.net/gallery/pinholeamsterdam/pic01.html

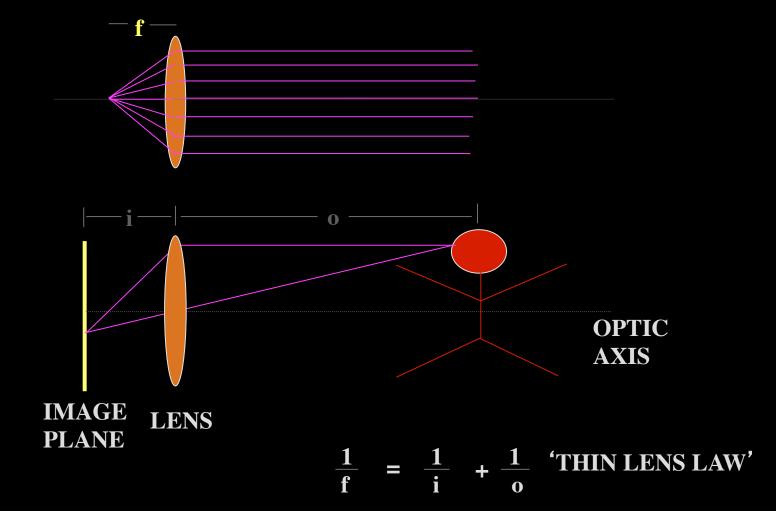


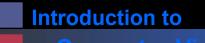
Equivalent mathematically



Thin Lens Model

Rays entering parallel on one side converge at focal point. Rays diverging from the focal point become parallel.





Coordinate System

Simplified Case:

- Origin of world and image coordinate systems coincide
- Y-axis aligned with y-axis
- X-axis aligned with x-axis
- Z-axis along the central projection ray

p(x,y)

Х

Image Coordinate System

(0,0)

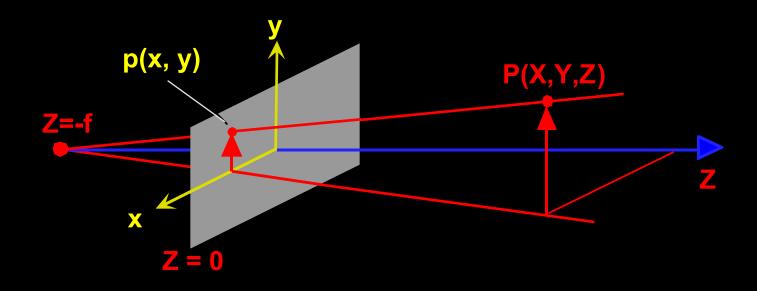
World Coordinate System Z

P(X,Y,Z)



Perspective Projection

Compute the image coordinates of p in terms of the world coordinates of P.



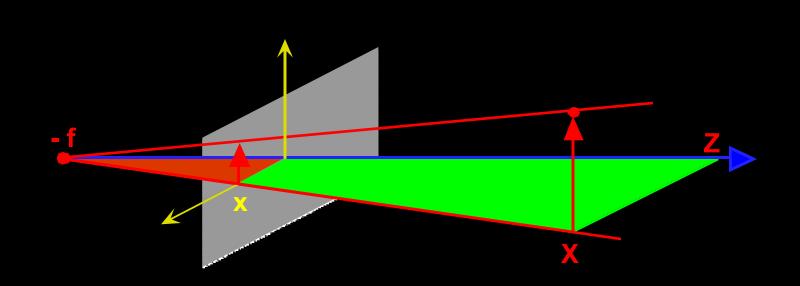
Look at projections in x-z and y-z planes



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X-Z Projection



By similar triangles: $\frac{x}{f} = \frac{X}{Z+f}$

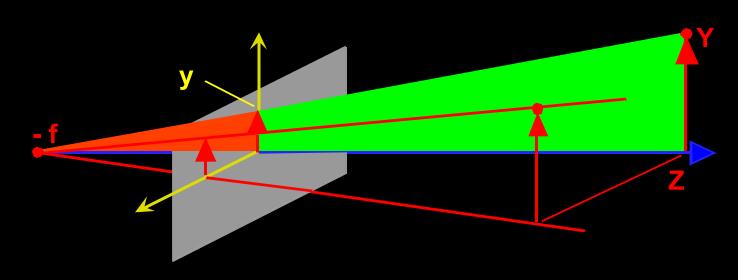
$$x = \frac{fX}{Z+f}$$



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Y-Z Projection



By similar triangles: $\frac{y}{f} = \frac{Y}{Z+f}$

$$y = \frac{fY}{Z+f}$$



Given point P(X,Y,Z) in the 3D world

The two equations:

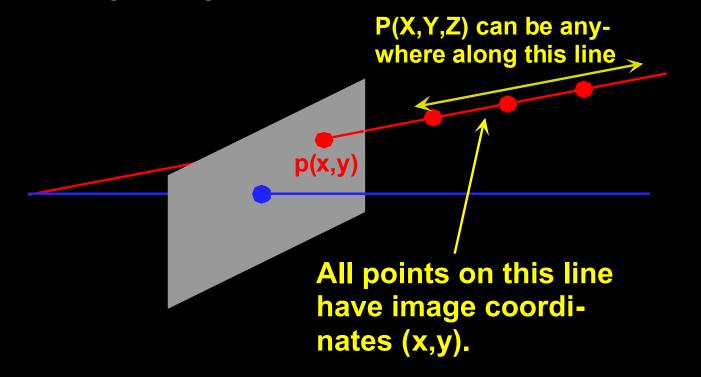
$$x = \frac{fX}{Z+f}$$
 $y = \frac{fY}{Z+f}$

transform world coordinates (X,Y,Z)
into image coordinates (x,y)

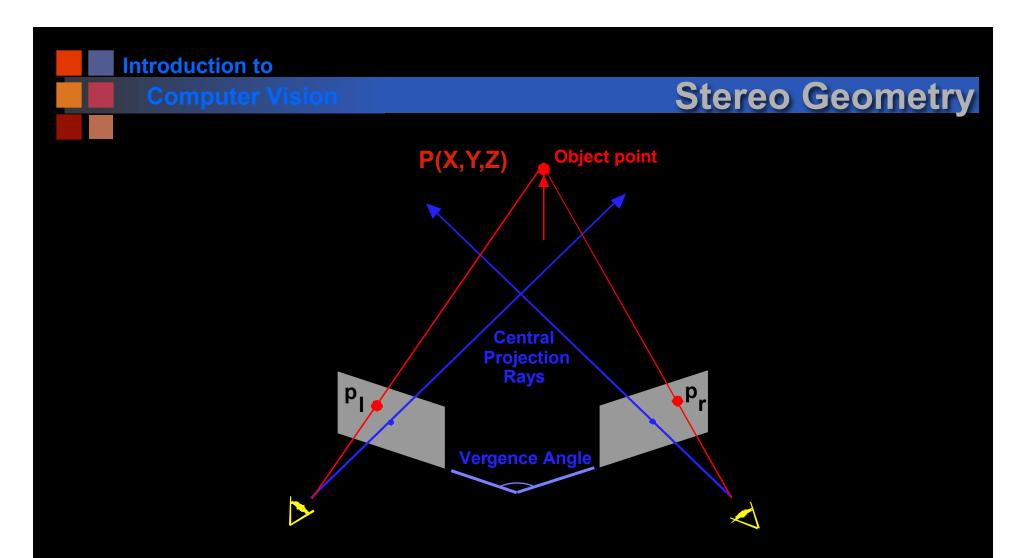


Reverse Projection

Given a center of projection and image coordinates of a point, it is not possible to recover the 3D depth of the point from a single image.



