

Photometric Stereo Recap + Lambertian

Computer Vision I CSE252A Lecture 8a

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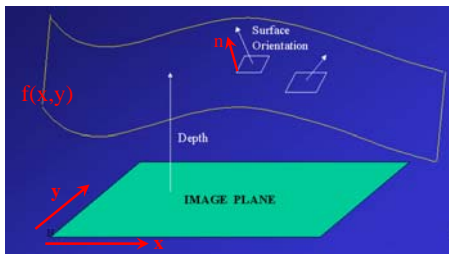
Announcements

- HW1 due today

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Coordinate system



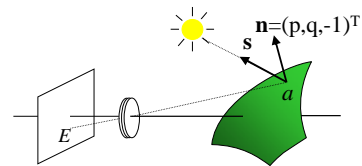
Gradient Space: (p,q)
 $p = \frac{\partial f}{\partial x}, q = \frac{\partial f}{\partial y}$

Normal vector
 $\mathbf{n} = \frac{\partial s}{\partial x} \times \frac{\partial s}{\partial y} = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, -1 \right)^T$
 $\hat{\mathbf{n}} = \frac{1}{\sqrt{p^2 + q^2 + 1}} (p, q, -1)^T$

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Reflectance map



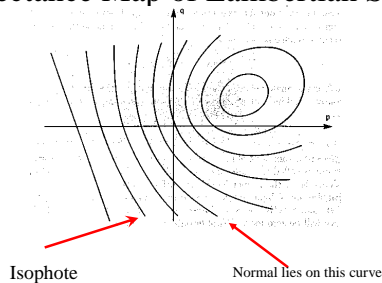
For known BRDF, fix light source direction/strength and projection direction

1. Then image irradiance is a function of only the direction of the surface normal.
2. In gradient space, we have $E(p,q)$ which is known as the reflectance map

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Reflectance Map of Lambertian Surface

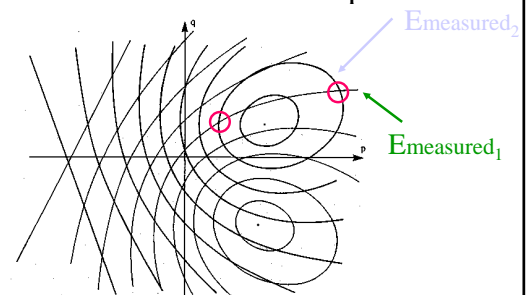


What does the value of one pixel in one image tell us?
It constrains normal to a curve

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Two Light Sources Two reflectance maps



Third image would disambiguate match

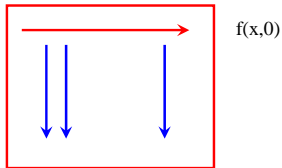
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Recovering the surface $f(x,y)$

Many methods: Simplest approach

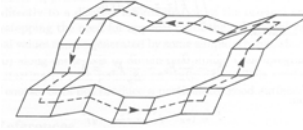
1. From estimate $\mathbf{n}=(n_x, n_y, n_z)$, $p=n_x/n_z$, $q=n_y/n_z$
2. Integrate $p=df/dx$ along a row $(x,0)$ to get $f(x,0)$
3. Then integrate $q=df/dy$ along each column starting with value of first row



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What might go wrong?



- Height $z(x,y)$ is obtained by integration along a curve from (x_0, y_0) .

$$z(x, y) = z(x_0, y_0) + \int_{(x_0, y_0)}^{(x, y)} (p dx + q dy)$$

- If one integrates the derivative field along any closed curve, one expects to get back to the starting value.
- Might not happen because of noisy estimates of (p,q)

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Integrability

If $f(x,y)$ is the height function, we expect that

$$\frac{\partial}{\partial y} \frac{\partial f}{\partial x} = \frac{\partial}{\partial x} \frac{\partial f}{\partial y}$$

In terms of estimated gradient space (p,q) , this means:

$$\frac{\partial p}{\partial y} = \frac{\partial q}{\partial x}$$

But since p and q were estimated independently at each point as intersection of curves on three reflectance maps, equality is not going to exactly hold

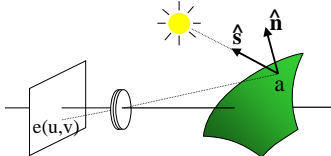


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II. Photometric Stereo: Lambertian Surface, Known Lighting

Lambertian Surface



At image location (u,v) , the intensity of a pixel $x(u,v)$ is:

$$e(u,v) = [a(u,v) \hat{\mathbf{n}}(u,v)] \cdot [s_0 \hat{\mathbf{s}}] = \mathbf{b}(u,v) \cdot \mathbf{s}$$

where

- $a(u,v)$ is the albedo of the surface projecting to (u,v) .
- $\mathbf{n}(u,v)$ is the direction of the surface normal.
- s_0 is the light source intensity.
- \mathbf{s} is the direction to the light source.

