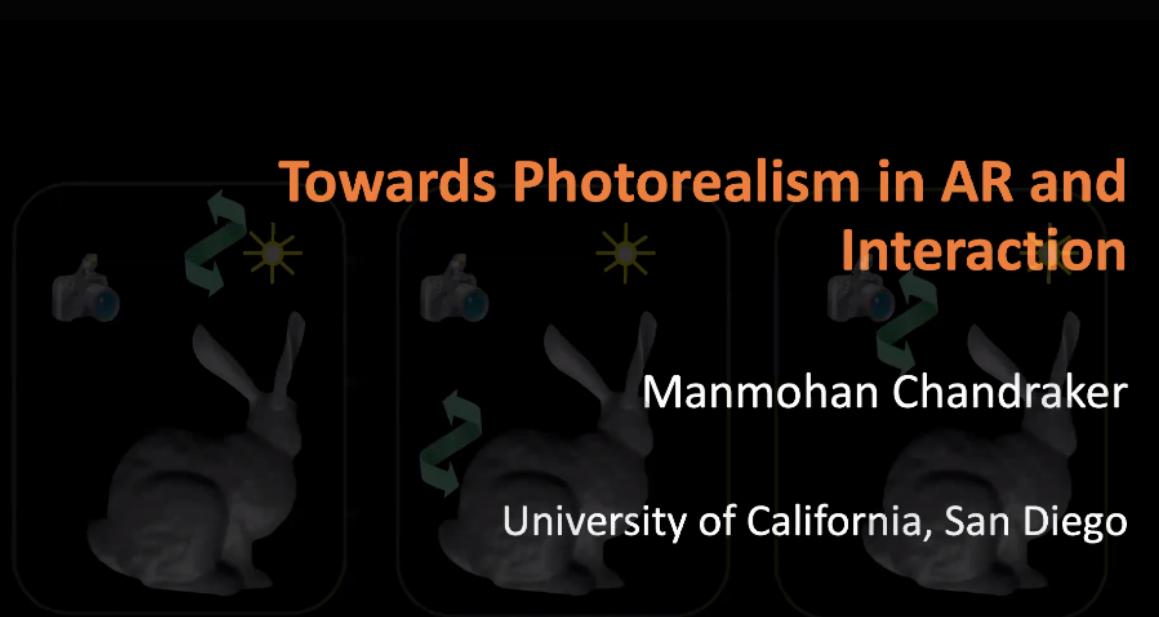


Day 17

Speaker: Prof. Manmohan Chandraker, University of California, San

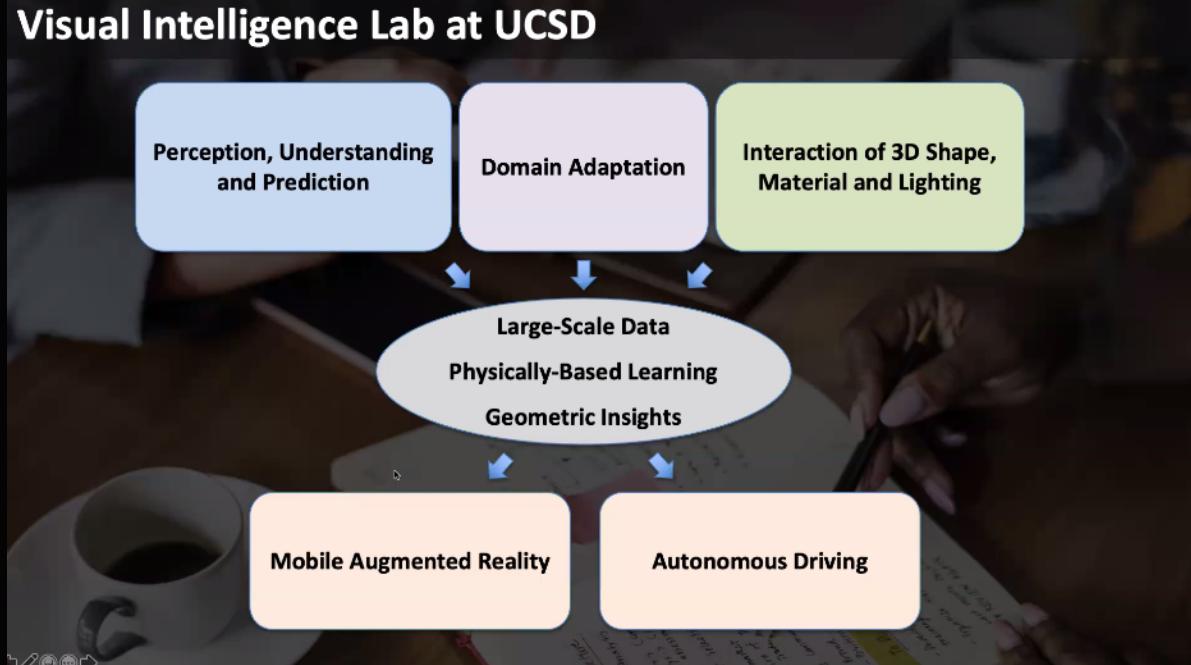
Title: Towards Photorealism in AR & Interaction  
Diego



