

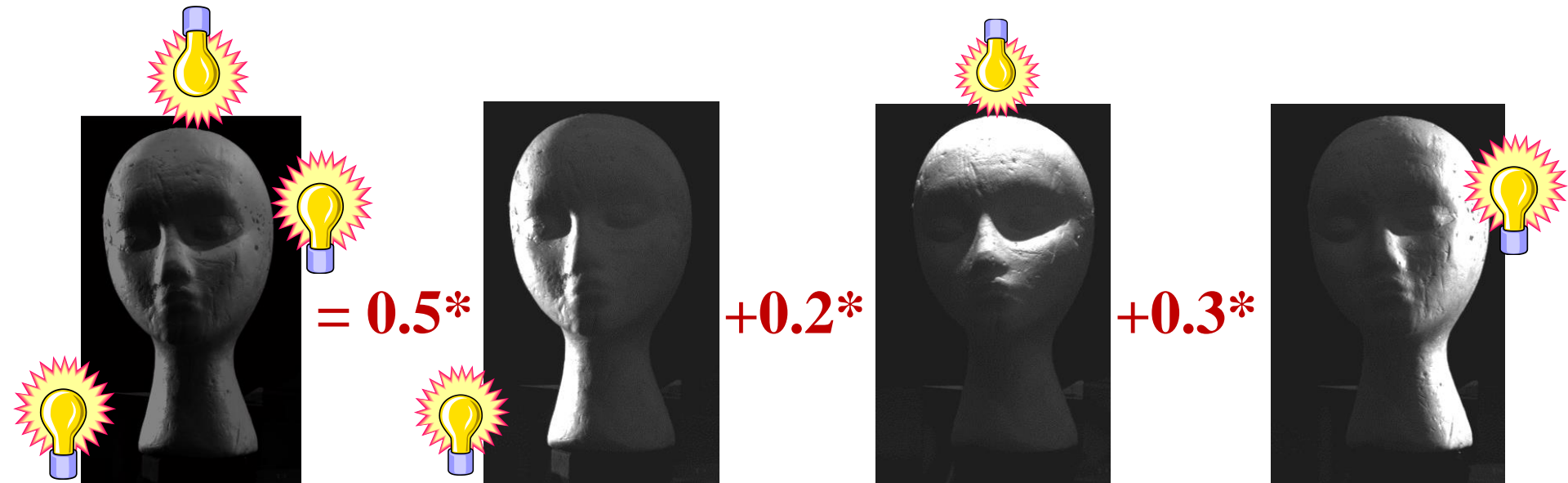
Lambertian model of reflectance II: harmonic analysis

Ronen Basri

Weizmann Institute of Science

Illumination cone

- What is the set of images of an object under different lighting, with any number of sources?
- Images are additive and non-negative
- This set, therefore, forms a *convex cone* in \mathbb{R}^p , p number of pixels (Belhumeur & Kriegman)



Illumination cone

- Cone characterization is generic, holds also with specularities, shadows and inter-reflections
- Unfortunately, representing the cone is complicated (infinite degrees of freedom)
- Cone is “thin” for Lambertian objects; indeed the illumination cone of many objects can be represented with few PCA vectors (Yuille et al.)

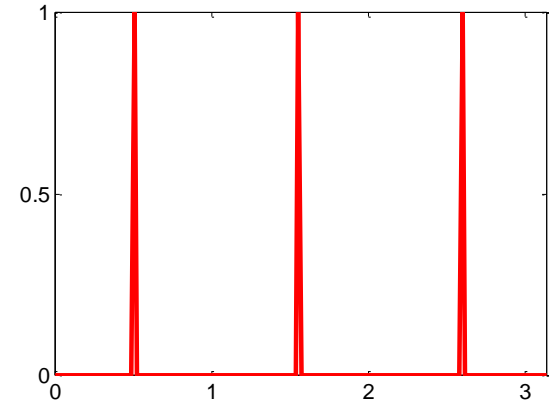
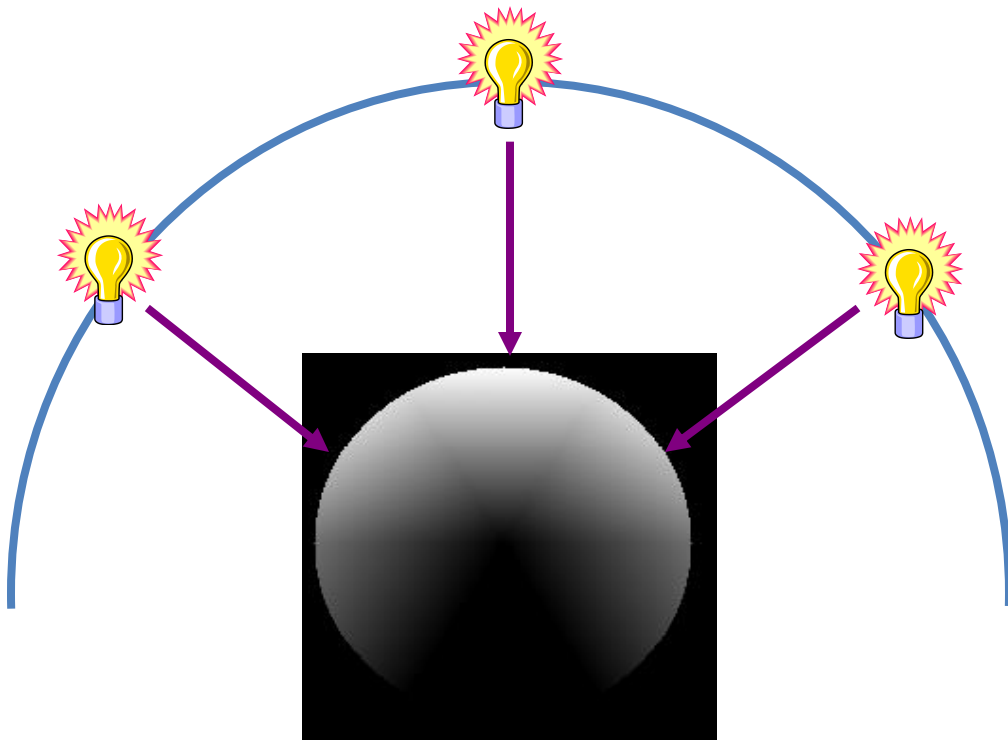
Illumination cone is often thin

