

Shape from Varying Illumination and Viewpoint

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Goal



- Given views of an object rotating in front of a light and a camera, where the lighting and object rotation is unknown, recover the structure

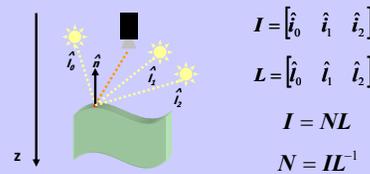
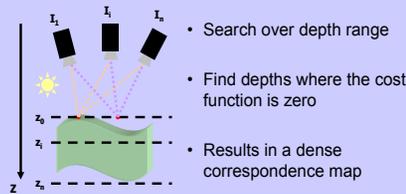
Contributions

- We recover shape under varying viewpoint and lighting by combining multi-view and photometric stereo
- We derive a multi-ocular photometric matching cost. By minimizing this using a graphcut, we find correspondence
- Once correspondence is found, photometric stereo recovers normals, and the depth map from correspondence and the normals are fused to recover high quality 3-D shape
- Relative to previous related work, we produce higher-quality reconstructions using a non-iterative method with fewer images

Summary

- Tomasi-Kanade Factorization to get projection matrices for each image
- Search over depths for each (x,y) point
 - Project point from the reference view to every other image
 - Get image patch centered around each projected point
 - Minimize Rank-3 approximation error using a graphcut
 - Perform photometric stereo and use the normals and the depth map to recover the final surface
- Key idea:
 - Rank-3 error is higher at the incorrect depth due to patches having inconsistent shading and albedo

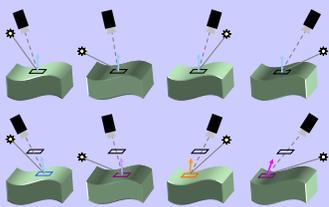
Combining Multi-view Stereo and Photometric Stereo



Multi-view and Photometric Stereo

Multi-view Stereo

Photometric Stereo

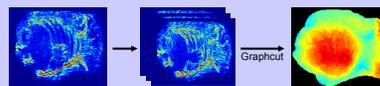


Rank-3 Approximation for pixel neighborhood:

$$I = [\hat{l}_0 \ \dots \ \hat{l}_n] \quad [U \ S \ V] = \text{svd}(I)$$

$$U' = U(:,1:3) \quad V' = V(1:3,:)$$

$$S' = S(1:3,1:3) \quad I' = U' S' V'^T$$

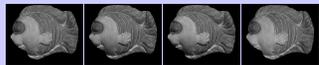


Rank-3 Approximation error for the pixel neighborhood:

$$e = \|I - I'\|^2$$

Compute for all pixels over a range of depths and find the depth map by minimizing the cost using a graphcut

Computing the Depth Map using a Photometric Cost Function



- Use depth map to align input images
- Perform Un-calibrated Photometric stereo:

$$\tilde{N} = \text{Pseudo-Normals} \quad \tilde{L} = \text{Pseudo-Lighting}$$

- Compute linear transform A such that:

$$\frac{\tilde{n}}{|\tilde{n}|} A = \nabla(\hat{z}) \quad \tilde{N} A = N$$

Recovering Normals

Error Function Minimized to Solve for the Final Surface:

$$J(S) = E^P + E^N + E^S$$

Position Error:

$$E^P = \lambda_1 \sum_{k=1}^m [S_k - z_k]^2$$

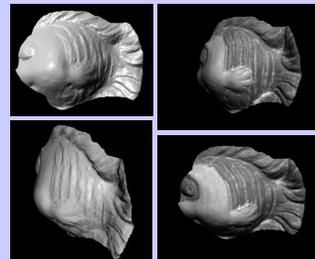
Normal Error:

$$E^N = (1 - \lambda_1) [\sum_{k=1}^m [T_k^x \cdot N_k^C]^2 + \sum_{k=1}^m [T_k^y \cdot N_k^C]^2]$$

Smoothness Constraint:

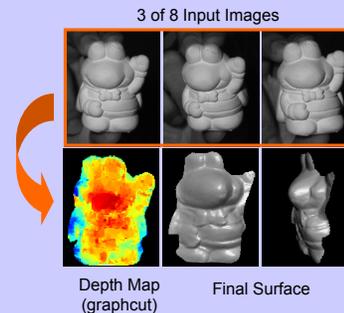
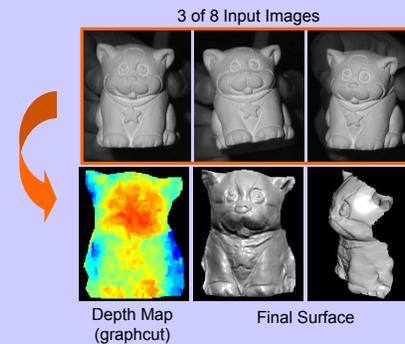
$$E^S = \lambda_2 \sum_{k=1}^m [(\nabla^2 S)(\sigma_k)]$$

Linear Least Squares

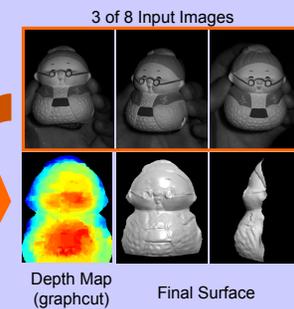
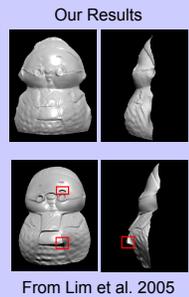


Solving for the Final Surface

Results



Comparison



Selected References

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