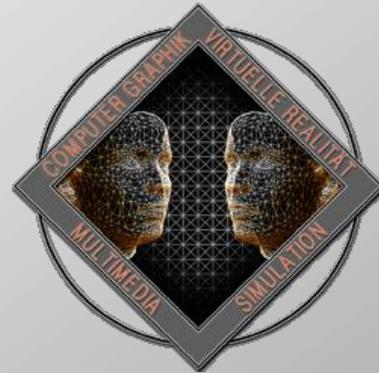


# Multi-View 3D Reconstruction of Highly-Specular Objects

Master Thesis

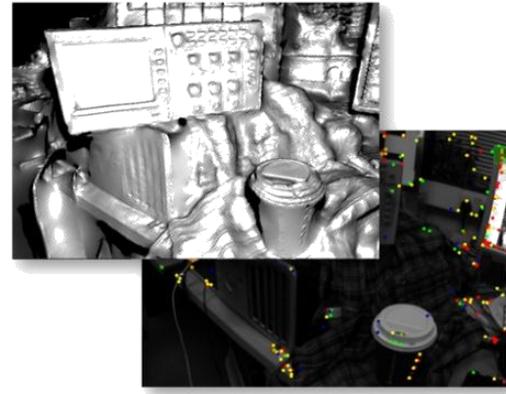


Author: Aljoša Ošep

Mentor: Michael Weinmann

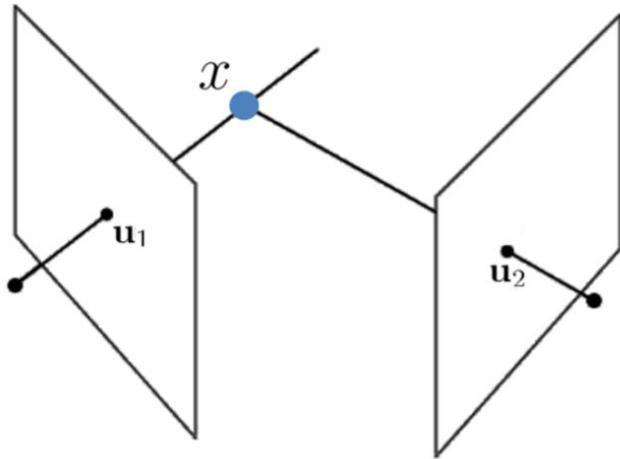
# Motivation

- **Goal:** faithful reconstruction of full 3D shape of an object
- **Current techniques:**



# Motivation

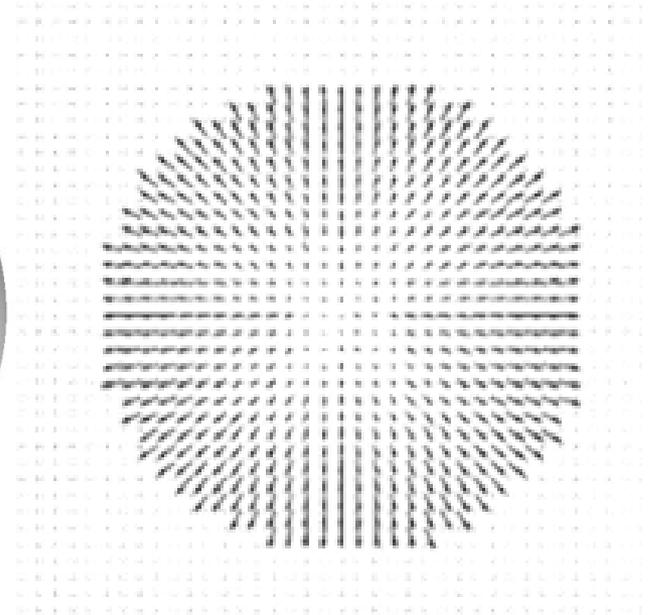
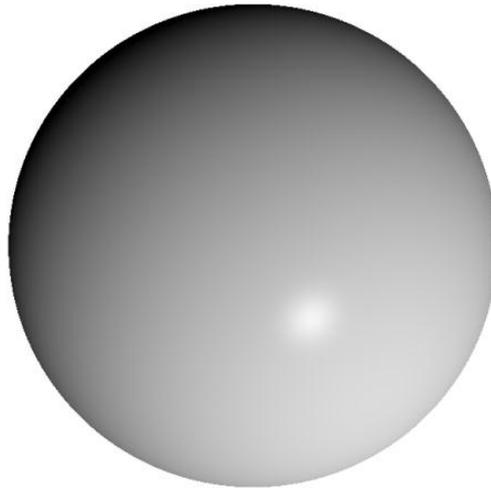
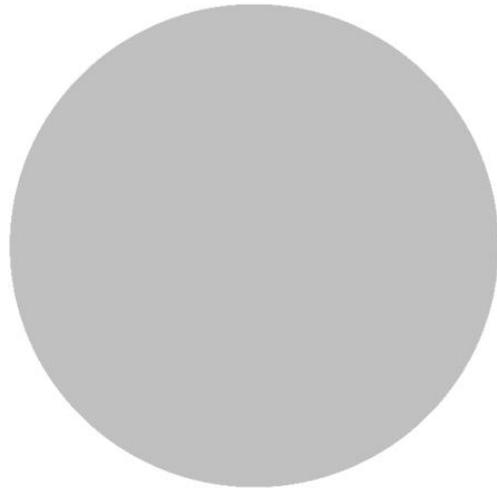
- **Challenge:** objects exhibiting a complex reflectance behavior
- **Focus:** on opaque+specular materials



# Introduction

---

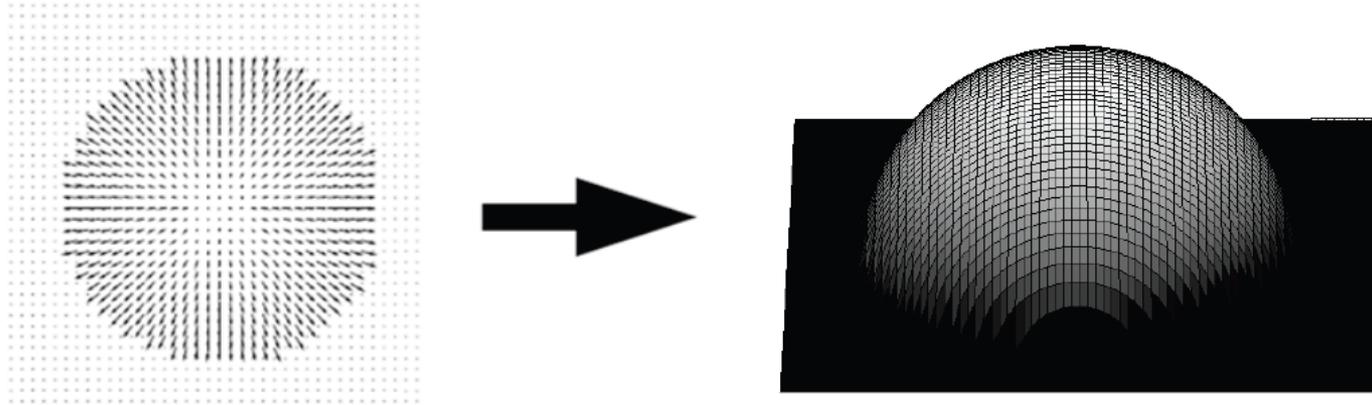
- **Observation:** shading is a powerful visual cue
- Provides surface orientation information



# Introduction

---

- Many techniques for surface normal field estimation using shading cues from single view



- How can information from several viewpoints be combined?
- Could 3D reconstruction of specular objects be addressed this way?



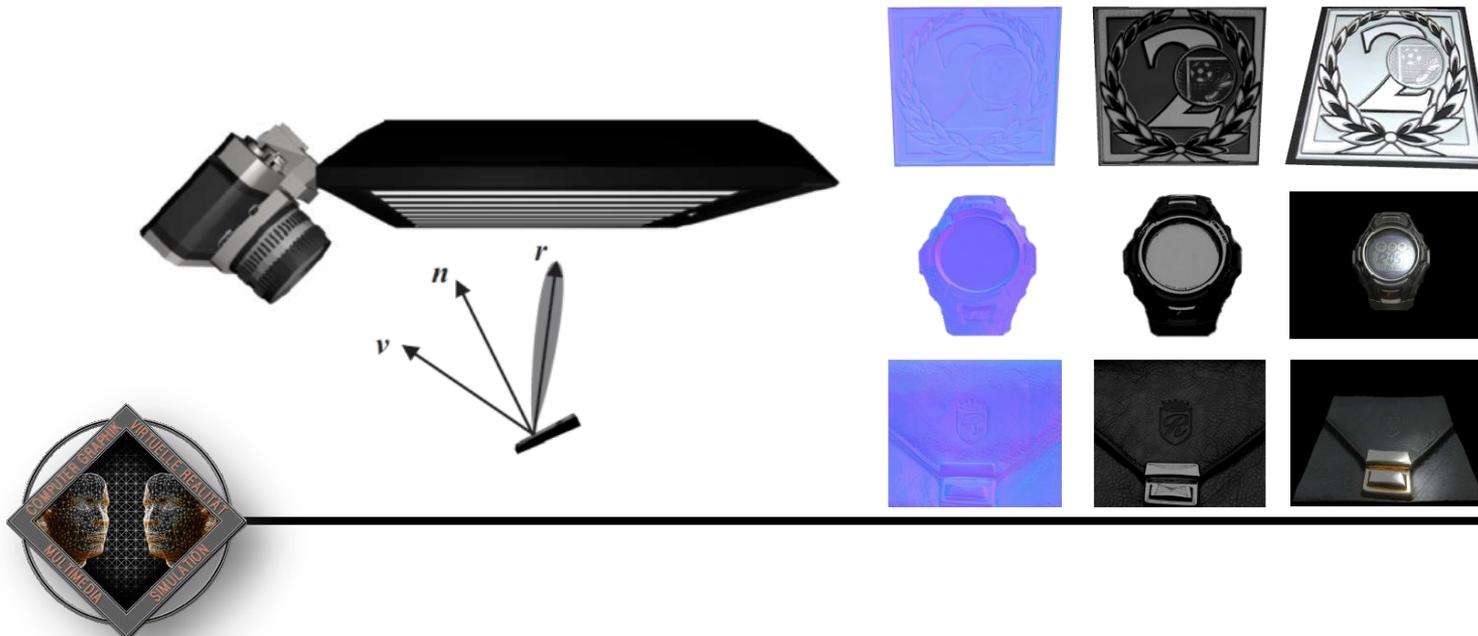
Image credits: N. Funk and Y.-H. Yang. Using a Raster Display Device for Photometric Stereo.