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Lambertian Photometric stereo

- If the light sources $\mathbf{s}_1, \mathbf{s}_2$, and \mathbf{s}_3 are **known**, then we **can** recover \mathbf{b} from as few as three images. (Photometric Stereo: Silver 80, Woodham81).

$$[e_1 \ e_2 \ e_3] = \mathbf{b}^T [\mathbf{s}_1 \ \mathbf{s}_2 \ \mathbf{s}_3]$$

- i.e., we measure e_1, e_2 , and e_3 and we know $\mathbf{s}_1, \mathbf{s}_2$, and \mathbf{s}_3 . We can then solve for \mathbf{b} by solving a linear system.

$$\mathbf{b}^T = [e_1 \ e_2 \ e_3] [\mathbf{s}_1 \ \mathbf{s}_2 \ \mathbf{s}_3]^{-1}$$

- Normal is: $\mathbf{n} = \mathbf{b}/|\mathbf{b}|$, albedo is: $|\mathbf{b}|$

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What if we have more than 3 Images? Linear Least Squares

$$[e_1 \ e_2 \ e_3] = \mathbf{b}^T [s_1 \ s_2 \ s_3]$$

Rewrite as

$$\mathbf{e} = \mathbf{S}\mathbf{b}$$

where

\mathbf{e} is n by 1

\mathbf{b} is 3 by 1

\mathbf{S} is n by 3

Let the residual be

$$\mathbf{r} = \mathbf{e} - \mathbf{S}\mathbf{b}$$

Squaring this:

$$\begin{aligned} |\mathbf{r}|^2 &= \mathbf{r}^T \mathbf{r} = (\mathbf{e} - \mathbf{S}\mathbf{b})^T (\mathbf{e} - \mathbf{S}\mathbf{b}) \\ &= \mathbf{e}^T \mathbf{e} - 2\mathbf{b}^T \mathbf{S}^T \mathbf{e} + \mathbf{b}^T \mathbf{S}^T \mathbf{S} \mathbf{b} \end{aligned}$$

$$\frac{\partial \mathbf{r}^2}{\partial \mathbf{b}} = 0 \quad \text{- zero derivative is a necessary condition for a minimum, or}$$

$$-2\mathbf{S}^T \mathbf{e} + 2\mathbf{S}^T \mathbf{S} \mathbf{b} = 0;$$

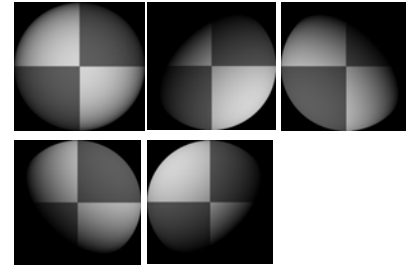
Solving for \mathbf{b} gives

$$\mathbf{b} = (\mathbf{S}^T \mathbf{S})^{-1} \mathbf{S}^T \mathbf{e}$$

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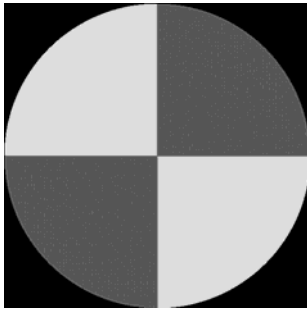
Input Images



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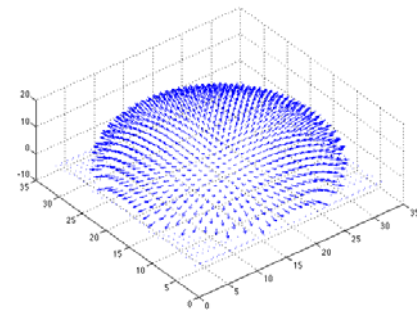
Recovered albedo



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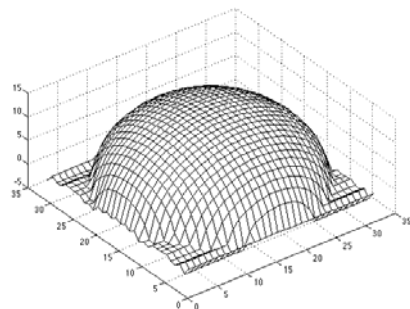
Recovered normal field



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Surface recovered by integration



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