# Photometric Stereo, Shape from Shading SfS F\&P Ch 5 (old) Ch 2 (new) 

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## Example

Reconstruct 3D Face Surface from unstructured image data


Kemelmacher et al., "Face Reconstruction from the Wild", ICCV 2011


## Photometric Stereo



## Depth from Shading?

Photometric Normals


## Photometric Stereo



Depth from Shading?
First step: Surface Normals from Shading


Second step:
Re-integration of surface from Normals


## Examples


http://www.youtube.com/watch?v=sfCQ7f7PMbc\&feature=related


Simulated voyage over the surface of Neptune's large moon Triton
http://www.youtube.com/watch?v=nwzVrC2GQXE


## Shape from Shading

Inverting the image formation process


Image formation $=$ "Shading from shape" (and light sources)

## Shape from Shading

## Authors: Emmanuel Prados and Olivier Faugeras

CVPR'2005, International Conference on Computer Vision and Pattern Recognition, San Diego, CA, USA, June 2005.

a)

b)

a) Synthetic image generated from the classical Mozart's face [Zhang-Tsai-etal:99]; b) reconstructed surface from a) by new algorithm; c) real image of a face; d)-e) reconstructed surface from c) by new algorithm.

## Photometric Stereo

- Assume:
- a local shading model
- 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)


## Setting for Photometric Stereo

Multiple images with different lighting (vs binocular/geometric stereo)


## Goal: 3D from One View and multiple Source positions

Input images



## Scene Results

## Needle Diagram: Surface Normals



## Albedo



## Projection model for surface recovery usually called a Monge patch



## Lambertian Reflectance Map

## LAMBERTIAN MODEL

$$
E=\rho<\mathbf{n}, \mathbf{n}_{\mathbf{L}}>=\rho \operatorname{Cos} \theta
$$



$$
\cos \theta=\frac{1+p p_{L}+q q_{L}}{\sqrt{1+p^{2}+q^{2}} \sqrt{1+{p_{L}^{2}}^{2}+{q_{L}^{2}}^{2}}}
$$

## REFLECTANCE MAP IS A VIEWER-CENTERED REPRESENTATION OF REFLECTANCE

$$
\left(f_{x}, f_{y},-1\right)=
$$



## REFLECTANCE MAP IS A VIEWER-CENTERED REPRESENTATION OF REFLECTANCE

$$
(-f x,-f y, 1)=(-p,-q, 1)
$$

p, q comprise a gradient or gradient space representation for local surface orientation.

Reflectance map expresses the reflectance of a material directly in terms of viewer-centered representation of local surface orientation.

## Reflectance Map

Shading on Lambertian surface - Overhead point source


## Reflectance Map (ps=0, qs=0)

The Reflectance Map - Lambertian surface from overhead source position

$$
R(p, q)=\frac{1}{\sqrt{p^{2}+q^{2}+1}}
$$




## Reflectance Map

## Shape from Shading

Shading on Lambertian surface - General point source

$$
I=\rho(\hat{N} \cdot \hat{L})=\rho \frac{-p \cdot L_{x}-q \cdot L_{y}+L_{z}}{\sqrt{p^{2}+q^{2}+1} \sqrt{L_{x}^{2}+L_{y}^{2}+L_{z}^{2}}}=\rho \frac{p \cdot p_{L}+q \cdot q_{L}+1}{\sqrt{p^{2}+q^{2}+1} \sqrt{p_{L}^{2}+q_{L}^{2}+1}}
$$



## Reflectance Map

The Reflectance Map - Lambertian surface from general source position

$$
R(p, q)=\frac{p \cdot p_{L}+q \cdot q_{L}+1}{\sqrt{p^{2}+q^{2}+1} \sqrt{p_{L}^{2}+q_{L}^{2}+1}}
$$



## Reflectance Map (General)



Figure 10-13. The reflectance map is a plot of brightness as a function of surface orientation. Here it is shown as a contour map in gradient space. In the case of a Lambertian surface under point-source illumination, the contours turn out to be nested conic sections. The maximum of $R(p, q)$ occurs at the point $(p, q)=\left(p_{s}, q_{s}\right)$, found inside the nested conic sections, while $R(p, q)=0$ all along the line on the left side of the contour map.