In general, at least two images of the same point taken from two different locations are required to recover depth.



- Depth obtained by triangulation
- Correspondence problem: p_l and p_r must correspond to the left and right projections of P, respectively.



Variability in appearance

- Consequences of image formation geometry for computer vision
 - What set of shapes can an object take on?
 - rigid
 - non-rigid
 - planar
 - non-planar
 - SIFT features
- Sensitivity to errors.



Brightness: informal notion used to describe both scene and image brightness.

Image brightness: related to energy flux incident on the image plane:

IRRADIANCE

Illuminance

Scene brightness: brightness related to energy flux emitted (radiated) from a surface.

> RADIANCE luminance



Light and Surfaces

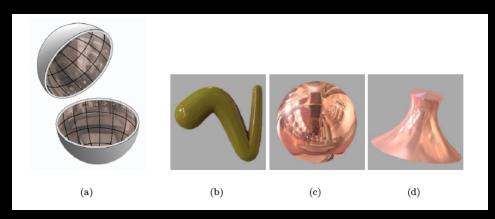
Reflection

- mirrors
- highlights
- specularities
- Scattering
 - Lambertian
 - matte
 - diffuse



Light sources

- Point source
- Extended source
- Single wavelength
- Multi-wavelength
- Uniform
- Non-uniform

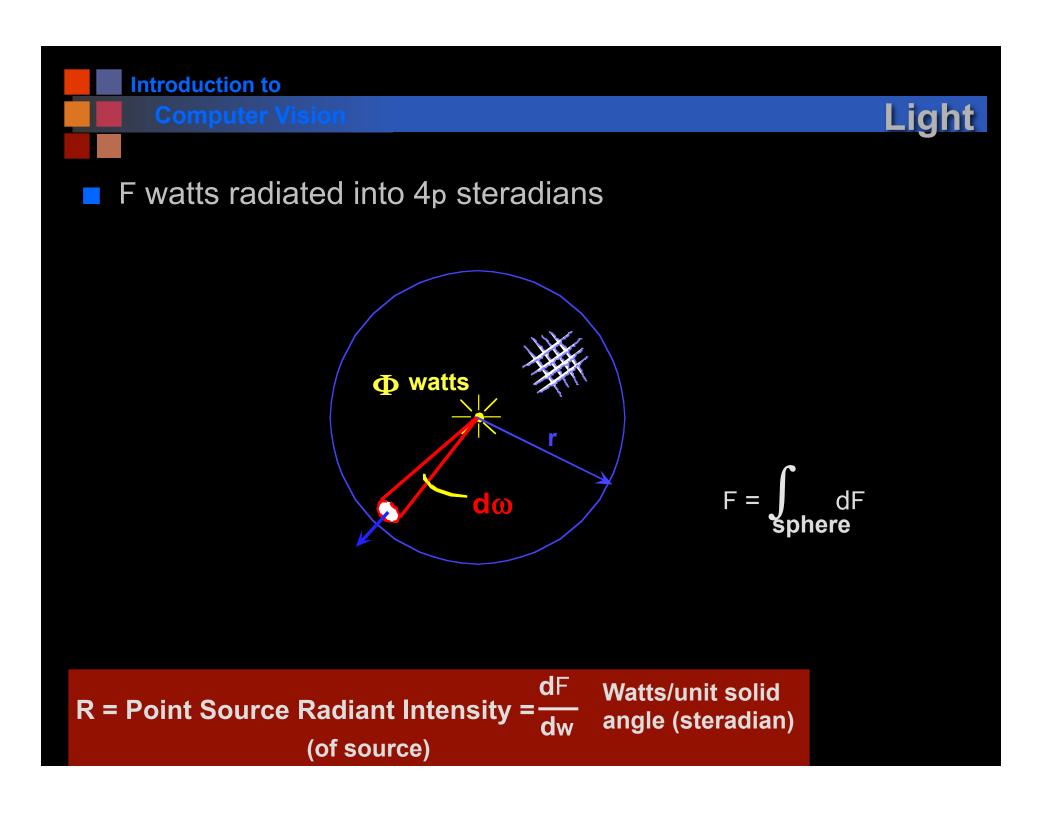




Linearity of Light

Linearity

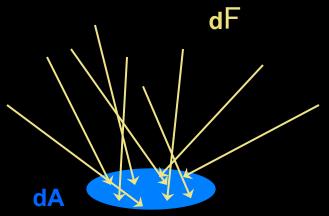
- definition
- For extended sources
- For multiple wavelengths
- Across time





Irradiance

Light falling on a surface from all directions.How much?



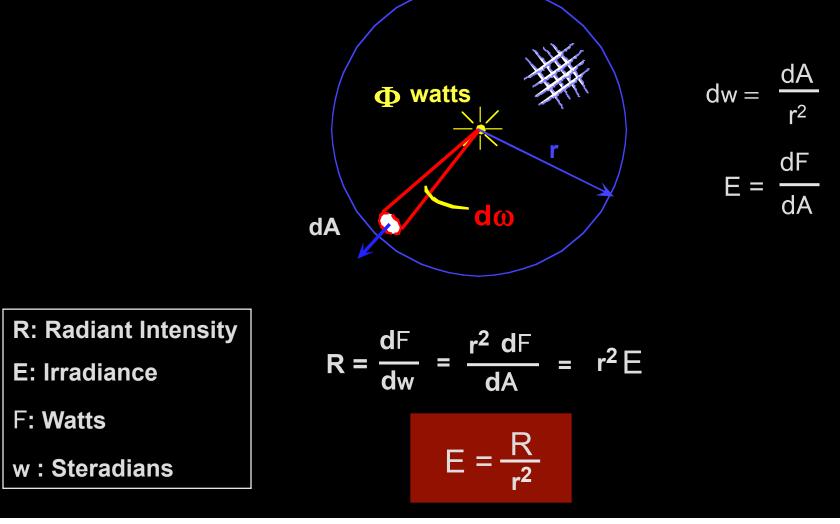
Irradiance: power per unit area falling on a surface.