	Ball	Face	Phone	Parrot
#1	48.2	53.7	67.9	42.8
#3	94.4	90.2	88.2	76.3
#5	97.9	93.5	94.1	84.7
#7	99.1	95.3	96.3	88.5
#9	99.5	96.3	97.2	90.7

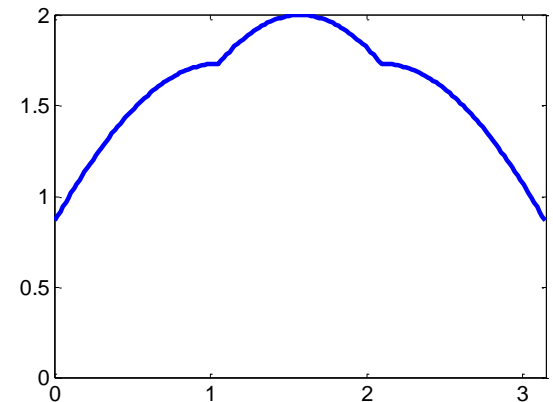
(Yuille et al.)

Lambertian reflectance is smooth

(Basri & Jacobs; Ramamoorthi & Hanrahan)

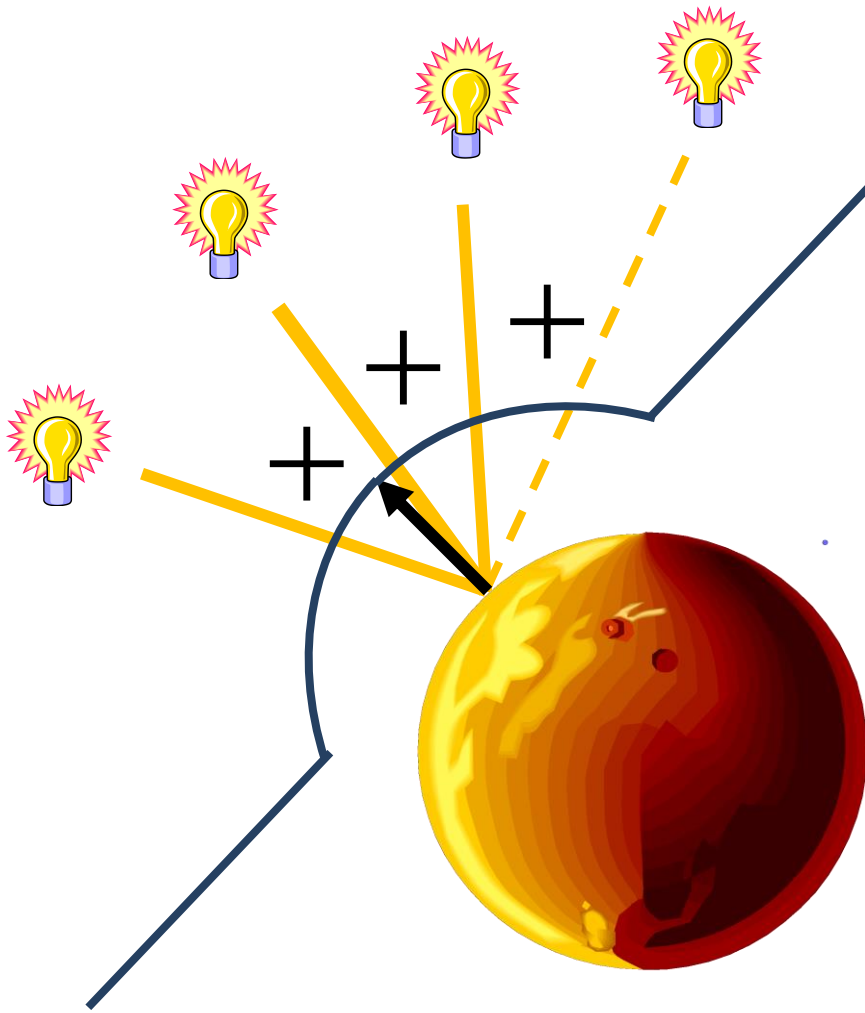


lighting



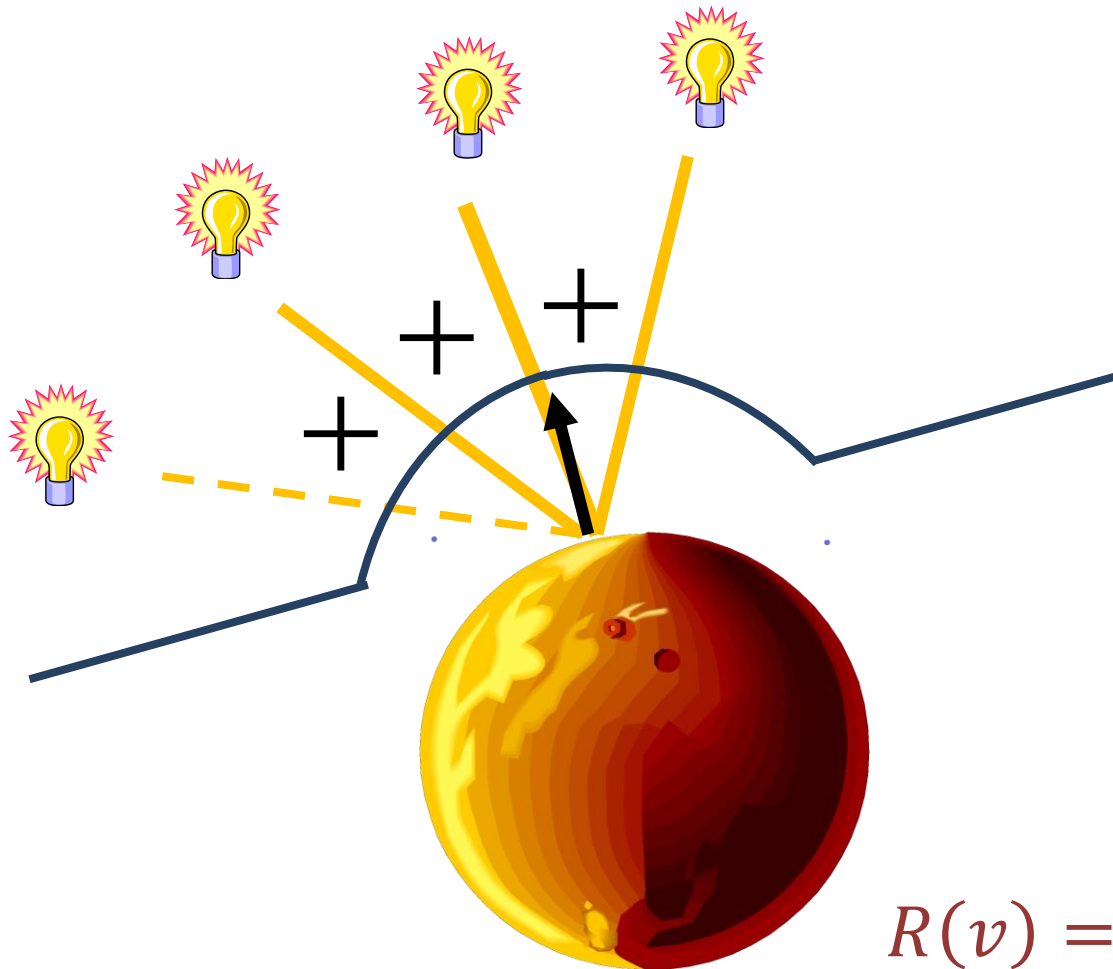
reflectance

Reflectance obtained with convolution



$$R(v) = \int_{S^2} k(u, v) l(u) du$$

Reflectance obtained with convolution



$$R(v) = \int_{S^2} k(u, v) l(u) du$$

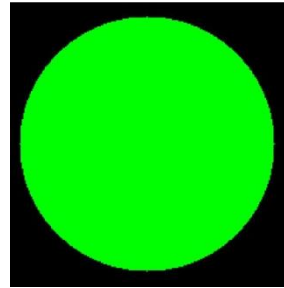
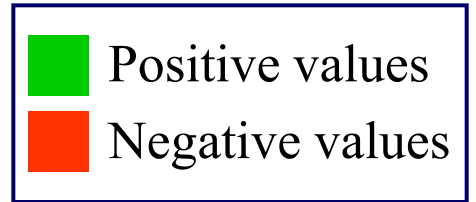
Spherical harmonics

$$Y_{nm}(\theta, \phi) = \sqrt{\frac{(2n+1)(n-|m|)!}{4\pi(n+|m|)!}} P_{nm}(\cos \theta) e^{im\phi}$$
$$p_{nm}(z) = \frac{(1-z^2)^{m/2}}{2^n n!} \frac{d^{n+m}}{dz^{n+m}} (z^2 - 1)^n$$