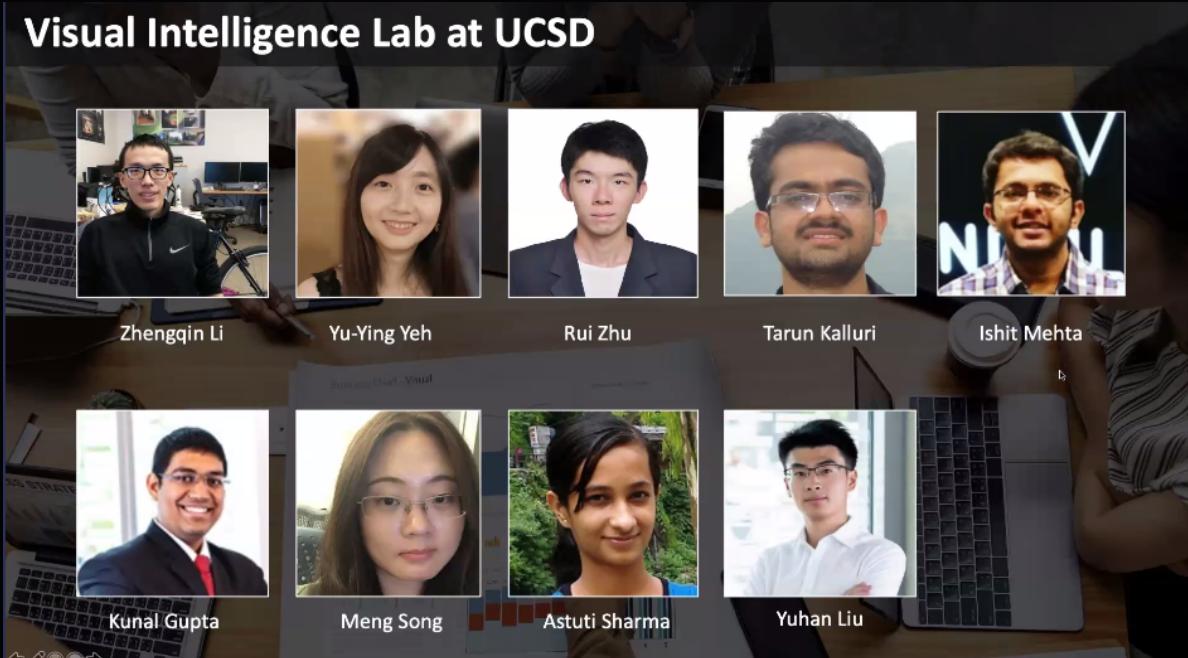
Acknowledgments: NSF, Adobe, Google, Cognex, Snap, BASF, Qualcomm



## Visual Intelligence Lab at UCSD



## Visual Intelligence Lab at UCSD



## Mobile Phones Are Making Technology More Accessible



## Visualize Glossy Object in Novel Lighting and Views



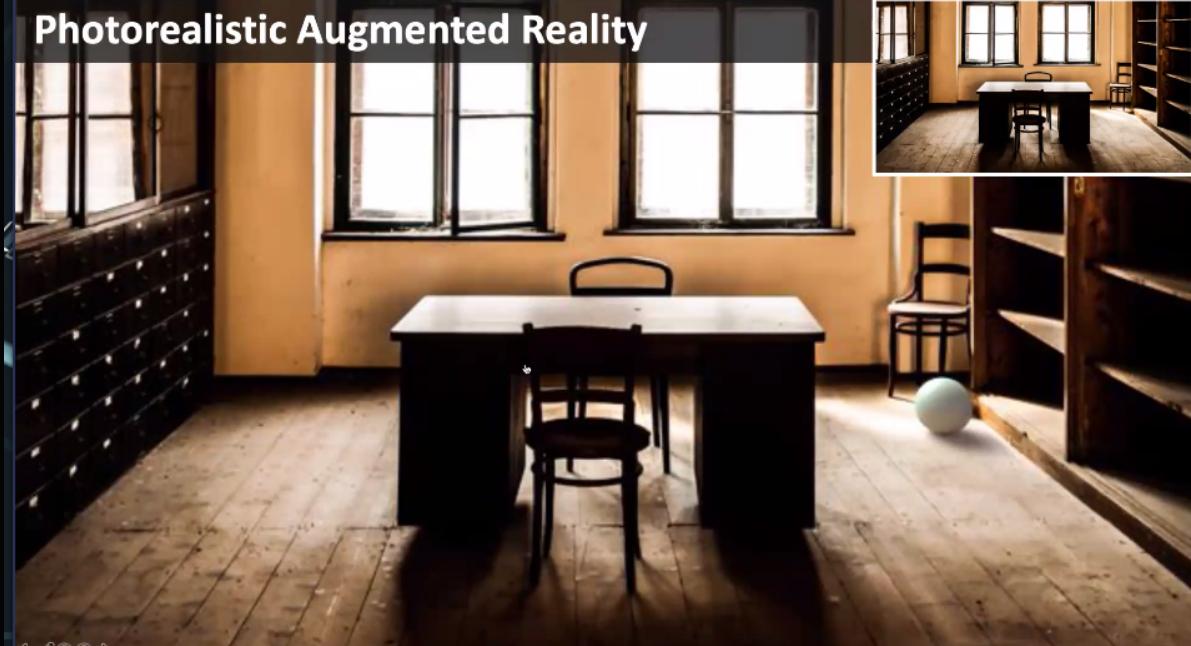
## Model 3D Content with Complex Light Transport



