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Machine Vision Laboratory The University of the West of England Bristol, UK

Photometric Stereo in 3D Biometrics

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<u>Overview</u>

- 1. What is photometric stereo?
- 2. Why photometric stereo?
- 3. The basic method
- 4. Advanced methods
- 5. Applications

<u>Overview</u>

- 1. What is photometric stereo?
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A method of estimating surface geometry using multiple light source directions



A method of estimating surface geometry using multiple light source directions



Captured images

Surface normals

A method of estimating surface geometry using multiple light source directions



A method of estimating surface geometry using multiple light source directions





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Depth map

Surface normals

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Why 3D?

Why PS in particular?

Why 3D? Why PS in particular?

- 1. Robust face recognition
- 2. Ambient illumination independence
- 3. Facilitates pose correction
- 4. Non-contact fingerprint analysis



Richer dataset in 3D

Why 3D? Why PS in particular?

- 1. Robust face recognition
- 2. Ambient illumination independence
- 3. Facilitates pose, expression correction
- 4. Non-contact fingerprint analysis





- Same day
- Same camera
- Same expression
- Same pose
- Same background
- Even same shirt
- Different illumination

Completely different image

Why 3D? Why PS in particular?

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Why 3D? Why PS in particular?

- 1. Robust face recognition
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- 3. Facilitates pose, expression correction
- 4. Non-contact fingerprint analysis



Why 3D?

Why PS in particular?

1. Captures reflectance data for 2D matching

2. Compares well to other methods

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
SFS	Good	Average	Poor	Good	Good	Good
GS	Good	Good	Average	Variable [*]	Good	Average
LT	Average	Good	Average	Average	Good	Poor
PPT	Poor	Poor	Good	Good	Poor	Average
PS	Good	Good	Average	Good	Good	Average

*Depending on correspondence density

- SFS Shape-from-shading
- GS Geometric stereo
- LT Laser triangulation
- PPT Projected pattern triangulation
- PS Photometric stereo

Shape-from-Shading

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
SFS	Good	Average	Poor	Good	Good	Good
GS	Good	Good	Average	$Variable^*$	Good	Average
LT	Average	Good	Average	Average	Good	Poor
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Estimate surface normals from shading patterns. \rightarrow More to follow...

Geometric stereo

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
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 $z = \frac{bf}{|dl - dr|}$

Geometric stereo

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[Li Wei, and Eung-Joo Lee, JDCTA 2010]

[Wang, et al. JVLSI 2007]

Laser triangulation

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
SFS	Good	Average	Poor	Good	Good	Good
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 $h = d \tan \theta$

Laser triangulation

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General case

Laser triangulation

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
SFS	Good	Average	Poor	Good	Good	Good
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Projected Pattern Triangulation

	Cost	Computation	Accuracy	Resolution	Calibration	Practicality
SFS	Good	Average	Poor	Good	Good	Good
GS	Good	Good	Average	$Variable^*$	Good	Average
LT	Average	Good	Average	Average	Good	Poor
PPT	Poor	Poor	Good	Good	Poor	Average
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Greyscale camera Pattern projector Colour camera Flash Greyscale camera





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<u>Overview</u>

- 1. What is photometric stereo?
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3. The Basic Method – Assumptions

Assumptions:

- 1. No cast/self-shadows or specularities
- 2. Greyscale/linear imaging
- 3. Distant and uniform light sources
- 4. Orthographic projection
- 5. Static surface
- 6. Lambertian reflectance







$I \propto \cos \theta = \mathbf{N} \cdot \mathbf{L}$

I = radiance (intensity)N = unit normal vector L = unit source vector θ = angle between N and L





Equation of a plane Ax + By + Cz + D = 0

Surface normal vector to plane

 $\mathbf{N} = \begin{bmatrix} A, B, C \end{bmatrix}^T$



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Equation of a plane Ax + By + Cz + D = 0

Surface normal vector to plane

 $\mathbf{N} = \begin{bmatrix} A, B, C \end{bmatrix}^T$

 $\frac{\partial z}{\partial x} = -\frac{A}{C} \qquad \frac{\partial z}{\partial y} = -\frac{B}{C}$

Rescaled surface normal

$$\mathbf{n} = \left[\frac{A}{C}, \frac{B}{C}, 1\right]^T = \left[-\frac{\partial z}{\partial x}, -\frac{\partial z}{\partial y}, 1\right]^T$$



3. The Basic Method – Reflectance Equation



Consider the imaging of an object with one light source.