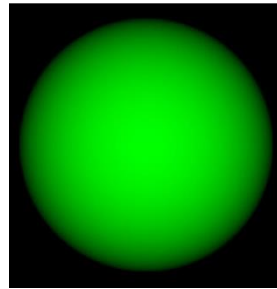
- Orthonormal basis for functions on the sphere
- n 'th order harmonics have $2n+1$ components
- Rotation = phase shift (same n , different m)
- In space coordinates: polynomials of degree n
- *Funk-Hecke convolution theorem*

Spherical harmonics

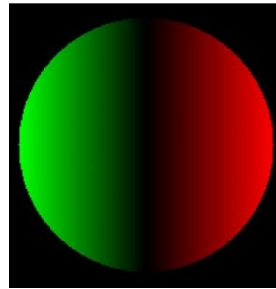
$$X^2 + Y^2 + Z^2 = 1$$



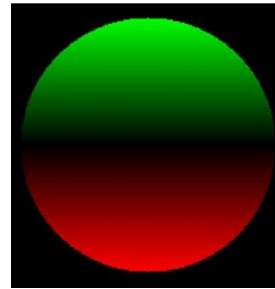
1



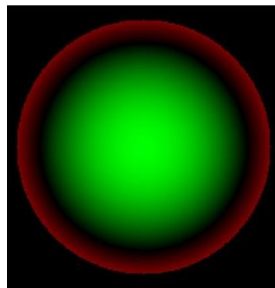
Z



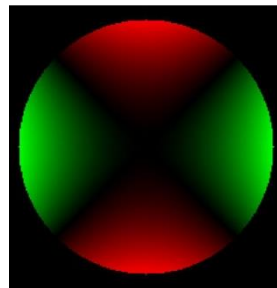
X



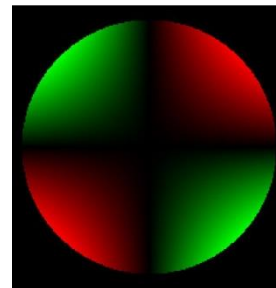
Y



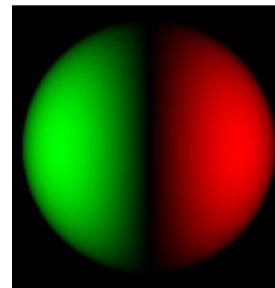
$3Z^2 - 1$



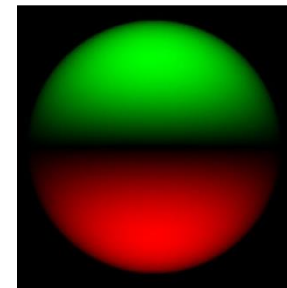
$X^2 - Y^2$



XY



XZ



YZ

Harmonic approximation

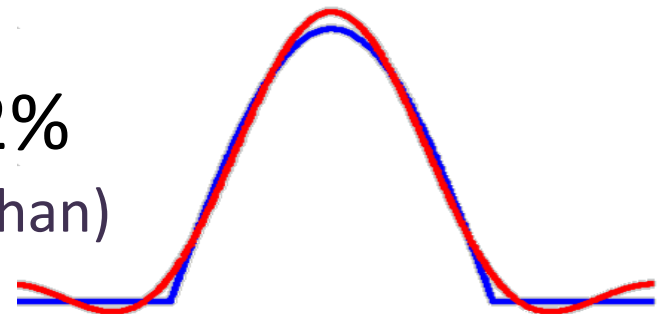
- Lighting, in terms of harmonics

$$\ell(\theta, \phi) = \sum_{n=0}^{\infty} \sum_{m=-n}^n l_{nm} Y_{nm}(\theta, \phi)$$