# Background

- Mesostructure from Specularity (Chen et al., CVPR '06)



- Gloss and Normal Map Acquisition of Mesostructures Using Gray Codes (Francken et al., ISVC '09)



# Background

- Specularity-Consistency based methods
  - Voxel Carving for Specular Surfaces (Bonfort et al., ICCV '03)
  - Dense 3D Reconstruction from Specularity Consistency (Nehab et al., CVPR '08)

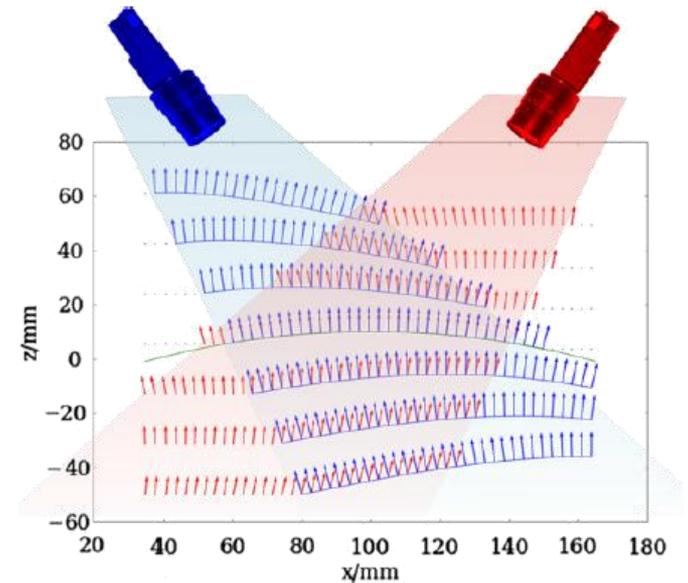
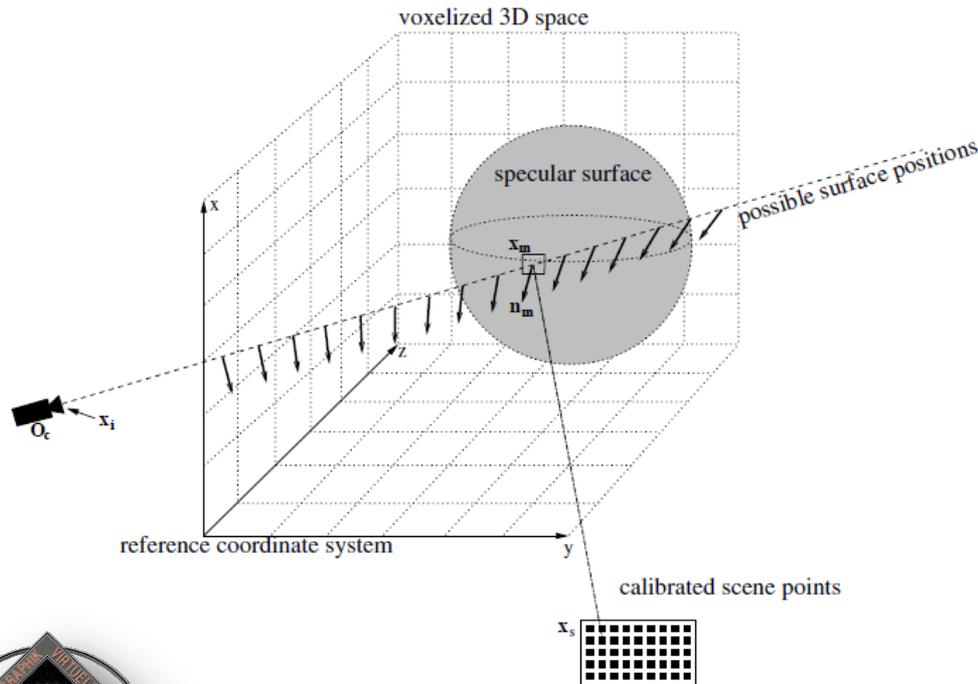


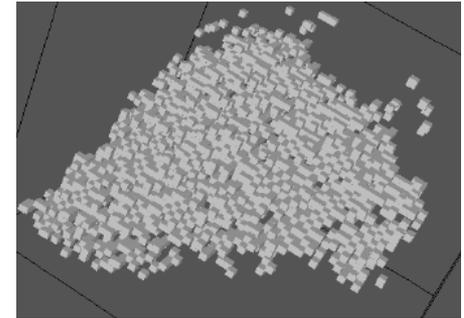
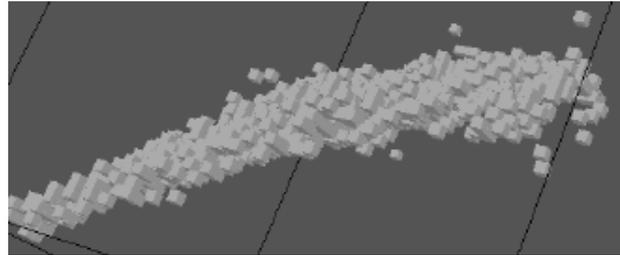
Image credits: T. Bonfort and P. Sturm. Voxel carving for specular surfaces (*left*), J. Balzer and S. Werling. Principles of Shape from Specular Reflection (*right*)

# Background

## ■ Results



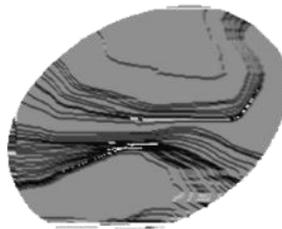
stereo rig  
printed target  
camera  
specular spoon



Real Quarter Dataset



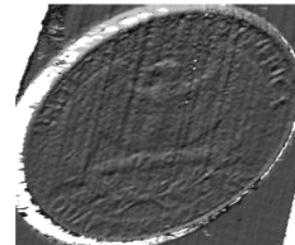
(a) Captured normals



(b) Captured positions



(c) Refined positions

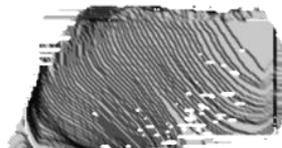


(d) Traditional scanner

Real Mirrored Sphere Dataset



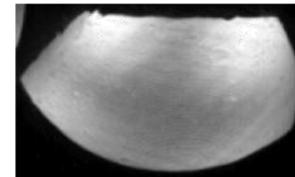
(e) Captured normals



(f) Captured positions



(g) Refined positions



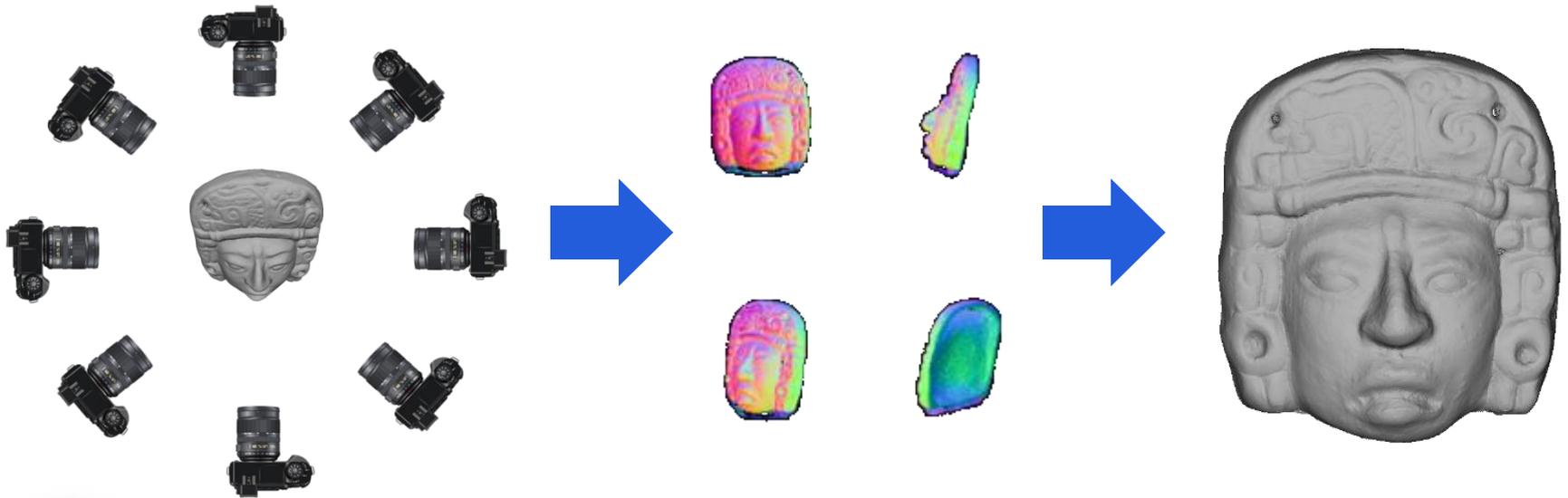
(h) Photograph



# Our Approach

---

- Estimate multi-view normal fields using structured environment
- Multi-view normal field integration problem
  - Need very robust algorithm!

