



北京航空航天大学  
BEIHANG UNIVERSITY



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CVPR VANCOUVER, CANADA

# NeuFace: Realistic 3D Neural Face Rendering from Multi-view Images

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Rendered



Geometry



Diffuse



Specular

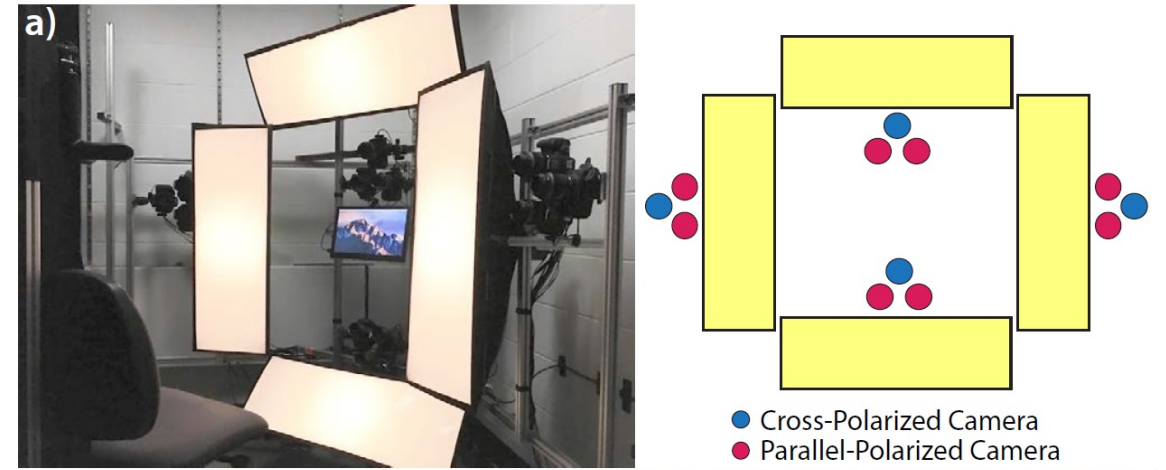


Relighting

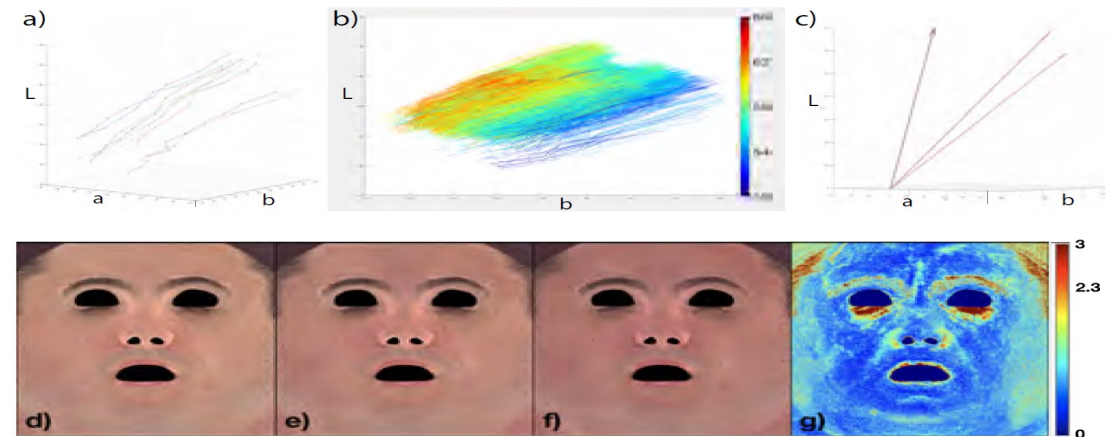
# Face Geometry & Appearance Capture



Active: OLAT/Multi-phased Light



Passive: Polarized Camera



Passive: Dynamic Blood Flow Input

NVIDIA. Realistic Digital Human Rendering with Omniverse RTX Renderer. SIGGRAPH 2021 Session.  
Jérémy Riviere, et.al. Single-Shot High-Quality Facial Geometry and Skin Appearance Capture. TOG 2020.  
Paulo Gotardo, et.al. Practical Dynamic Facial Appearance Modeling and Acquisition. TOG 2018.