## Reflectance Map



Figure 10-13. The reflectance map is a plot of brightness as a function of surface orientation. Here it is shown as a contour map in gradient space. In the case of a Lambertian surface under point-source illumination, the contours turn out to be nested conic sections. The maximum of $R(p, q)$ occurs at the point $(p, q)=\left(p_{s}, q_{s}\right)$, found inside the nested conic sections, while $R(p, q)=0$ all along the line on the left side of the contour map.

## Multiple Images = Multiple Maps

## Can isolate $\mathrm{p}, \mathrm{q}$ as contour intersection



Figure 10-21. In the case of a Lambertian surface illuminated successively by two different point sources, there are at most two surface orientations that produce a particular pair of brightness values. These are found at the intersection of the corresponding contours in two superimposed reflectance maps.

## Example: Two Views



Still not unique for certain intensity pairs.

## Constant Albedo



## Varying Albedo

## Solution Forsyth \& Ponce:

For each point source, we know the source vector (by assumption). We assume we know the scaling constant of the linear camera (k). Fold the normal ( $\mathbf{N}$ ) and the reflectance ( $\boldsymbol{\rho}(\mathrm{x}, \mathrm{y})$ ) into one vector $\mathbf{g}$, and the scaling constant and source vector

- Out of shadow:

$$
\begin{aligned}
I(x, y) & =k B(\boldsymbol{x}) \\
& =k B(x, y) \\
& =\underbrace{k} \rho(x, y) \boldsymbol{N}(x, y) \cdot\left(\boldsymbol{S}_{1}\right. \\
& =\boldsymbol{g}(x, y) \cdot \boldsymbol{V}_{1}
\end{aligned}
$$

- In shadow:

$$
I(x, y)=0
$$ into another $\mathbf{V}_{\mathbf{j}}$.

where $\boldsymbol{g}(x, y)=\rho(x, y) \boldsymbol{N}(x, y)$ and $\boldsymbol{V}_{1}=k \boldsymbol{S}_{1}$, where $k$ is the constant connecting the camera response to the input radiance.

## Multiple Images:

## Linear Least Squares Approach

- Combine albedo and normal
- Separate lighting parameters
- More than 3 images => overdetermined system

$$
\begin{aligned}
& \mathcal{V}=\left(\begin{array}{c}
\boldsymbol{V}_{1}^{T} \\
\boldsymbol{V}_{2}^{T} \\
\ldots \\
\boldsymbol{V}_{n}^{T}
\end{array}\right) \quad \boldsymbol{i}(x, y)=\left\{I_{1}(x, y), I_{2}(x, y), \ldots, I_{n}(x, y)\right\}^{T} \\
& \boldsymbol{i}(x, y)=\mathcal{V} \boldsymbol{g}(x, y) \\
& \boldsymbol{g} \text { is obtained by solving this linear system: } \bar{g}(\mathbf{x}, \mathrm{y})=V^{-1} \mathbf{i}(\mathrm{x}, \mathrm{y})
\end{aligned}
$$

- How to calculate albedo $\rho$ and $\bar{N}$ ?

$$
\begin{aligned}
& \bar{g}(x, y)=\rho(x, y) \bar{N}(\mathrm{x}, \mathrm{y}) \\
& \rightarrow \bar{N}=\frac{\bar{g}}{|\bar{g}|}, \quad \rho(x, y)=|\bar{g}|
\end{aligned}
$$

## Example LLS Input



Problem: Some regions in some images are in the shadow (no image intensity).