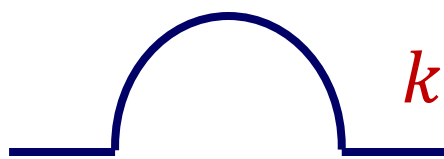
- Reflectance

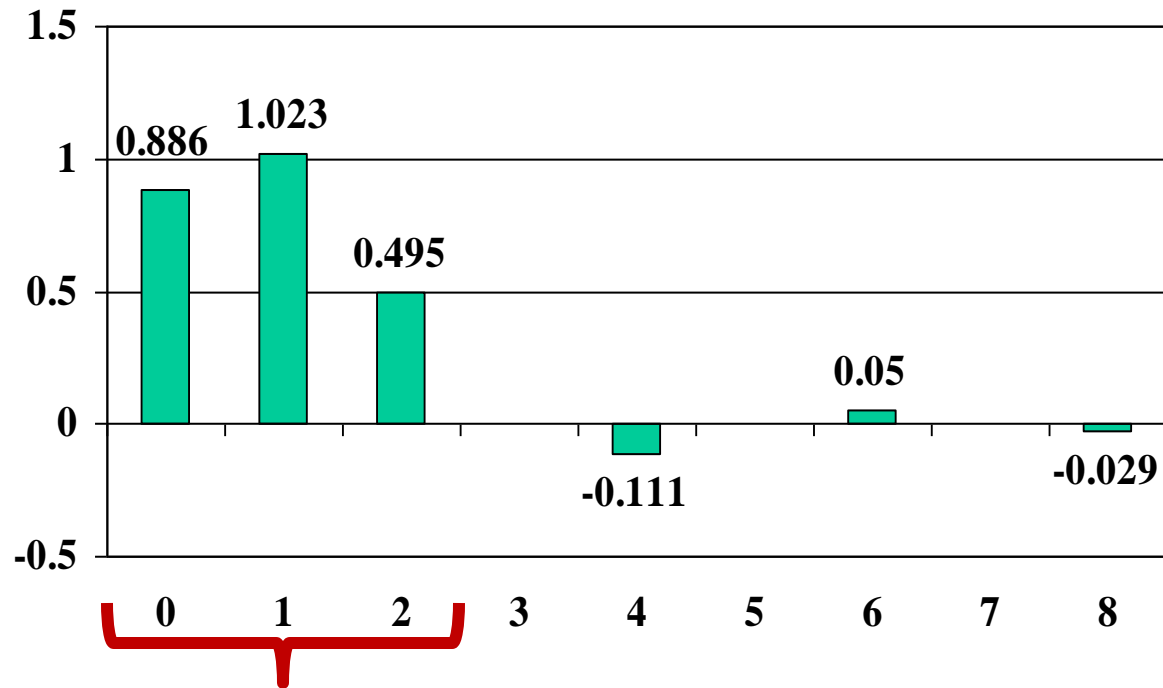
$$r(\theta, \phi) = k * \ell \approx \sum_{n=0}^2 \sum_{m=-n}^n k_n l_{nm} Y_{nm}(\theta, \phi)$$

- Approximation accuracy, 99.2%
(Basri & Jacobs; Ramamoorthi & Hanrahan)



Harmonic transform of kernel


$$k(\theta) = \max(\cos \theta, 0) = \sum_{n=0}^{\infty} k_n Y_{n0}$$



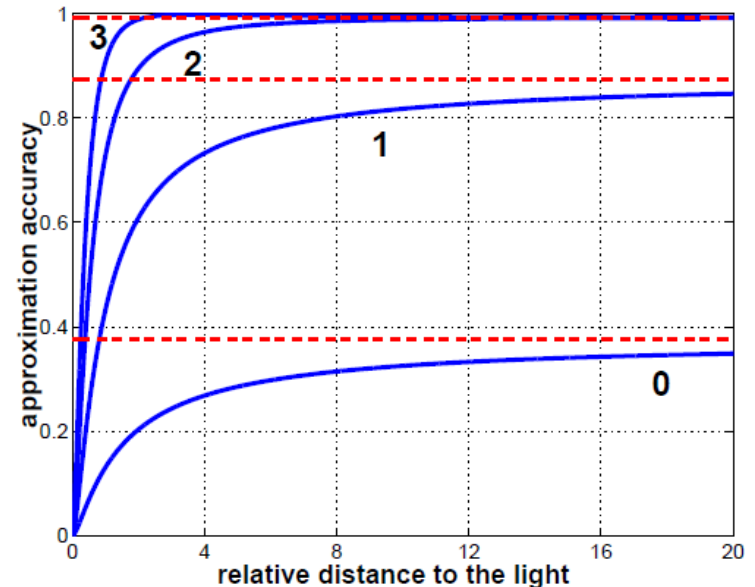
99.2%

Subspace approximation

- Up to 2nd order:
 - 9 basis images
 - Accuracy: 99.2%
- Up to 1st order:
 - 4 basis images: ambient + point source
 - Accuracy: 87.5%
- In practice, due to self occlusions ~98% can be achieved with just 6 basis images (Ramamoorthi)