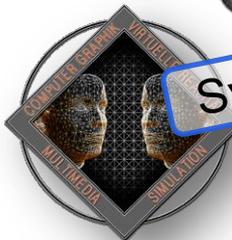
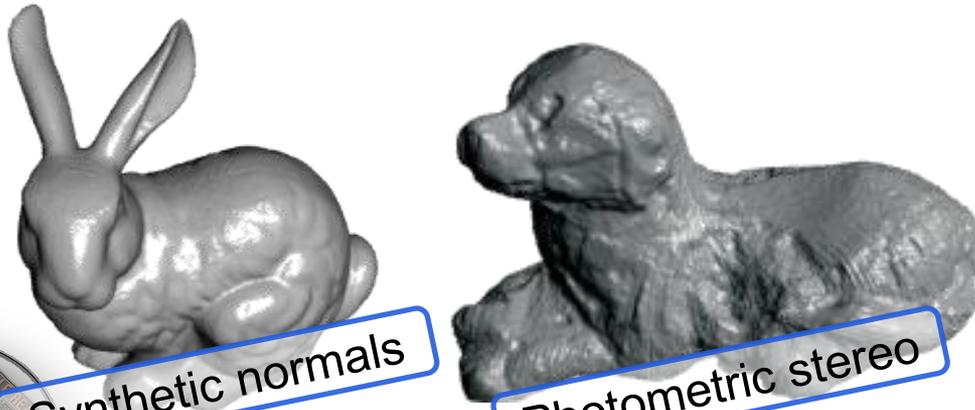


# Multi-View Normal Field Integration

- Chang et al., CVPR'07
  - Level sets



- Master thesis of Z. Dai
  - MRF approach



# Overview

---

- I. Multi-View Normal Field Integration
- II. Multi-View Shape-from-Specularity
- III. Evaluation
- IV. Conclusion and Future Work



---

I. Multi-View Normal Field Integration

II. Multi-View Shape-from-Specularity

III. Evaluation

IV. Conclusion and Future Work



# Problem Statement

## ■ Given:

- $\kappa_C$  calibrated cameras  $C_i$
- Projection matrices  $P_i = K_i [R_i | t_i]$
- Normal fields  $\mathcal{N}_i$

$$i = 1 \dots \kappa_C$$

## ■ Goal:

- Reconstruction of surface  $\partial S$

## ■ Problem:

- Inferring coordinates of all surface point given normal fields estimates



# Challenges

---

- Noise
- Outliers
- Systematic errors
- Holes
- Initial guess (visual hull) in practice difficult to compute
- Implementation concerns
  - Fine-detail reconstruction



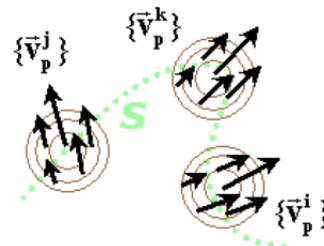
# Variational Approach

- Solve variational problem:

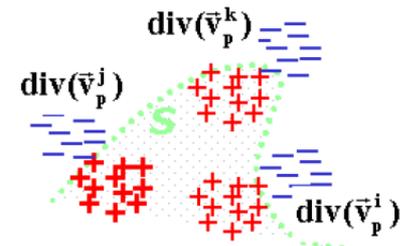
$$E(\partial S) = \lambda_1 \underbrace{\int_{\partial S} dA}_{E_1} - \lambda_2 \underbrace{\int_{\partial S} \langle c\mathbf{N}, \mathbf{n} \rangle dA}_{E_2},$$

- $\mathbf{N}(\mathbf{x})$  ... vector field, reconstructed from normal fields
  - $c(\mathbf{x})$  ... surface consistency
- Solving minimal surface problems

- Active contour
- Level sets
- Graph cuts
- **Convex relaxation**



(a) Flux



(b) Divergence

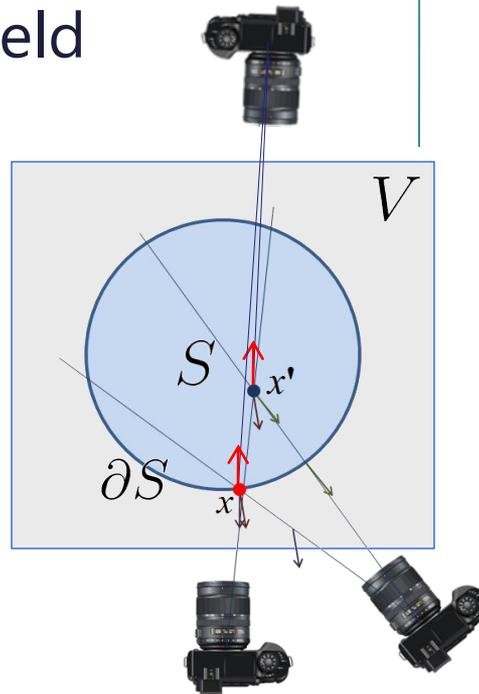


Image credits: V. Lempitsky and Y. Boykov: Global Optimization for Shape Fitting

# Vector Field Computation: Idea

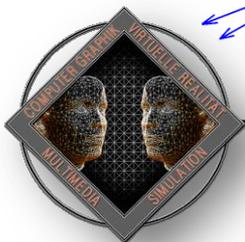
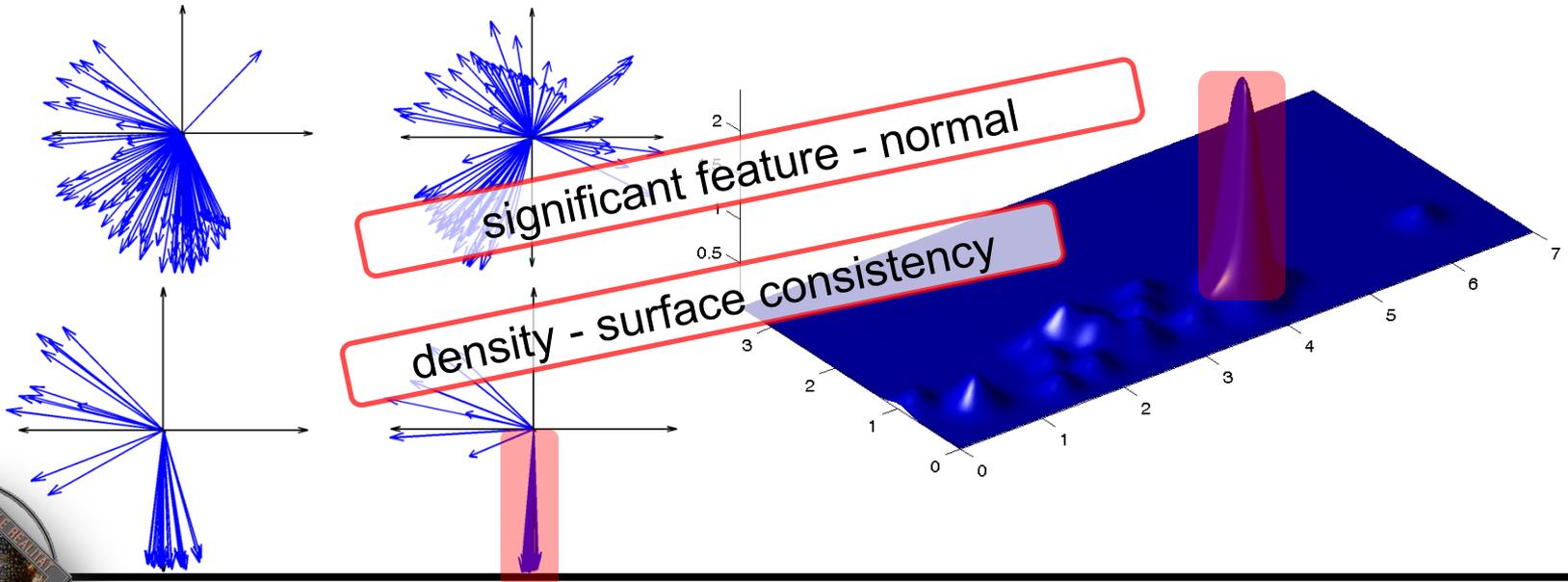
- Core of our approach
- Key questions:
  - Surface consistency measure  
 $c(\mathbf{x}) : \mathbf{x} \mapsto \mathbb{R}$
  - Value of vector field  
 $\mathbf{N}(\mathbf{x}) : \mathbf{x} \mapsto \mathbb{S}^2$

$$\mathbf{x} \in V \subset \mathbb{R}^3$$



# Vector Field Computation

- Back-project normal fields into volume  $V$
- Map back-projected normals to feature space
- Density estimation to find the patterns - normal
  - Based on discrete, back-projected normal samples
  - Non-parametric method essential



