# Vision and Image Processing: Shading, Photometric Stereo

#### François Lauze

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## Plan for today and January 3rd

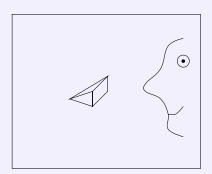
- Image Formation and reflectance.
- Lighting Models.
- The Photometric Stereo Problem.

### **Outline**

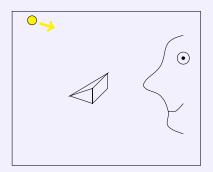
What is Photometric Stereo

2 Lighting Models

3 Photometric Stered



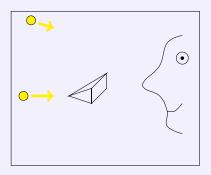
- 1 fixed camera + 1 "fixed" scene
- m lightings



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Slide by Y. Quéau

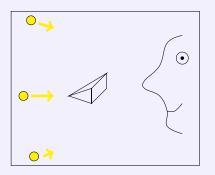


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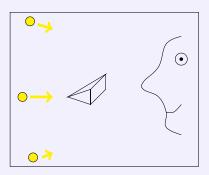
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- 1 fixed camera + 1 "fixed" scene
- m lightings

#### Goal:

3D-reconstruction of the scene from the 2D images



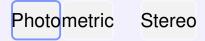
Slide by Y. Quéau







Photometric Stereo



ullet φῶς/φωτός : (phõs, gen. photós): light



- μέτρον (métron): measure

## **Photometric**



- φῶς/φωτός : (phõs, gen. photós): light
- μέτρον (métron): measure
- στερεός (stereós): solid/volume

## Photometric Stereo

μέτρον (métron): measure

στερεός (stereós): solid/volume

Suggest. Volume recovery from measured light. Needed:

## Photometric Stereo

- μέτρον (métron): measure
- στερεός (stereós): solid/volume

#### Suggest. Volume recovery from measured light. Needed:

- Understand reflectance: how is light reflected from an object.
- How can we measure it.
- How object geometry is linked to light.

### **Outline**

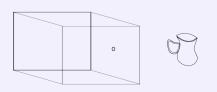
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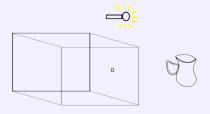
Photometric Stereo



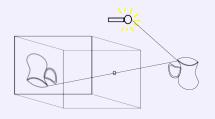




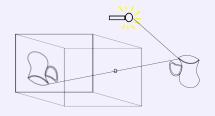
- Object
- Camera



- Object
- Camera
- Light source



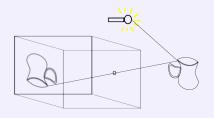
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#### Ingredients

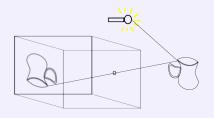
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 Image formation inside camera: when light, scene and camera parameters known: reflectance function.



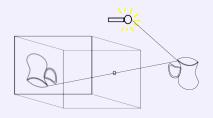
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- BTW: Camera detectors react almost truly linearly to received luminance.



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- Image formation inside camera: when light, scene and camera parameters known: reflectance function.
- BTW: Camera detectors react almost truly linearly to received luminance.
- Can image formation model give enough information about the object surface to reconstruct it?

## **Materials and Light**



## **Materials and Light**

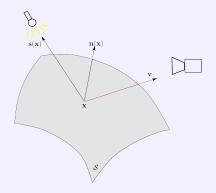


From now, only opaque objects.

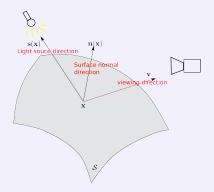
#### **Matte vs Brilliant**



## **Bidirectional Reflectance Distribution Function – BRDF**



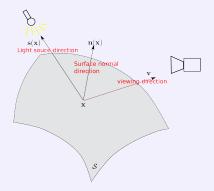
#### **Bidirectional Reflectance Distribution Function – BRDF**