Scope

- Harmonic representations handle convex, lambertian objects with multiple light sources (including attached shadows)
- Harmonic representations do not model cast shadows and inter-reflections
- Accuracy is maintained for fairly close light sources
- Representing specular objects may require a very large basis



Applications

- We can use this theory to predict novel appearances under new lighting
- Harmonic lighting theory has led to applications in
 - Face recognition
 - Photometric stereo
 - 3D reconstruction with prior
 - Motion analysis

“Harmonic faces”



ρ

□ Positive values
■ Negative values



ρn_z



ρn_x



ρn_y

ρ Albedo
 n Surface normal
 $n = (n_x, n_y, n_z)$



$\rho(3n_z^2 - 1)$



$\rho(n_x^2 - n_y^2)$



$\rho n_x n_y$



$\rho n_x n_z$



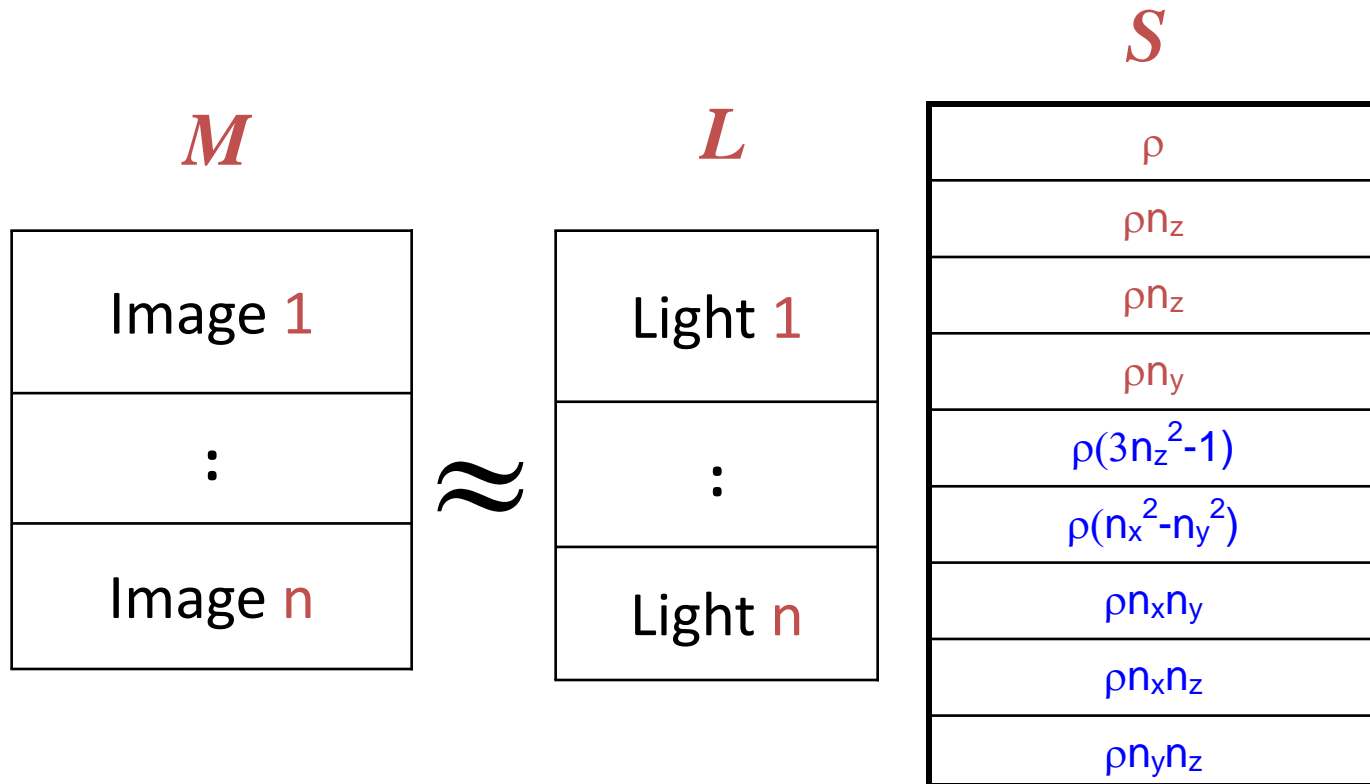
$\rho n_y n_z$

Non-negative light

- We can enforce in addition that light is non-negative, by projecting the illumination cone onto the harmonic space
- Closed-form constraints for 1st order approximation
- Sampling method, or Toeplitz matrix (Shirdhonkar & Jacobs) for higher orders