# Feature Space Analysis

- Histogram method
- Kernel density estimation
- **Mean-Shift clustering**

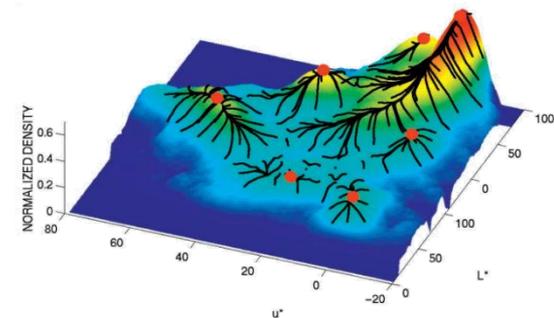
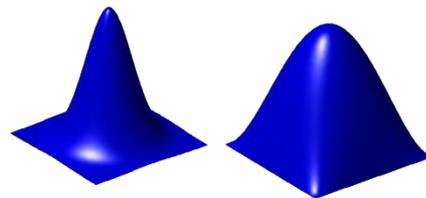
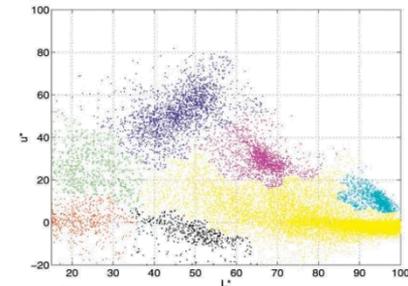
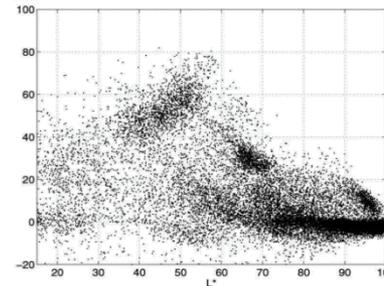
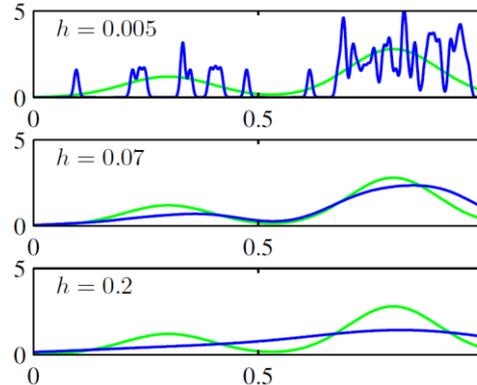
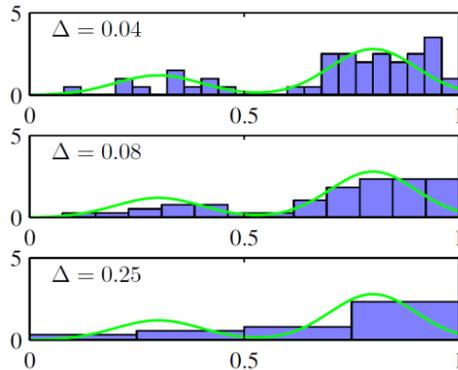
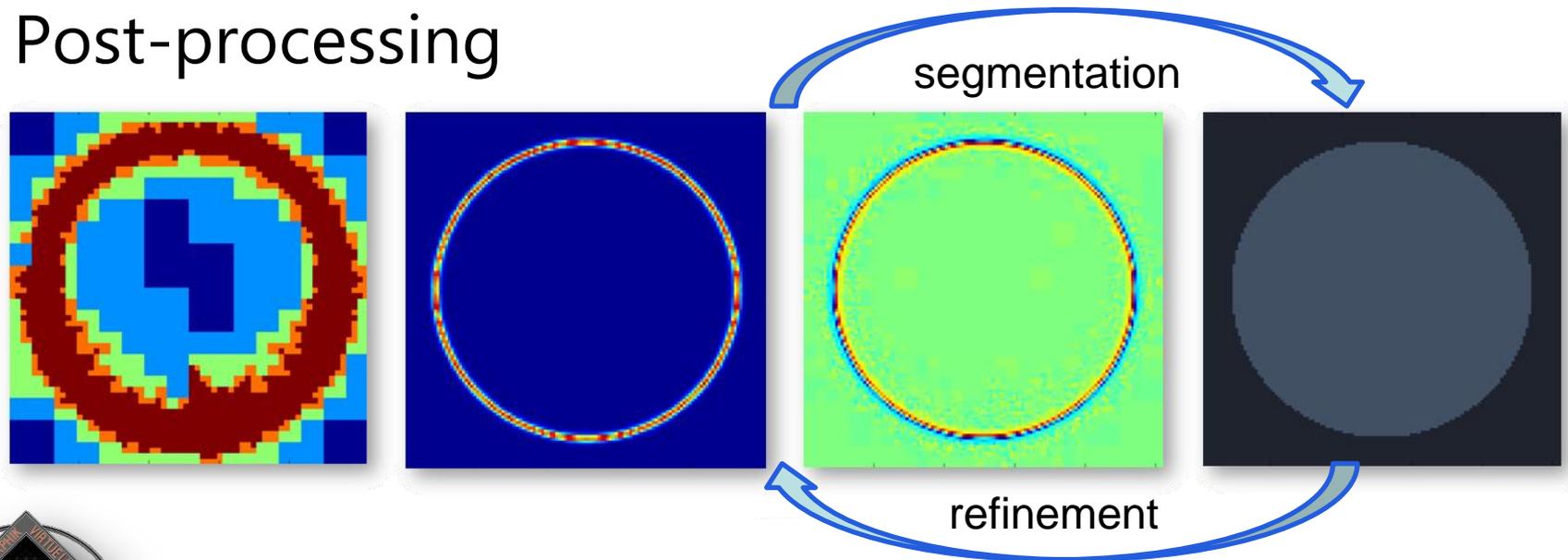


Image credits: Christopher M. Bishop. Pattern Recognition and Machine Learning (Information Science and Statistics) (left), D. Comaniciu and P. Meer. Mean shift: A robust approach toward feature space analysis (right),

# Implementation

- Octree-based discretization of bounding volume  $V$
- Initial refinement strategy
- Continuous Max-Flow based volume segmentation
- Iterative scheme
- Post-processing



---

I. Multi-View Normal Field Integration

**II. Multi-View Shape-from-Specularity**

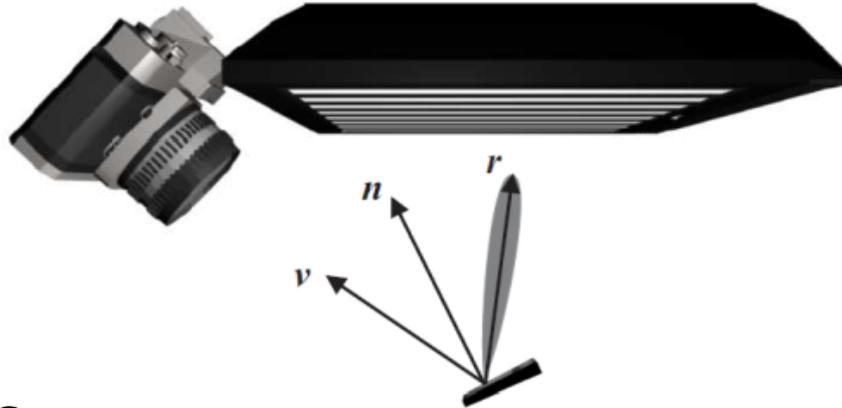
III. Evaluation

IV. Conclusion and Future Work



# Multi-View Shape-from-Specularity

- How to compute normal fields of specular objects?



- Challenges

- How to lit object fully?
- Distant light assumption violation
- How to reliably decode patterns?
- Screen calibration

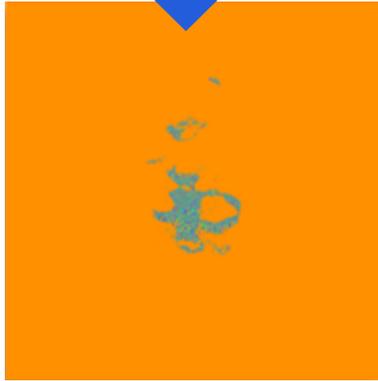
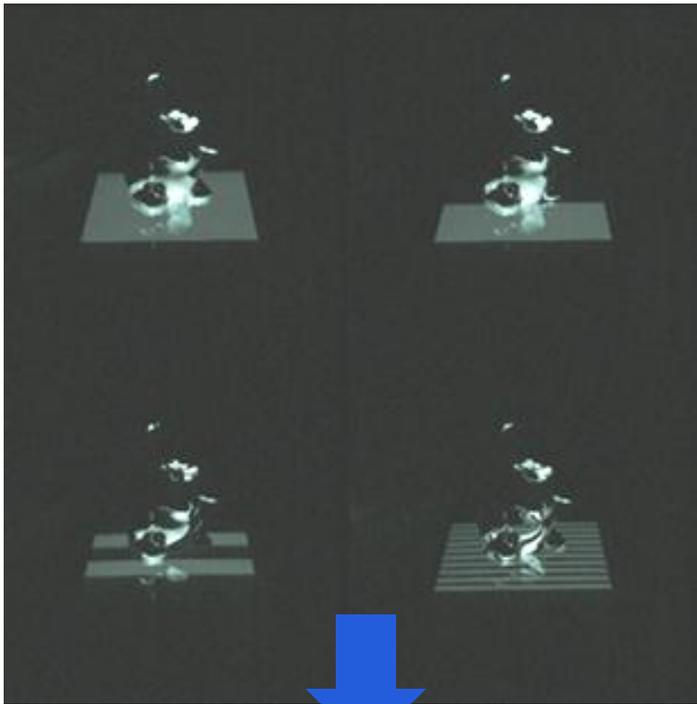
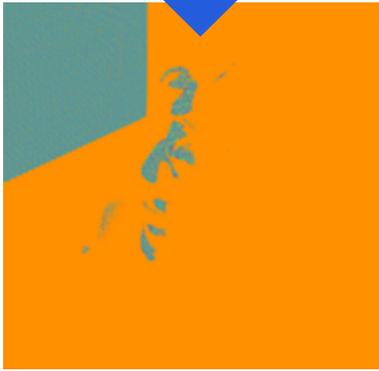
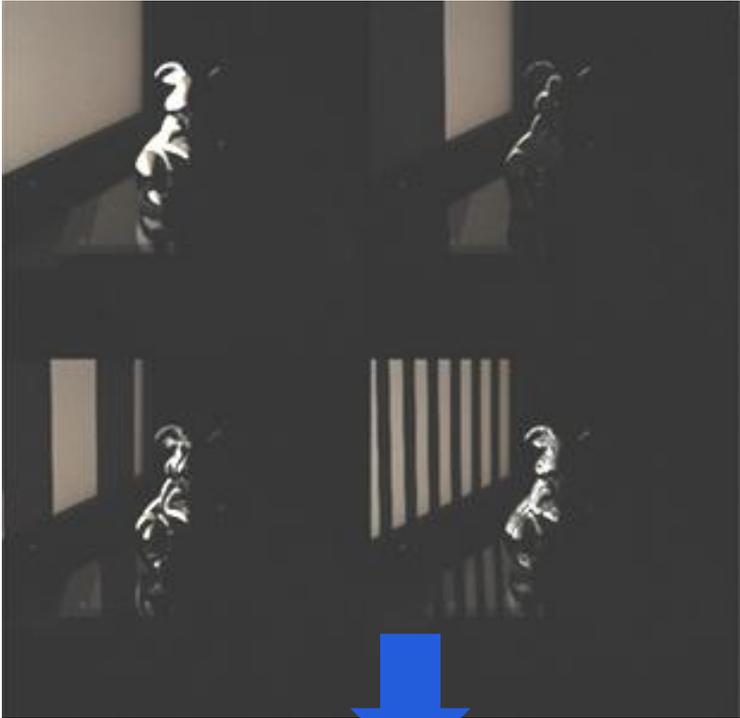


Image credits: Y. Francken, T. Cuypers, T. Mertens, and P. Bekaert: Gloss and normal map acquisition of mesostructures using gray codes.

