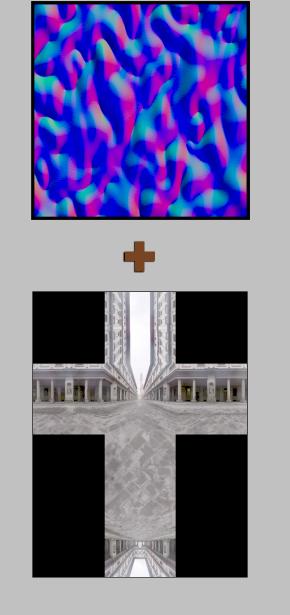


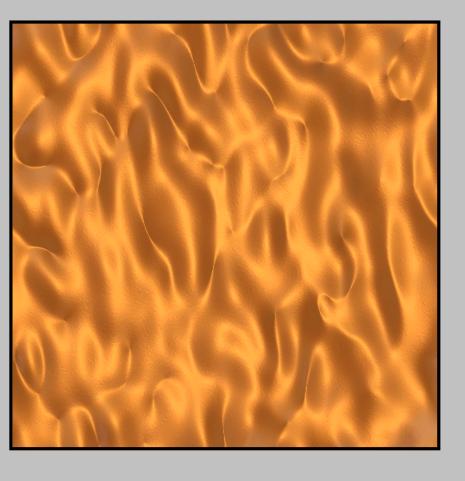
Sparse BRDF approximation using compressive sensing

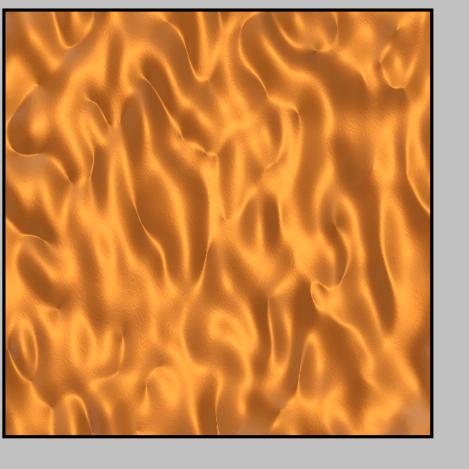
Benoît Zupancic - Cyril Soler

How much information do we need to reconstruct a BRDF?

Overview





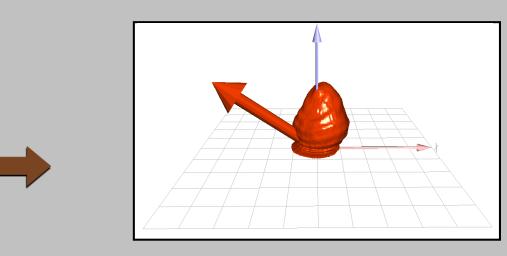


Synthetic input image

Take measurements and solve

 $x = argmin||Ax - y||_{2}^{2} + \lambda |x|_{1}$ where A is the measurement matrix,

x the BRDF representation y the vector of measurements



Reconstucted lobes

Context

BRDF acquisition is tedious (4D Data): either explicit methods are difficult to setup and produce high volume Data, or acquisition methods based on parametric models reduce dimensionnality of BRDF space, with no guaranty that a given model can ever fit particular measurements.

Contribution

We experiment a new acquisition method from a single image, knowing the normals and illumination. We used synthetic input images (as a proof of concept) but we aim at using real data in the future.

Method

We work with isotropic BRDF, representing them into a fourier basis : Given an incoming angle θ_i , the lobe is expressed spherical harmonics, and coefficients are decomposed into cosines functions of θ_i :

$$\sum_{k} \sum_{l,m} a_{k,l,m} cos(\frac{k\theta_i}{n}) Y_l^m(\theta_o, \phi_o)$$

Then we solve a L₂ - L₁ problem to fit the coefficients with an input image.

Compressive sensing

We use the SpaRSA algorithm to fit the BRDF. [Wright and al. 2009]

Starting from a random x_0 , the method iterates :

$$x_{n+1} = S(x_n + \frac{1}{\alpha_n} A^T(y - Ax_n), \frac{\lambda}{\alpha_n}) \text{ with } \alpha_n = \frac{||Ax_n||_2^2}{||x_n||_2^2}$$

Where S, the is the soft-threshold operator. This algorithm finds the sparsest solution to the fitting problem.

 λ is the sparsity weight. Most computation costs rely on the precomputation of the matrix A.

Results





Reconstructed material

Original Material



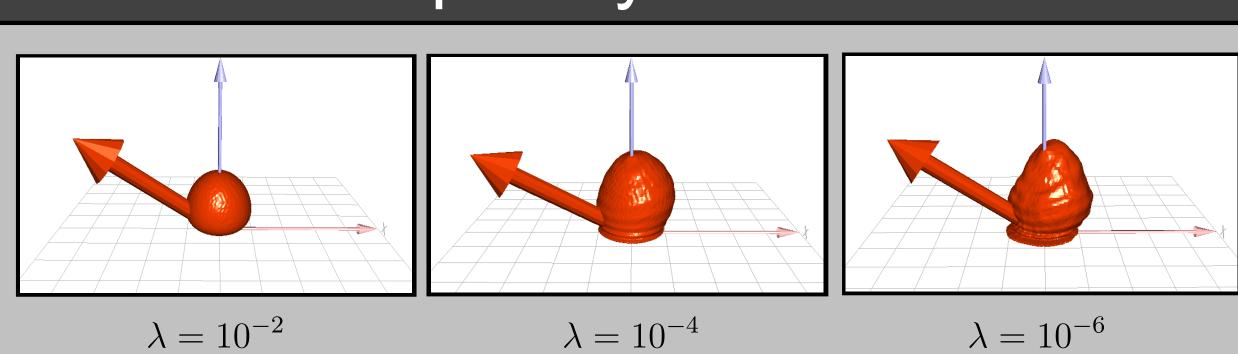


Reconstructed material

Original Material

We produced input images using real materials from MERL Database [Matusik and al. 2003]. Here blue-fabric and gold-paint are presented. Output images are produced using geometry from Stanford repository. Rendering with acquired BRDF is on the left and ground truth on the right.

Control of sparsity



From left to right: sparsity factor decreases, making the solution less sparse and getting closer to a classical gradient descent solution.

Conclusion

Sparse reconstruction is possible with a moderate amount of input data.

Future work : glossy/specular BRDF.

Simultaneous acquisition of BRDF and normals