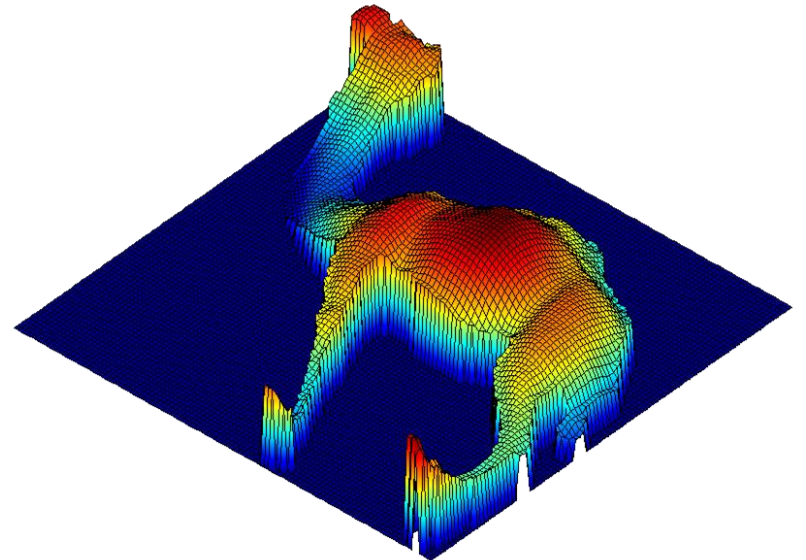
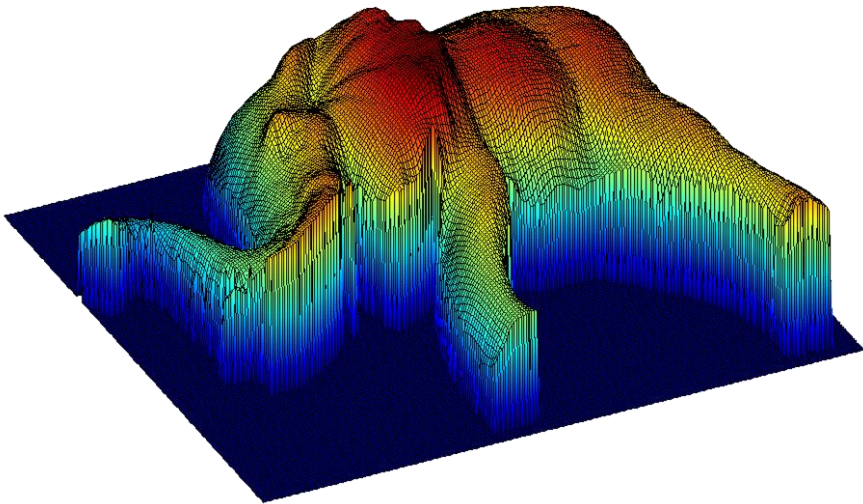
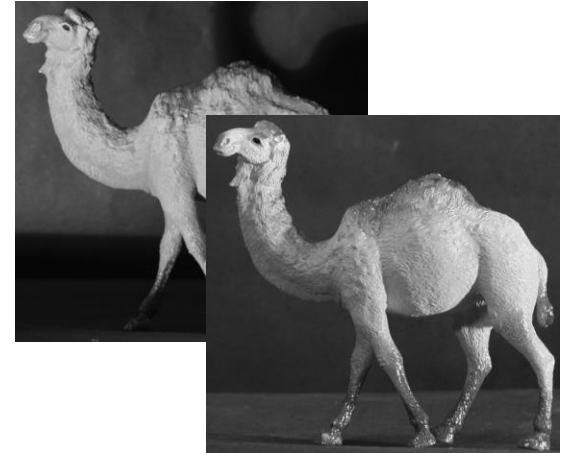
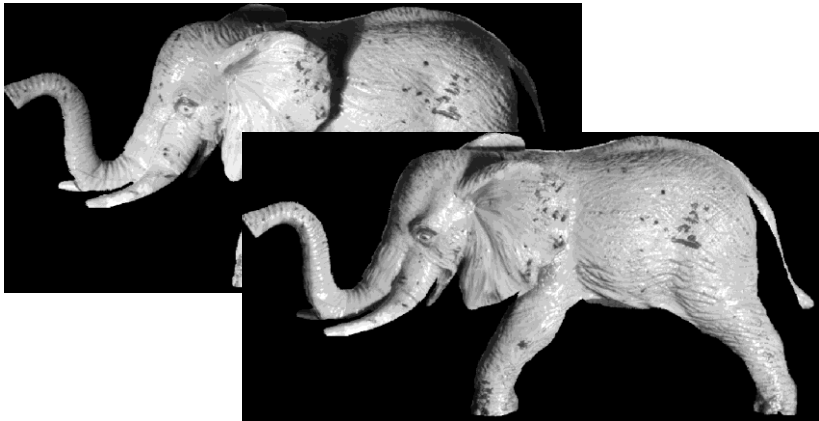
Photometric stereo

(Basri, Jacobs & Kemelmacher)



SVD recovers *L* and *S* up to an $(r \times r)$ ambiguity

Photometric stereo



Reconstruction with a prior

(Kemelmacher & Basri)

- Given just one image SFS is impractical
- Reconstruction is possible when a prior is available
- Energy

$$\min_{l, \rho, Z} \iint_{\Omega} D + S$$

- Data term

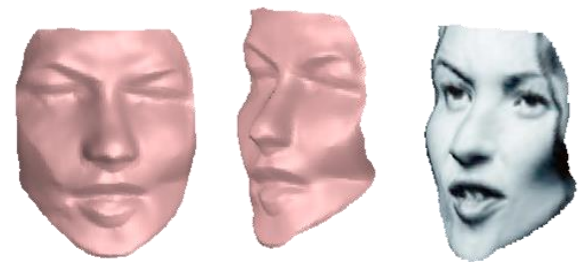
$$D = (I - \rho l^T Y(\hat{n}))^2$$

- Regularization

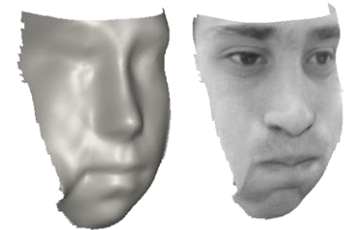
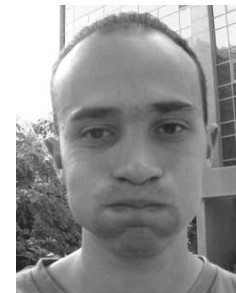
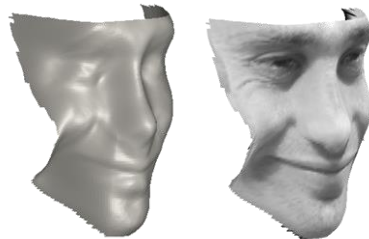
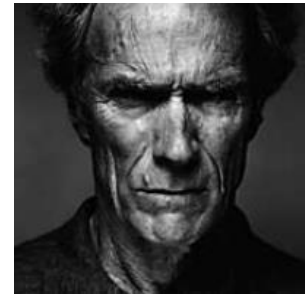
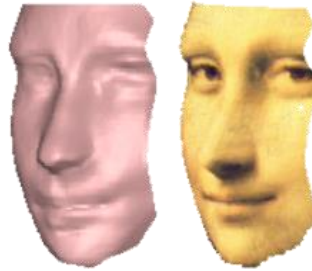
$$S = \lambda_1 \left(\Delta(Z - Z_{prior}) \right)^2 + \lambda_2 \left(\Delta(\rho - \rho_{prior}) \right)^2$$

- Solve as a linear PDE

Reconstruction with a prior



More...