3. The Basic Method – Reflectance Equation



Lambertian reflection

$$E = \rho L \cos \theta$$
$$E = \text{emittance}$$
$$\rho = \text{albedo}$$

L = irradiance

<u>3. The Basic Method – Reflectance Equation</u>



Lambertian reflection

- $E = \rho L \cos \theta$
- E = emittance
- $\rho = albedo$
- L = irradiance

Take scalar product of light source vector and normal vector to obtain $\cos\theta$:

$$pp_s + qq_s + 1 = \sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}\cos\theta$$
Perform substitution with Lambert's law $E = \rho L \cos \theta$

$$E = \rho L \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

Perform substitution with Lambert's law $E = \rho L \cos \theta$

$$E = \rho L \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

If ρ is re-defined and the camera response is linear, then

$$I = \rho' \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

Perform substitution with Lambert's law $E = \rho L \cos \theta$

$$E = \rho L \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

If ρ is re-defined and the camera response is linear, then

$$I = \rho \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

Known: p_s , q_s , I

1 equation, 3 unknowns

Unknown: p, q, ρ

 \rightarrow Insoluble

$$I = \rho \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

For a given intensity measurement, the values of pand q are confined to one of the lines (circles here) in the graph.

 $p_s = q_s = 0$





$$I = \rho \frac{pp_s + qq_s + 1}{\sqrt{p^2 + q^2 + 1}\sqrt{p_s^2 + q_s^2 + 1}}$$

For a given intensity measurement, the values of pand q are confined to one of the lines (circles here) in the graph.

 $p_{s} = q_{s} = 0.5$



This is the result for a different light source direction.





We can overcome the cone ambiguity using several light source directions: <u>Photometric Stereo</u>

3. The Basic Method – Two Sources



Two sources: normal confined to two points – Or fully constrained if the albedo is known

3. The Basic Method – Three Sources



Three sources: fully constrained

3. The Basic Method – Matrix Representation

$$I = \rho \cos \theta$$
$$I = \rho \mathbf{N} \cdot \mathbf{s} = \rho \begin{bmatrix} s_x \\ s_y \\ s_z \end{bmatrix}^T \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$$

- N = Unit surface normal
- s = Unit light source vector

(using slightly different symbols to aid clarity)

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- N = Unit surface normal
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(using slightly different symbols to aid clarity)

$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \rho \begin{bmatrix} s_x^1 & s_y^1 & s_z^1 \\ s_x^2 & s_y^2 & s_z^2 \\ s_x^3 & s_y^3 & s_z^3 \end{bmatrix} \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$$

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$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \rho \begin{bmatrix} s_x^1 & s_y^1 & s_z^1 \\ s_x^2 & s_y^2 & s_z^2 \\ s_x^3 & s_y^3 & s_z^3 \end{bmatrix} \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix} \qquad \begin{bmatrix} s_x^1 & s_y^1 & s_z^1 \\ s_x^2 & s_y^2 & s_z^2 \\ s_x^3 & s_y^3 & s_z^3 \end{bmatrix}^{-1} \begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \rho \begin{bmatrix} N_x \\ N_y \\ N_z \end{bmatrix}$$

<u>3. The Basic Method – Matrix Representation</u>

$$\begin{bmatrix} s_{x}^{1} & s_{y}^{1} & s_{z}^{1} \\ s_{x}^{2} & s_{y}^{2} & s_{z}^{2} \\ s_{x}^{3} & s_{y}^{3} & s_{z}^{3} \end{bmatrix}^{-1} \begin{bmatrix} I_{1} \\ I_{2} \\ I_{3} \end{bmatrix} = \rho \begin{bmatrix} N_{x} \\ N_{y} \\ N_{z} \end{bmatrix} = \begin{bmatrix} m_{x} \\ m_{y} \\ m_{z} \end{bmatrix}$$

Recall that, based on the surface gradients, $\mathbf{n} = [p, q, -1]^T$ $\mathbf{N} = \left[\frac{p}{\sqrt{p^2 + q^2 + 1}}, \frac{q}{\sqrt{p^2 + q^2 + 1}}, \frac{-1}{\sqrt{p^2 + q^2 + 1}}\right]^T$

