

CS5670 : Computer Vision

Noah Snavely

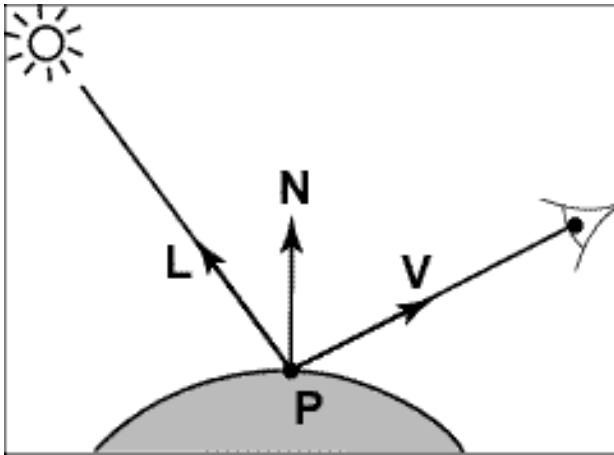
Photometric stereo



Announcements

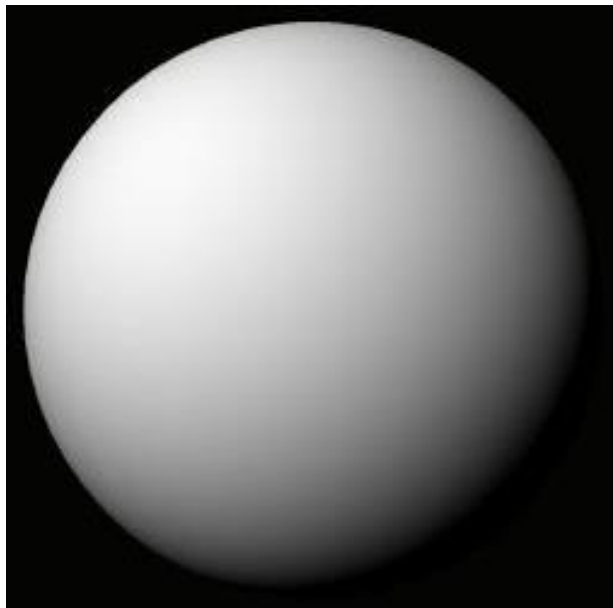
- Project 3 code due tomorrow, 3/29, by 11:59pm to CMS
 - Artifact due Friday, 3/30, by 11:59pm to CMS

Recap: Lambertian (Diffuse) Reflectance



$$I = k_d \mathbf{N} \cdot \mathbf{L}$$

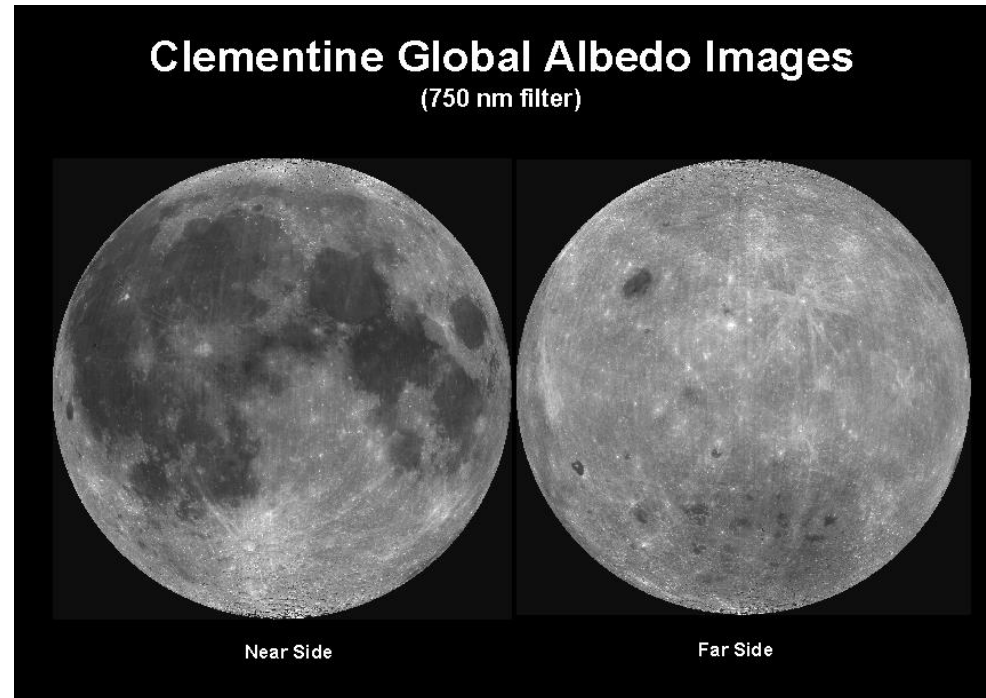
- I : observed image intensity
- k_d : object albedo
- \mathbf{N} : surface normal
- \mathbf{L} : light source direction



Lambertian sphere with constant albedo lit by a directional light source

Sample albedos

Surface	Typical albedo
Fresh asphalt	0.04 ^[4]
Open ocean	0.06 ^[5]
Worn asphalt	0.12 ^[4]
Conifer forest (Summer)	0.08, ^[6] 0.09 to 0.15 ^[7]
Deciduous trees	0.15 to 0.18 ^[7]
Bare soil	0.17 ^[8]
Green grass	0.25 ^[8]
Desert sand	0.40 ^[9]
New concrete	0.55 ^[8]
Ocean ice	0.5–0.7 ^[8]
Fresh snow	0.80–0.90 ^[8]

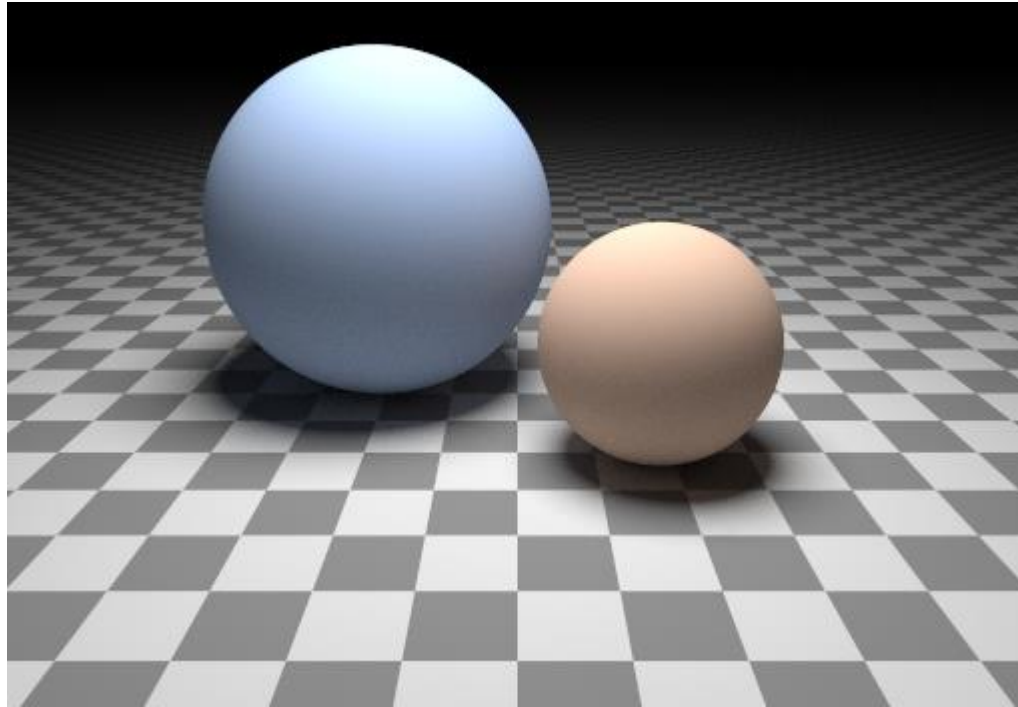


Objects can have varying albedo and albedo varies with wavelength

Source:

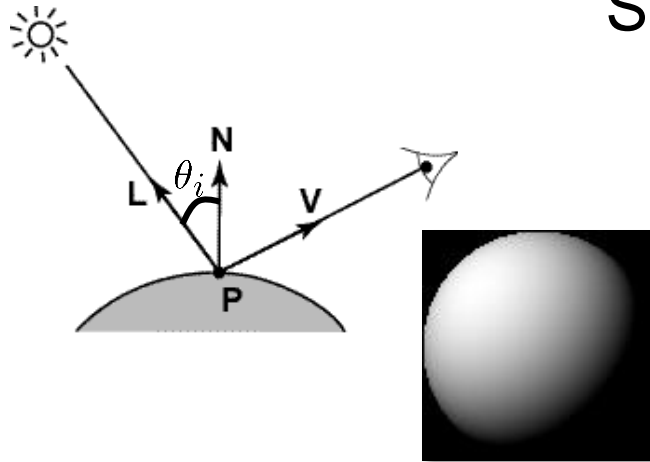
<https://en.wikipedia.org/wiki/Albedo>

Can we determine shape from lighting?



- Are these spheres?
 - Or just flat discs painted with varying albedo?

Shape from shading



Suppose $k_d = 1$

$$\begin{aligned} I &= k_d \mathbf{N} \cdot \mathbf{L} \\ &= \mathbf{N} \cdot \mathbf{L} \\ &= \cos \theta_i \end{aligned}$$

You can directly measure angle between normal and light source

- Not quite enough information to compute surface shape
- But can be if you add some additional info, for example
 - assume a few of the normals are known (e.g., along silhouette)
 - constraints on neighboring normals—“integrability”
 - smoothness
- Hard to get it to work well in practice
 - plus, how many real objects have constant albedo?

Application: Detecting composite photos

Fake photo

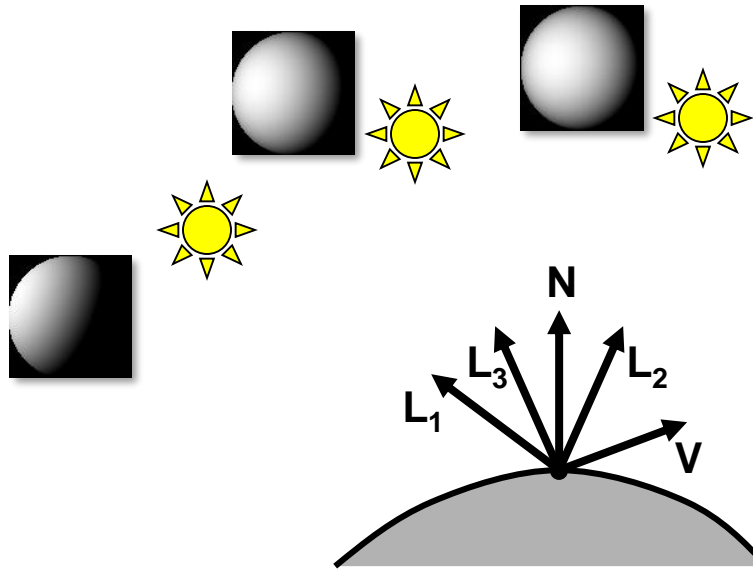


Real photo



Let's take more than one photo!

Photometric stereo



$$I_1 = k_d \mathbf{N} \cdot \mathbf{L}_1$$

$$I_2 = k_d \mathbf{N} \cdot \mathbf{L}_2$$

$$I_3 = k_d \mathbf{N} \cdot \mathbf{L}_3$$

Can write this as a matrix equation:

$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = k_d \begin{bmatrix} \mathbf{L}_1^T \\ \mathbf{L}_2^T \\ \mathbf{L}_3^T \end{bmatrix} \mathbf{N}$$

Solving the equations

$$\underbrace{\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix}}_{\mathbf{I} \quad 3 \times 1} = \underbrace{\begin{bmatrix} \mathbf{L}_1^T \\ \mathbf{L}_2^T \\ \mathbf{L}_3^T \end{bmatrix}}_{\mathbf{L} \quad 3 \times 3} \underbrace{k_d \mathbf{N}}_{\mathbf{G} \quad 3 \times 1}$$

$$\mathbf{G} = \mathbf{L}^{-1} \mathbf{I}$$

$$k_d = \|\mathbf{G}\|$$

$$\mathbf{N} = \frac{1}{k_d} \mathbf{G}$$

Solve one such linear system **per pixel** to solve for that pixel's surface normal

More than three lights

Get better results by using more lights

$$\begin{bmatrix} I_1 \\ \vdots \\ I_n \end{bmatrix} = \begin{bmatrix} \mathbf{L}_1 \\ \vdots \\ \mathbf{L}_n \end{bmatrix} k_d \mathbf{N}$$

Least squares solution:

$$\begin{aligned} \mathbf{I} &= \mathbf{L}\mathbf{G} \\ \mathbf{L}^T \mathbf{I} &= \mathbf{L}^T \mathbf{L}\mathbf{G} \\ \mathbf{G} &= (\mathbf{L}^T \mathbf{L})^{-1} (\mathbf{L}^T \mathbf{I}) \end{aligned}$$

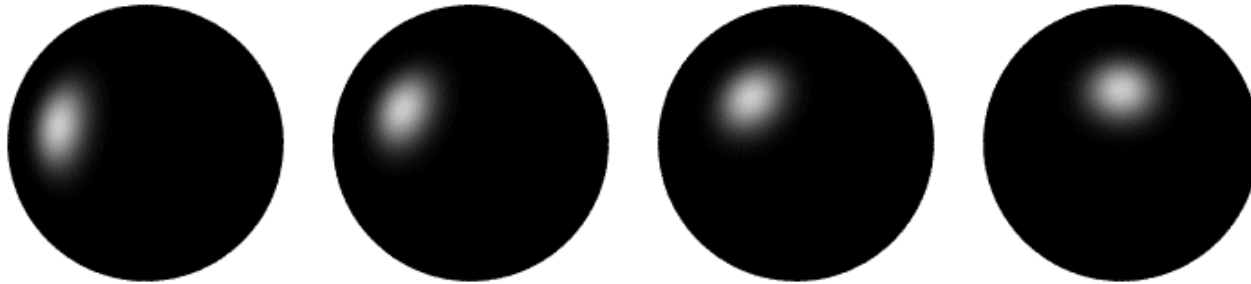
Solve for \mathbf{N} , k_d as before

What's the size of $\mathbf{L}^T \mathbf{L}$?



