

FAKULTÄT FÜR INFORMATIK

Faculty of Informatics

Diplomarbeitspräsentation



Interactive 3D Reconstruction and BRDF Estimation for Mixed Reality Environments

Masterstudium: Computergraphik & Digitale Bildverarbeitung

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Scope and Motivation



typical mixed reality system

Requires to have **geometry**, **material** and **lighting** information about the real world scene

- All of these can change dynamically in mixed reality
- Estimations need to be done in real time since interaction should

| Contribution | | |
|---|----|--|
| We developed a BRDF estimation algorithm that | -6 | |
| runs at interactive frame rates | | |
| can handle dynamic scenes like moving objects, insertion or removal of objects and lighting changes | | |

be possible

desired mixed reality system



does not need any pre-processing

One reason for the significant speed-up was

a novel GPU K-Means implementation using MIP maps

Results

We achieve ~2 estimations per second*, which is a huge speed-up compared to 20 minutes for a single one and we thus have proved that realtime estimations are possible. Additionally, our approach is able to handle changes of the lighting and the geometry.

Left is a Phong rendering of the estimated parameters and right an integration of the technique into a mixed reality system. Note the red color bleeding on the virtual Buddha and the real objects and that the pocket lamp is in fact turned off. Apart from the Buddha nothing was modeled, so all the geometry and color information needed for the rendering was estimated by the algorithm.

* on GTX 580 and Core i7-2600



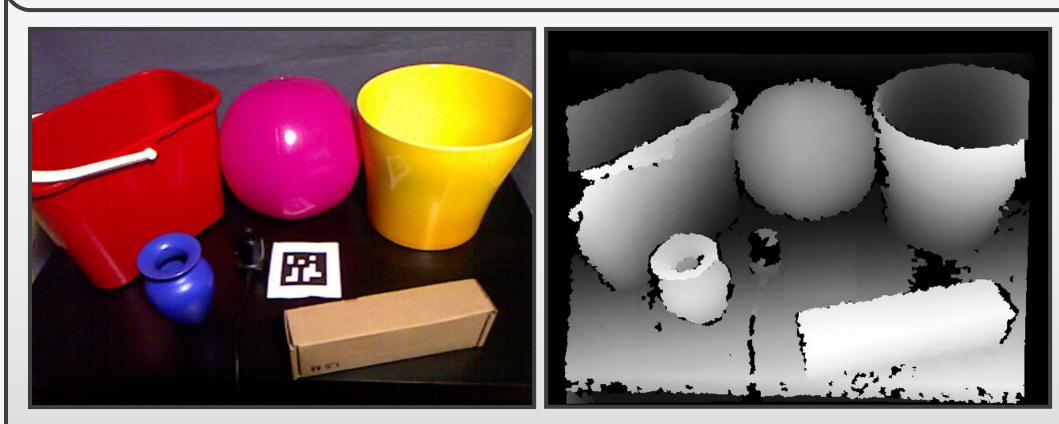


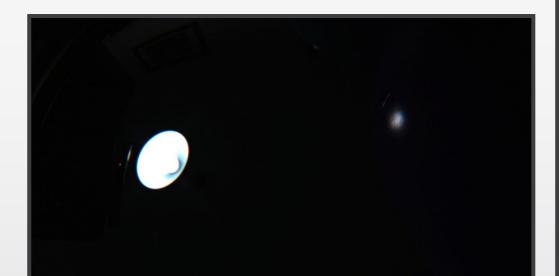
Algorithm

Based on the Phong reflection model as stated below, the goal is to estimate the diffuse (k_{d}) and specular (k_{s}, n_{s}) characteristics of all the objects in the scene:

 $I = k_a I_a + \sum_{l=1}^{n} I_l [k_d (N \cdot L_l) + k_s (V \cdot R_l)^{ns}]$

Data Acquisition





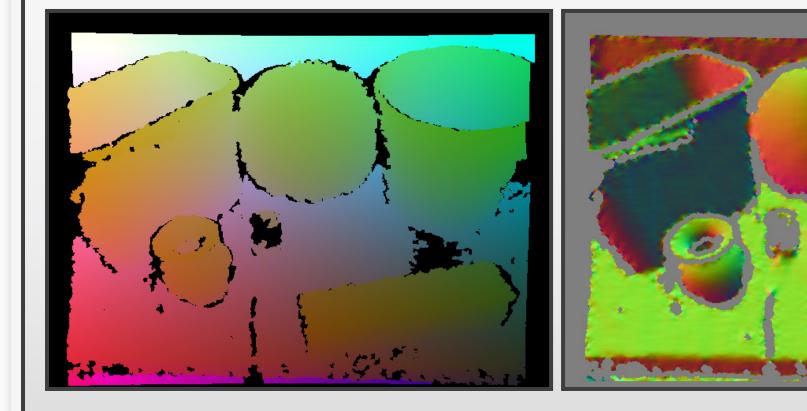
Using the Microsoft Kinect sensor color (left) and depth (right) data is acquired.



The environment is observed using a fish-eye lens camera.



Data Extraction



The depth data is used to calculate 3D point locations (left) and point normals (right).

Importance sampling of the environment map is used to place point lights on a hemisphere.

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Specular and Diffuse Estimation





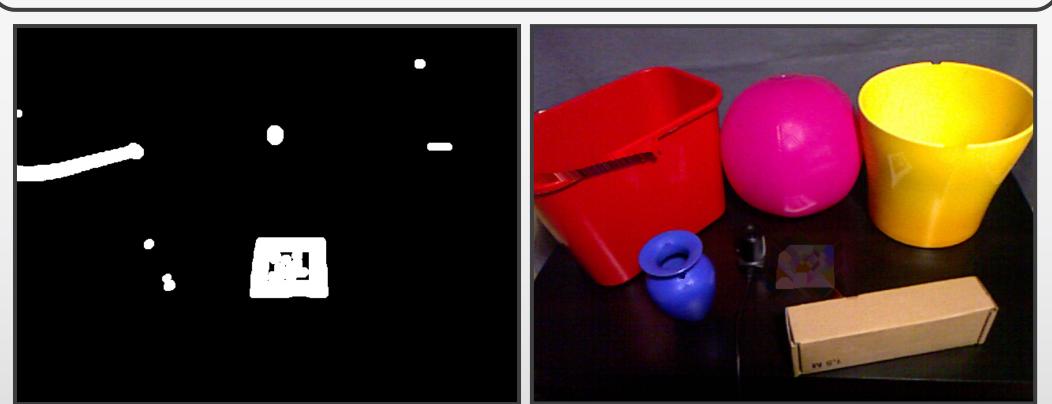
Specular estimation using nonlinear least squares per cluster.

 $Fj = \sum \left[I_{i \, KINECT} - I_{i \, RENDERING} \right]^2$

Diffuse estimation per pixel.

 $k_{d} = \frac{(I - k_{a}I_{a} - \sum_{l=1}^{n} I_{l}k_{s}(V \cdot R_{l})^{ns}}{\sum_{l=1}^{n} I_{l}(N \cdot L_{l})}$

Highlight Removal



Highlight pixels are detected as those having high brightness and low saturation values yielding a highlight mask (left). This mask is used to apply a simple inpainting technique to obtain a specular free image (right).

Inverse Diffuse Shading



Inverse diffuse shading is used to flatten objects yielding a more uniform color distribution.

Clustering



Similar colors are assumed to belong to similar materials and are therefore clustered together in order to estimate common specular reflection characteristics.

Clustering is done using a novel

GPU K-Means implementation. In contrast to almost all other implementations of K-Means our method runs entirely on the GPU.

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