# Proposed Setup

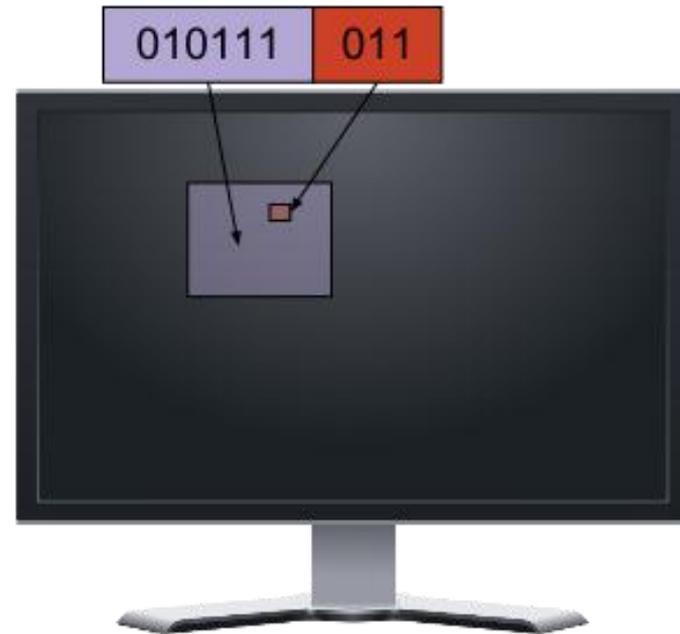


# Capturing the Data



# Computing Light Maps

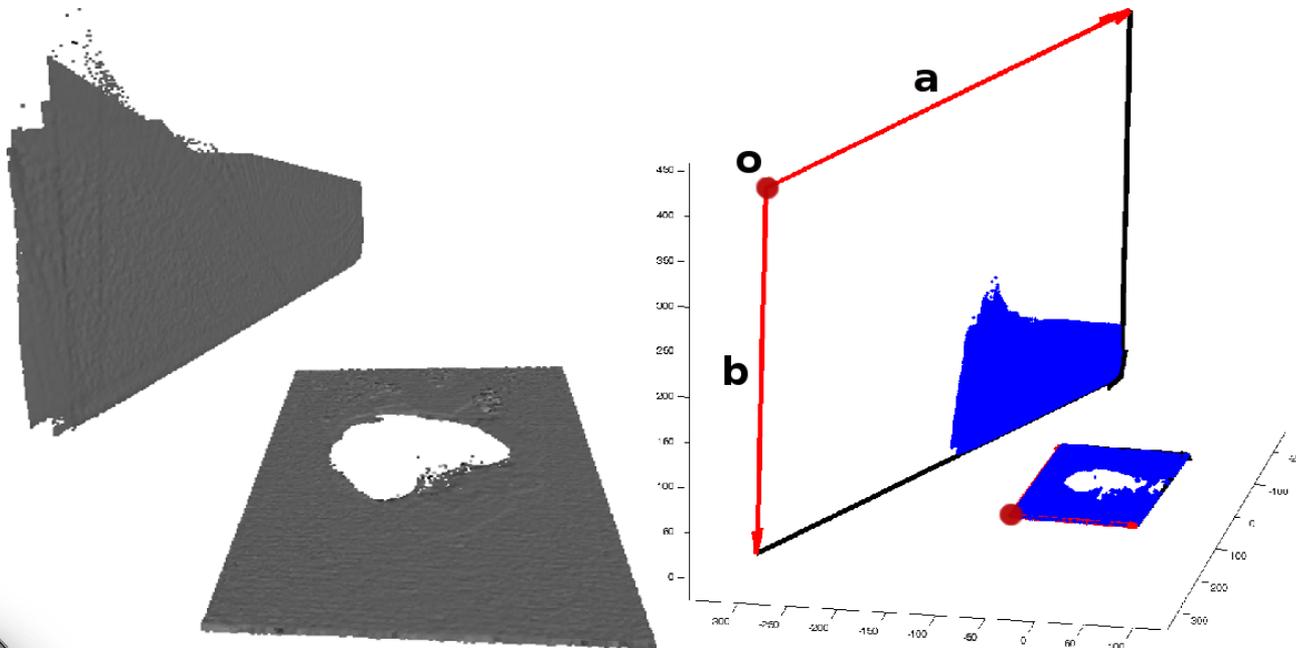
- Fuzzy decoding



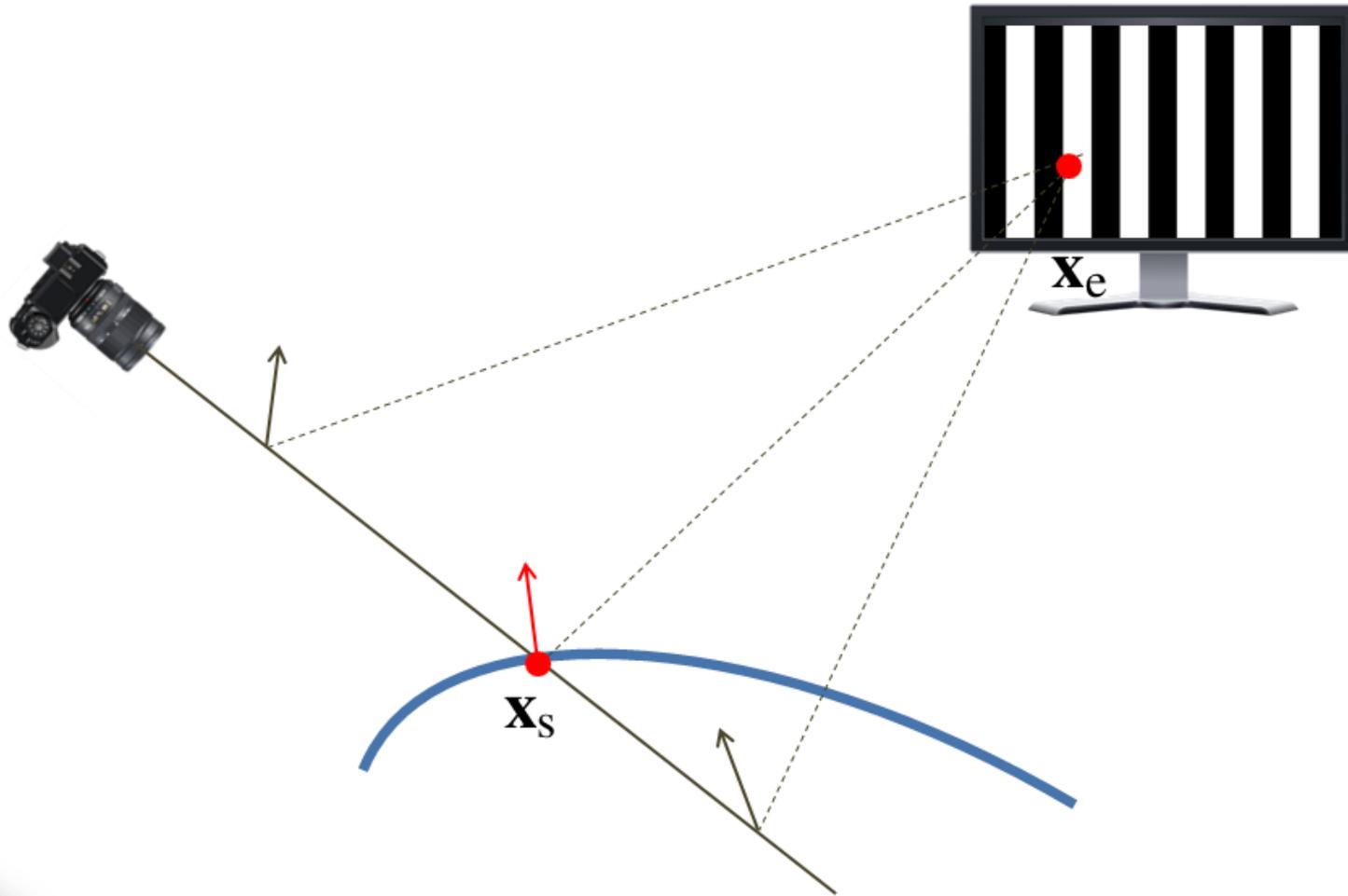
# Screen Calibration

- Structured pattern based triangulation

$$Q = \sum_{m=1}^M (\mathbf{p}_m - (\mathbf{o} + q_{x,m}\mathbf{a} + q_{y,m}\mathbf{b}))^2$$



# Normal-Depth Ambiguity



# Reconstruction

---

- Input are light-maps and not normal fields
- Project labels to octree corners and compute normal hypotheses
- Reconstruct vector field and compute surface consistency
- Fit surface to vector field