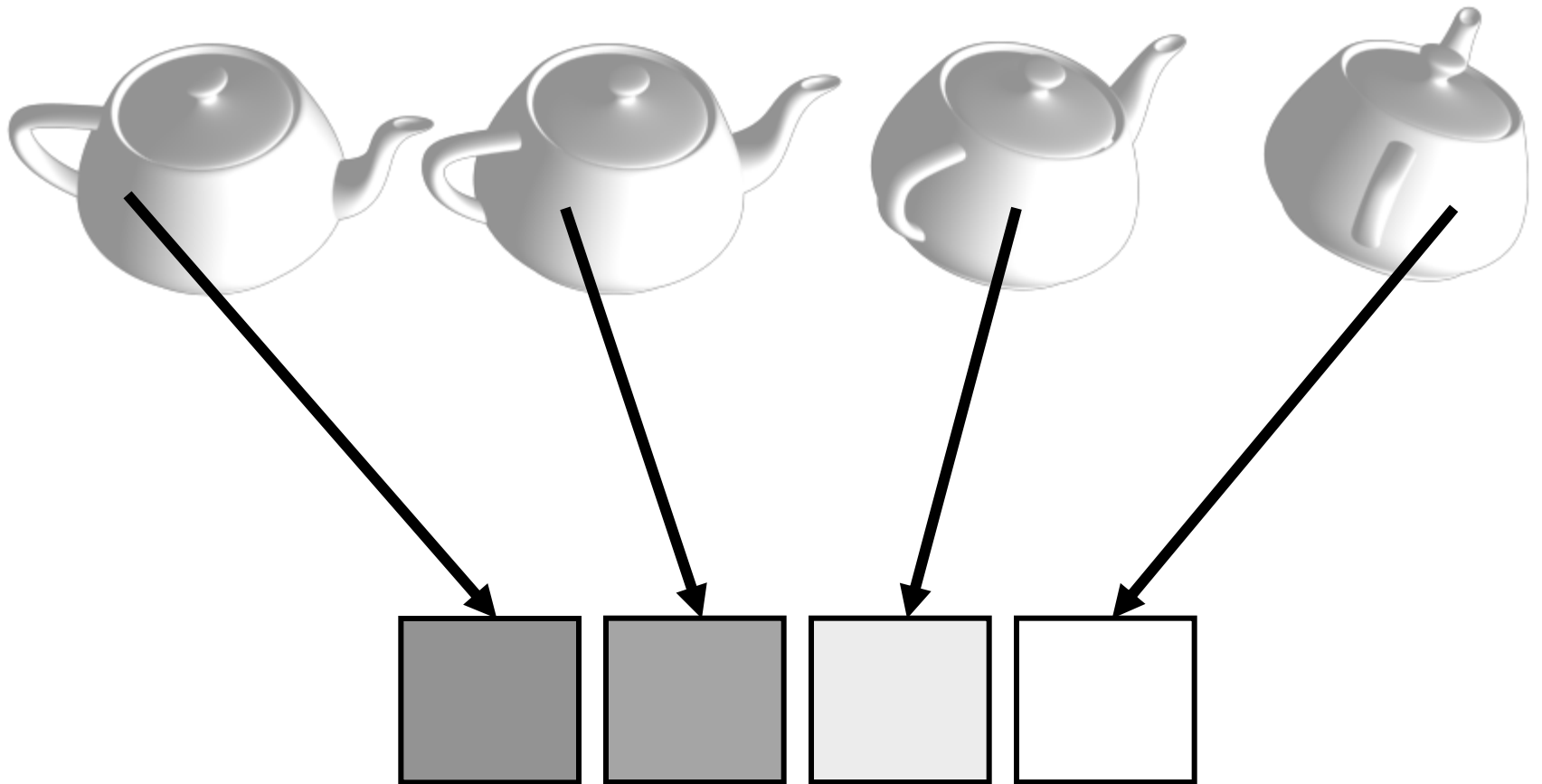


Mooney faces



(Kemelmacher, Nadler & B, CVPR 2008)

Motion + lighting



Motion + lighting

(Basri & Frolova)