# Face Geometry & Appearance Capture

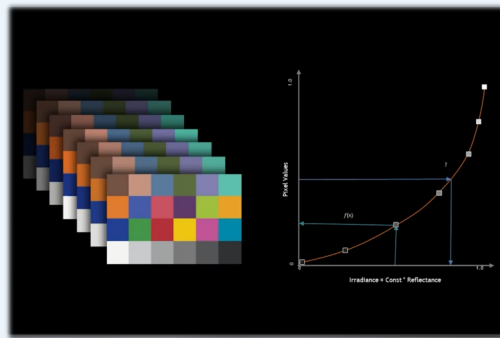
## DATA ACQUISITION

- ~8400 images with time-multiplexed, multi-phased polarized and constant illuminations
- ~180GB of data



## RADIOMETRIC CALIBRATION

- Recovering sensor response curve
- Calibrating color by capturing color chart with multiple illuminations



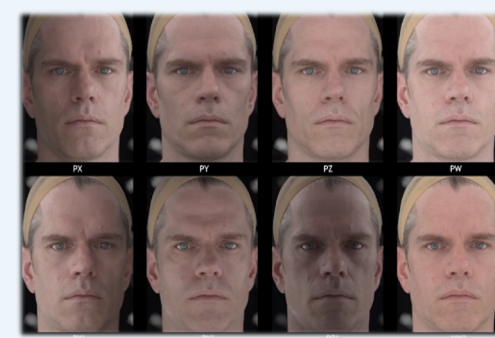
## MULTI-VIEW RECONSTRUCTION

- Correspondence finding, depth fusion and surface reconstruction
- Mesh parameterization and texture projection



## MATERIAL ESTIMATION

- Light estimation
- Decomposing reflectance by polarized images (or dynamic images)