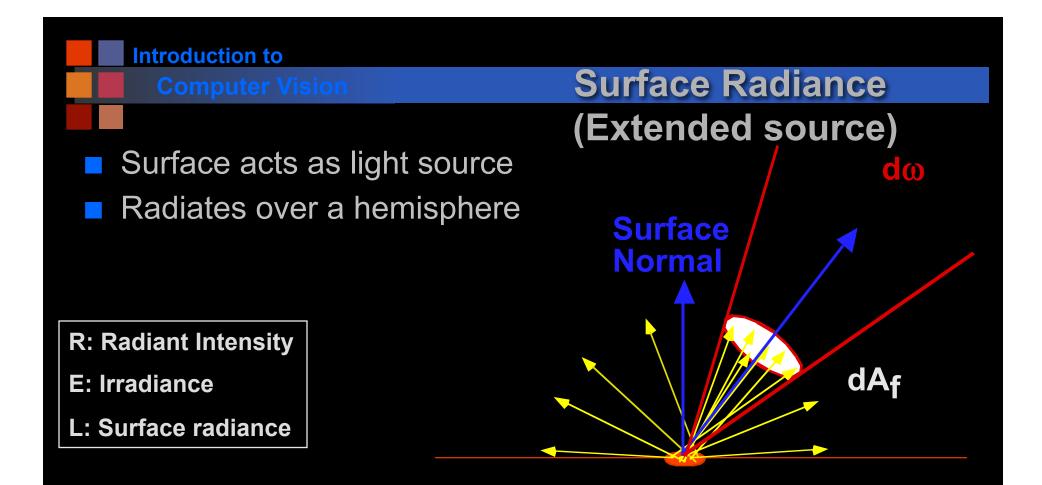
Irradiance E =
$$\frac{dF}{dA}$$
 watts/m²



Inverse Square Law

Relationship between point source radiance (radiant intensity) and irradiance





Surface Radiance: power per unit foreshortened area emitted into a solid angle

$$L = \frac{dF}{dA_{f}dw}$$

(watts/m² steradian)



Pseudo-Radiance

Consider two definitions:

• Radiance:

power per unit foreshortened area emitted into a solid angle

Pseudo-radiance

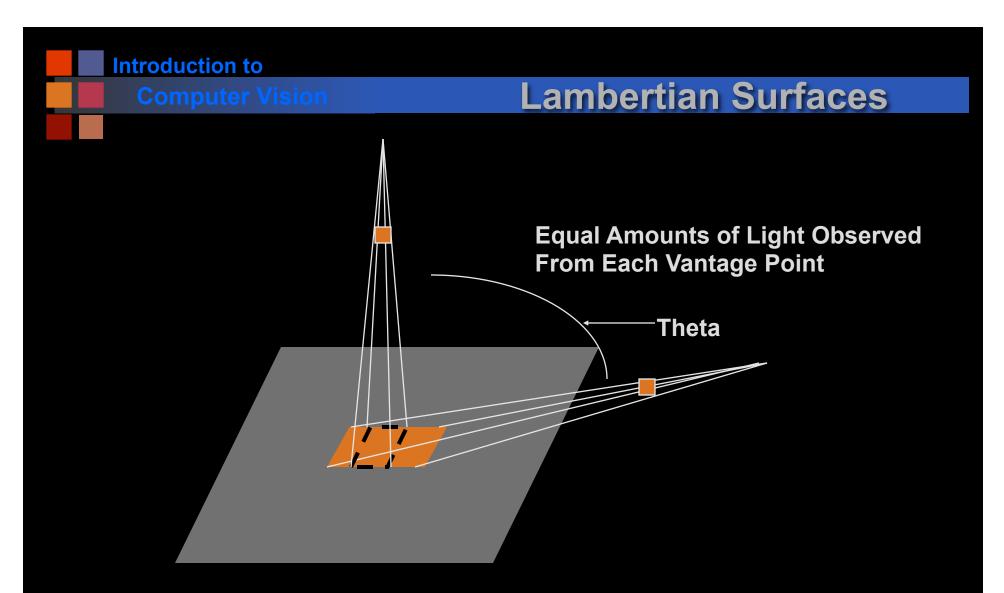
power per unit area emitted into a solid angle

- Why should we work with radiance rather than pseudoradiance?
 - Only reason: Radiance is more closely related to our intuitive notion of "brightness".



Lambertian Surfaces

- A particular point P on a Lambertian (perfectly matte) surface appears to have the same brightness no matter what angle it is viewed from.
 - Piece of paper
 - Matte paint
- Doesn't depend upon incident light angle.
- What does this say about how they emit light?

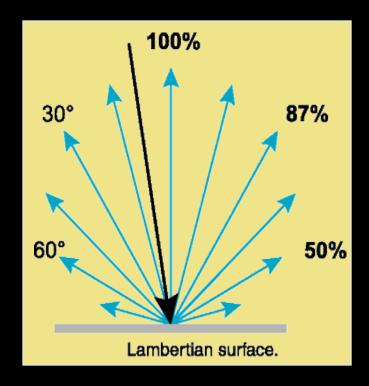


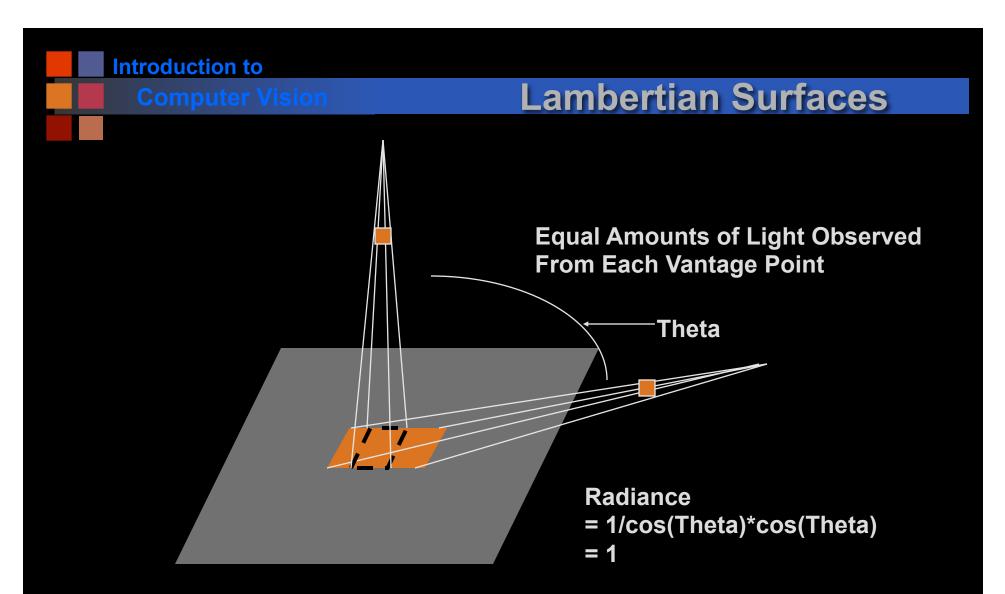
Area of black box = 1 Area of orange box = 1/cos(Theta) Foreshortening rule.