- Given 2 images

$$I(p) = \rho l^T \hat{n} \quad J(p') = \rho l^T R \hat{n}$$

- Take ratio to eliminate albedo

$$\frac{J(p')}{I(p)} = \frac{l^T R \hat{n}}{l^T \hat{n}}$$

- If motion is small we can represent $J(p')$ using a Taylor expansion around p

Small motion

- We obtain a PDE that is quasi linear in z

$$az_x + bz_y = c$$

- Where

$$a(x, y, z) = l_1(I_\theta - zJ_x) - l_3I$$

$$b(x, y, z) = l_2(I_\theta - zJ_x)$$

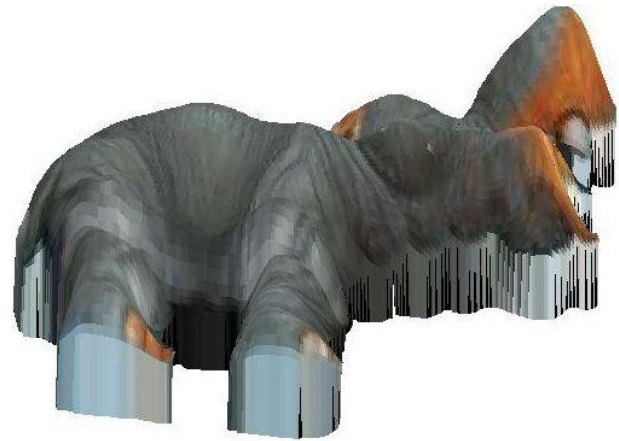
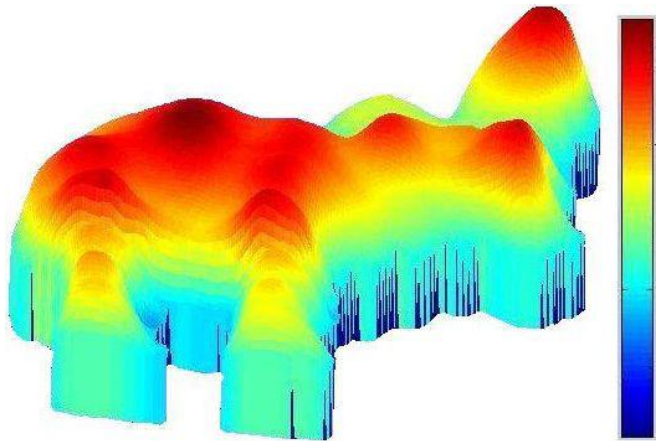
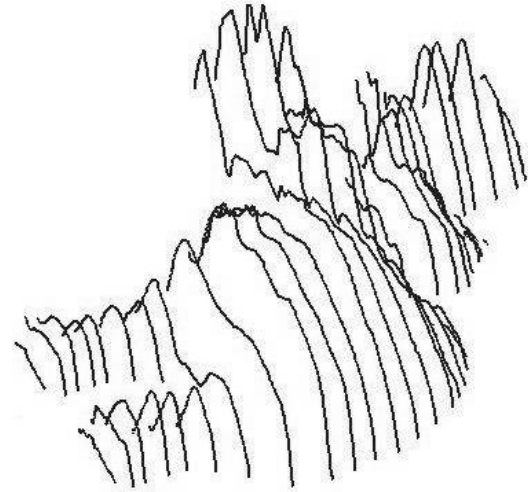
$$c(x, y, z) = -l_3(I_\theta - zJ_x) - l_1I$$

with

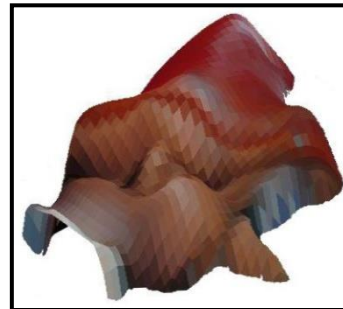
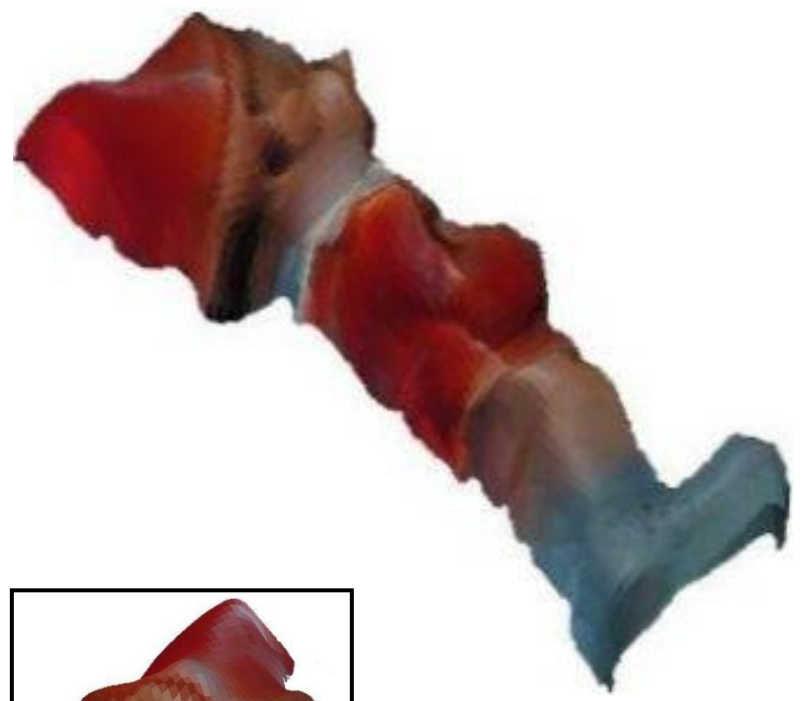
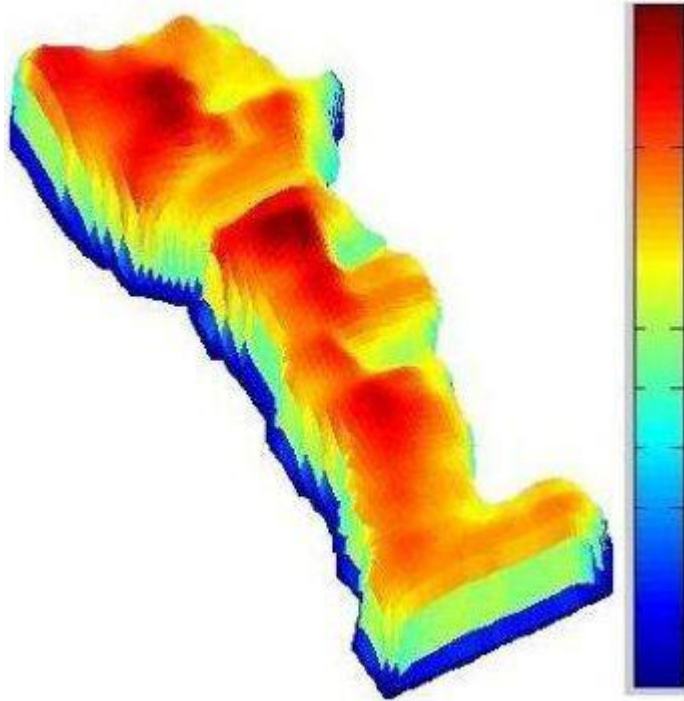
$$I_\theta = \frac{J - I}{\theta}$$

- Can be solved with continuation (characteristics)

Reconstruction



More reconstructions



Conclusion

- Understanding the effect of lighting on images is challenging, but can lead to better interpretation of images
- Harmonic analysis allows to model complex lighting in a linear model
- Various applications in recognition and reconstruction
- We only looked at Lambertian objects...