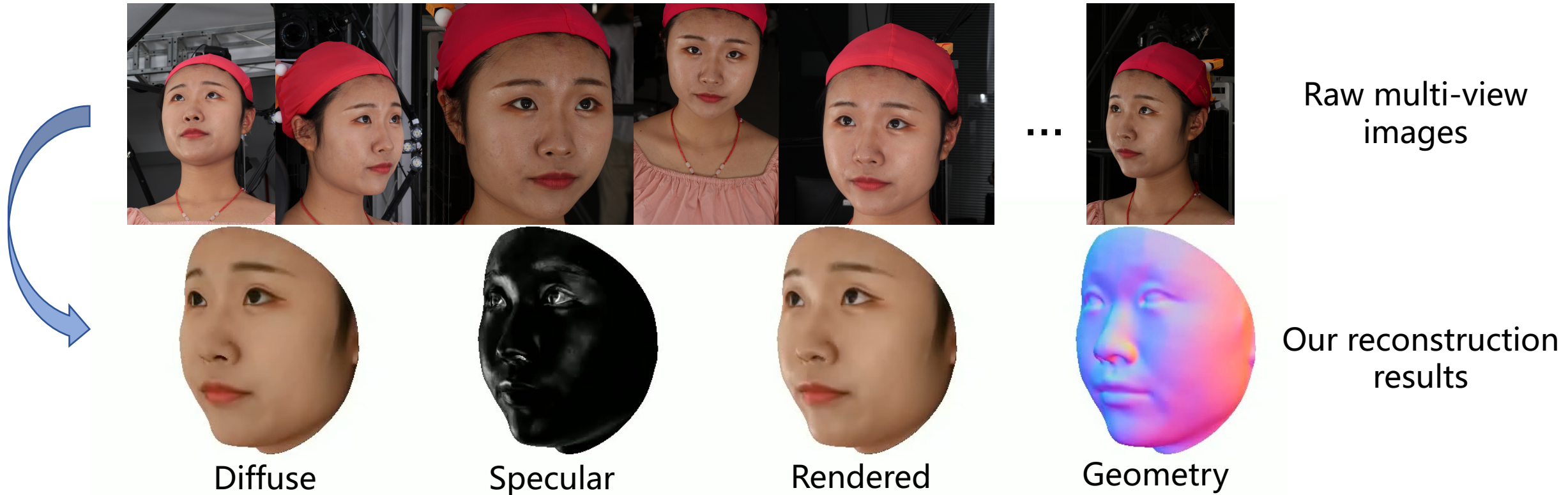


### ➤ An elaborately designed workflow:

- Depending on the expertise of the engineers with **significant manual efforts**;
- The multi-step process inevitably brings **diverse optimization goals**.



# Our Goal



- **Recovering facial geometry and appearance from multi-view images:**
  - **Uncalibrated, unpolarized** multi-view RGB images (~40)
  - **Unknow illumination** (but nearly white)
- **High-quality results with a simple and complete end-to-end workflow.**



# Challenges

## Complicated Material

- The multi-layered facial skin leads to complex view dependent and spatially-varying highlights.

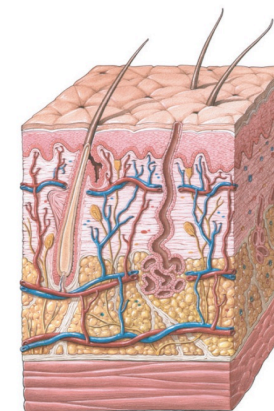
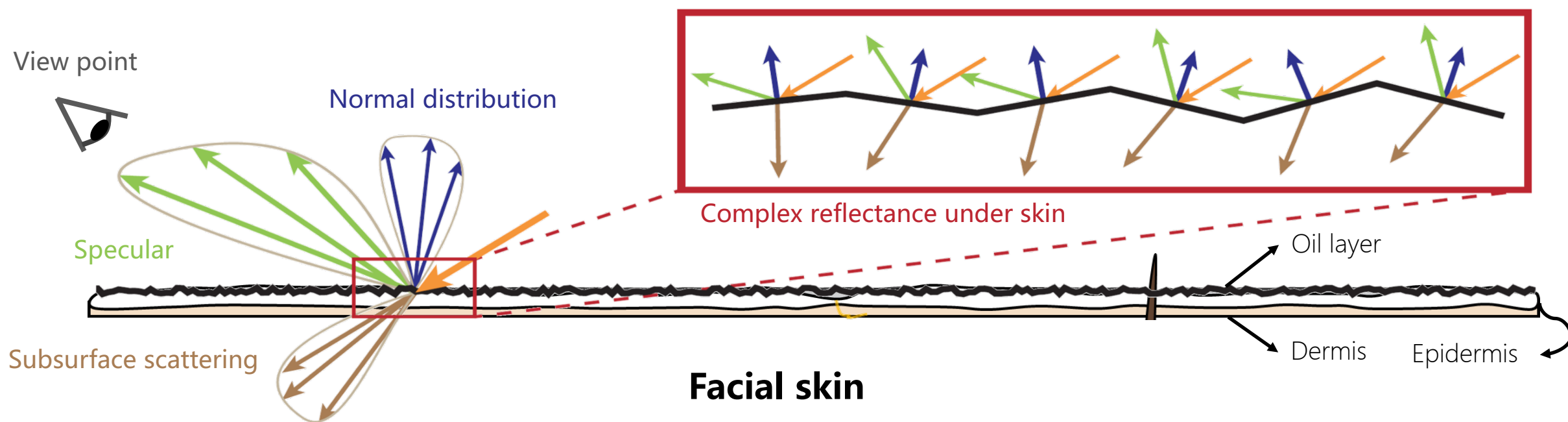


Image courtesy of A.D.A.M.