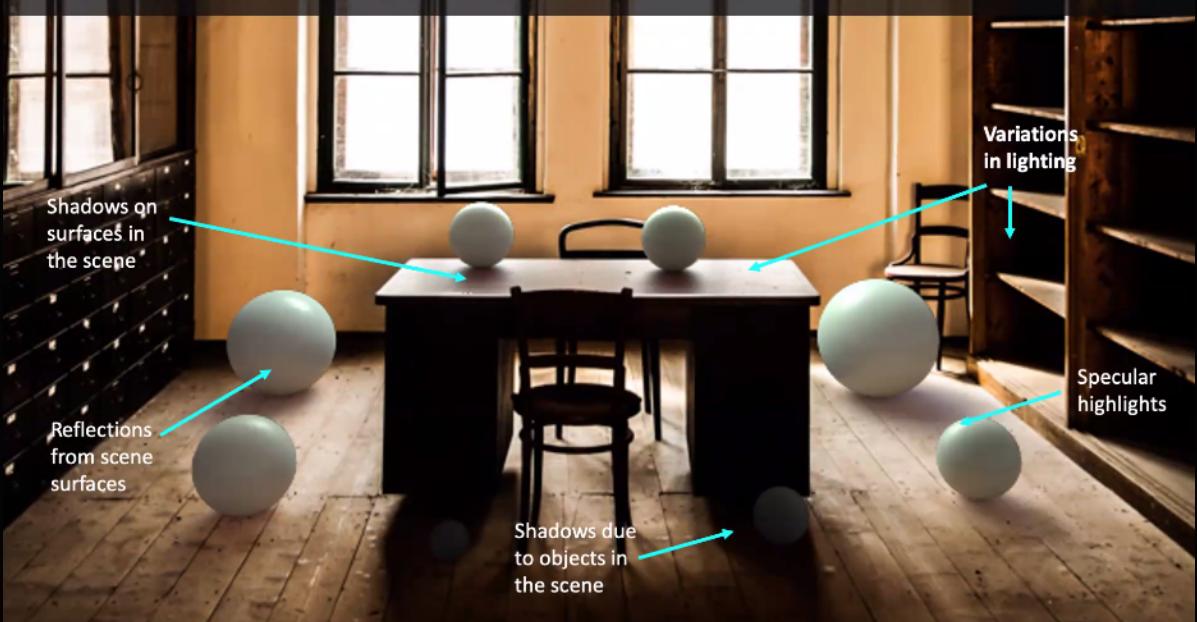
## Photorealistic Augmented Reality



## Photorealistic Augmented Reality



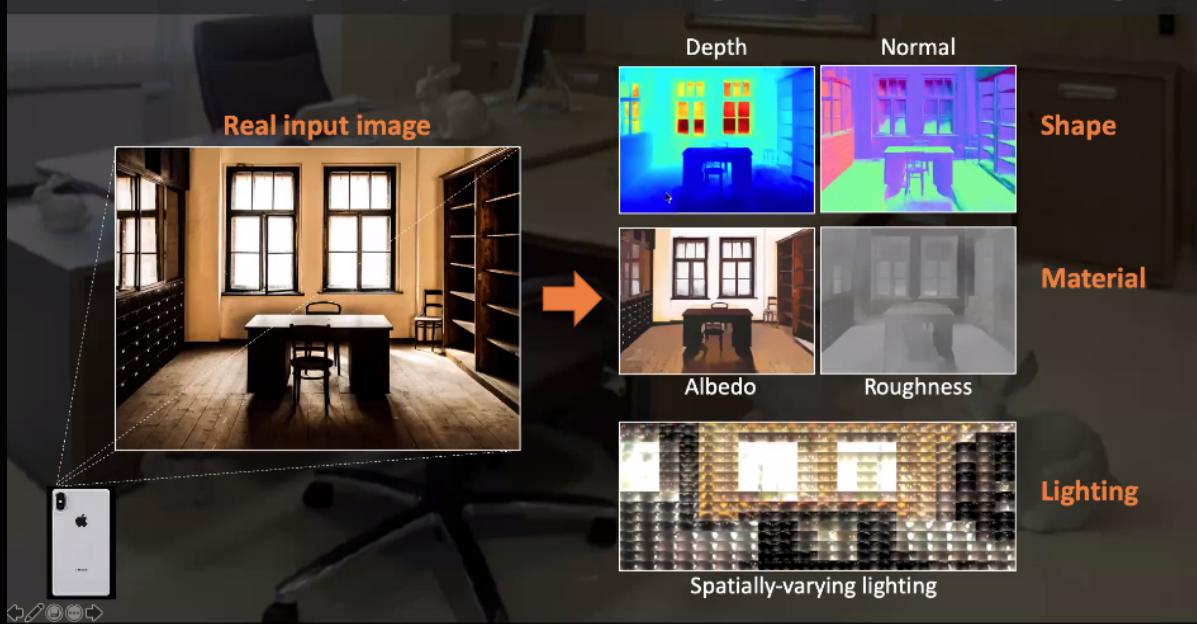
## The Demands of Photorealism



## Image Formation is a Complex Process



## Inverse Rendering: Shape, Material and Lighting from a Single Image



## Component Problems of Inverse Rendering

### Geometry Estimation



[Newcombe 2011, Eigen 2015]

### Material Estimation



[Bi 2011, Bell 2015]

### Lighting Estimation

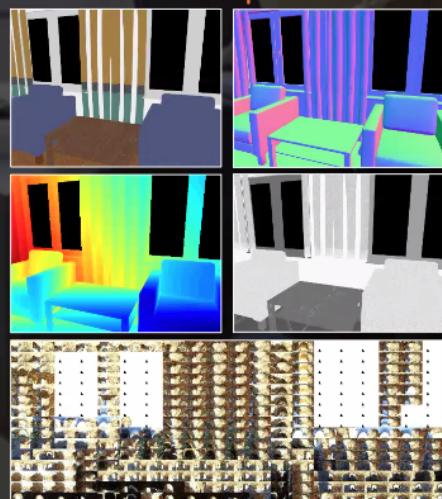


[LeGendre 2019, Garon 2019]