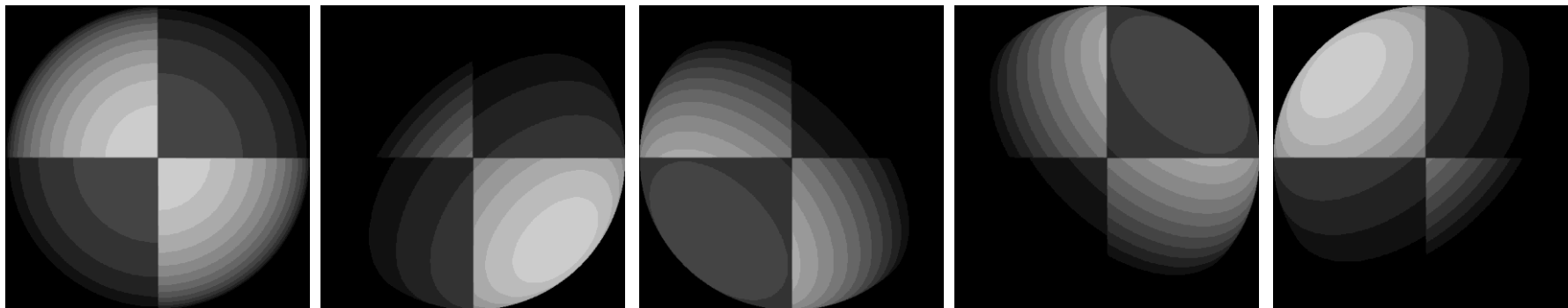
Computing light source directions

Trick: place a chrome sphere in the scene



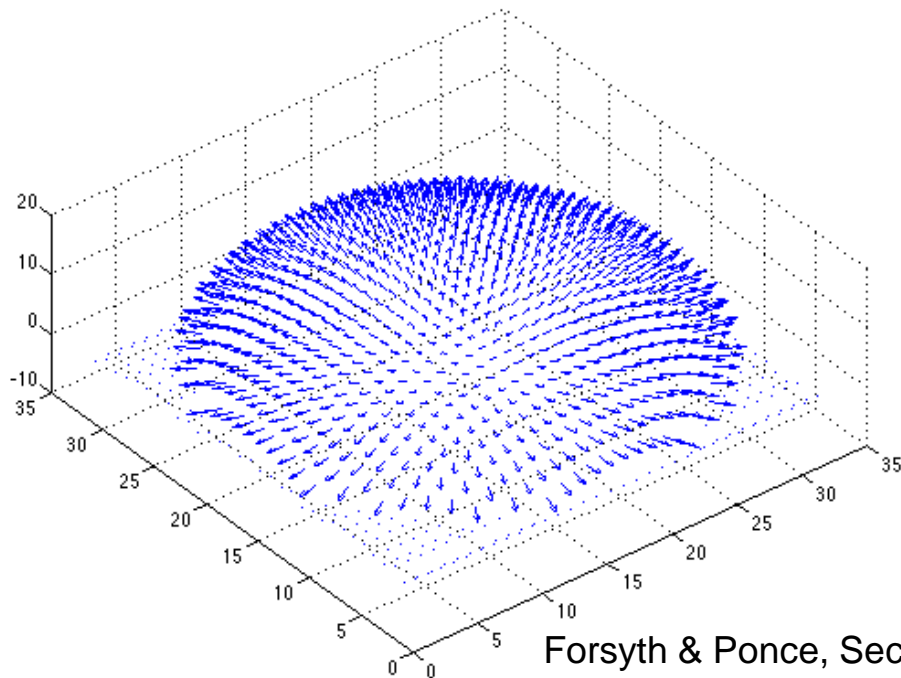
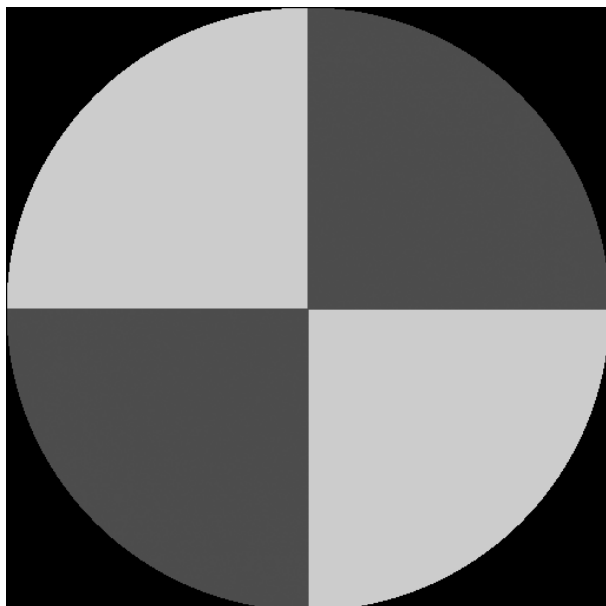
- the location of the highlight tells you where the light source is

Example



Recovered albedo

Recovered normal field



Depth from normals

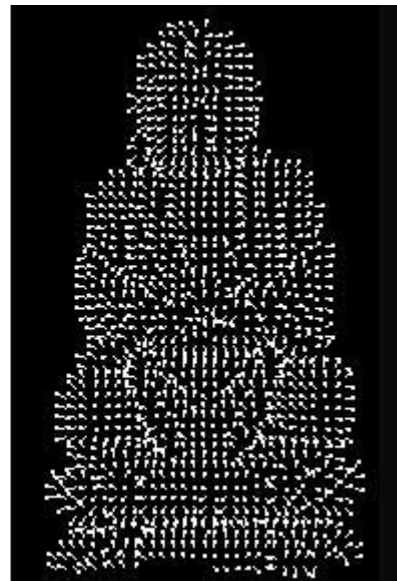
- Solving the linear system per-pixel gives us an estimated surface normal for each pixel



Input photo



Estimated normals

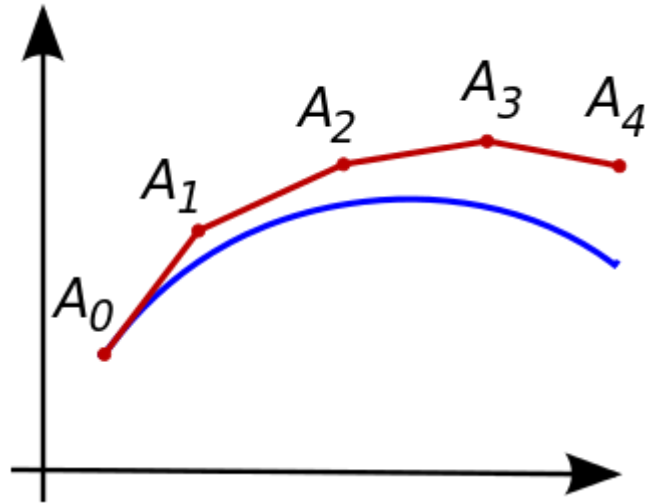


Estimated normals
(needle diagram)