Lambertian Surfaces

Relative magnitude of light scattered in each direction. Proportional to cos (Theta).



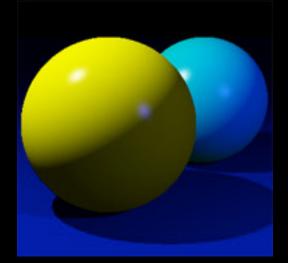


Area of black box = 1 Area of orange box = 1/cos(Theta) Foreshortening rule.



The BRDF

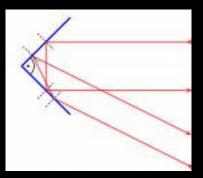
The bidirectional reflectance distribution function.



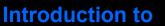






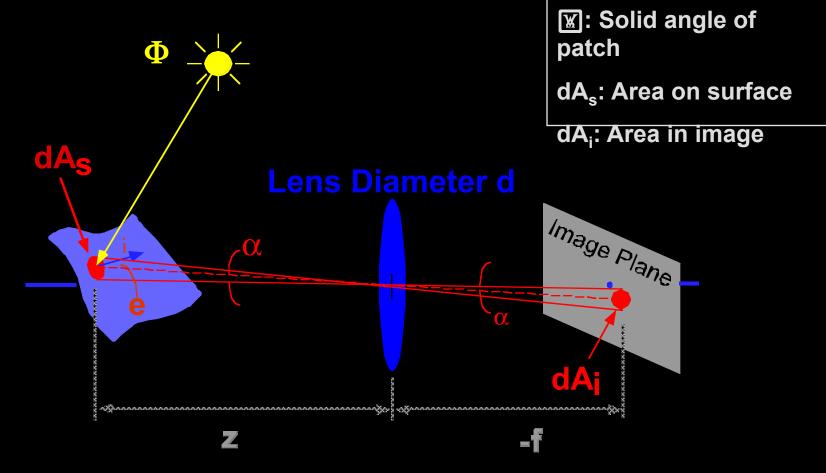


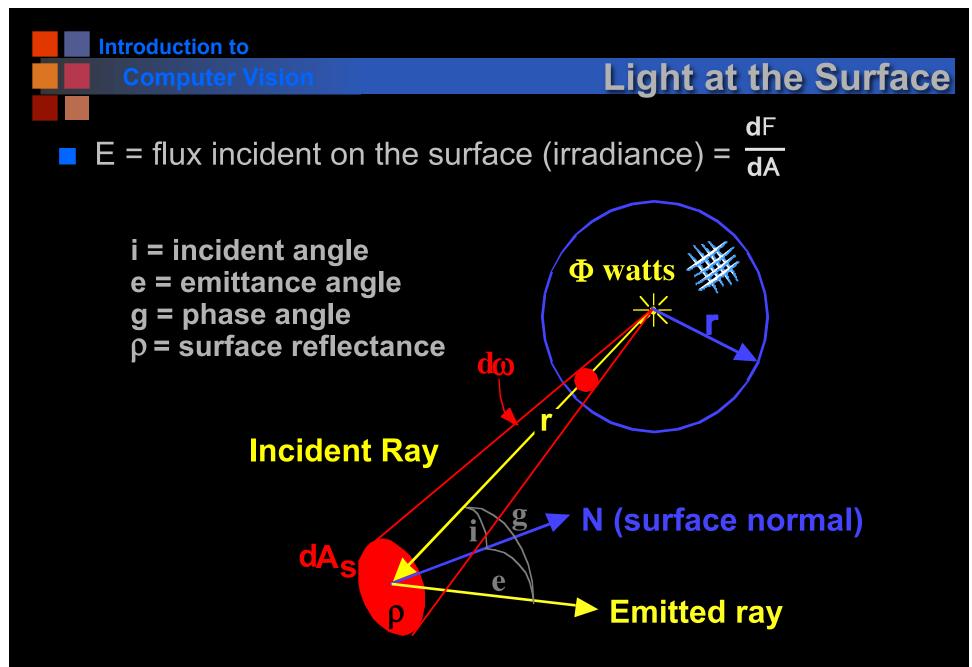




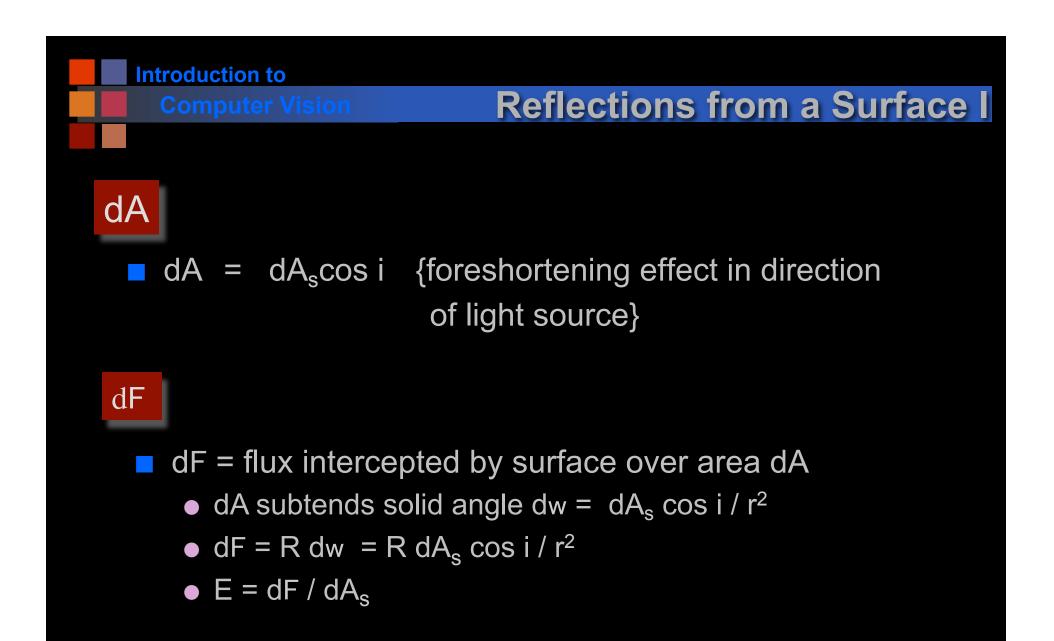
Geometry

Goal: Relate the radiance of a surface to the irradiance in the image plane of a simple optical system.