## Ambiguities of Inverse Rendering

### Scene Components

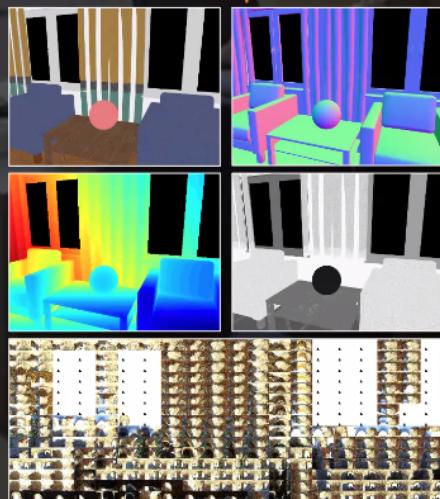


### Input Image



## Ambiguities of Inverse Rendering

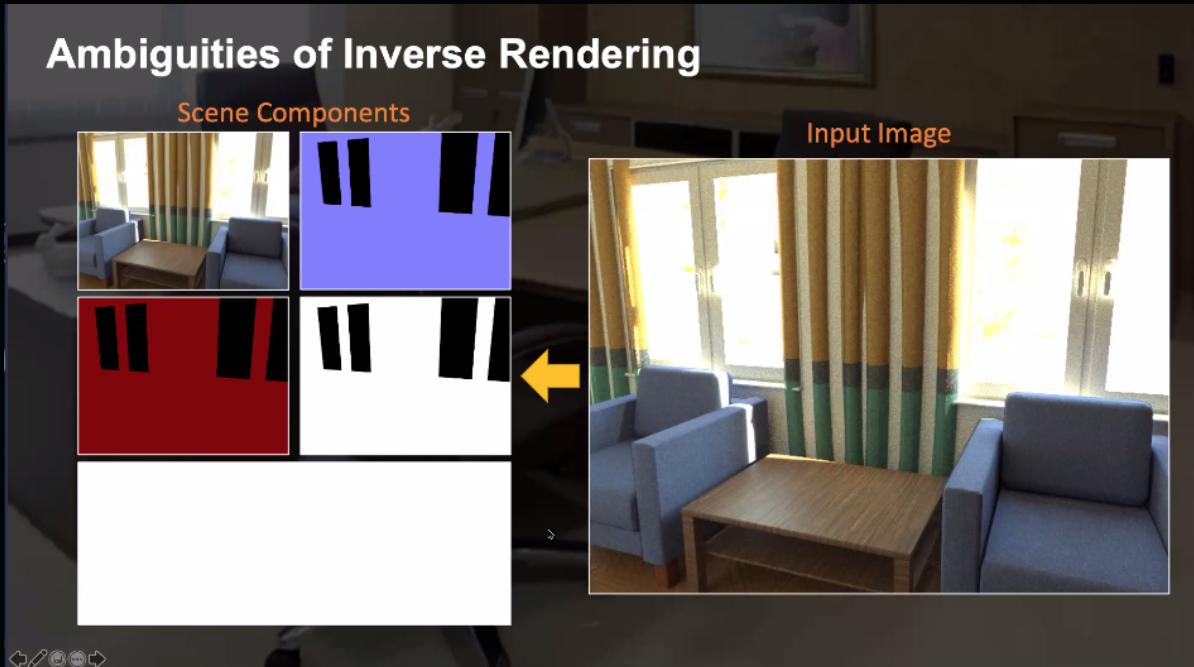
### Scene Components



### Augmented Image



## Ambiguities of Inverse Rendering



## Approaches to Inverse Rendering



## Approaches to Inverse Rendering

**Physics-based**

Theoretical guarantees, but usually restrictive models

Diagram: A light source (v) emitting light rays onto a surface with normal vector n and sampling vector s. Results: A 4x4 grid of spheres showing various lighting effects, and a 2x2 grid of faces and a 3D model of a sheep.

[Chandraker 2014, Barron 2015, Lombardi 2015]

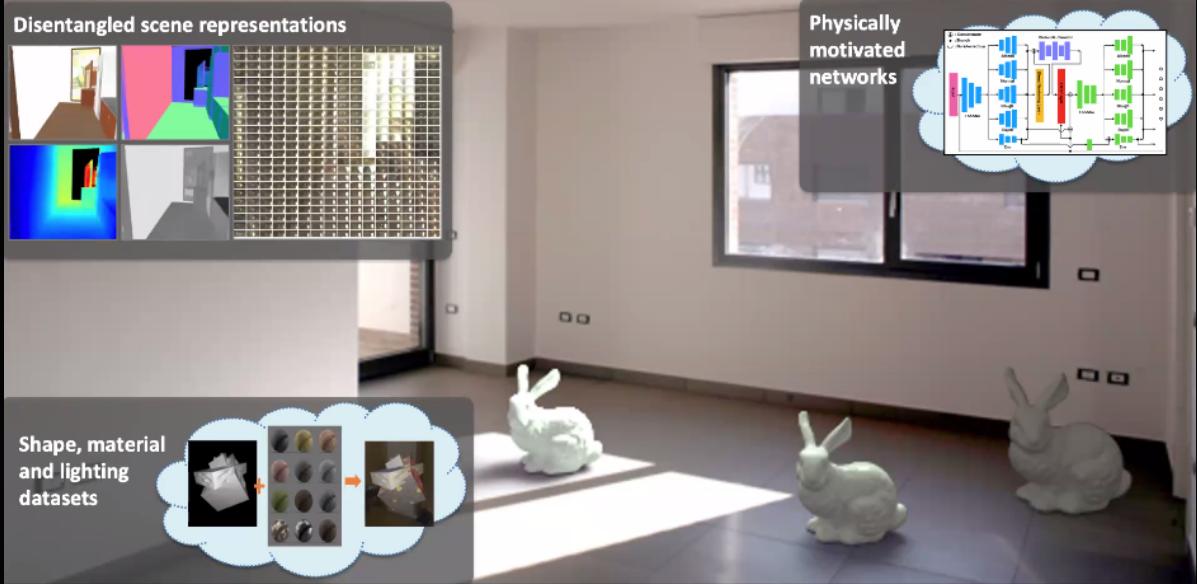
**Measurement-based**

High photorealism, but expensive acquisition

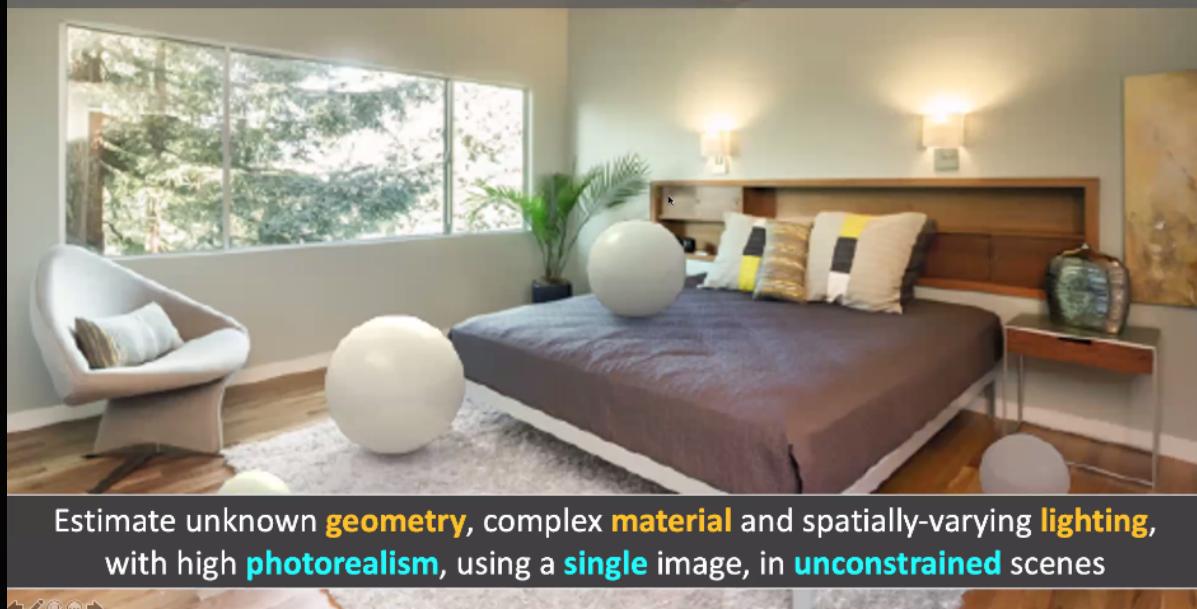
Diagram: A complex rig with many cameras and projectors. Results: A large sphere reflecting a room scene with coordinates (60,40,35), (10,17,19), and (Actual projector output (compensated)).