## Dealing with Shadows (Missing Info)

or each point source, we know the source vector (by assumption). We assume we know the scaling constant of the linear camera. Fold the normal and the reflectance into one vector g , and the scaling constant and source vector into another Vj

- Out of shadow:
$I_{j}(x, y)=k B(x, y)$
$=k \rho(x, y)\left(\mathbf{N}(x, y) \bullet \mathbf{S}_{j}\right)$
$=\mathbf{g}(x, y) \bullet \mathbf{V}_{j}$
- In shadow:
$\mathrm{I}_{\mathrm{j}}(\mathrm{x}, \mathrm{y})=0$
No partial shadow


## Matrix Trick for Complete Shadows

- Matrix from Image Vector:

$$
\mathcal{I}(x, y)=\left(\begin{array}{cccc}
I_{1}(x, y) & \ldots & 0 & 0 \\
0 & I_{2}(x, y) & \ldots & 0 \\
\ldots & & & \\
0 & 0 & \ldots & I_{n}(x, y)
\end{array}\right)
$$

- Multiply LHS and RHS with diag matrix

$$
\begin{aligned}
& \mathcal{I} \boldsymbol{i}=\mathcal{I} \mathcal{V} \boldsymbol{g}(x, y) \\
& \left(\begin{array}{c}
I_{1}^{2}(x, y) \\
I_{2}^{2}(x, y) \\
. \\
I_{n}^{2}(x, y) \\
\mid
\end{array}\right)=\left(\begin{array}{cccc}
I_{1}(x, y) & 0 & . & 0 \\
0 & I_{2}(x, y) & . . & . . \\
. & . . & . . & 0 \\
0 & \ddot{0} & 0 & I_{n}(x, y)
\end{array}\right)\left(\begin{array}{c}
\mathbf{v}_{1}^{T} \\
\mathbf{v}_{2}^{T} \\
. \\
\mathbf{V}_{n}^{T}
\end{array}\right) \uparrow \mathbf{g ( x , y )} \\
& \text { Known }
\end{aligned}
$$

$\Rightarrow$ Relevant elements of the left vector and the matrix are zero at points that are in shadow.

## Obtaining Normal and Albedo

- Given sufficient sources, we can solve the previous equation (most likely need a least squares solution) for $\mathbf{g}(x, y)$.
- Recall that $\mathbf{N}(x, y)$ is the unit normal.
- This means that $\rho(x, y)$ is the magnitude of $\mathbf{g}(\mathrm{x}, \mathrm{y})$.
- This yields a check
- If the magnitude of $\mathbf{g}(x, y)$ is greater than 1 , there's a problem.
- And $\mathbf{N}(x, y)=\mathbf{g}(x, y) / \rho(x, y)$.


## Example LLS Input



## Example LLS Result

- Reflectance / albedo:



## Recap

- Obtain normal / orientation, no depth


2

## Goal

## Shape as surface with depth and normal, so far only normal




## Recovering a surface from normals - 1

Recall the surface is written as

$$
(x, y, f(x, y))
$$

This means the normal has the form:

$$
N(x, y)=\left(\frac{1}{\sqrt{f_{x}^{2}+f_{y}^{2}+1}}\right)\left(\begin{array}{c}
-f_{x} \\
-f_{y} \\
1
\end{array}\right)
$$

- Then we obtain values for the partial derivatives of the surface:

$$
\begin{aligned}
& f_{x}(x, y)=\left(g_{1}(x, y) / g_{3}(x, y)\right) \\
& f_{y}(x, y)=\left(g_{2}(x, y) / g_{3}(x, y)\right)
\end{aligned}
$$

## Recovering a surface from normals - 2

- Recall that mixed second partials are equal --- this gives us an integrability check. We must have:

$$
\begin{aligned}
& \frac{\partial\left(g_{1}(x, y) / g_{3}(x, y)\right)}{\partial y}= \\
& \frac{\partial\left(g_{2}(x, y) / g_{3}(x, y)\right)}{\partial x}
\end{aligned}
$$

- We can now recover the surface height at any point by integration along some path, e.g.

$$
\begin{array}{r}
f(x, y)= \\
\int_{0}^{x} f_{x}(s, y) d s+ \\
\int_{0}^{y} f_{y}(x, t) d t+c
\end{array}
$$

## 展

## Height Map from Integration



How to integrate?

## Possible Solutions

- Engineering approach: Path integration (Forsyth \& Ponce)
- In general: Calculus of Variation Approaches
- Horn: Characteristic Strip Method
- Kimmel, Siddiqi, Kimia, Bruckstein: Level set method
- Many others ....