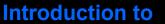




We need to determine dF and dA



Surface Irradiance: $E = R \cos i / r^2$

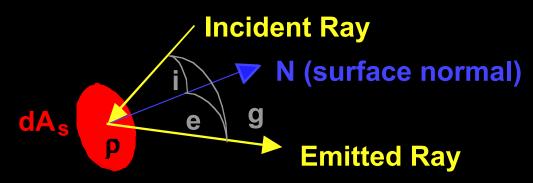


Reflections from a Surface II

Now treat small surface area as an emitter

....because it is bouncing light into the world

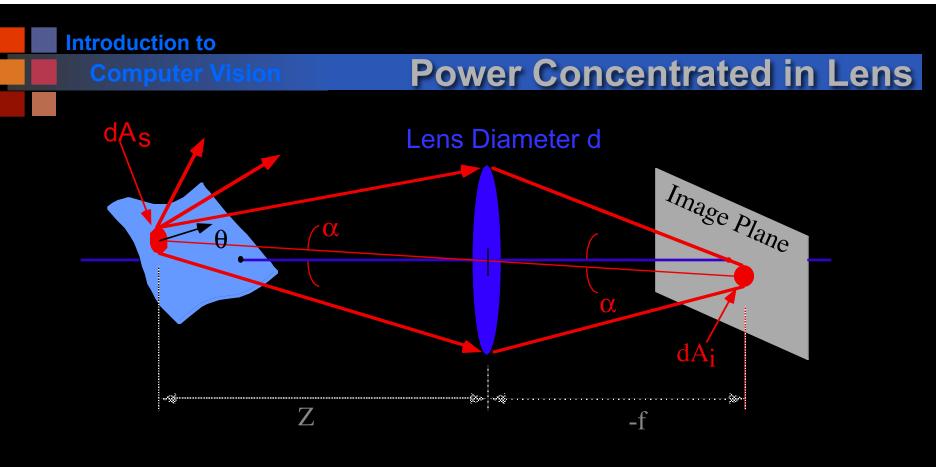
How much light gets reflected?



- E is the surface irradiance
- L is the surface radiance = luminance
- They are related through the surface reflectance function:

$$\frac{L_s}{E} = r(i,e,g,l)$$

May also be a function of the wavelength of the light



 $L_s = \frac{d^2F}{dA_sdw}$

Luminance of patch (known from previous step)

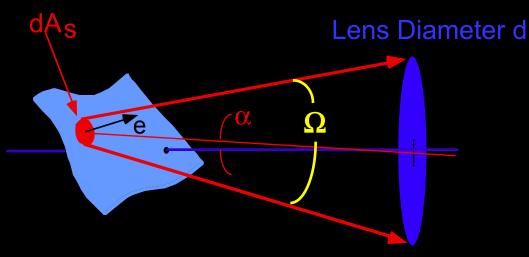
What is the power of the surface patch as a source in the direction of the lens?

$$d^{2}F = L_{s}dA_{s}dw$$



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Through a Lens Darkly

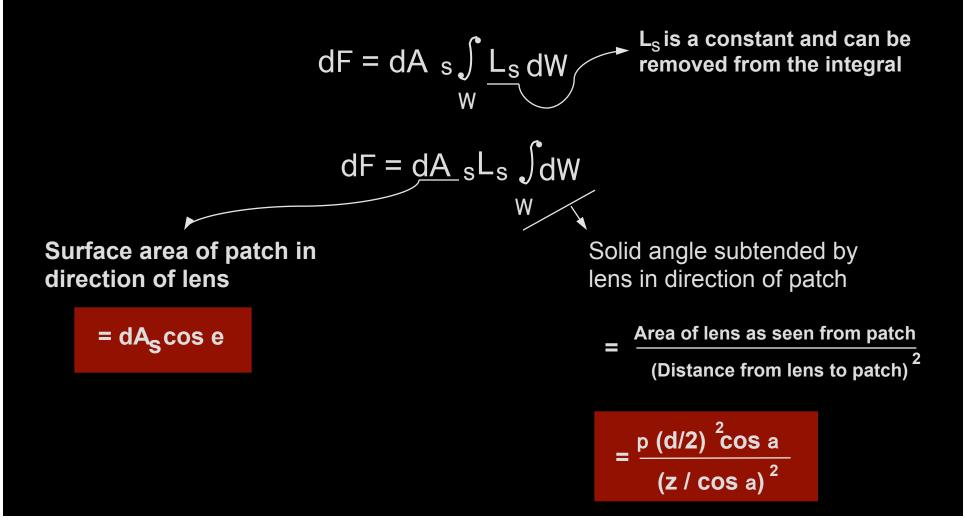


In general:

- L_s is a function of the angles i and e.
- Lens can be quite large
- Hence, must integrate over the lens solid angle to get dF

