

Computer Vision

Spring 2006 15-385,-685

Instructor: S. Narasimhan

Wean 5403

T-R 3:00pm – 4:20pm

Lecture #13

Announcements

Homework 4 will be out today. Due 4/4/06.
Please start early.

Midterm stats: A range → 40+, B range → 30+

40+ → 13 students

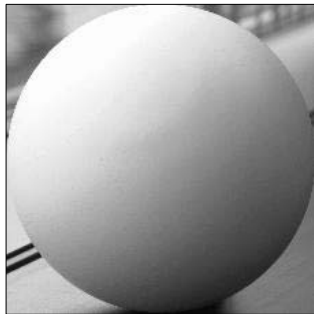
30+ → 9 students

Below 30 → 6 students

Shape from Shading

Lecture #13

Image Intensity and 3D Geometry



- Shading as a cue for shape reconstruction
- What is the relation between intensity and shape?
 - Reflectance Map

Reflectance Map - RECAP

- Relates image irradiance $I(x,y)$ to surface orientation (p,q) for given source direction and surface reflectance
- Lambertian case:

k : source brightness

ρ : surface albedo (reflectance)

c : constant (optical system)

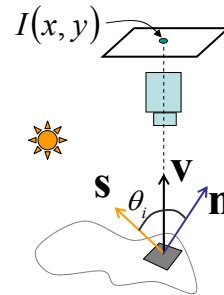


Image irradiance:

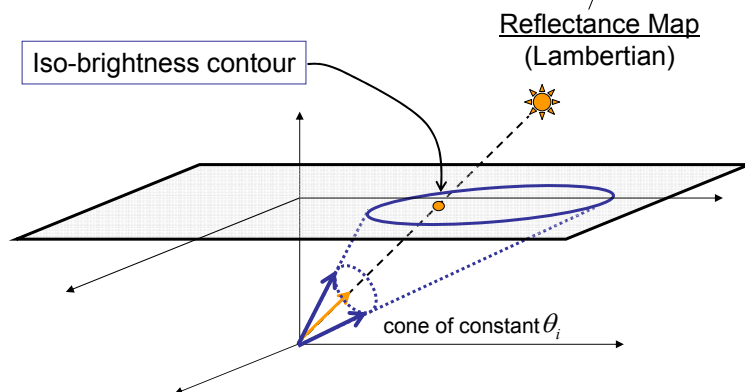
$$I = \frac{\rho}{\pi} kc \cos \theta_i = \frac{\rho}{\pi} kc \mathbf{n} \cdot \mathbf{s}$$

Let $\frac{\rho}{\pi} kc = 1$ then $I = \cos \theta_i = \mathbf{n} \cdot \mathbf{s}$

Reflectance Map - RECAP

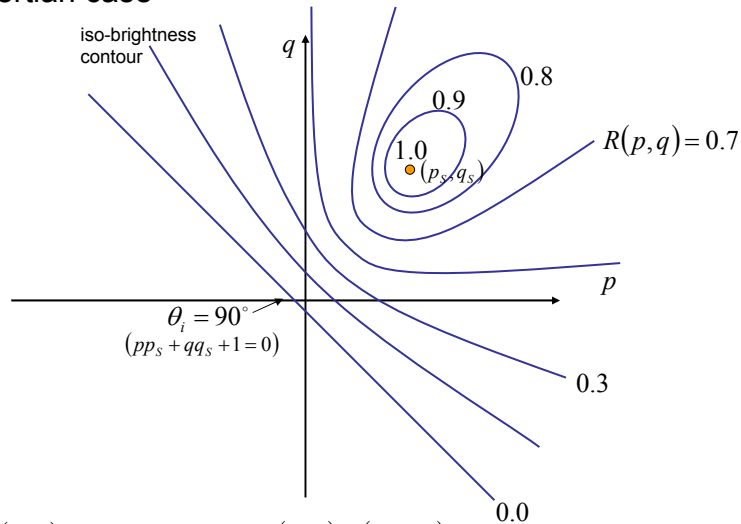
- Lambertian case

$$I = \cos \theta_i = \mathbf{n} \cdot \mathbf{s} = \frac{(pp_s + qq_s + 1)}{\sqrt{p^2 + q^2 + 1} \sqrt{p_s^2 + q_s^2 + 1}} = R(p, q)$$



Reflectance Map - RECAP

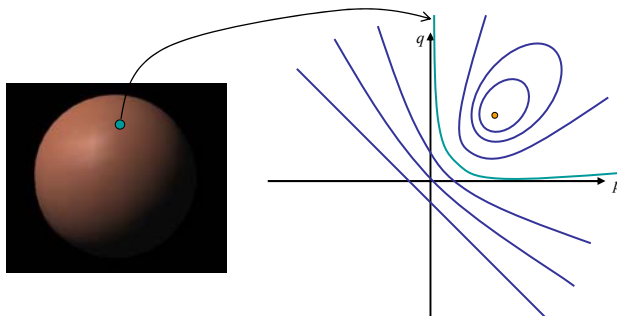
- Lambertian case



Note: $R(p, q)$ is maximum when $(p, q) = (p_s, q_s)$

Shape from a Single Image?