[Debevec 1998, Chandraker 2011, Chang 2017]

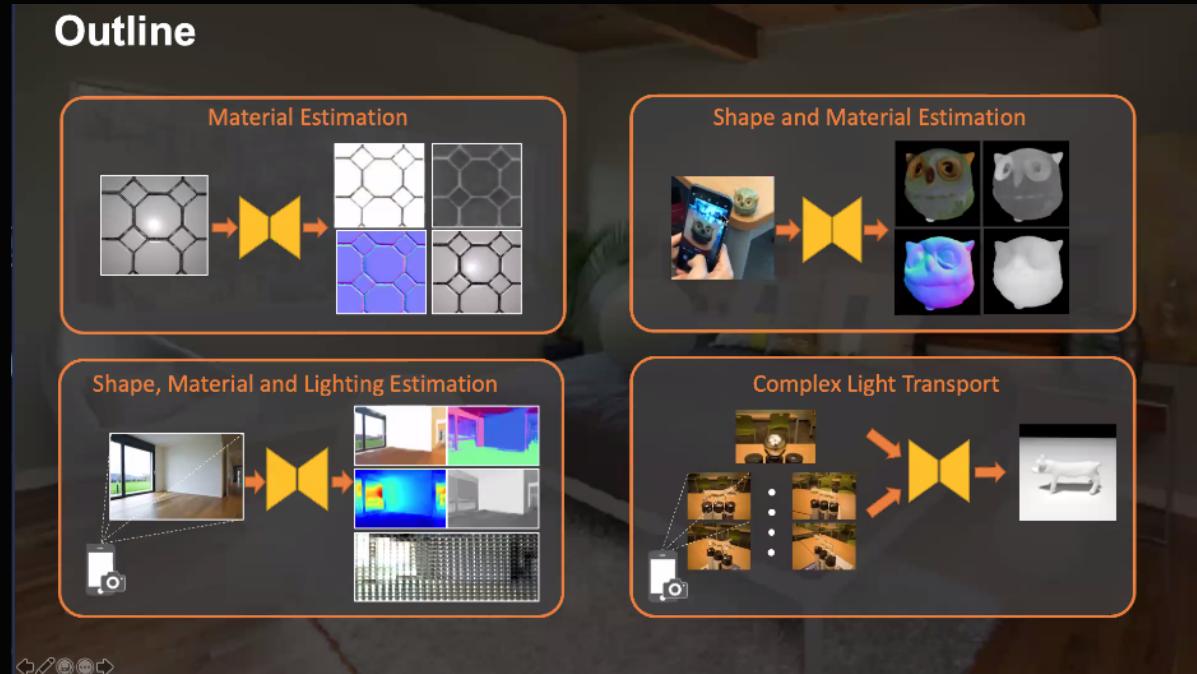
# Physically-Motivated Learning for Inverse Rendering



## Photorealism in AR: Canonical Challenge of Inverse Rendering



## Outline



# Photorealism in AR: Canonical Challenge of Inverse Rendering

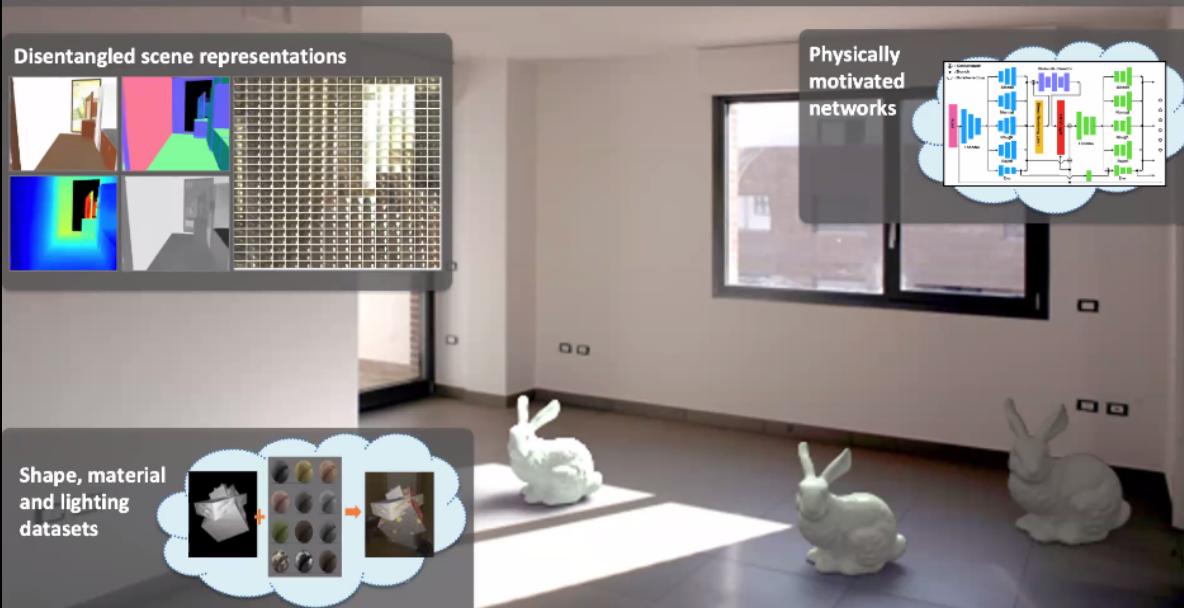


Estimate unknown **geometry**, complex **material** and spatially-varying **lighting**, with high **photorealism**, using a **single** image, in **unconstrained** scenes