$$dF = dA_{s} \int_{W} L_{s} dW$$

Lens diameter is small relative to distance from patch



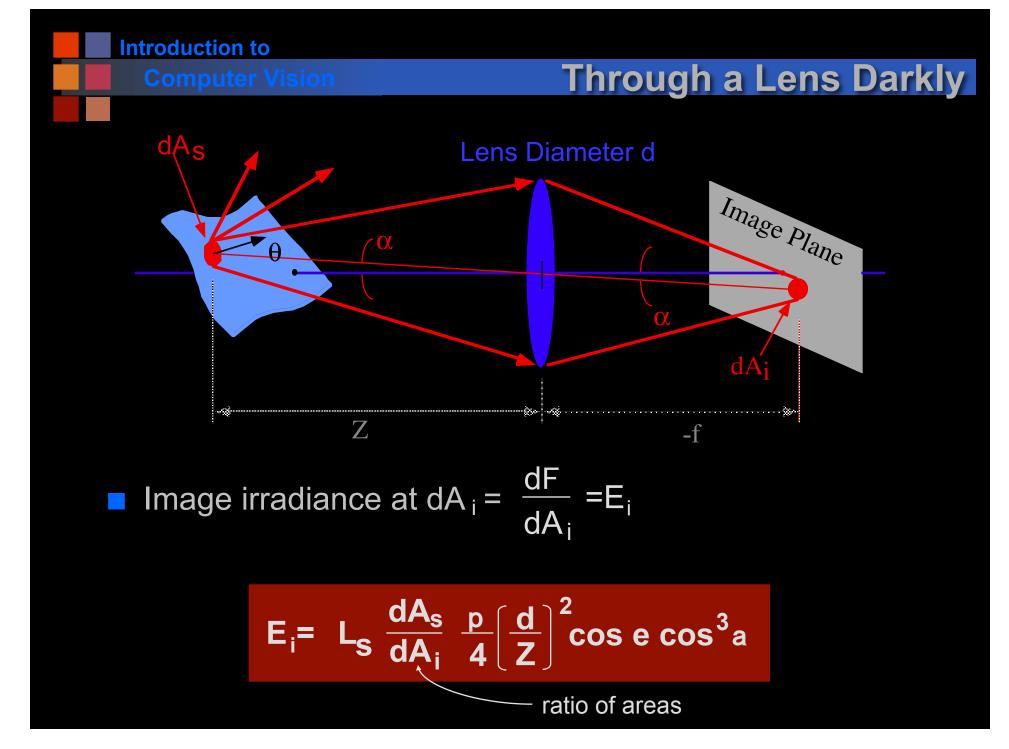


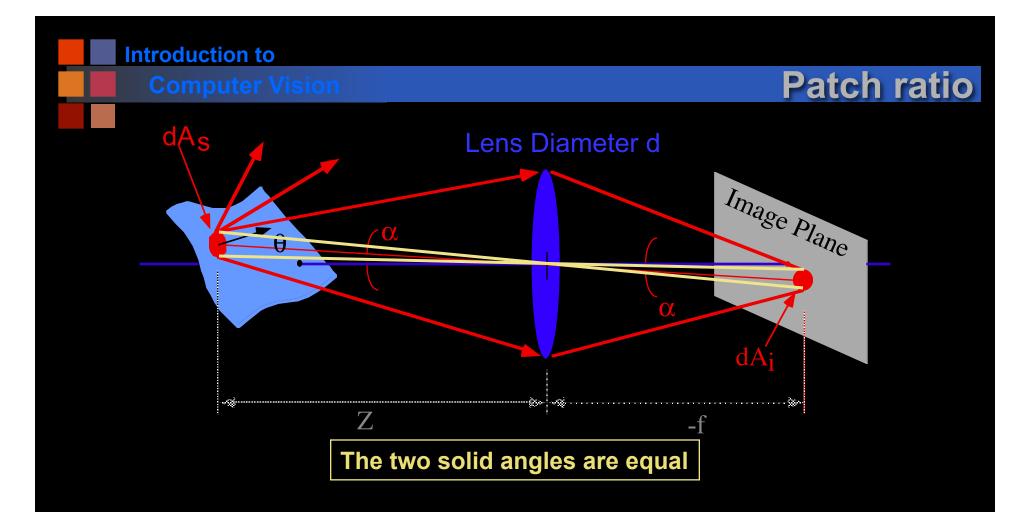
$$dF = dA_{s} \int_{W}^{L_{s}} dW$$
$$= dA_{s} \cos e L_{s} \frac{p (d/2)^{2} \cos a}{(z / \cos a)^{2}}$$

Power concentrated in lens:

dF =
$$\frac{p}{4} L_s dA_s \left[\frac{d}{Z}\right]^2 \cos e \cos^3 a$$

Assuming a lossless lens, this is also the power radiated by the lens as a source.





$$\frac{dA_{s}\cos e}{(Z/\cos a)^{2}} = \frac{dA_{i}\cos a}{(-f/\cos a)^{2}} \qquad \qquad \frac{dA_{s}}{dA_{i}} = \frac{\cos a}{\cos e} \left(\frac{Z}{-f}\right)^{2}$$

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The Fundamental Result

Source Radiance to Image Sensor Irradiance:

$$\frac{dA_{s}}{dA_{i}} = \frac{\cos a}{\cos e} \left[\frac{Z}{-f}\right]^{2}$$

$$E_i = L_s \frac{dA_s}{dA_i} \frac{p}{4} \left[\frac{d}{Z}\right]^2 \cos e \cos \frac{3}{a}$$

$$E_i = L_s \frac{\cos a}{\cos e} \left(\frac{Z}{-f}\right)^2 \frac{p}{4} \left(\frac{d}{Z}\right)^2 \cos e \cos \frac{3}{a}$$

$$E_{i} = L_{s} \frac{p}{4} \left[\frac{d}{-f}\right]^{2} \cos^{4}a$$



Radiometry Final Result

$$E_{i} = L_{s} \frac{p}{4} \left(\frac{d}{-f}\right)^{2} \cos^{4}a$$

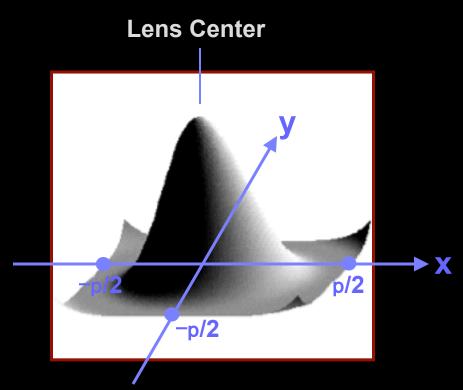
Image irradiance is a function of:

- Scene radiance L_s
- Focal length of lens f
- Diameter of lens d
 - f/d is often called the 'effective focal length' of the lens
- Off-axis angle a

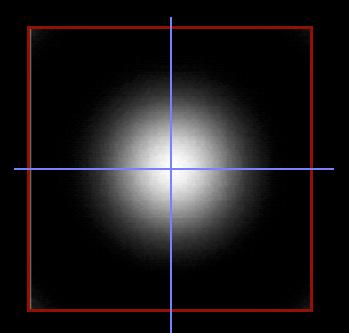


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Cos⁴ a Light Falloff



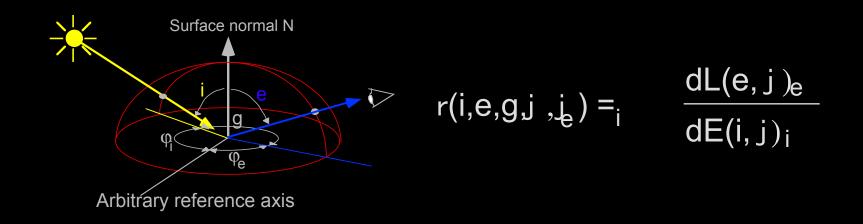
Top view shaded by height





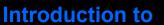
Limitation of Radiometry Model

Surface reflection r can be a function of viewing and/or illumination angle



r may also be a function of the wavelength of the light source

Assumed a point source (sky, for example, is not)

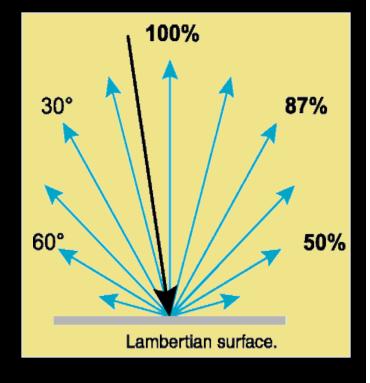


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Lambertian Surfaces

The BRDF for a Lambertian surface is a constant

- function of cos e due to the foreshortening effect
- k is the 'albedo' of the surface
- Good model for diffuse surfaces
- Other models combine diffuse and specular components (Phong, Torrance-Sparrow, Oren-Nayar)





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Ron Dror's thesis

Reflection parameters

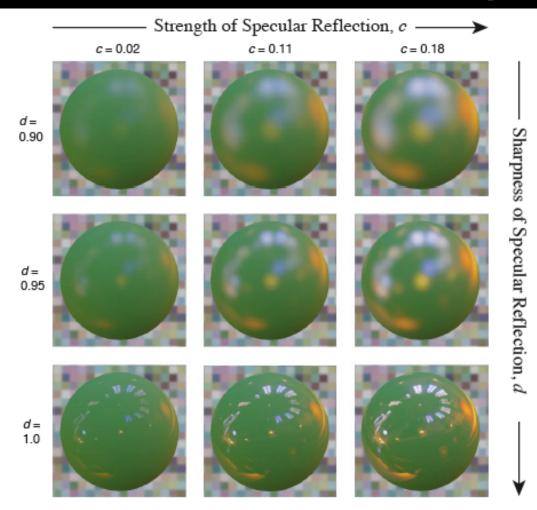


Figure 2.6. Grid showing range of reflectance properties used in the experiments for a particular real-world illumination map. All the spheres shown have an identical diffuse component. In Pellacini's reparameterization of the Ward model, the specular component depends on the c and d parameters. The strength of specular reflection, c, increases with ρ_s , while the sharpness of specular reflection, d, decreases with α . The images were rendered in *Radiance*, using the techniques described in Appendix B.