- How can we compute depth from normals?
 - Normals are like the “derivative” of the true depth

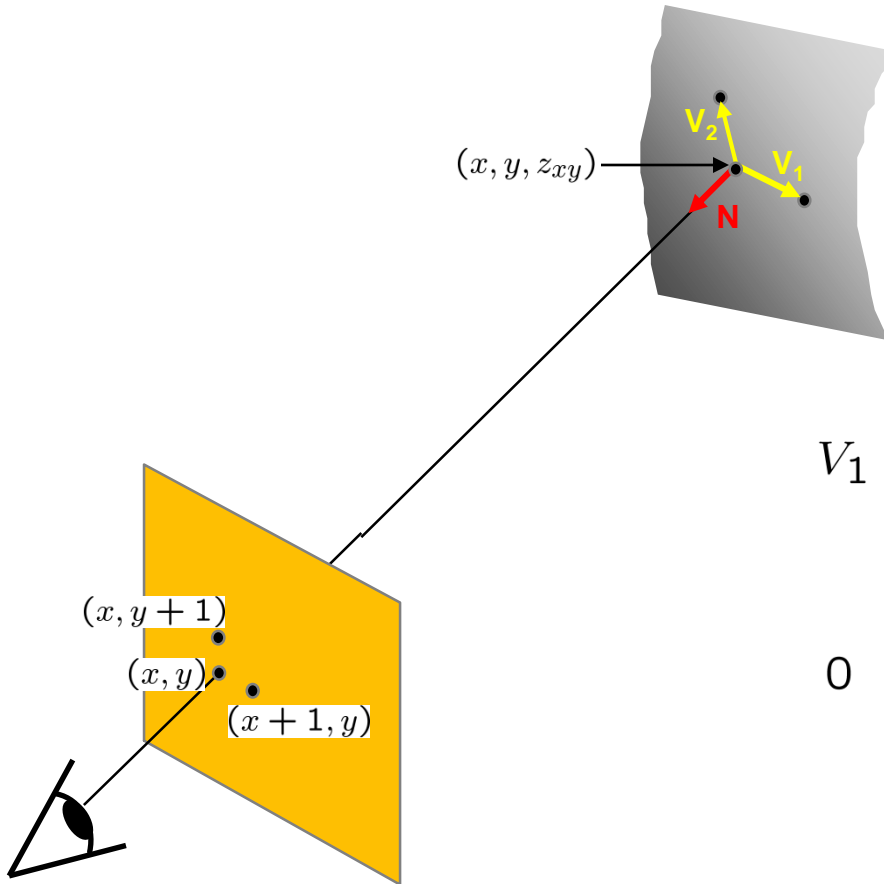
Normal Integration

- Integrating a set of derivatives is easy in 1D
 - (similar to Euler's method from diff. eq. class)



- Could just integrate normals in each column / row separately
- Instead, we formulate as a linear system and solve for depths that *best agree with the surface normals*

Depth from normals



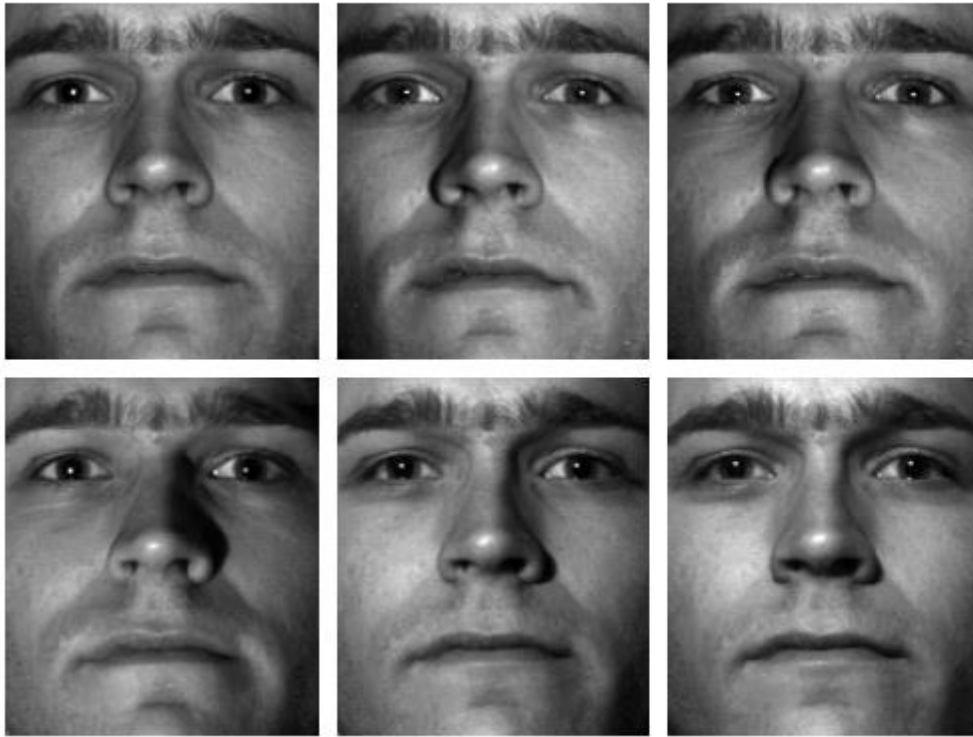
$$\begin{aligned}V_1 &= (x + 1, y, z_{x+1,y}) - (x, y, z_{xy}) \\ &= (1, 0, z_{x+1,y} - z_{xy})\end{aligned}$$

$$\begin{aligned}0 &= N \cdot V_1 \\ &= (n_x, n_y, n_z) \cdot (1, 0, z_{x+1,y} - z_{xy}) \\ &= n_x + n_z(z_{x+1,y} - z_{xy})\end{aligned}$$

Get a similar equation for V_2

- Each normal gives us two linear constraints on z
- compute z values by solving a matrix equation