## Approaches to Inverse Rendering



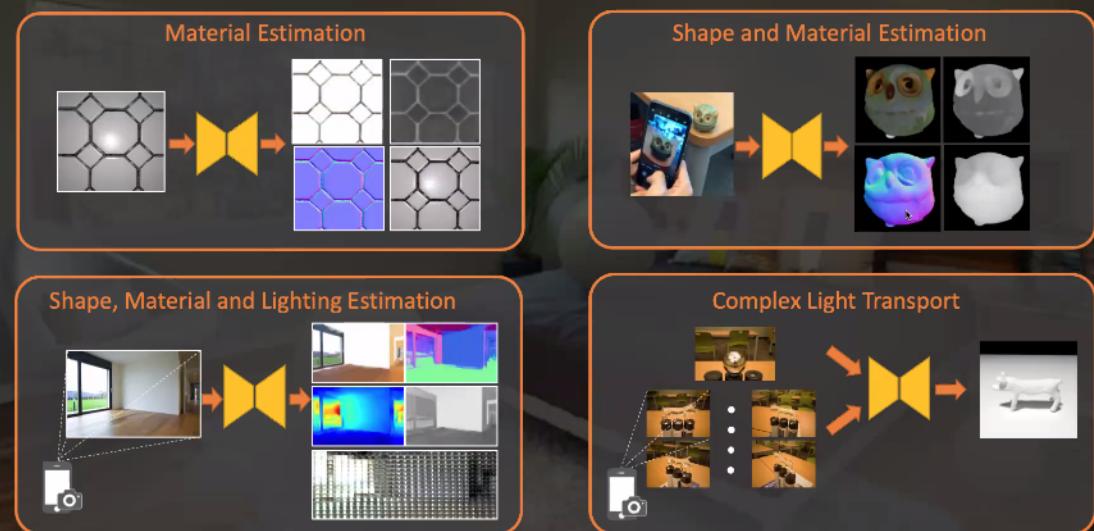
## Physically-Motivated Learning for Inverse Rendering



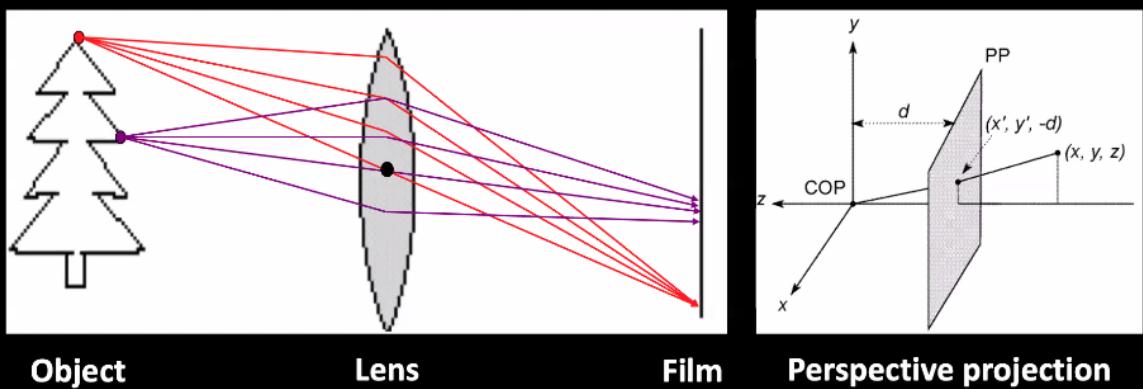
# Photorealism in AR: Canonical Challenge of Inverse Rendering



## Outline



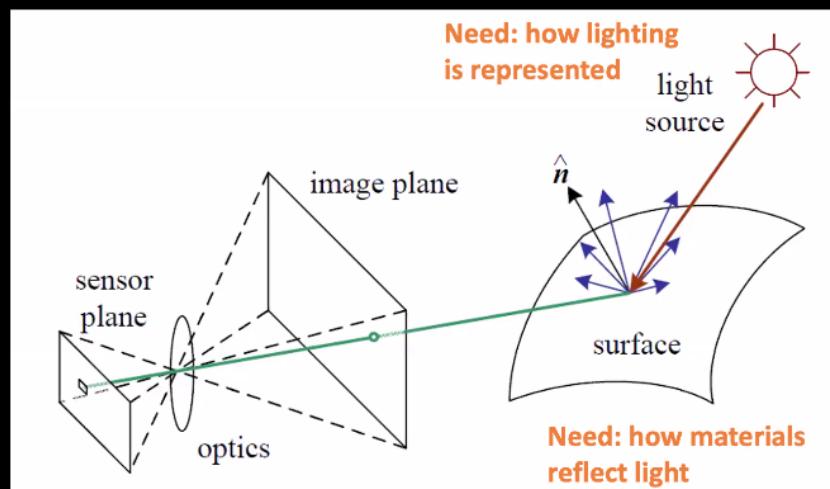
## Image Formation: Geometric



# Geometric Models Do Not Explain Appearance



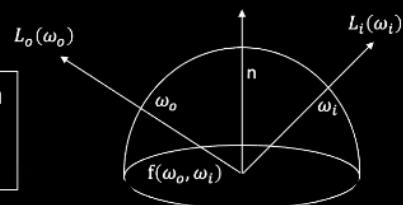
## Image Formation: Photometric



## Background: BRDF

**Bidirectional Reflectance Distribution Function**

- Ratio of outgoing and incident energies
- Depends on viewing and lighting directions



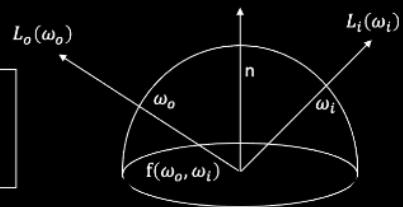
b



## Background: BRDF

### Bidirectional Reflectance Distribution Function

- Ratio of outgoing and incident energies
- Depends on viewing and lighting directions