## Shape by Integation (Forsyth\&Ponce)

- The partial derivative gives the change in surface height with a small step in either the x or the y direction
- We can get the surface by summing these changes in height along some path.

$$
f(x, y)=\oint_{C}\left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}\right) \cdot d l+c
$$

For example, we can reconstruct the surface at $(u, v)$ by starting at $(0,0)$, summing the $y$-derivative along the line $x=0$ to the point $(0, v)$, and then summing the $x$-derivative along the line $y=v$ to the point $(u, v)$

$$
f(u, v)=\int_{0}^{v} \frac{\partial f}{\partial y}(0, y) d y+\int_{0}^{u} \frac{\partial f}{\partial x}(x, v) d x+c
$$

```
Obtain many images in a fixed view under different illuminants
Determine the matrix }\mathcal{V}\mathrm{ from source and camera information
Create arrays for albedo, normal (3 components),
    p (measured value of }\frac{\partialf}{\partialx}\mathrm{ ) and
    q (measured value of }\frac{\partialf}{\partialy}\mathrm{ )
For each point in the image array
    Stack image values into a vector i
    Construct the diagonal matrix }\mathcal{I
    Solve \mathcal{IV}\boldsymbol{g}=\mathcal{I}\boldsymbol{i}
        to obtain g}\mathrm{ for this point
    albedo at this point is |}\boldsymbol{g
    normal at this point is }\frac{g}{|}
    p at this point is }\frac{\mp@subsup{N}{1}{}}{\mp@subsup{N}{3}{}
    q at this point is \frac{N2}{N}
end
Check: is ( }\frac{\partialp}{\partialy}-\frac{\partialq}{\partialx}\mp@subsup{)}{}{2}\mathrm{ small everywhere?
top left corner of height map is zero
for each pixel in the left column of height map
    height value=previous height value + corresponding q value
end
for each row
    for each element of the row except for leftmost
        height value = previous height value + corresponding p value
    end
end
```


## Simple Algorithm Forsyth \& Ponce

Problem: Noise and numerical (in)accuracy are added up and result in distorted surface.

## Solution: Choose several different integration paths, and build average height map.

## Mathematical Property: Integrability

- Smooth, C2 continuous surface:

$$
\begin{aligned}
& Z(x, y)_{x y}=Z(x, y)_{y x} \\
& \quad \Rightarrow \frac{\partial p}{\partial y}=\frac{\partial q}{\partial x}
\end{aligned}
$$

$\Rightarrow$ check if $\left(\frac{\partial p}{\partial y}-\frac{\partial q}{\partial x}\right)^{2}$ is small

## SHAPE FROM SHADING (Calculus of Variations Approach)

- First Attempt: Minimize error in agreement with Image Irradiance Equation over the region of interest:

$$
\iint_{\text {object }}(I(x, y)-R(p, q))^{2} d x d y
$$

## SHAPE FROM SHADING (Calculus of Variations Approach)

- Better Attempt: Regularize the Minimization of error in agreement with Image Irradiance Equation over the region of interest:

$$
\iint_{\text {object }} p_{x}^{2}+p_{y}^{2}+q_{x}^{2}+q_{y}^{2}+\lambda(I(x, y)-R(p, q))^{2} d x d y
$$

## Linear Approaches for SFS

- Linear approaches reduce the non-linear problem into a linear through the linearization of the image irradiance equation.
- The idea is based on the assumption that the lower order components in the reflectance map dominate. Therefore, these algorithms only work well under this assumption.


## Simple Scenario

- We will be concerned with the simplest scenario, where the following assumptions hold;
- Camera; orthographic projection.
- Surface reflectivity; Lambertian surface
- Known/estimated illumination direction.
- Known/estimated surface albedo/
- The optical axis is the $Z$ axis of the camera and the surface can be parameterized as $Z=Z(x, y)$.
- The image irradiance (amount of light received by the camera to which the gray-scale produced is directly proportional) can be defined as follows;

$$
\begin{equation*}
E(x, y)=R_{\rho, I}(p, q)=\rho \mathbf{I}^{T} \mathbf{n}=\frac{\rho}{\sqrt{1+p^{2}+q^{2}}} \mathbf{I}^{T}[-p,-q, 1]^{T} \tag{A}
\end{equation*}
$$