- Given a single image of an object with known surface reflectance taken under a known light source, can we recover the shape of the object?
- Given $R(p, q)$ (p_s, q_s and surface reflectance) can we determine (p, q) uniquely for each image point?

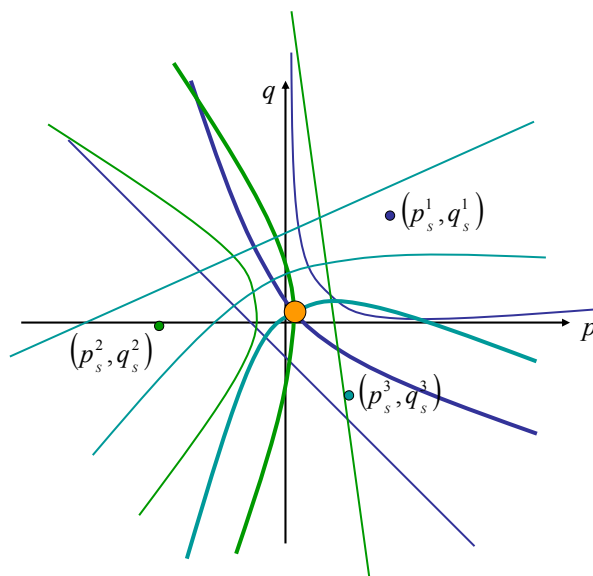


NO

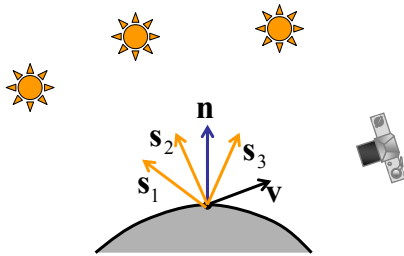
Solution

- Take more images
 - Photometric stereo (previous class)
- Add more constraints
 - Shape-from-shading (this class)

Photometric Stereo



Photometric Stereo



Lambertian case:

$$I = \frac{\rho}{\pi} kc \cos \theta_i = \rho \mathbf{n} \cdot \mathbf{s} \quad \left(\frac{kc}{\pi} = 1 \right)$$

Image irradiance:

$$I_1 = \rho \mathbf{n} \cdot \mathbf{s}_1$$

$$I_2 = \rho \mathbf{n} \cdot \mathbf{s}_2$$

$$I_3 = \rho \mathbf{n} \cdot \mathbf{s}_3$$

- We can write this in matrix form:

$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \rho \begin{bmatrix} \mathbf{s}_1^T \\ \mathbf{s}_2^T \\ \mathbf{s}_3^T \end{bmatrix} \mathbf{n}$$

Solution

- Take more images
 - Photometric stereo (previous class)
- Add more constraints
 - Shape-from-shading (this class)

Human Perception

- Our brain often perceives shape from shading.
- Mostly, it makes many assumptions to do so.
- For example:

Light is coming from above (sun).

Biased by occluding contours.

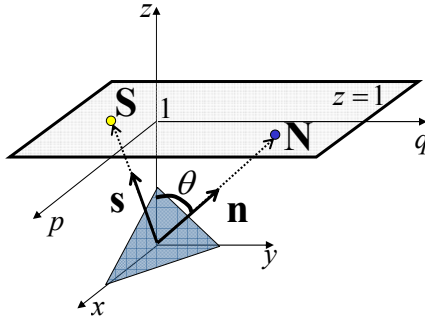
by V. Ramachandran

See Ramachandran's work on Shape
from Shading by Humans

<http://psy.ucsd.edu/chip/ramabio.html>

Stereographic Projection

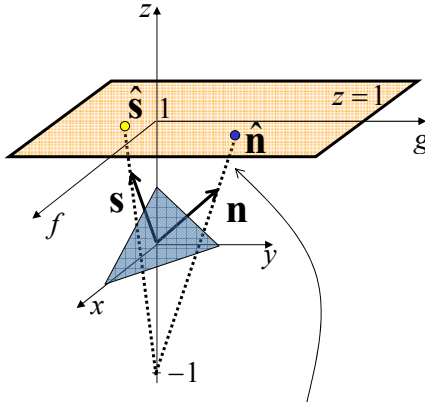
(p,q) -space (gradient space)