Traditional assumptions: diffuse, Lambertian BRDF



Real-world materials: complex, unknown BRDF



$$\text{Lambertian BRDF : } \frac{\rho}{\pi}$$

$$\text{Microfacet BRDF : } \frac{\rho}{\pi} + \frac{D(\omega_o, \omega_i, R)F(\omega_o, \omega_i)G(\omega_o, \omega_i, R)}{4(\omega_o \cdot n)(\omega_i \cdot n)}$$



## Microfacet BRDF

- Microfacet theory: Assume the surface is made up of microscopic grooves



- Assume the faces of these grooves (called microfacets) are perfect reflectors

- Take into account 3 phenomena



## Microfacet BRDF

### Without self-shadowing



### With self-shadowing



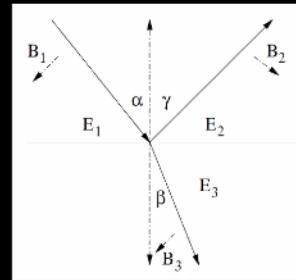
# Microfacet BRDF

Reflection greater at more oblique angle



[Lafortune, Foo, Torrance, Greenberg, SIGGRAPH 1997]

- Beam splits into reflected and refracted
- Determined by Snell's Law



$$\begin{aligned}\sin \alpha &= n \sin \beta \\ \alpha &= \gamma \\ r_{\perp} &= \frac{\cos \alpha - n \cos \beta}{\cos \alpha + n \cos \beta} \\ r_{\parallel} &= \frac{n \cos \alpha - \cos \beta}{n \cos \alpha + \cos \beta}\end{aligned}$$

## Microfacet BRDF

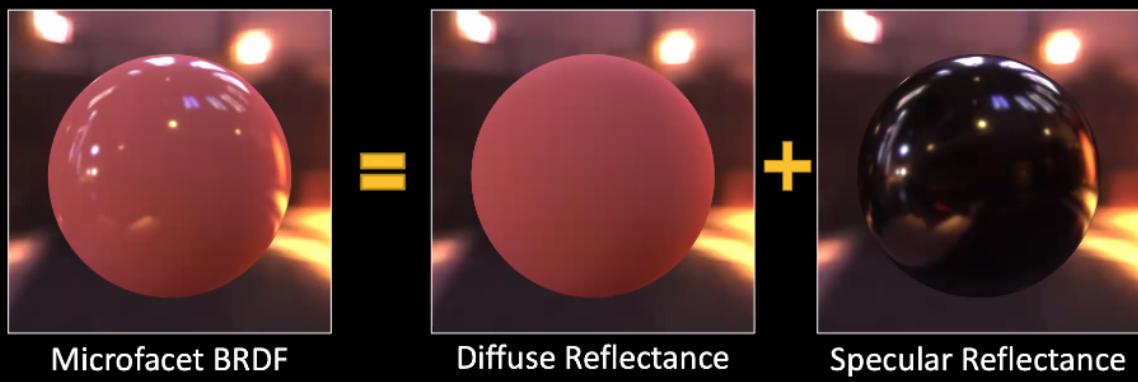
$$f(\omega_o, \omega_i) = \frac{\rho}{\pi} + \frac{D(\omega_o, \omega_i, R)F(\omega_o, \omega_i)G(\omega_o, \omega_i, R)}{4(\omega_o \cdot n)(\omega_i \cdot n)}$$

Annotations for the Microfacet BRDF equation:

- Distribution function**: Percentage of microfacets oriented towards light
- Fresnel term**: Accounts for wavelength
- Geometric attenuation**: Reduces output based on extent of shadowing
- Diffuse albedo**: Specular roughness
- How much of macroscopic surface is visible to viewer
- How much of macroscopic surface is visible to light source

## Background: BRDF

$$\text{Microfacet BRDF : } f(\omega_o, \omega_i) = \frac{\rho}{\pi} + \frac{D(\omega_o, \omega_i, R)F(\omega_o, \omega_i)G(\omega_o, \omega_i, R)}{4(\omega_o \cdot n)(\omega_i \cdot n)}$$



# Local and Global Illumination

- Local illumination

- Light directly from light sources
- No shadows or interreflections



- Global illumination

- Effect of light sources, interactions with rest of scene and participating media
- Hard and soft shadows
- Reflections, refractions and scattering
- Diffuse and glossy interreflections



## Background: Lighting

- Light as a function of direction, from entire environment



## Background: Lighting

- Light as a function of direction, from entire environment
- Captured by photographing a chrome steel or mirror sphere
- Accurate only for one point, but nearly same at all scene points for distant lighting

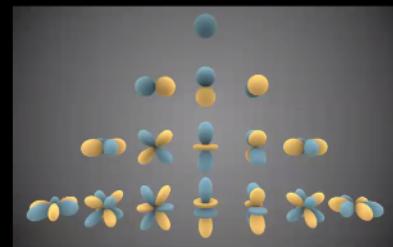


[Blinn and Newell 1976, Miller and Hoffman, 1984]

## Background: Spherical Harmonics Lighting

Distant lighting can be approximated by N order spherical harmonics:

$$L_i(\omega_i) = \sum_{l=1}^N \sum_{m=-l}^l K_{lm} Y_{lm}(\omega_i)$$

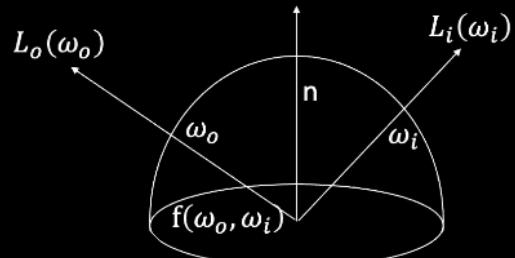


Larger N can preserve higher angular frequency



## Image Formation: Rendering Equation

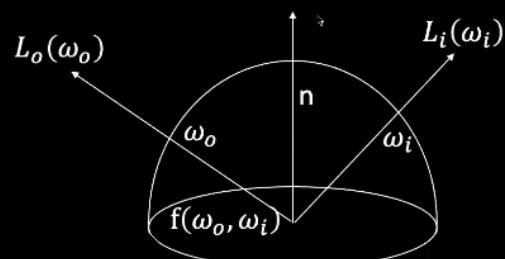
$$L_o(\omega_o) = \int_{\Omega} L_i(\omega_i) f(\omega_o, \omega_i) (\omega_i \cdot n) d\omega_i$$