 Luminance emitted by punctual object x on a surface S with normal direction n(x) at x, in emission direction v characterized by spherical angles (θ<sub>e</sub>, φ<sub>e</sub>) w.r.t n(x):

$$L(\mathbf{x}, \theta_{e}, \varphi_{e}) = \int_{\theta_{i}=0}^{\frac{\pi}{2}} \int_{\varphi_{i}=0}^{2\pi} \kappa(\mathbf{x}, \theta_{i}, \varphi_{i}, \theta_{e}, \varphi_{e}) \overline{L}(\theta_{i}, \varphi_{i}) \sin \theta_{i} \cos \theta_{i} d\theta_{i} d\varphi.$$

#### **Bidirectional Reflectance Distribution Function – BRDF**



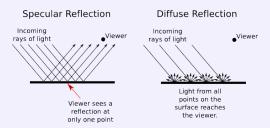
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• Nice formula but I won't explain what it means!



## Specular Vs. Matte Objects



Two standard reflection models: specular: mirror like surface, diffuse: rough surface (at very small scale): Lambertian model. Others, especially useful in Computer Graphics.

#### **Diffuse Reflection**

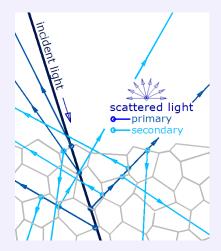
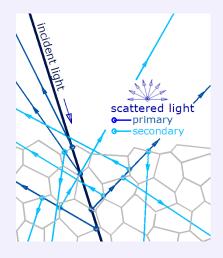


Image Source GianniG46, Wikipedia

#### **Diffuse Reflection**



Rough surface at micro-scale.

- Light bounces.
- Reflections in all directions
- Some light is absorbed.
- Only a percentage of light energy is reemitted.

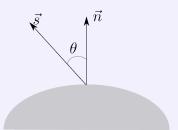
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#### Reflectance

- Linearised Lambertian model:  $I(\mathbf{p}) = \rho(\mathbf{x})\mathbf{s}(\mathbf{x}) \cdot \mathbf{n}(\mathbf{x})$
- $\rho(\mathbf{x})$  is the *albedo* at  $\mathbf{x}$  material light absorption property,  $\rho \in [0, 1]$ . Assumes matter material such as chalk...

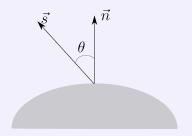
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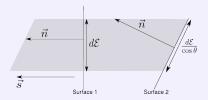
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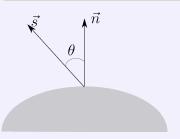
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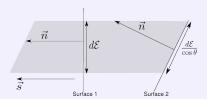




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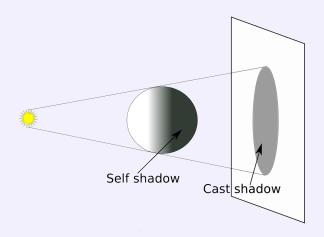




#### In words

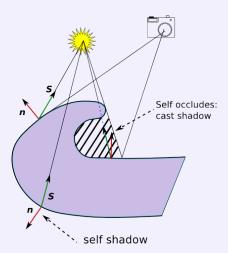
 Local orientation of object w.r.t. light: In surface 1: surface area matches ray "section". In surface 2: surface area larger than ray section, but receive same amount of light.

#### **Shadows**



- Self shadow: surface is behind the light source.  $\mathbf{s} \cdot \mathbf{n} \leq 0$ .
- Cast shadow: part of the scene occludes another part.

## Shadows again



- Lambert's law and self-shadows:  $I = \rho \max(\mathbf{s} \cdot \mathbf{n}, 0)$ .
- Cast shadows: Non local phenomenon, Lambert's law is local...



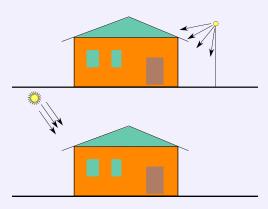
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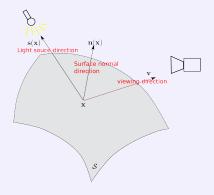
Photometric Stereo

#### **Type of Light Sources**



- Top: near light source, radial.
- Bottom: far light source: parallel. Our choice in these lectures.
- Other types?

### Shape From Shading (SfS) – B. Horn 1970



- Use Lambert's Law to gain information on visible surface via normal vector n(x).
- Need link between surface equations and normal vector.