---

I. Multi-View Normal Field Integration

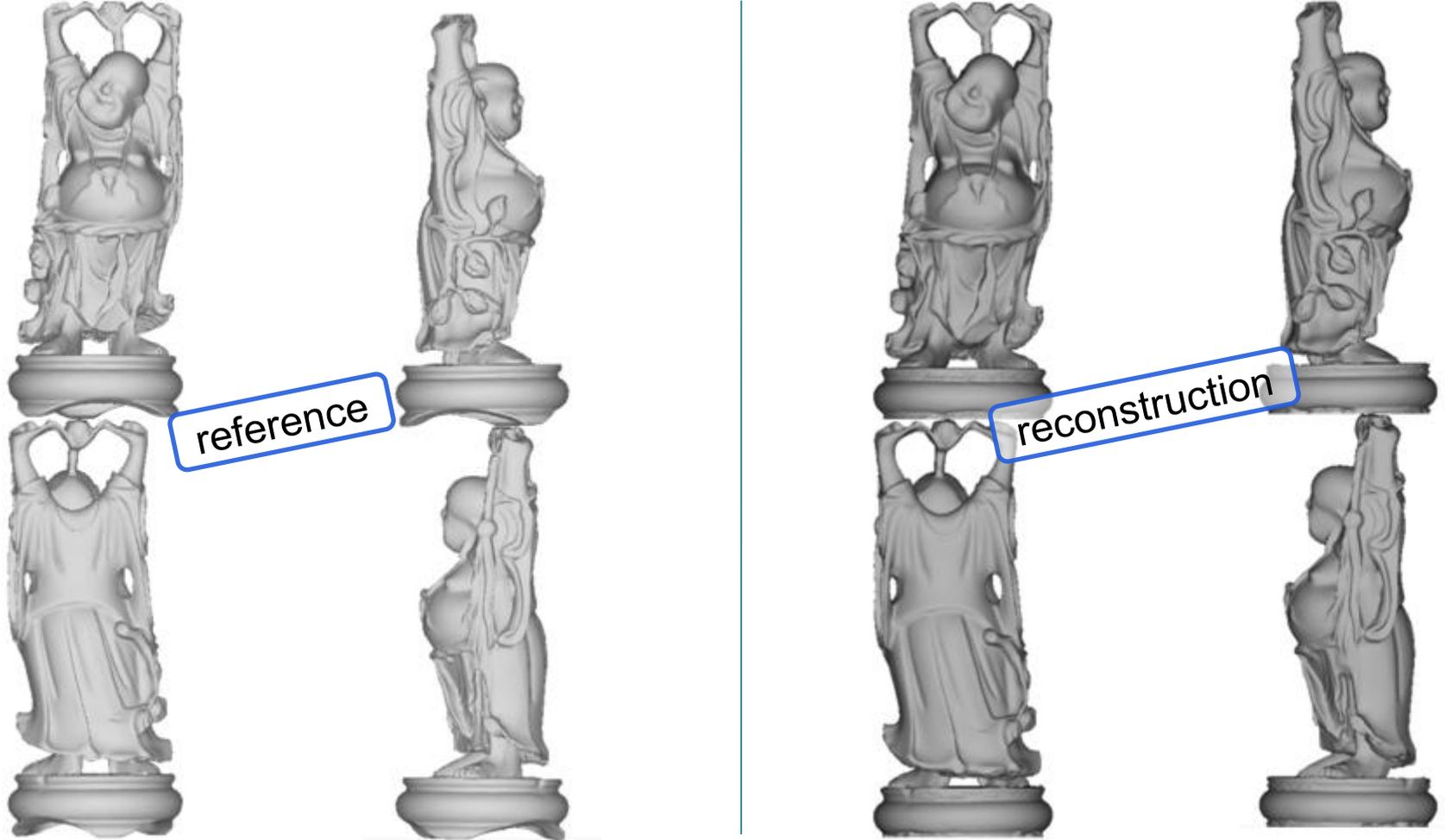
II. Multi-View Shape-from-Specularity

**III. Evaluation**

IV. Conclusion and Future Work



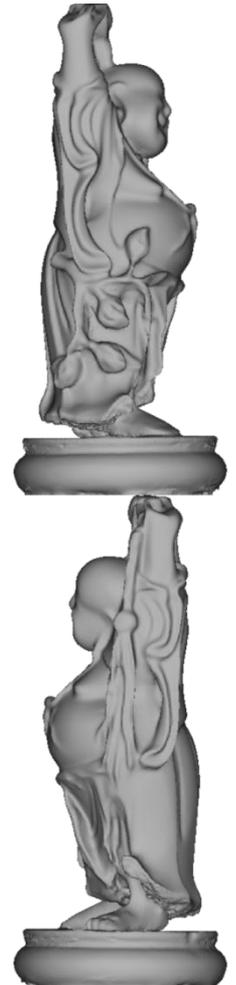
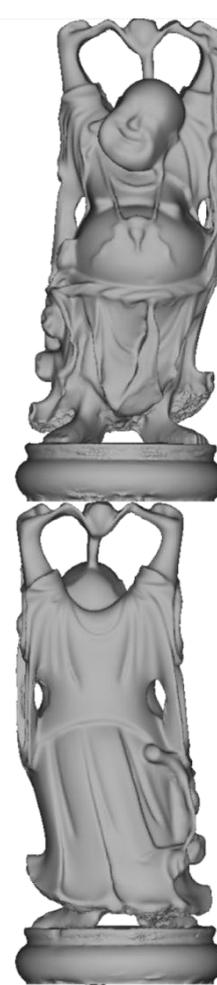
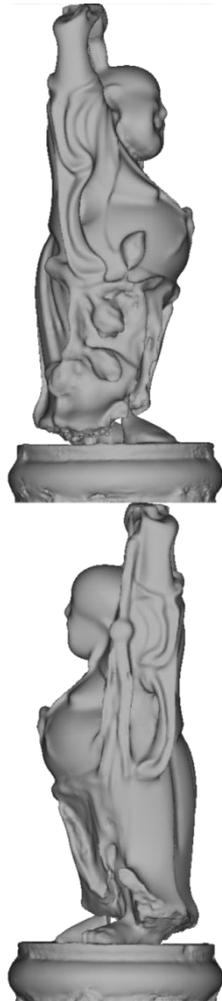
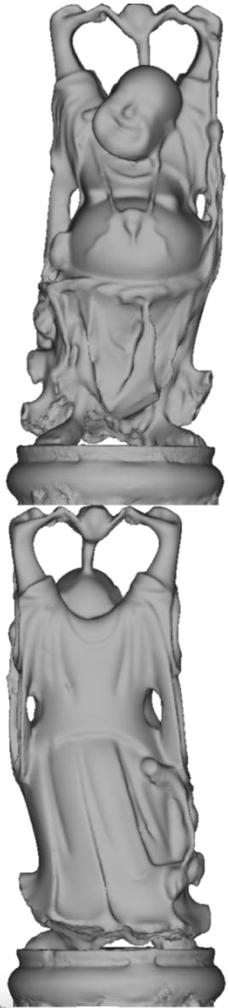
# Synthetic Normal Fields



Happy Buddha, 75 cameras



# Synthetic Normal Fields

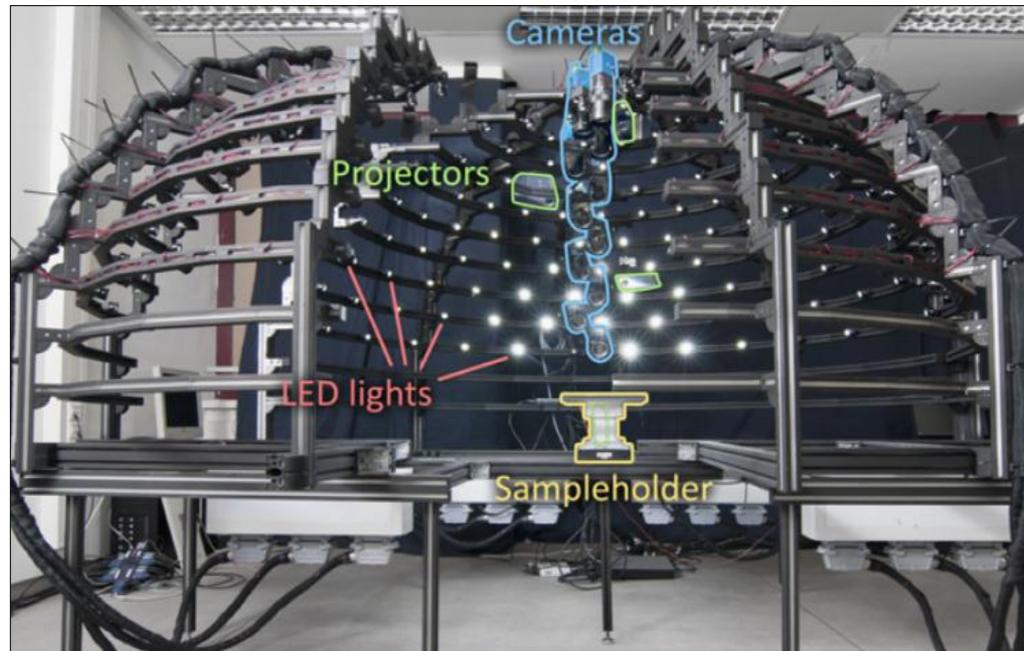


Happy Buddha, 10 (*left*) and 20 (*right*) cameras



# Real Data: Photometric Stereo

- Classic least-squares photometric stereo
  - Simple thresholding prior to fitting
- 6 cameras, 12 rotations, 72 views
- 198 images for computation of single normal field