Real World Light Variation

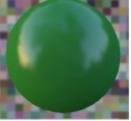
Real World Illuminations



(a) "Beach"



(b) "Building"



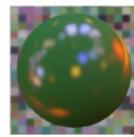
(c) "Campus"



(d) "Eucalyptus"



(e) "Galileo"



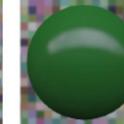
(f) "Grace"



(g) "Kitchen"



(h) "St. Peter's"



(i) "Uffizi"

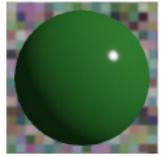




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Fake Light and Real Light

Artificial Illuminations



(a) Point source



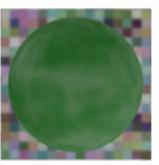
(b) Multiple points



(c) Extended



(d) White noise



(e) Pink noise



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Simple and Complex Light

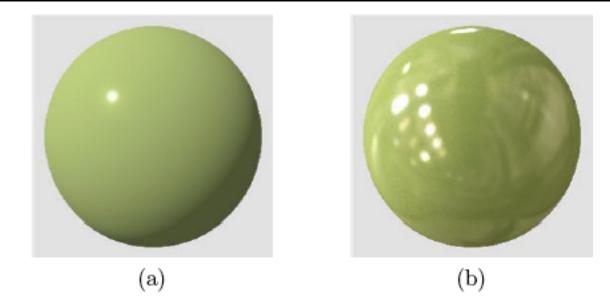
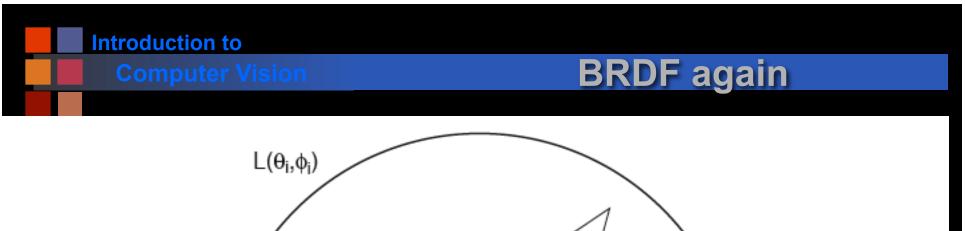


Figure 2.9. (a) A shiny sphere rendered under illumination by a point light source. (b) The same sphere rendered under photographically-acquired real-world illumination. Humans perceive reflectance properties more accurately in (b).



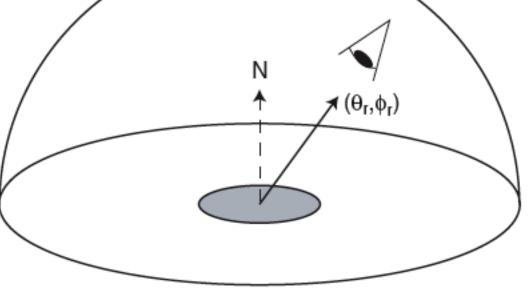


Figure 3.1. A viewer observes a surface patch with normal N from direction (θ_r, ϕ_r) . $L(\theta_i, \phi_i)$ represents radiance of illumination from direction (θ_i, ϕ_i) . The coordinate system is such that N points in direction (0, 0).

$$B(\theta_r, \phi_r) = \int_{\phi_i=0}^{2\pi} \int_{\theta_i=0}^{\pi/2} L(\theta_i, \phi_i) f(\theta_i, \phi_i; \theta_r, \phi_r) \cos \theta_i \sin \theta_i \, d\theta_i \, d\phi_i,$$



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Likelihood of material

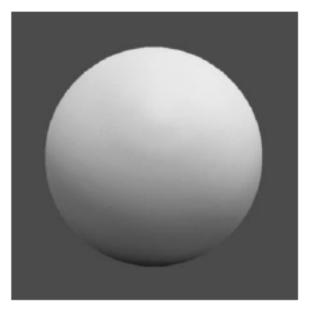


Figure 3.6. A photograph of a matte sphere, shown against a uniform gray background. This image could also be produced by a chrome sphere under appropriate illumination, but that scenario is highly unlikely.

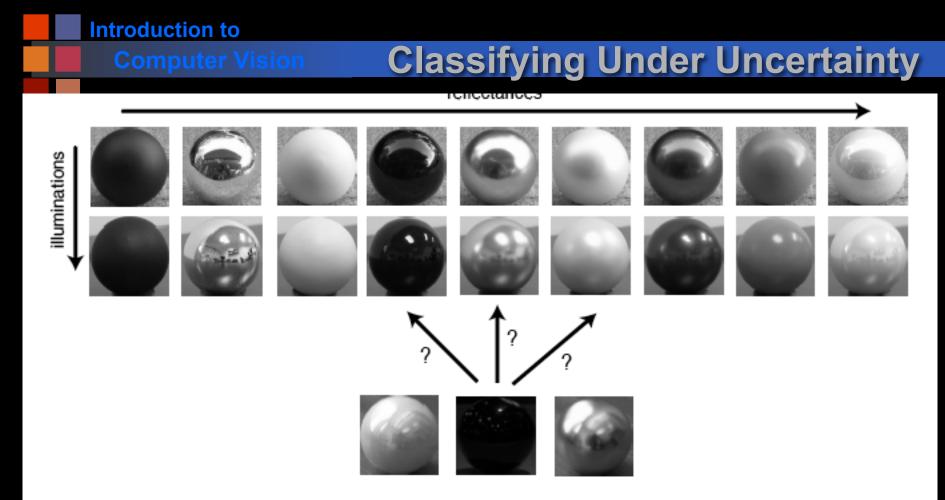
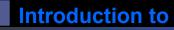


Figure 3.7. The problem addressed by a classifier of Chapter 6, illustrated using a database of photographs. Each of nine spheres was photographed under seven different illuminations. We trained a nine-way classifier using the images corresponding to several illuminations, and then used it to classify individual images under novel illuminations.