## **Settings**

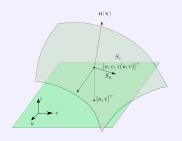
• Representation of the surface. Assume surface parameterized by  $(u, v) \mapsto S(u, v) \in \mathbb{R}^3$ . Even better: depth map:

$$S(u, v) = [u, v, z(u, v)]^T$$
: Monge Patch.

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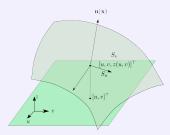
Tangent vectors at  $\mathbf{x}(u, v) = [u, v, z(u, v)]^{\top}$ 

$$\frac{\partial \mathcal{S}}{\partial u} = \mathcal{S}_u = \begin{bmatrix} 1 \\ 0 \\ \frac{\partial z}{\partial u} \end{bmatrix}, \quad \frac{\partial \mathcal{S}}{\partial v} = \mathcal{S}_v \begin{bmatrix} 0 \\ 1 \\ \frac{\partial z}{\partial v} \end{bmatrix},$$

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• Normal vector  $\mathbf{n}(u, v) = \mathbf{n}(\mathbf{x}(u, v))$ 

$$\mathbf{n}(\mathbf{x}) = \frac{\mathcal{S}_u \times \mathcal{S}_v}{|\mathcal{S}_u \times \mathcal{S}_v|} = \frac{1}{\sqrt{|\nabla z|^2 + 1}} \begin{bmatrix} -z_u \\ -z_v \\ 1 \end{bmatrix}$$

#### **Camera Model**

- Two Choices: pinhole and orthographic.
- Orthographic Camera Model.
  - $[x = u, y = v, z] \mapsto [u, v]$ : orthographic projection.
  - Formula from previous slide: Coordinates of normal vector in orthographic projection:

$$\mathbf{n} = \frac{1}{\sqrt{|\nabla z|^2 + 1}} \begin{bmatrix} -z_u \\ -z_v \\ 1 \end{bmatrix}$$

- Pinhole Camera model.
  - Projection

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} \mapsto \begin{bmatrix} u = \frac{fx}{z} \\ v = \frac{fy}{z} \end{bmatrix}$$

Normal vector in camera coordinates: complicated!

$$\mathbf{n} = \frac{1}{\sqrt{|\nabla z|^2 + \left(\frac{z + [u,v]\nabla z}{f}\right)^2}} \begin{bmatrix} -z_u \\ -z_v \\ \frac{z + [u,v]\nabla z}{f} \end{bmatrix}$$

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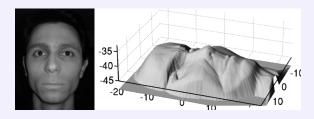
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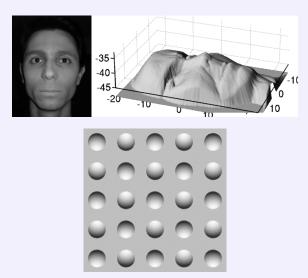
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Complicated Math and Research Topics

# **Examples, Problems**



# **Examples, Problems**



# SfS (Counter)Example



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# SfS (Counter)Example







## Why Does It Go Wrong

Assume Light **s** known and constant (far light source with known intensity).Per Pixel:

- Number of unknowns:
  - Normal vector  $\mathbf{n}(u, v)$ , 3 components  $[\mathbf{n}_1, \mathbf{n}_2, \mathbf{n}_3]$ , but

$$\|\mathbf{n}\| = 1 : \mathbf{n}_1^2 + \mathbf{n}_2^2 + \mathbf{n}_3^2 = 1.$$

2 degrees of freedom (DoF).

- Albedo  $\rho(u, v)$ : 1 value, 1 DoF.
- Total: 3 DoF.
- Known information per pixel:
  - Reflectance  $I(u, v) = \rho(u, v)\mathbf{s} \cdot \mathbf{n}(u, v)$ : 1 equation linking 3 unknowns.

- Remaining DoFs: 2.
- For unambiguous solution, need remaining DoF = 0.
- In counterexample, 1DoF removed by assuming  $\rho(u, v) \equiv 1$ : Wrong!

• How to remove Degrees of Freedom?