# Real Data: Photometric Stereo

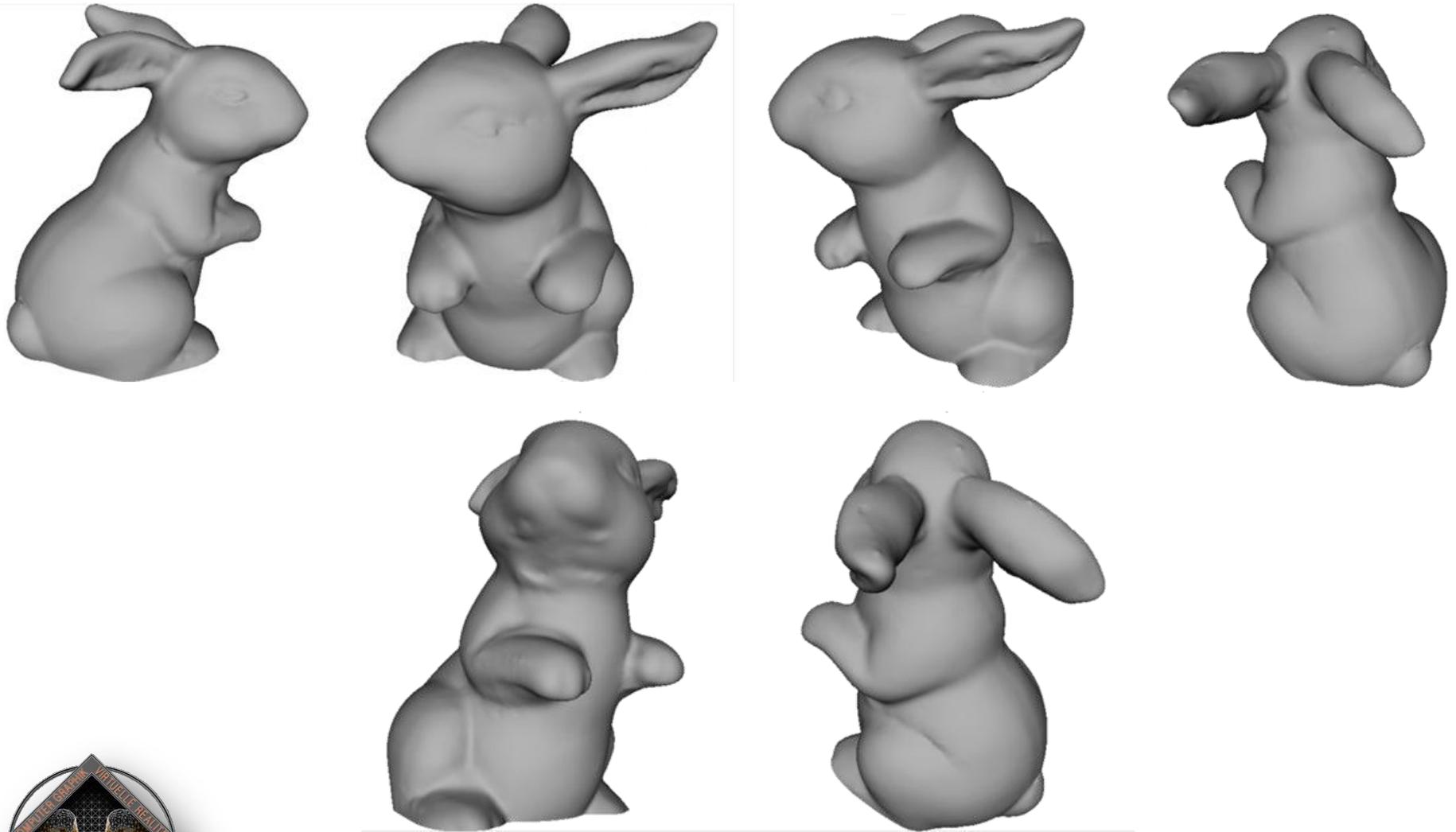


# Real Data: Shape-from-Specularity

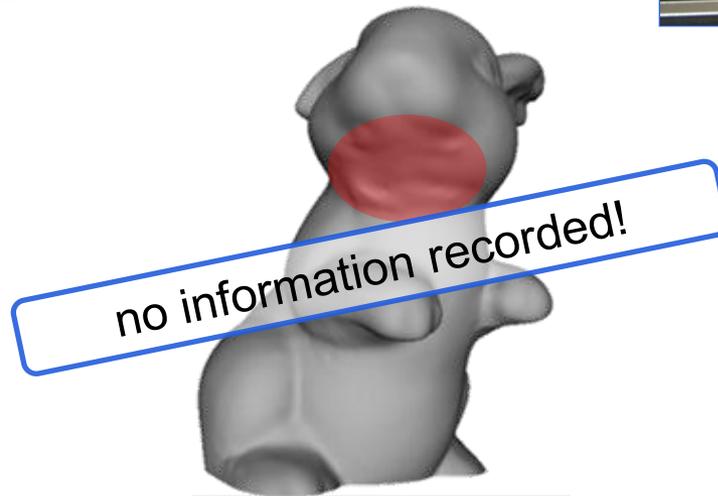
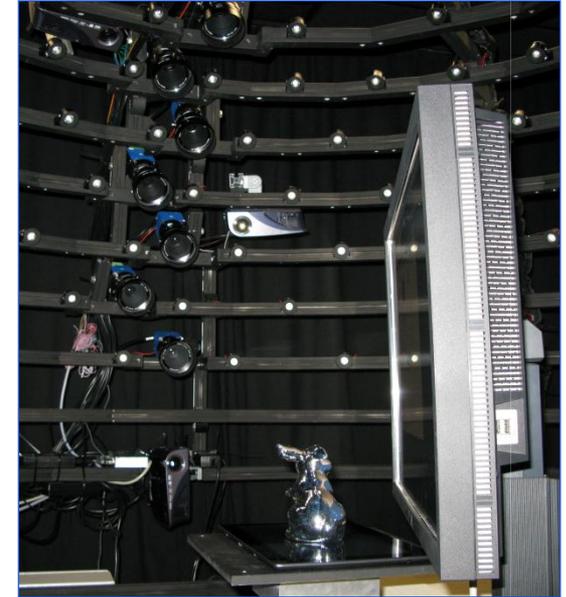
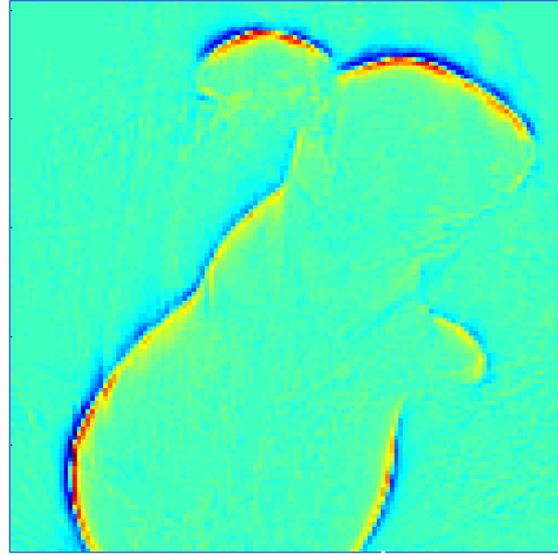
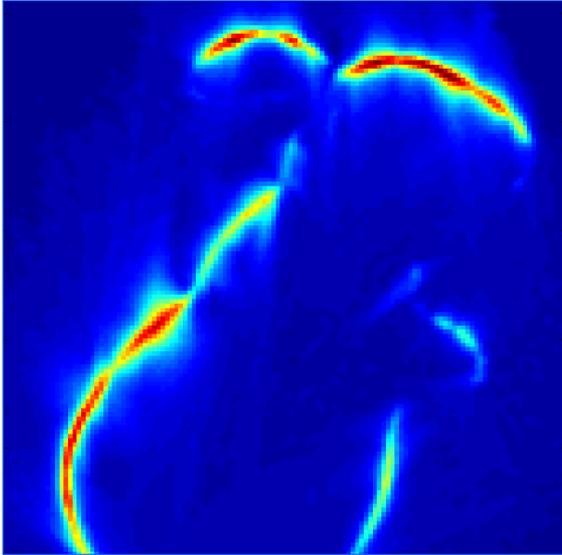
- 10 cameras, 24 rotations, 240 views, two sources of structured illumination
- Mirror bunny



# Real Data: Shape-from-Specularity

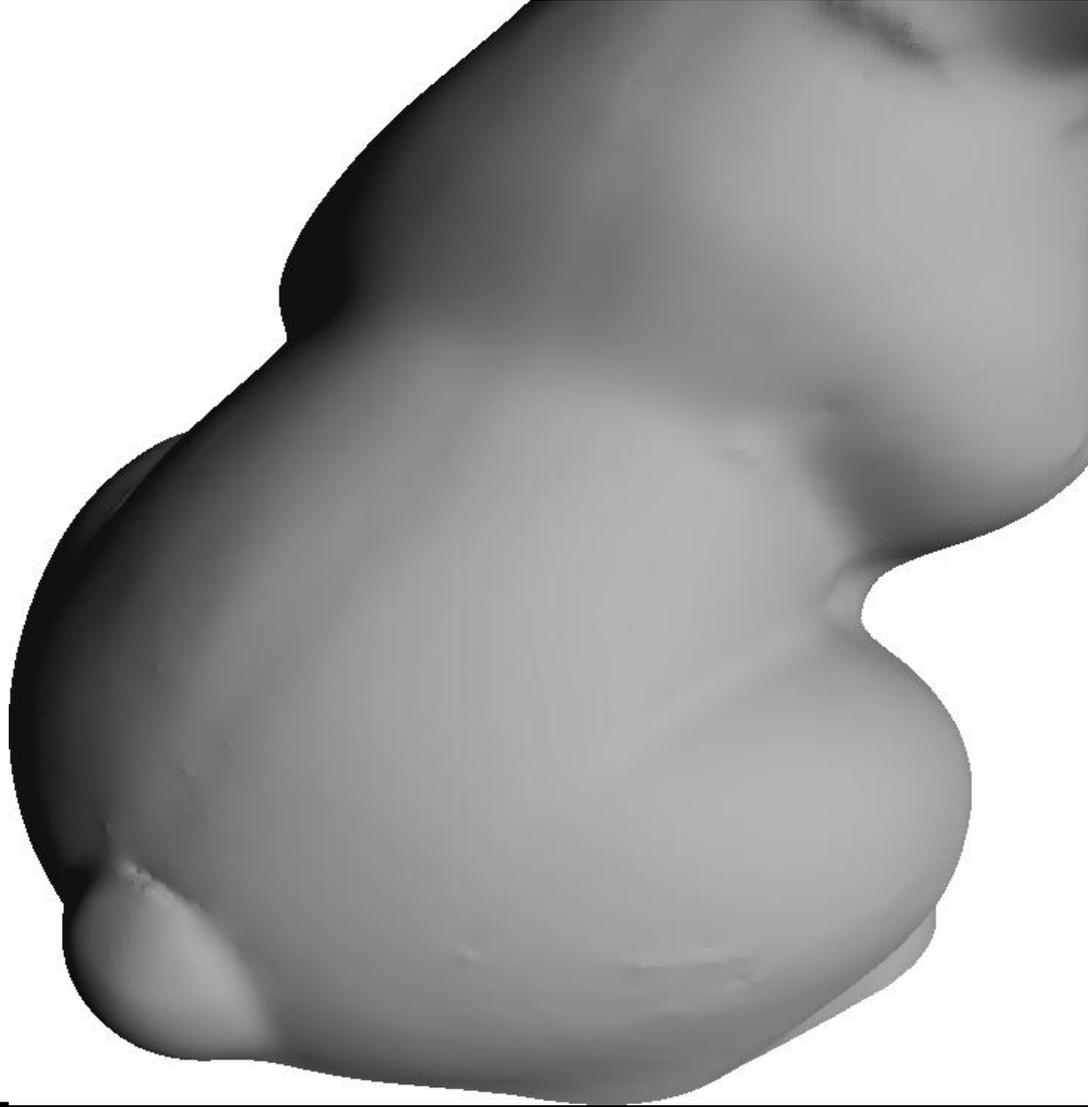


# Real Data: Shape-from-Specularity



# Real Data: Shape-from-Specularity

---



---

I. Multi-View Normal Field Integration

II. Multi-View Shape-from-Specularity

III. Evaluation

IV. Conclusion and Future Work



# Conclusion

---

- New, robust multi-view normal field integration algorithm
  - No initial guess (visual hull) needed
  - First results, demonstrated on captured data
  - Efficient numerical techniques
- New dome-based method for reconstruction of highly specular objects
  - Display screens as sources of structured lighting
  - Normal computation and integration based approach
  - State-of-the-art results



# Future Work

---

- Testing integration algorithm using more general normal estimation techniques
- Coding the light pattern using different coding strategies
- Weight normals coming from specular surfaces according to uncertainty of source of illumination
- Parallelization potential



---

Thank you for your attention!