# Challenges

## Complicated Material

- The multi-layered facial skin leads to complex view dependent and spatially-varying highlights.
- The exploited physical priors are incapable of describing human face 🙄:
  - Physical priors: Phong<sup>1</sup>, Torrance and Sparrow\*, and Disney BRDF model<sup>2</sup>

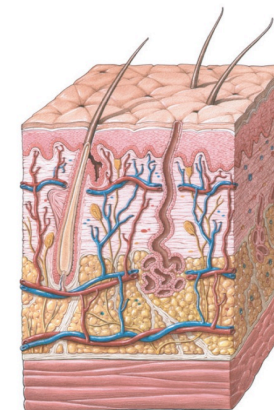
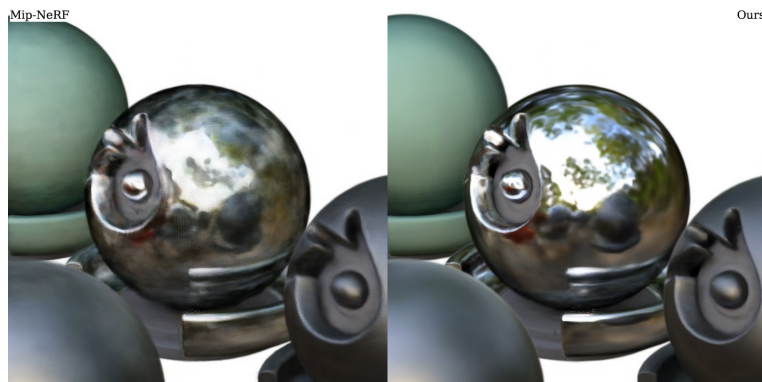
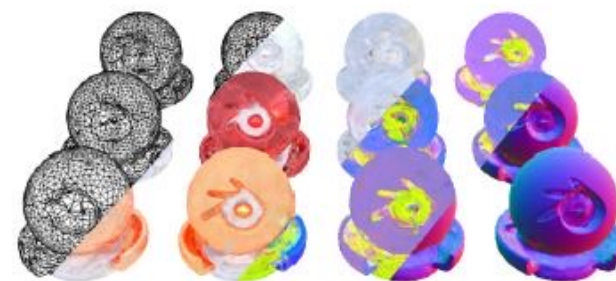


Image courtesy of A.D.A.M.



<sup>1</sup>Ref-NeRF (CVPR 2022)



Mesh /  $k_d$  /  $k_{orm}$  /  $n$



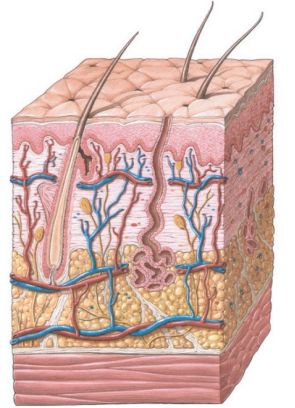
Extracted probe

<sup>2</sup>DIFFREC (CVPR 2022)