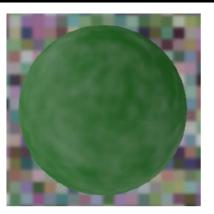


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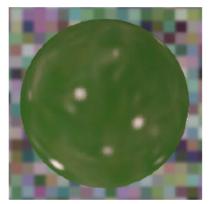
More Fake Light



(a) Original



(b) $1/f^2$ power spectrum

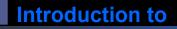


(c) Heeger and Bergen texture



(d) Portilla and Simoncelli texture

Figure 4.14. Spheres of identical reflectance properties rendered under a photographically-acquired illumination map (a) and three synthetic illumination maps (b-d). The illumination in (b) is Gaussian noise with a $1/f^2$ power spectrum. The illumination in (c) was synthesized with the procedure of Heeger and Bergen [43] to match the pixel histogram and marginal wavelet histograms of the illumination in (a). The illumination in (d) was synthesized using the technique of Portilla and Simoncelli, which also enforces conditions on the joint wavelet histograms. The illumination map of (a) is due to Debevec [24].



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Illumination Maps

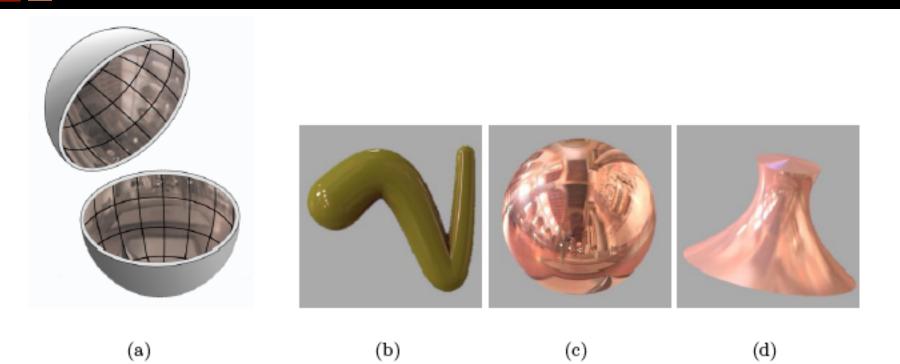


Figure 5.2. (a) A photographically-acquired illumination map, illustrated on the inside of a spherical shell. The illumination map is identical to that of Figure 4.1d. (b-d) Three surfaces of different geometry and reflectance rendered under this illumination map using the methods of Appendix B.



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Classification

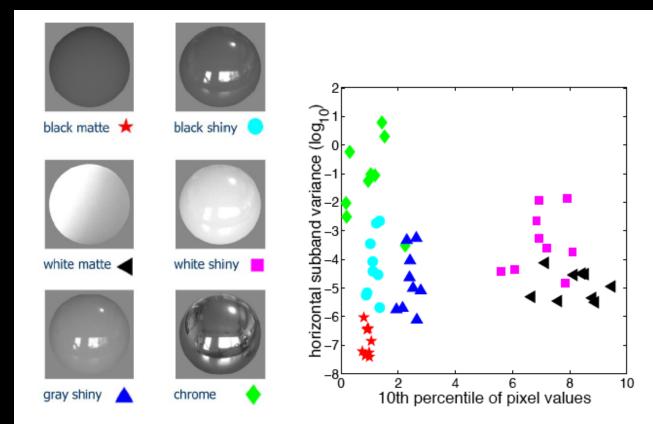
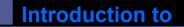


Figure 5.11. At left, synthetic spheres of 6 different reflectances, each rendered under one of Debevec's illumination maps. Ward model parameters are as follows: black matte, $\rho_d = .1$, $\rho_s = 0$; black shiny, $\rho_d = .1$, $\rho_s = .1$, $\alpha = .01$; white matte, $\rho_d = .9$, $\rho_s = 0$; white shiny, $\rho_d = .7$, $\rho_s = .25$, $\alpha = .01$; chrome, $\rho_d = 0$, $\rho_s = .75$, $\alpha = 0$; gray shiny, $\rho_d = .25$, $\rho_s = .05$, $\alpha = .01$. We rendered each sphere under the nine photographically-acquired illuminations depicted in Figure 2.7 and plotted a symbol corresponding to each in the two-dimensional feature space at right. The horizontal axis represents the 10th percentile of pixel intensity, while the vertical axis is the log variance of horizontally-oriented QMF wavelet coefficients at the second-finest scale, computed after geometrically distorting the original image as described in Section 6.1.2.



Computer Vision

Classification Continued

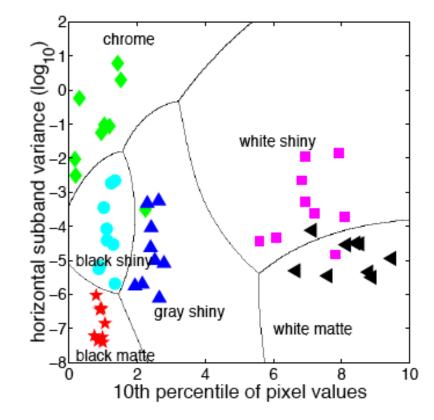


Figure 5.12. The curved lines separate regions assigned to different reflectances by a simple classifier based on two image features. The training examples are the images described in Figure 5.11. The classifier is a one-versus-all support vector machine, described in Section 6.1.1. Using additional image features improves classifier performance.

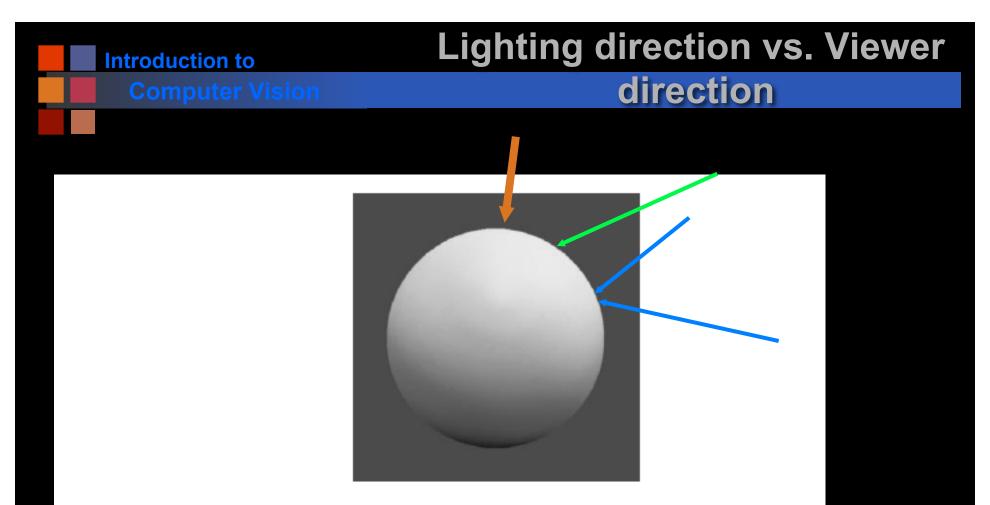


Figure 3.6. A photograph of a matte sphere, shown against a uniform gray background.

For Lambertian surface:

Viewer direction is irrelevant

Lighting direction is very relevant



Computer Vision