# Structured Light 3D Scanning

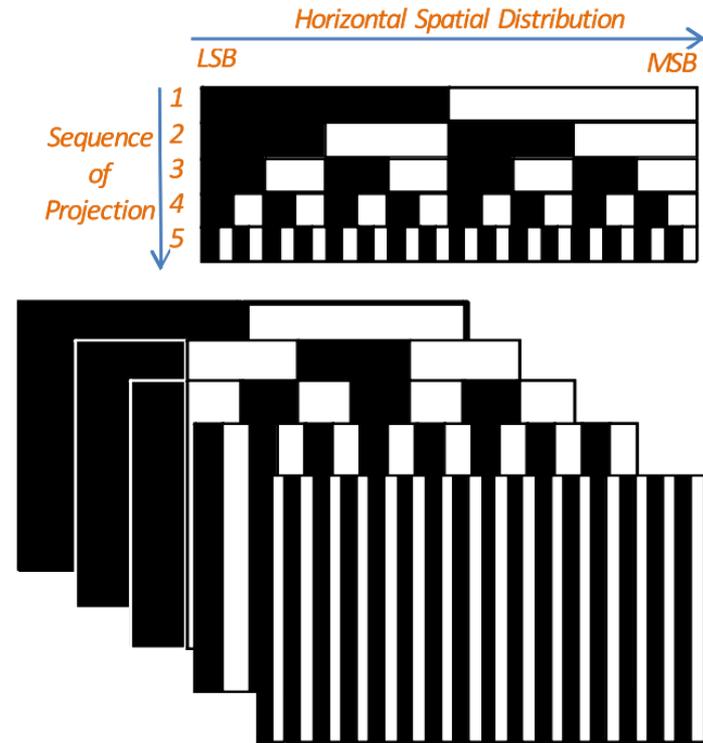
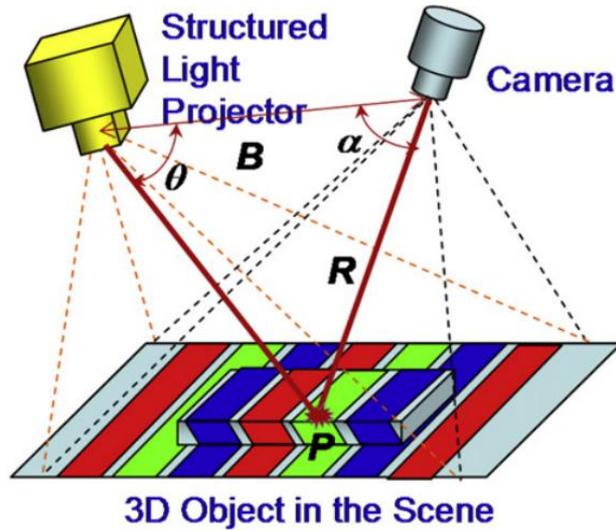


Image credits: J. Geng: Structured-Light 3D Surface Imaging: A Tutorial

# Photometric Stereo

- Lambertian assumption:

$$J(\mathbf{u}_X) = k_X \cdot (\mathbf{n}_X^T \cdot \mathbf{l})$$

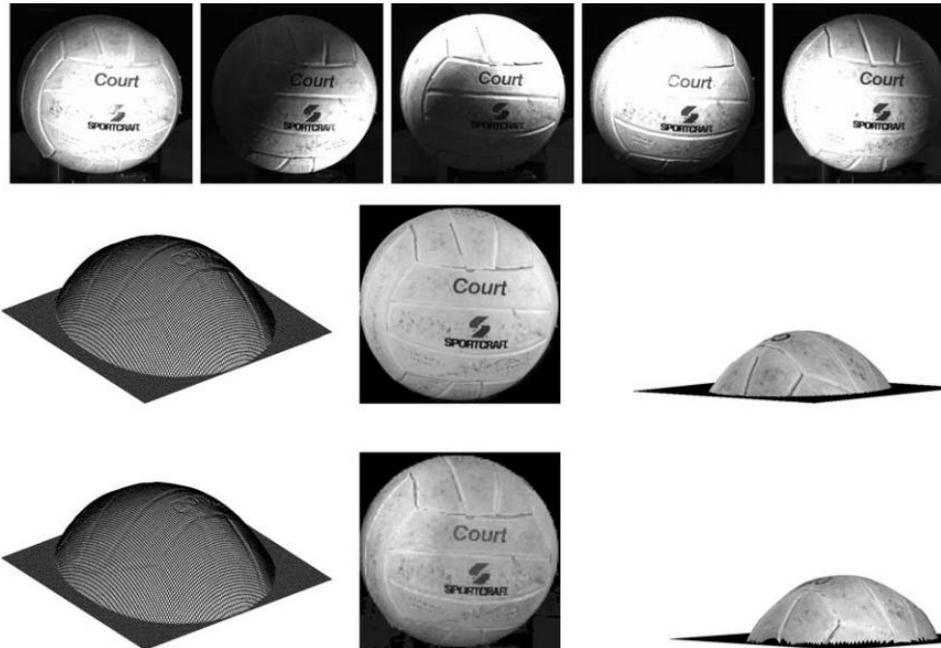
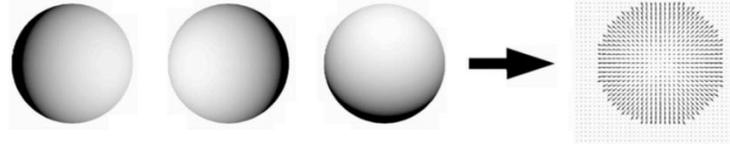


Image credits: R. Basri, D. Jacobs, I. Kemelmacher: Photometric Stereo with General, Unknown Lighting  
(bottom image)

# Convex Relaxation

---

$$E(\gamma) = \lambda_1 \int_V \|\nabla \gamma\| \, dV - \lambda_2 \int_V (\nabla \cdot (c\mathbf{N})) \gamma \, dV$$

*w.r.t*  $\gamma : V \mapsto [0, 1]$



# Reflectance Models

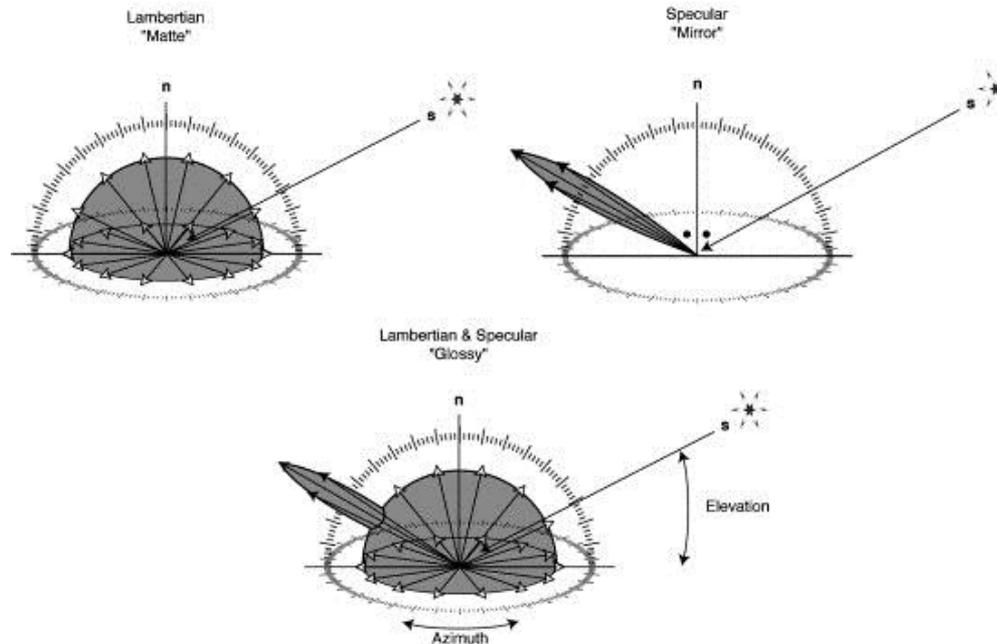
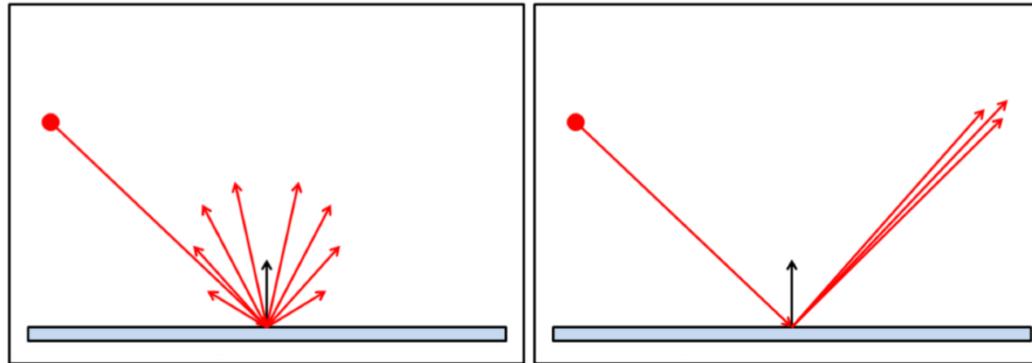


Image credits: H. T. Nefs, J. J. Koenderink, A. M.L. Kappers: Shape-from-Shading for Matte and Glossy Objects (*bottom*).