- Eq(A) is the typical starting point of many shape from shading techniques, yet it is of a great mathematical complexity, it is a non-linear partial differential equation in $p$ $=p(x, y)$ and $q=q(x, y)$, which are the gradients of the unknown surface $Z=Z(x, y)$


## Pentland's Approach

- Under the assumptions of :
- Lambertian surface,
- orthographic projections,
- the surface being illuminated by distant light sources, and
- the surface is not self-shadowing,
- Pentland defined the image irradiance equation as follows;

$$
E(x, y)=R(p, q)=\frac{\rho\left(i_{x} p+i_{y} q-i_{z}\right)}{\sqrt{1+p^{2}+q^{2}}}=\frac{p \sin \sigma \cos \tau+q \sin \sigma \sin \tau+\cos \sigma}{\sqrt{1+p^{2}+q^{2}}}
$$

Where light source direction is defined as:

$$
I=[\sin \sigma \cos \tau, \sin \sigma \sin \tau, \cos \sigma]^{T}
$$

Pentland, A., "Shape Information From Shading: A Theory About Human Perception," Computer Vision., Second International Conference on , vol., no., pp.404-413, 5-8 Dec 1988.

## Pentland's Approach



Pentland, A., "Shape Information From Shading: A Theory About Human Perception," Computer Vision., Second International Conference on , vol., no., pp.404-413, 5-8 Dec 1988.

## Shah's Approach




## Deformable Models: SNAKES

Geodesic Snake formulated as PDE


Curve evolves over time

Normal direction
to curve

## Deformable Models: SNAKES

Geodesic Snake:


Curve evolves over time

Normal direction to curve


Mathematical solution is circle

## Deformable Models: SNAKES

Geodesic Snake:



Curve evolves over time

Normal direction
to curve

# Another Solution to SFS: Kimmel, Siddiqi, Kimia, Bruckstein 

Proposed Solution: Evolve a curve such that it tracks the height contours of $z(x, y)$.
[Kimmel et al., IJCV95]

Height climbed while progressing a distance $|\Delta C|$ in the direction $\hat{n}$ in the $(x, y)$ plane is given by $|\Delta C|=|\Delta z| \cot (\alpha)$.

Let $z$ denote time in the course of evolution,
 i.e., $z=t$. Since $E=\rho \lambda \cos (\alpha)$, we have

$$
\begin{equation*}
\left|\frac{\Delta C}{\Delta t}\right|=\cot (\alpha)=\frac{E / \rho \lambda}{\sqrt{1-(E / \rho \lambda)^{2}}} . \tag{11}
\end{equation*}
$$

$$
\sin (\alpha)=\sqrt{1-\left(\frac{E}{\delta \lambda}\right)^{2}}
$$

pdf document

## Kimmel, Siddiqi, Kimia, Bruckstein

Proposed Solution: Evolve a curve such that it tracks the height contours of $z(x, y)$.
[Kimmel et al., IJCV95]

The curve evolution equation is:

$$
\begin{cases}\frac{\partial \mathcal{C}}{\partial t} & =\frac{E / \rho \lambda}{\sqrt{1-E^{2} /(\rho \lambda)^{2}}} \cdot \hat{n} \\ \mathcal{C}(s, 0) & =\mathcal{C}_{0}(s)\end{cases}
$$

## Kimmel, Siddiqi, Kimia, Bruckstein

Examples - Pyramids


## Kimmel, Siddiqi, Kimia, Bruckstein

Examples - Three Mountains



## Application Area: Geography








## Application: Braille Code



## Abbildung 3 :

Oben links: Messanordnung mit einer Kamera und vier blauen LED-Leuchtfeldern.
Unten links: Ausschnitt einer 「altschachtel mit Blindenschrift-「ragung.
Rechts: 3D-Bild nach SFS-Analyse. Darunter ist ein Hohenprofil durch drei Braille-Punkte dargestellt.

# Mars Rover Heads to a New Crater NYT Sept 22, 2008 



## Limitations

- Controlled lighting environment
- Specular highlights?
- Partial shadows?
- Complex interrreflections?
- Fixed camera
- Moving camera?
- Multiple cameras?
=> Another approach: binocular / geometric stereo