<u>3. The Basic Method – Matrix Representation</u>

$$\begin{bmatrix} s_{x}^{1} & s_{y}^{1} & s_{z}^{1} \\ s_{x}^{2} & s_{y}^{2} & s_{z}^{2} \\ s_{x}^{3} & s_{y}^{3} & s_{z}^{3} \end{bmatrix}^{-1} \begin{bmatrix} I_{1} \\ I_{2} \\ I_{3} \end{bmatrix} = \rho \begin{bmatrix} N_{x} \\ N_{y} \\ N_{z} \end{bmatrix} = \begin{bmatrix} m_{x} \\ m_{y} \\ m_{z} \end{bmatrix}$$

Recall that, based on the surface gradients, $\mathbf{n} = \begin{bmatrix} p, q, -1 \end{bmatrix}^T$ $\mathbf{N} = \begin{bmatrix} \frac{p}{\sqrt{p^2 + q^2 + 1}}, \frac{q}{\sqrt{p^2 + q^2 + 1}}, \frac{-1}{\sqrt{p^2 + q^2 + 1}} \end{bmatrix}^T$

Substituting and re-arranging these equations gives us

$$p = -\frac{m_x}{m_z}, \quad q = -\frac{m_y}{m_z}, \quad \rho = \sqrt{m_x^2 + m_y^2 + m_z^2}$$

That is, the surface normals/gradient and albedo have been determined.

If the three light source vectors are co-planar, then the light source matrix becomes non-singular – i.e. cannot be inverted, and the equations are insoluble.

$$p = -\frac{m_x}{m_z}, \quad q = -\frac{m_y}{m_z}, \quad \rho = \sqrt{m_x^2 + m_y^2 + m_z^2}$$

That is, the surface normals/gradient and albedo have been determined.

3. The Basic Method – Example results







Surface normals

Albedo map More on applications of this later...

What about the depth?

Recall the relation between surface normal and gradient:

$$\mathbf{n} = \left[\frac{\partial z}{\partial x}, \frac{\partial z}{\partial y}, -1\right]^{T}$$

Memory jogger from calculus: $p = \frac{dz}{dx}$ $z = \int dz = \int p dx$ $z = \sum \delta z = \sum p \delta x$

Recall the relation between surface normal and gradient:

$$\mathbf{n} = \left[\frac{\partial z}{\partial x}, \frac{\partial z}{\partial y}, -1\right]^T$$

Memory jogger
from calculus:
$$p = \frac{dz}{dx}$$

 $z = \int dz = \int p dx$
 $z = \sum \delta z = \sum p \delta x$

So the height can be determined via an integral (or summation in the discrete world of computer vision). This can be written in several ways:

$$z(P) = z(P_0) + \int_{P_0}^{P} (p \, dx + q \, dy)$$
$$z(u, v) = \int_{0}^{u} q(0, y) \, dy + \int_{0}^{v} p(x, v) \, dx + c$$
$$z(x, y) = \oint_{C} (p, q) \, dl + c$$

So we can generate a height map by summing up individual normal components. Simple, right?



So we can generate a height map by summing up individual normal components. Simple, right?

WRONG!!!

- Depends on the path taken
- Is affected by noise
- Cannot handle discontinuities
- Suffers from non-integrable regions

(there's some overlap here)

$$\frac{\partial^2 z}{\partial x \partial y} = \frac{\partial^2 z}{\partial y \partial x}$$

Further details are beyond the scope of this talk.



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4. Advanced Methods

Recall the assumptions of standard PS:

- 1. No cast/self-shadows or specularities
- 2. Greyscale imaging
- 3. Distant and uniform light sources
- 4. Orthographic projection
- 5. Static surface
- 6. Lambertian reflectance

4. Advanced Methods

Recall the assumptions of standard PS:

- 1. No cast/self-shadows or specularities
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- 5. Static surface
- 6. Lambertian reflectance

The two highlighted assumptions are the two most problematic, and studied, assumptions.