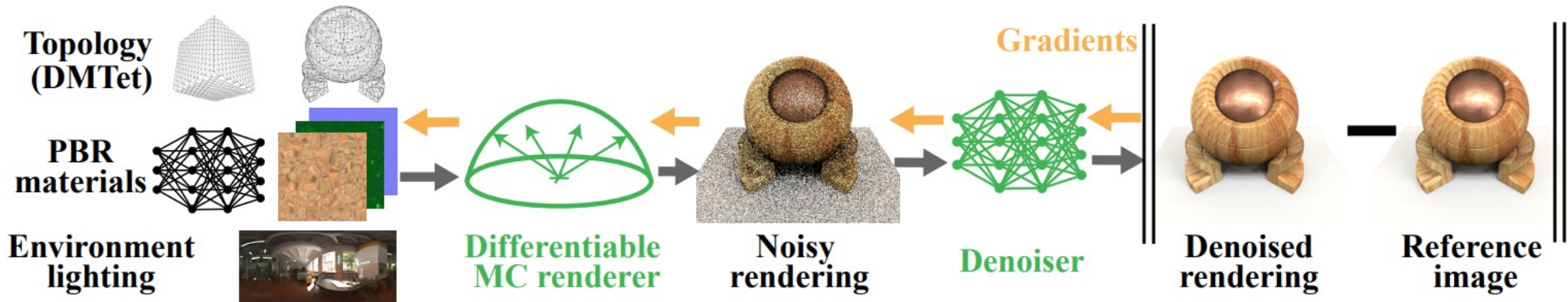
# Challenges

## Complicated Material

- The multi-layered facial skin leads to complex view dependent and spatially-varying highlights.
- The exploited physical priors are incapable of describing human face 🙄:
  - Physical priors: Phong<sup>1</sup>, Torrance and Sparrow\*, and Disney BRDF model<sup>2</sup>
- Solving the rendering equation is computationally-expensive 🙄:
  - Monte Carlo sampling is typically required.

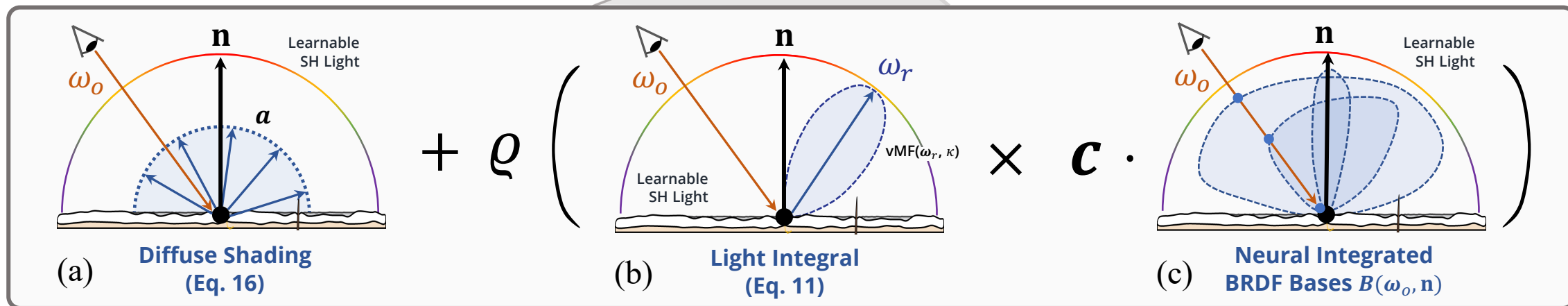
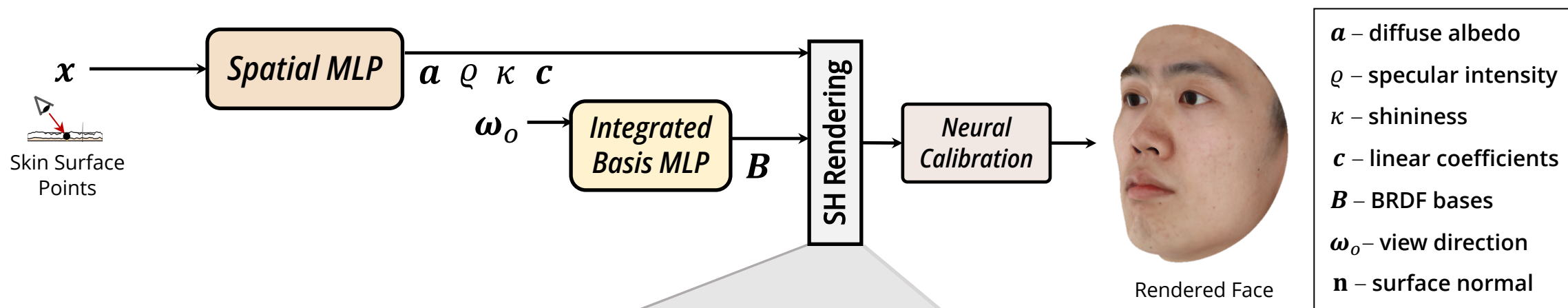