Problem

(p,q) can be infinite when $\theta = 90^\circ$

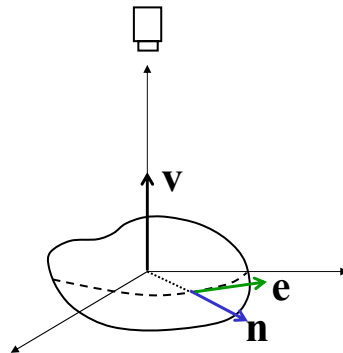
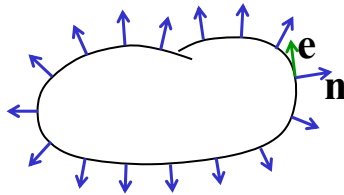
(f,g) -space



$$f = \frac{2p}{1 + \sqrt{1 + p^2 + q^2}} \quad g = \frac{2q}{1 + \sqrt{1 + p^2 + q^2}}$$

Redefine reflectance map as $R(f,g)$

Occluding Boundaries



$$\mathbf{n} \perp \mathbf{e}, \quad \mathbf{n} \perp \mathbf{v} \quad \therefore \quad \mathbf{n} = \mathbf{e} \times \mathbf{v} \quad \mathbf{e} \text{ and } \mathbf{v} \text{ are known}$$

The \mathbf{n} values on the occluding boundary can be used as the boundary condition for shape-from-shading

Image Irradiance Constraint

- Image irradiance should match the reflectance map

Minimize

$$e_i = \iint_{\text{image}} (I(x, y) - R(f, g))^2 dx dy$$

(minimize errors in image irradiance in the image)

Smoothness Constraint

- Used to constrain shape-from-shading
- Relates orientations (f, g) of neighboring surface points

Minimize

$$e_s = \iint_{\text{image}} (f_x^2 + f_y^2) + (g_x^2 + g_y^2) dx dy$$

(f, g) : surface orientation under stereographic projection

$$f_x = \frac{\partial f}{\partial x}, f_y = \frac{\partial f}{\partial y}, g_x = \frac{\partial g}{\partial x}, g_y = \frac{\partial g}{\partial y}$$

(penalize rapid changes in surface orientation f and g over the image)

Shape-from-Shading

- Find surface orientations (f, g) at all image points that minimize

$$e = e_s + \lambda e_i$$

weight
 ↓
 λ
 ↑
 smoothness
 constraint
 ↑
 image irradiance
 error

Minimize

$$e = \iint_{\text{image}} (f_x^2 + f_y^2) + (g_x^2 + g_y^2) + \lambda (I(x, y) - R(f, g))^2 dx dy$$

Numerical Shape-from-Shading

- Smoothness error** at image point (i, j)

$$s_{i,j} = \frac{1}{4} \left((f_{i+1,j} - f_{i,j})^2 + (f_{i,j+1} - f_{i,j})^2 + (g_{i+1,j} - g_{i,j})^2 + (g_{i,j+1} - g_{i,j})^2 \right)$$

Of course you can consider more neighbors (smoother results)

- Image irradiance error** at image point (i, j)

$$r_{i,j} = (I_{i,j} - R(f_{i,j}, g_{i,j}))^2$$

Find $\{f_{i,j}\}$ and $\{g_{i,j}\}$ that minimize

$$e = \sum_i \sum_j (s_{i,j} + \lambda r_{i,j})$$

(Ikeuchi & Horn 89)

Numerical Shape-from-Shading

Find $\{f_{i,j}\}$ and $\{g_{i,j}\}$ that minimize $e = \sum_i \sum_j (s_{i,j} + \lambda r_{i,j})$

If $f_{k,l}$ and $g_{k,l}$ minimize e , then $\frac{\partial e}{\partial f_{k,l}} = 0, \frac{\partial e}{\partial g_{k,l}} = 0$

$$\frac{\partial e}{\partial f_{k,l}} = 2(f_{k,l} - \bar{f}_{k,l}) - 2\lambda(I_{k,l} - R(f_{k,l}, g_{k,l})) \frac{\partial R}{\partial f} \bigg|_{f_{k,l}} = 0$$

$$\frac{\partial e}{\partial g_{k,l}} = 2(g_{k,l} - \bar{g}_{k,l}) - 2\lambda(I_{k,l} - R(f_{k,l}, g_{k,l})) \frac{\partial R}{\partial g} \bigg|_{g_{k,l}} = 0$$

where $\bar{f}_{k,l}$ and $\bar{g}_{k,l}$ are 4-neighbors average around image point (k,l)

$$\bar{f}_{k,l} = \frac{1}{8}(f_{i+1,j} + f_{i,j+1} + f_{i-1,j} + f_{i,j-1})$$

$$\bar{g}_{k,l} = \frac{1}{8}(g_{i+1,j} + g_{i,j+1} + g_{i-1,j} + g_{i,j-1}) \quad (\text{Ikeuchi \& Horn 89})$$