4. Advanced Methods – Shadows and Specularities

Recall that:

p

$$\begin{bmatrix} s_{x}^{1} & s_{y}^{1} & s_{z}^{1} \\ s_{x}^{2} & s_{y}^{2} & s_{z}^{2} \\ s_{x}^{3} & s_{y}^{3} & s_{z}^{3} \end{bmatrix}^{-1} \begin{bmatrix} I_{1} \\ I_{2} \\ I_{3} \end{bmatrix} = \rho \begin{bmatrix} N_{x} \\ N_{y} \\ N_{z} \end{bmatrix} = \begin{bmatrix} m_{x} \\ m_{y} \\ m_{z} \end{bmatrix}$$
$$= -\frac{m_{x}}{m_{z}}, \quad q = -\frac{m_{y}}{m_{z}}, \quad \rho = \sqrt{m_{x}^{2} + m_{y}^{2} + m_{z}^{2}}$$

Suppose we have a fourth light source

- Apply the above equation for each combination of three light sources.
 → Four normal estimates
- Where error in estimates > camera noise an anomaly is present.

 \rightarrow Ignore one light source here

[Coleman and Jain-esque]

Due to the linear dependence of light source vectors, we know that

 $a_1\mathbf{s}_1 + a_2\mathbf{s}_2 + a_3\mathbf{s}_3 + a_4\mathbf{s}_4 = 0$

For some combination of constants a_1 , a_2 , a_3 and a_4 .

[Barsky and Petrou]

Due to the linear dependence of light source vectors, we know that

$$a_1 \mathbf{s}_1 + a_2 \mathbf{s}_2 + a_3 \mathbf{s}_3 + a_4 \mathbf{s}_4 = 0$$

For some combination of constants a_1 , a_2 , a_3 and a_4 .

Multiply by albedo and take scalar product with surface normal:

$$a_1\rho(\mathbf{s}_1\cdot\mathbf{N}) + a_2\rho(\mathbf{s}_2\cdot\mathbf{N}) + a_3\rho(\mathbf{s}_3\cdot\mathbf{N}) + a_4\rho(\mathbf{s}_4\cdot\mathbf{N}) = 0$$

Due to the linear dependence of light source vectors, we know that

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$$\Rightarrow \quad a_1 I_1 + a_2 I_2 + a_3 I_3 + a_4 I_4 = 0$$

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Multiply by albedo and take scalar product with surface normal:

$$a_1 \rho(\mathbf{s}_1 \cdot \mathbf{N}) + a_2 \rho(\mathbf{s}_2 \cdot \mathbf{N}) + a_3 \rho(\mathbf{s}_3 \cdot \mathbf{N}) + a_4 \rho(\mathbf{s}_4 \cdot \mathbf{N}) = 0$$
$$\Rightarrow \quad a_1 I_1 + a_2 I_2 + a_3 I_3 + a_4 I_4 = 0$$

Where this is satisfied (subject to the confines of camera noise) the pixel is not specular, not in shadow and follows Lambert's law.

[Barsky and Petrou]

4. Advanced Methods – Lambert's Law / Gauges

Attempt to overcome lighting non-uniformities and non-Lambertian behaviour using gauges.



$$\mathbf{I}^{(\text{scene})} = \begin{bmatrix} I_1^{(\text{scene})} & I_2^{(\text{scene})} & \dots & I_M^{(\text{scene})} \end{bmatrix}^T$$

$$\mathbf{I}^{(\text{gauge})} = \begin{bmatrix} I_1^{(\text{gauge})} & I_2^{(\text{gauge})} \dots & I_M^{(\text{gauge})} \end{bmatrix}^T$$

Image scene in the presence of a "gauge" object of known geometry and uniform albedo.

Seek a mapping between normals on the object with normals on the gauge.

$$\mathbf{I}^{(\text{scene})}[i] \mapsto \mathbf{I}^{(\text{gauge})}[j] : \mathbf{N}^{(\text{scene})}[i] = \mathbf{N}^{(\text{gauge})}[j]$$



[Leitão et al.]

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Face recognition

Weapons detection

Archaeology

Skin analysis

Fingerprint recognition

5. Applications – Face recognition



Recognition rates (on 61 subjects)

2D Photograph: 91.2% Surface normals: 97.5%



Computer monitor [Schindler]



5. Applications – Fingerprint recognition



Demo



Conclusion

- 1. What is photometric stereo? Shape / normal estimation from multiple lights
- 2. Why photometric stereo? Cheap, high resolution, efficient
- 3. The basic method Covered
- 4. Advanced methods Introduced
- 5. Applications

Faces, weapons, fingerprints

2rd September 2010

Photometric Stereo in 3D Biometrics

Thank You

Questions?



University of the West of England

Machine Vision Laboratory