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- k far (parallel) light sources  $s^1, ..., s^k$ : k equations

$$\begin{cases} I^{1}(u, v) &= \rho(u, v)\mathbf{s}^{1} \cdot \mathbf{n}(u, v) \\ I^{2}(u, v) &= \rho(u, v)\mathbf{s}^{2} \cdot \mathbf{n}(u, v) \\ \dots & \dots \\ I^{k}(u, v) &= \rho(u, v)\mathbf{s}^{k} \cdot \mathbf{n}(u, v) \end{cases}$$
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- Answer is geometric!
- From (u, v) → n(u, v) to surface? Integration of normals: out of scope;
   Matlab / Python functions will be provided.



• Woodham idea: Normal + Albedo Simultaneously

$$\mathbf{m} := \rho \mathbf{n}, \quad \rho = \|\mathbf{m}\|, \mathbf{n} = \frac{\mathbf{m}}{\|\mathbf{m}\|}$$

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• Number of solutions?

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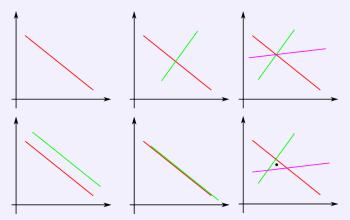
- Problem if  $\rho = 0$ : absolutely black object shadow.
- k Lambert's laws for a given pixel at [u, v]

$$\underbrace{\begin{bmatrix} I_1 \\ I_2 \\ \vdots \\ I_k \end{bmatrix}}_{\text{Image vector } \mathbf{I}, k \times 1} = \underbrace{\begin{bmatrix} \mathbf{s}_1^1 & \mathbf{s}_1^2 & \mathbf{s}_1^3 \\ \mathbf{s}_2^1 & \mathbf{s}_2^2 & \mathbf{s}_2^3 \\ \vdots & \vdots & \vdots \\ \mathbf{s}_k^1 & \mathbf{s}_k^2 & \mathbf{s}_k^3 \end{bmatrix}}_{\text{Light matrix } \mathbf{M}_{\mathbf{S}}, k \times 3} \begin{bmatrix} \mathbf{m}_1 \\ \mathbf{m}_2 \\ \mathbf{m}_3 \end{bmatrix}$$

- Number of solutions?
- Can vary from none to a lot!

### Linear Algebra Again

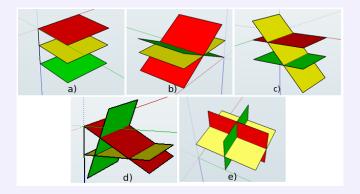
System of equations in 2 unknown.



- Need at least 2 independent equations! ...but not 3!
- For 3 or more: concept of "closest solution in least squares".



#### In 3D



Same light source, incompatible measurements, b) coplanar light sources, compatible measurements, c) 2 light source, 3 incompatible measurements, d) coplanar light sources, incompatible measurements, e) 3 non coplanar light sources

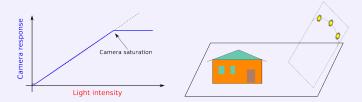
Picture from Guillermo Bautista, mathandmultimedia.com

- Parameters / devices:
  - Measurement errors: Camera?
  - Coplanar light sources: example?
- Reflectance
  - Lambert's law only valid for matte materials: specularities.
  - Shadows / penumbra: non black cast shadow areas.
- More?

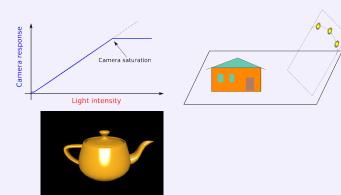
- Parameters / devices:
  - Measurement errors: Camera?
  - Coplanar light sources: example?
- Reflectance
  - Lambert's law only valid for matte materials: specularities.
  - Shadows / penumbra: non black cast shadow areas.
- More?



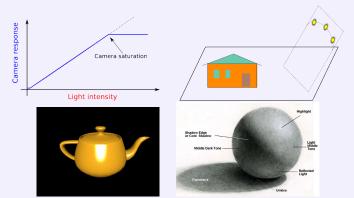
- Parameters / devices:
  - Measurement errors: Camera?
  - Coplanar light sources: example?
- Reflectance
  - Lambert's law only valid for matte materials: specularities.
  - Shadows / penumbra: non black cast shadow areas.
- More?