Results



from Athos Georghiades

Results



Extension

Photometric Stereo from Colored Lighting

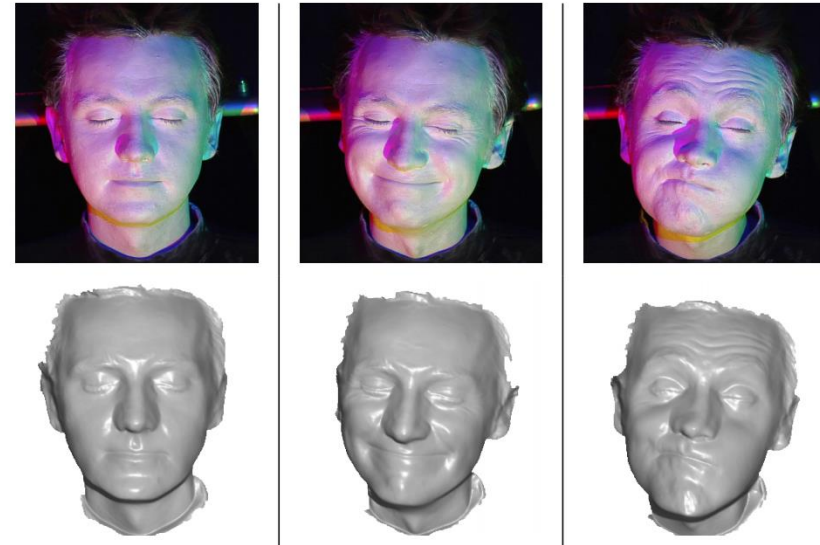
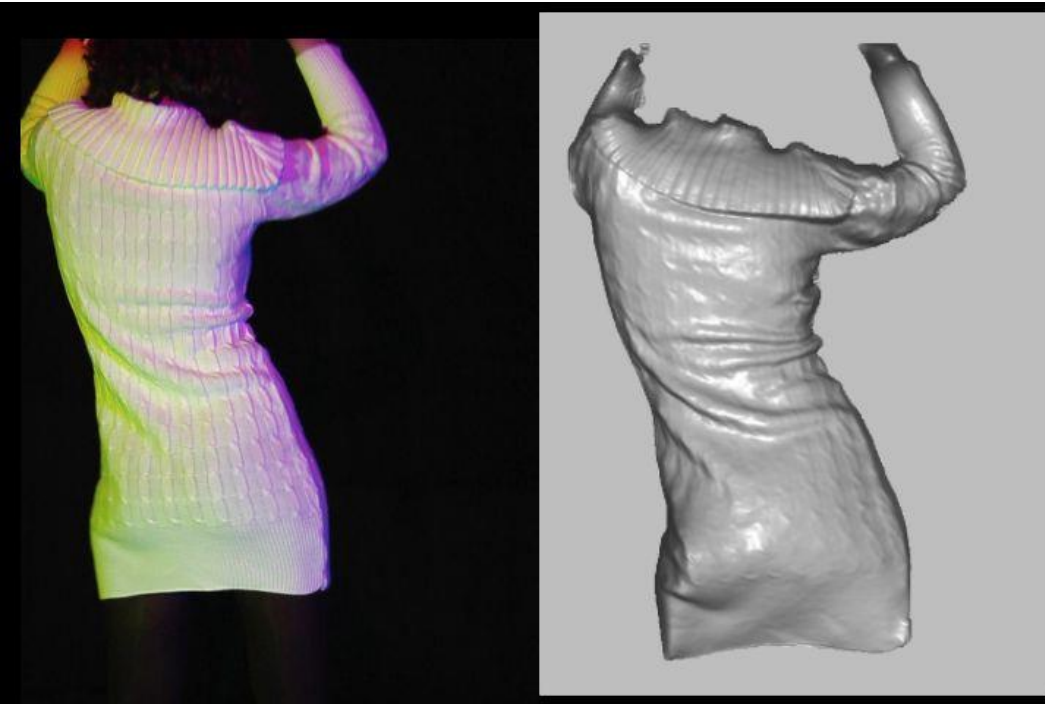


Fig. 2. Applying the original algorithm to a face with white makeup. Top: example input frames from video of an actor smiling and grimacing. Bottom: the resulting integrated surfaces.

Video Normals from Colored Lights

Gabriel J. Brostow, Carlos Hernández, George Vogiatzis, Björn Stenger, Roberto Cipolla
[IEEE TPAMI](#), Vol. 33, No. 10, pages 2104-2114, October 2011.

Questions?

For now, ignore specular reflection



And Refraction...

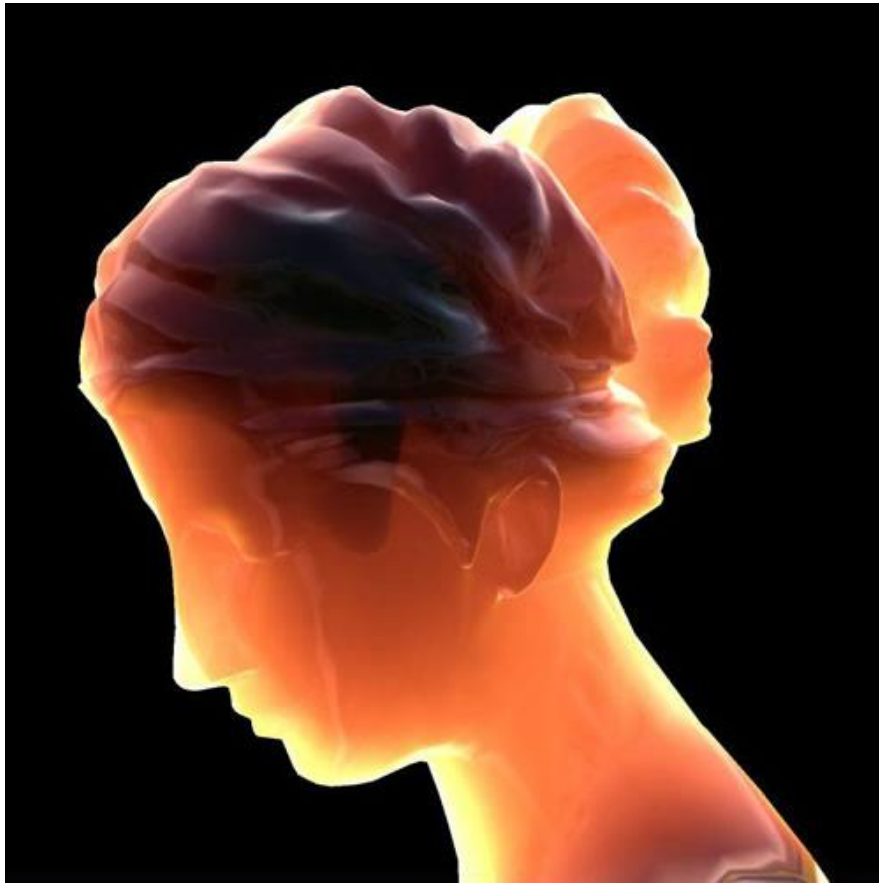


And Interreflections...



Slides from Photometric Methods for 3D Modeling, Matsushita, Wilburn, Ben-Ezra

And Subsurface Scattering...