*Image courtesy of A.D.A.M.*

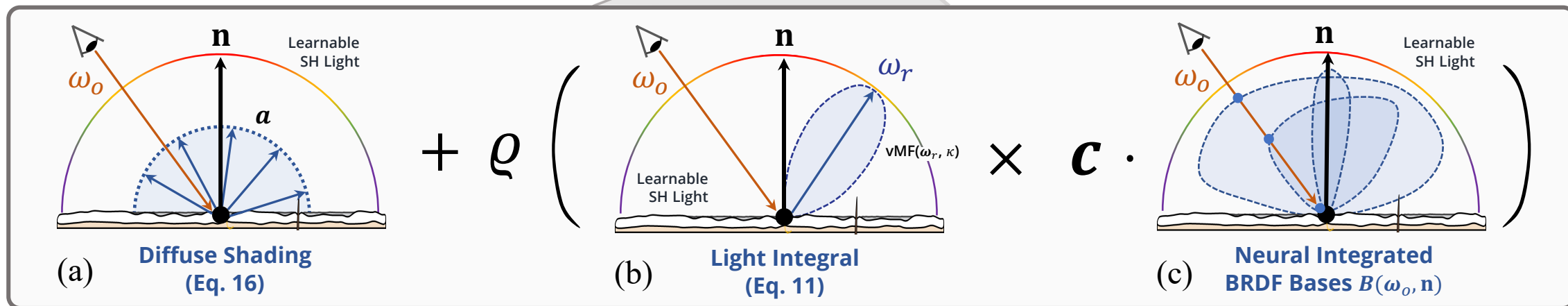
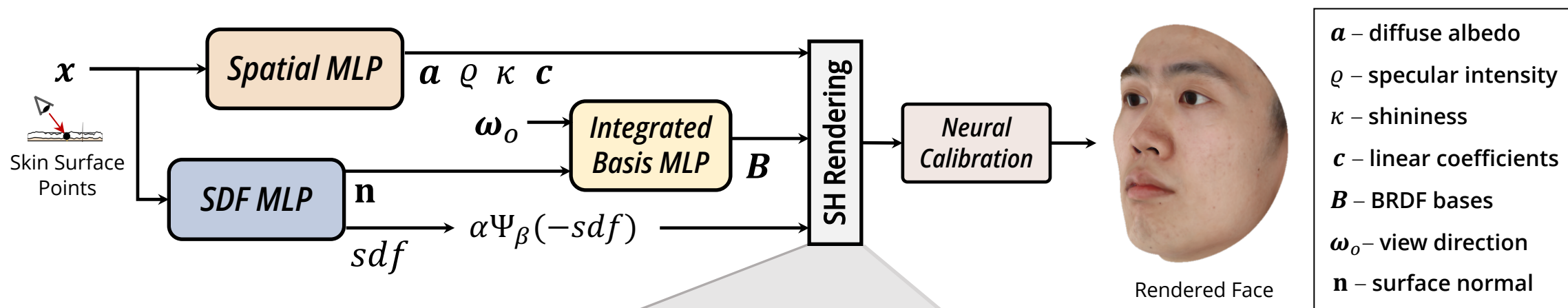




# NeuFace Overview



# NeuFace Overview



$$L_o(x, \omega_o) = \frac{\alpha(x)}{\pi} \sum_{l=0}^{\infty} \sum_{m=-l}^l A_{lm} c_{lm} Y_{lm}(\mathbf{n}) + \varrho \cdot c(x) \cdot B(\omega_o, \mathbf{n}) \sum_{l=0}^{\infty} \sum_{m=-l}^l e^{-\frac{l(l+1)\rho}{2}} c_{lm}(x, \omega_o) Y_{lm}(\omega_r)$$

