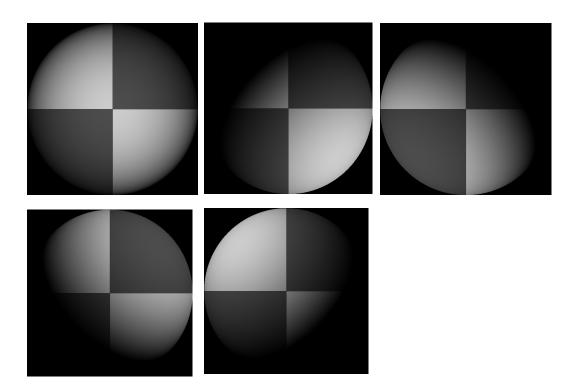
Shape from shading



- Can we find the normals at every point?
 - Under-constrained! (Only have one piece of data per pixel, but we need 2 or 3 (if we need to estimate albedo as well)).
 - We can can impose regularization (smoothness) and consider boundary conditions
- Do normals give us shape?
 - Normals are not shape, but they can be related to the partial derivatives of the shape as a function (x, y, f(x,y))
 - The partial derivatives must satisfy integrability constraints.
 Random normals do not come from a continuous surface!

- Shape from shading is hard! Consider an easier problem.
- Suppose that we have a number of known point sources, and we have successive pictures taken with each one used in turn.



- Shape from shading is hard! Consider an easier problem.
- Suppose that we have a number of known point sources, and we have successive pictures taken with each one used in turn.
- Let g(x,y) be the **unknown** surface normal times the albedo (for the point in the world corresponding to image point (x,y).
- Let V_i be the light source direction, i, times a scalar embodying the light source magnitude and camera sensitivity (**known**).
- Let $I_i(x,y)$ be image intensity (**measured**).

Then
$$I_i(x,y) = \mathbf{V}_i \bullet \mathbf{g}(x,y)$$
 (Lambert's law)

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So how to solve for the surface?

Simpler---how to solve for g(x,y)?

How many lights do we need (assuming that albedo is not known)?

$$I_i(x,y) = \mathbf{V}_i \bullet \mathbf{g}(x,y)$$
 (Lambert's law)

Hopefully this looks like the i'th row of matrix, V, multiplied by a vector, $\mathbf{g}(\mathbf{x},\mathbf{y})$.

Thus combining the conditions given by each light, i, we get

$$i = Vg$$

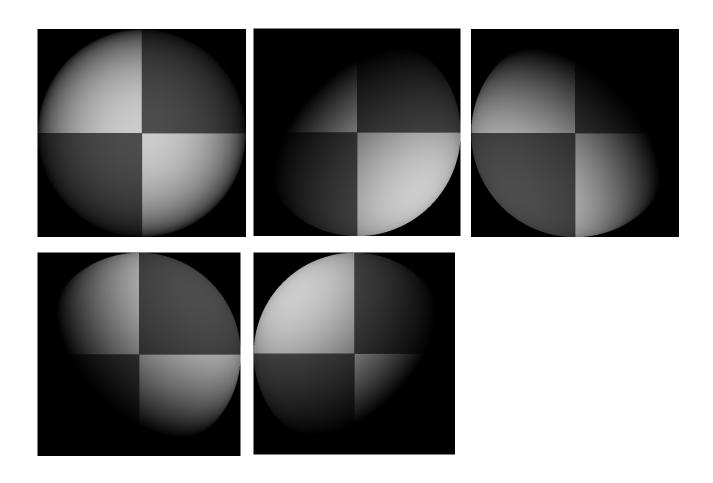
Where the ith element of **i** is $I_i(x,y)$ and the ith row of V is V_i

Since g has three elements, we need at least 3 lights.

If the number of lights is more than than 3, then use least squares!

You should understand the construction of this problem.

Example figures



Dealing with shadows

Each point is in K images (one for each light)

If $I_i(x,y)$ is in shadow (too dark), then ignore it.

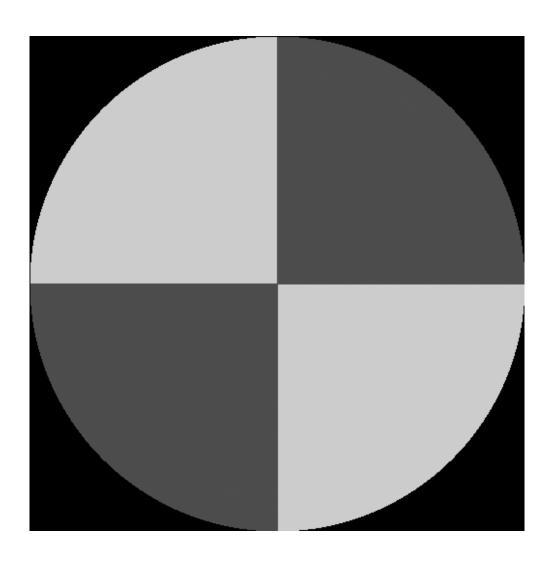
As in the book, we can simplify this in a program by multiplying both sides by a diagonal matrix with the image intensities on the diagonal.

This approach weights the equations according to image intensity, and so pixels in shadow are ignored (weight is zero). This changes the impact of the non-zero ones also (possibly for the better, depending on your error model).

Dealing with shadows

$$\begin{pmatrix} I_1^2(x,y) \\ I_2^2(x,y) \\ \vdots \\ I_n^2(x,y) \end{pmatrix} = \begin{pmatrix} I_1(x,y) & 0 & \dots & 0 \\ 0 & I_2(x,y) & \dots & \dots \\ \vdots \\ 0 & \dots & \dots & 0 \\ 0 & \dots & 0 & I_n(x,y) \end{pmatrix} \begin{pmatrix} \mathbf{V}_1^T \\ \mathbf{V}_2^T \\ \vdots \\ \mathbf{V}_n^T \end{pmatrix} \mathbf{g}(x,y)$$
 unknown image intensity shadow => 0 known light becomes squared

Recovered reflectance



Recovered normal field