Limitations

Bigger problems

- doesn't work for shiny things, semi-translucent things
- shadows, inter-reflections

Smaller problems

- camera and lights have to be distant
- calibration requirements
 - measure light source directions, intensities
 - camera response function

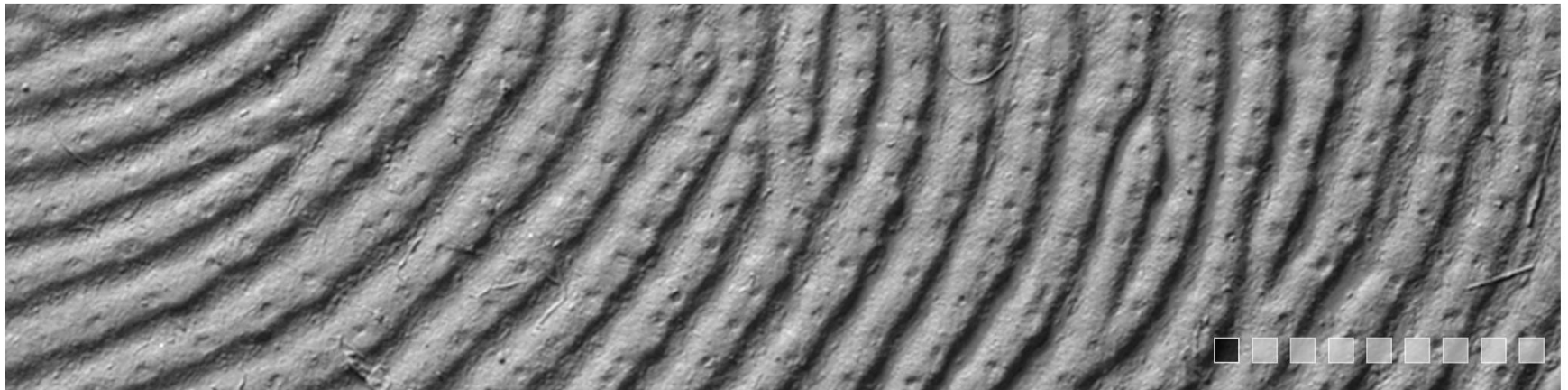
Newer work addresses some of these issues

Some pointers for further reading:

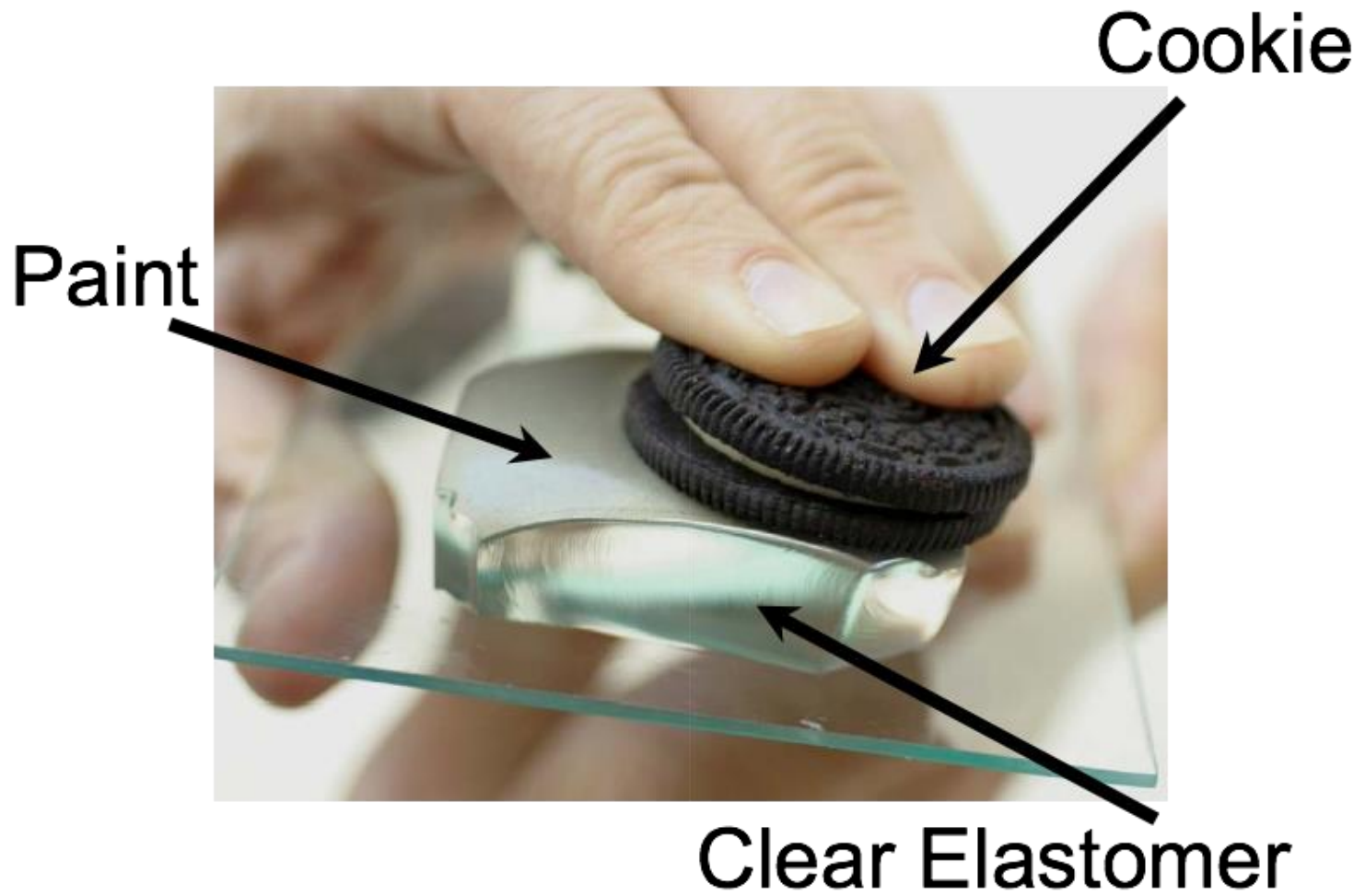
- Zickler, Belhumeur, and Kriegman, "[*Helmholtz Stereopsis: Exploiting Reciprocity for Surface Reconstruction*](#)." IJCV, Vol. 49 No. 2/3, pp 215-227.
- Hertzmann & Seitz, "[*Example-Based Photometric Stereo: Shape Reconstruction with General, Varying BRDFs*](#)." IEEE Trans. PAMI 2005

GELSIGHT

[HOME](#) [PRODUCTS](#) [VIDEOS](#) [IMAGES](#) [PAPERS](#) [NEWS](#) [ABOUT US](#) [CONTACT](#)