Numerical Shape-from-Shading

$$\frac{\partial e}{\partial f_{k,l}} = 2(f_{k,l} - \bar{f}_{k,l}) - 2\lambda(I_{k,l} - R(f_{k,l}, g_{k,l})) \frac{\partial R}{\partial f} \bigg|_{f_{k,l}} = 0 \quad \frac{\partial e}{\partial g_{k,l}} = 2(g_{k,l} - \bar{g}_{k,l}) - 2\lambda(I_{k,l} - R(f_{k,l}, g_{k,l})) \frac{\partial R}{\partial g} \bigg|_{g_{k,l}} = 0$$

Update rule

$$f_{k,l}^{n+1} = \bar{f}_{k,l}^n + \lambda(I_{k,l} - R(f_{k,l}^n, g_{k,l}^n)) \frac{\partial R}{\partial f} \bigg|_{f_{k,l}^n}$$

$$g_{k,l}^{n+1} = \bar{g}_{k,l}^n + \lambda(I_{k,l} - R(f_{k,l}^n, g_{k,l}^n)) \frac{\partial R}{\partial g} \bigg|_{g_{k,l}^n}$$

- Use known (f, g) values on the occluding boundary to constrain the solution (boundary conditions)
- Compare $(f_{k,l}^{n+1}, g_{k,l}^{n+1})$ with $(f_{k,l}^n, g_{k,l}^n)$ for convergence test
- As the solution converges, increase λ to remove the smoothness constraint

(Ikeuchi & Horn 89)

Calculus of Variations

Minimize

$$e = \iint_{\text{image}} F(f, g, f_x, f_y, g_x, g_y) dx dy$$

$$F = (f_x^2 + f_y^2) + (g_x^2 + g_y^2) + \lambda (I(x, y) - R(f, g))^2$$

Euler equations for F

(read Horn A.6)

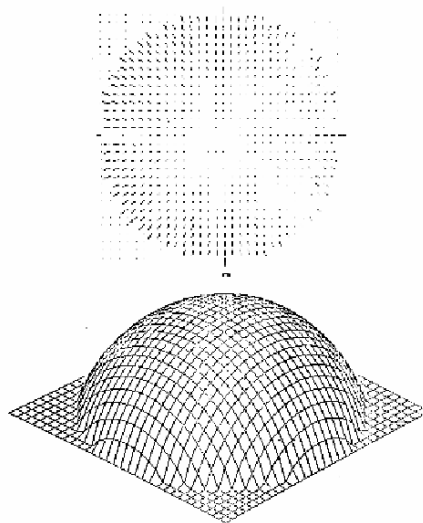
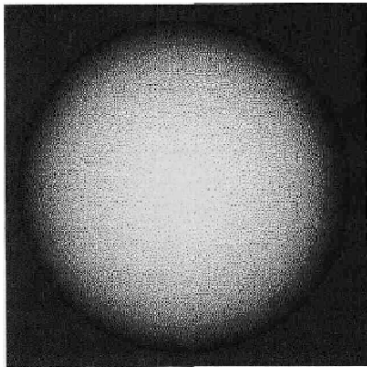
$$F_f - \frac{\partial}{\partial x} F_{f_x} - \frac{\partial}{\partial y} F_{f_y} = 0, F_g - \frac{\partial}{\partial x} F_{g_x} - \frac{\partial}{\partial y} F_{g_y} = 0$$

Euler equations for shape-from-shading

$$\nabla^2 f = -\lambda (I(x, y) - R(f, g)) \frac{\partial R}{\partial f}, \nabla^2 g = -\lambda (I(x, y) - R(f, g)) \frac{\partial R}{\partial g}$$

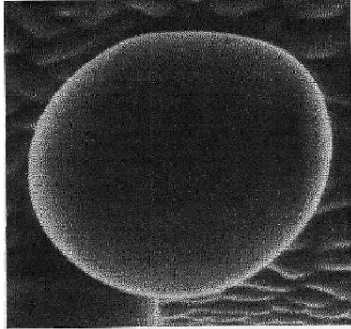
Solve this coupled pair of second-order partial differential equations
with the occluding boundary conditions!

Results



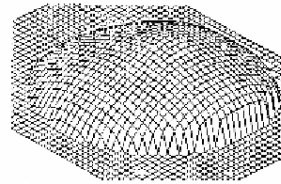
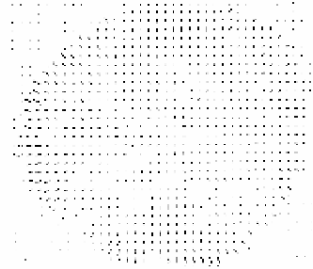
by Ikeuchi and Horn

Results



Scanning Electron Microscope image
(inverse intensity)

by Ikeuchi and Horn



Next Two Classes

- Binocular Stereo
- Reading: Horn, Chapter 13.