- Parameters / devices:
  - Measurement errors: Camera?
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- Parameters / devices:
  - Measurement errors: Camera?
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- More?



## **Solving For Normals: Pseudo-Inverse Approach**

Assume 3 or more non coplanar light sources.

- Equation  $I = M_s \mathbf{m}$  may not have solution if  $M_s$  is  $k \times 3$ , k > 3.
- Via Moore-Penrose Pseudo-Inverse  $M_s^{\dagger}$ , solution of

**m** such that 
$$\|\mathbf{I} - M_{\mathbf{s}}\mathbf{m}\|^2 = \min : \mathbf{m} = M_{\mathbf{s}}^{\dagger}\mathbf{I}$$
.

#### Pros.

 Good news: Matlab pinv function, Python pinv function in package numpy.linalg.

#### Cons.

Does not separate wrong and accurate measurements.

### Solving For Normals: Equations Selection - I

#### In a nutshell.

Per pixel: find best constraints: 3 "good measurements" I<sup>a</sup>, I<sup>b</sup>, I<sup>c</sup> and corresponding "good lights" s<sup>a</sup>, s<sup>b</sup>, s<sup>c</sup>.

$$\underbrace{ \begin{bmatrix} I^{a} \\ I^{b} \\ I^{c} \end{bmatrix}}_{\mathbf{I}_{abc}} = \underbrace{ \begin{bmatrix} \mathbf{S}_{1}^{a} & \mathbf{s}_{2}^{a} & \mathbf{S}_{3}^{a} \\ \mathbf{S}_{1}^{b} & \mathbf{S}_{2}^{b} & \mathbf{S}_{3}^{b} \\ \mathbf{S}_{1}^{c} & \mathbf{S}_{2}^{c} & \mathbf{S}_{3}^{c} \end{bmatrix}}_{M_{abc}} \underbrace{ \begin{bmatrix} \mathbf{m}^{1} \\ \mathbf{m}^{2} \\ \mathbf{m}^{3} \end{bmatrix}}_{\mathbf{I}_{abc}}$$

#### Good measurements.

- I<sub>i</sub> good measurement: not too small (shadow, penumbra) and not too large (saturation, potential specularity).
- Good light sources:  $\det M_{abc}$  as large as possible.

## **Solving For Normals: Equations Selection - II**

#### Pros.

Best system of equations per pixel.

#### Cons.

- Lack of spatial coherence: different lights can be chosen for neighbor pixels.
- Need thresholds for intensities: parameters of the algorithm.
- More complicated to code.

#### Size Matters.

- For relatively small k: equation selection can be a good idea.
- For very large k (1000, more...) pseudo-inverse very good: statistical reason.

## Algorithm in a Nutshell

#### Input.

• k known parallel light sources  $\mathbf{s}_1, \dots \mathbf{s}_k$ . k recorded images  $l_1, \dots, l_k$ .

#### Normals and Albedo recovery.

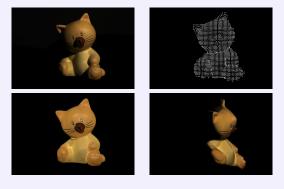
- For each (valid) pixel [u, v] in image domain:
  - **1** Solve  $\mathbf{m}(u, v)$  either via pseudo-inverse or equation selection.
  - Get albedo and normal:

$$\rho(u,v) = \|\mathbf{m}(u,v)\|, \quad \mathbf{n}(u,v) = \frac{1}{\rho(u,v)}\mathbf{m}(u,v).$$

#### Surface Recovery.

- Get surface from normals via surface integration.
- May "paint surface" with albedo.

#### **Process**



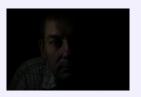
Top left: 1 of the input images, right: normal field. Bottom left: albedo, right: a 3D reconstruction.

# Mezigue



# Mezigue



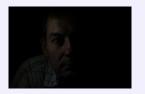






# Mezigue











#### Literature, available in Abaslon

- Horn, Shape from shading 1970
- Woodham, Photometric Stereo, 1980

# **Summary**

- φῶς/φωτός
- μέτρον
- στερεός

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