Johnson and Adelson, 2009



Johnson and Adelson, 2009



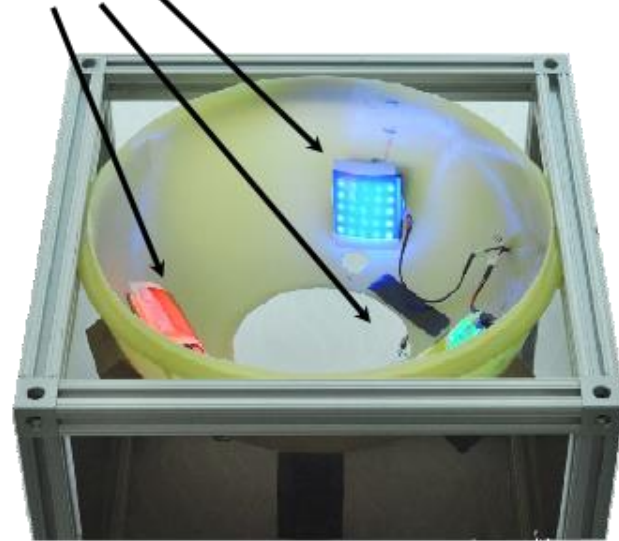


Lights, camera, action

Sensor



Lights



Camera



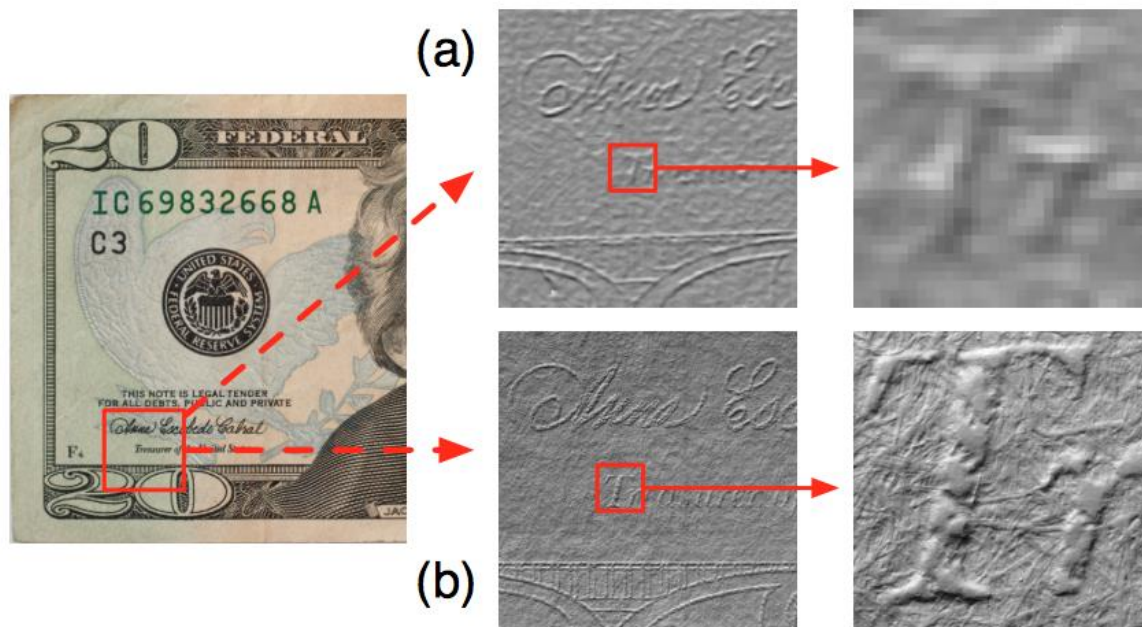
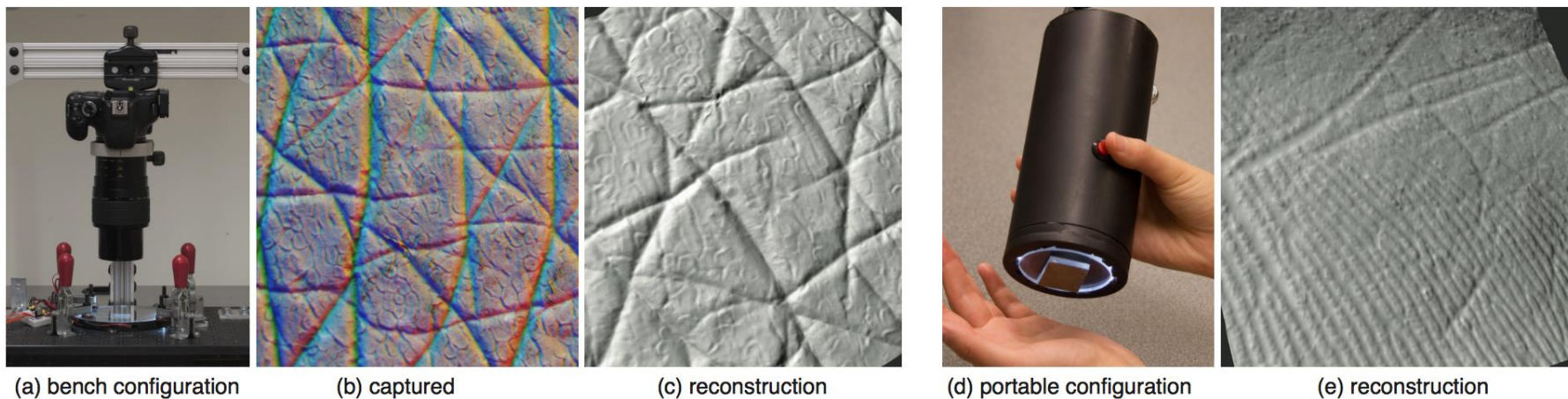
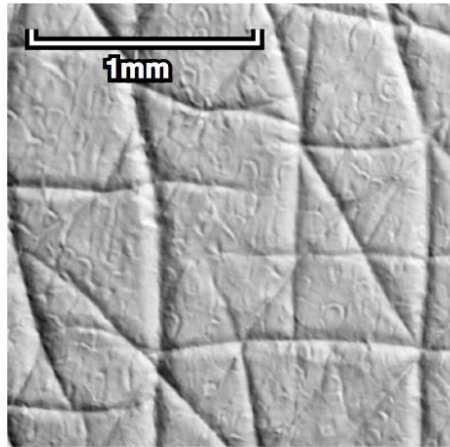
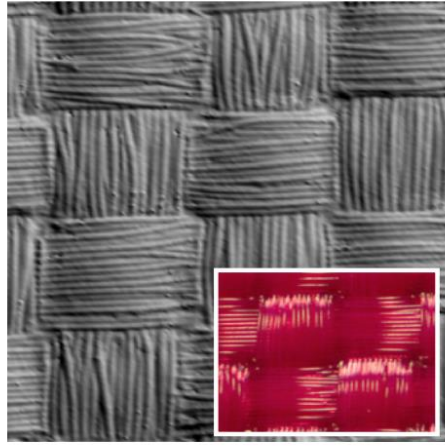


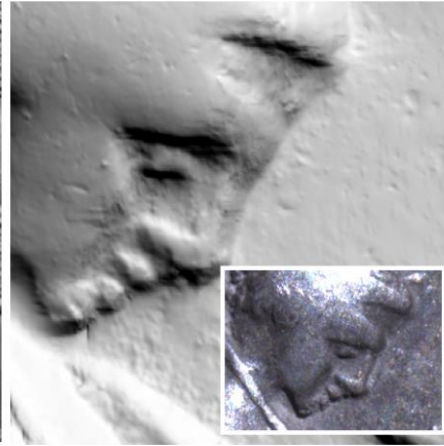
Figure 7: Comparison with the high-resolution result from the original retrographic sensor. (a) Rendering of the high-resolution \$20 bill example from the original retrographic sensor with a close-up view. (b) Rendering of the captured geometry using our method.