$n$	Normal direction, geometry
$\omega_i$	Direction of incoming light
$\omega_o$	Direction of outgoing light
$\Omega$	Hemisphere
$f(\omega_o, \omega_i)$	Bidirectional reflectance distribution function (BRDF)
$L_o(\omega_o)$	Outgoing radiance, the observation
$L_i(\omega_i)$	Incoming radiance

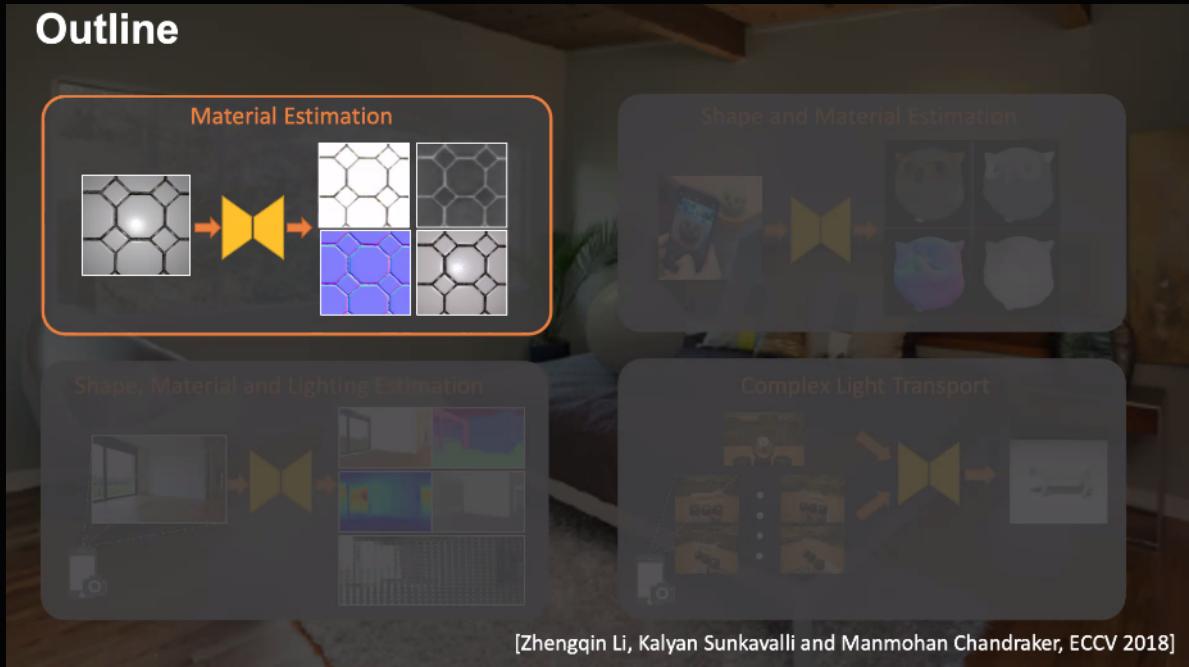
## Image Formation: Rendering Equation

$$L_o(\omega_o) = \int_{\Omega} L_i(\omega_i) f(\omega_o, \omega_i) (\omega_i \cdot n) d\omega_i$$

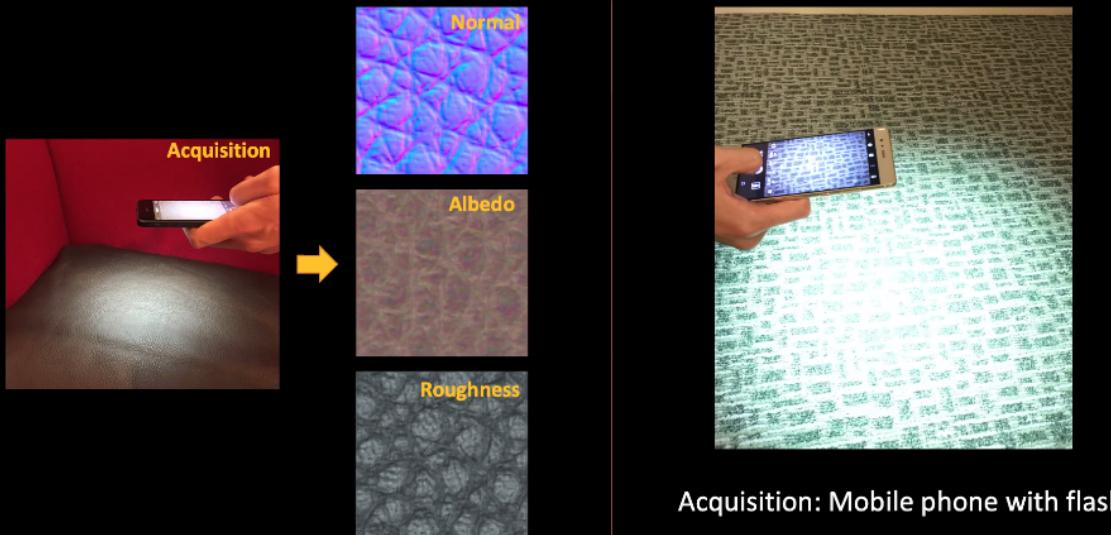


$n$	<b>Normal direction, geometry</b>
$\omega_i$	Direction of incoming light
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$\Omega$	Hemisphere
$f(\omega_o, \omega_i)$	<b>Bidirectional reflectance distribution function (BRDF)</b>
$L_o(\omega_o)$	Outgoing radiance, the observation
$L_i(\omega_i)$	<b>Incoming radiance</b>

# Outline



## Material Estimation from a Single Image



## Large-Scale Dataset of Complex Materials



Homogeneous, isotropic, 100 materials



Large-scale, but not exclusively materials and only coarse properties available