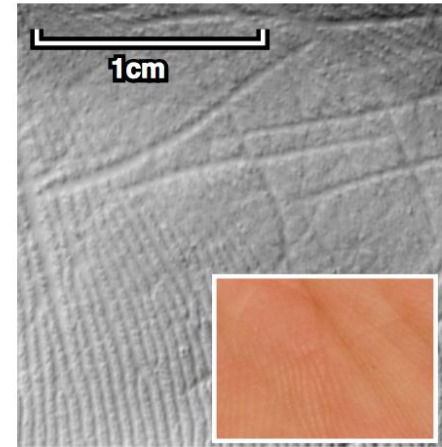
human skin



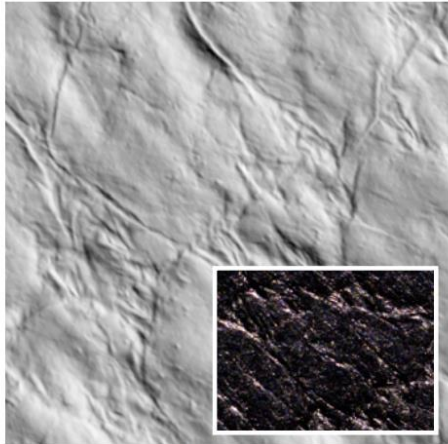
nylon fabric



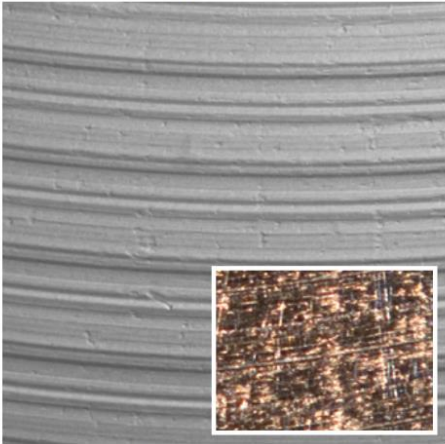
Greek coin



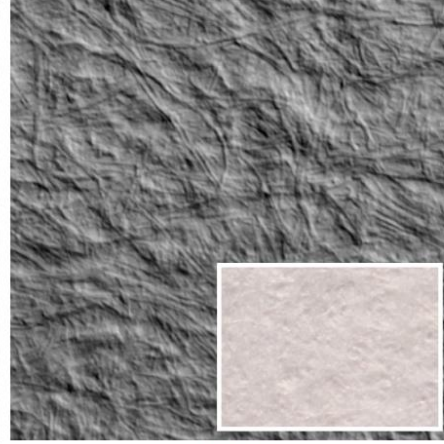
human skin



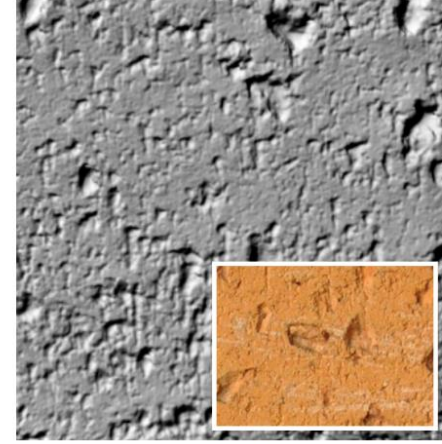
leather



vertically milled metal



paper

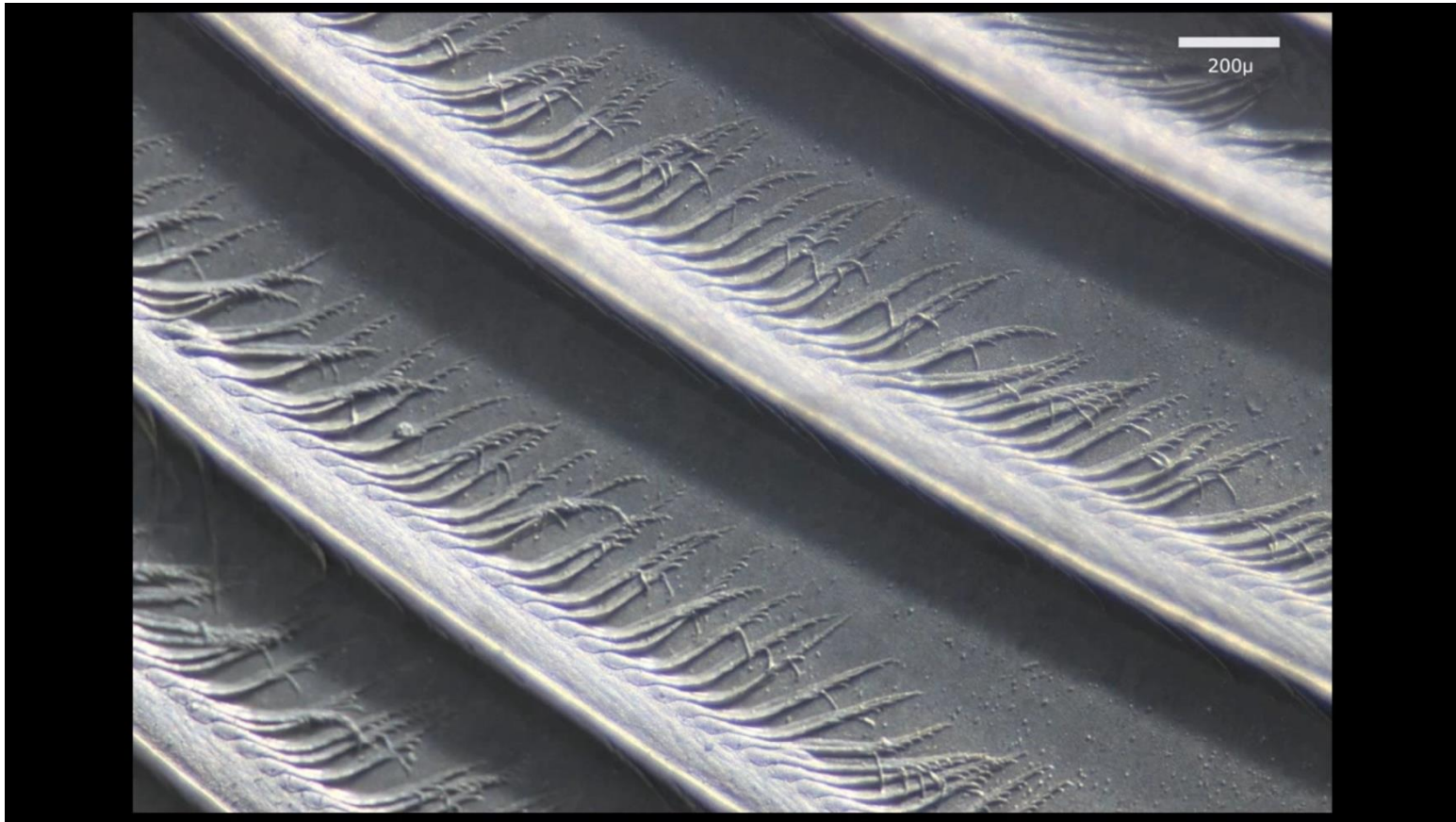


brick

(a) bench configuration

(b) portable configuration

Figure 9: Example geometry measured with the bench and portable configurations. Outer image: rendering under direct lighting. Inset: macro photograph of original sample. Scale shown in upper left. Color images are shown for context and are to similar, but not exact scale.



Sensing Surfaces with GelSight



kimoatmit



138,850 views

<https://www.youtube.com/watch?v=S7gXih4XS7A>

Questions?