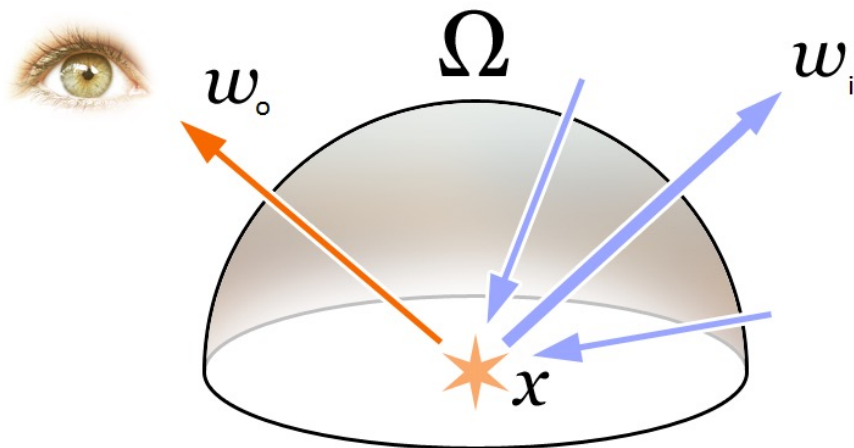
# Method

- A novel physically-based neural rendering framework.

$$L_o(x, \omega_o) = \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i + \varrho \int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

↑
↓
↑

Diffuse term  $L_d$ 
Specular term  $L_s$





# Step 1. Split Integral

$$L_o(x, \omega_o) = \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i + \rho \int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

↑
↓
↑

Diffuse term  $L_d$

$$\int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

$$= \int_{\Omega} f_s(x, \omega_i, \omega_o) (\omega_i \cdot \mathbf{n}) d\omega_i \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$

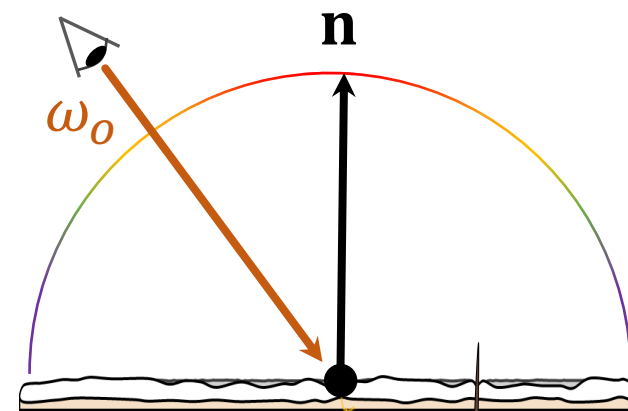
Material integral

Light integral

**Split-sum approximation (from Unreal Engine)**

Specular intensity  $\rho$

Specular term  $L_s$



## Step 2. Material Integral

$$L_o(x, \omega_o) = \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i + \rho \int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

↑
↓

Diffuse term  $L_d$ 
Specular term  $L_s$

↑
Specular intensity  $\rho$

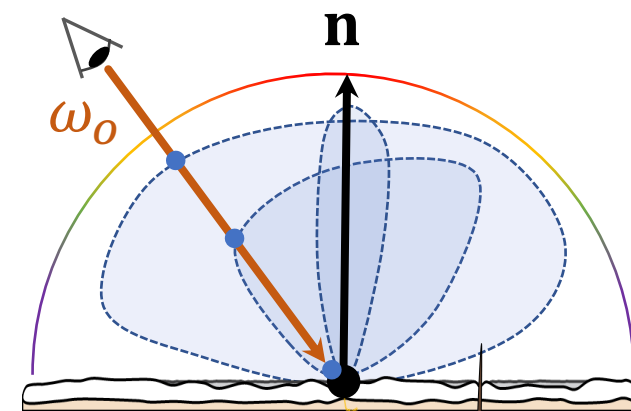
$$\int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

$$= \int_{\Omega} f_s(x, \omega_i, \omega_o) (\omega_i \cdot \mathbf{n}) d\omega_i \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$

Material integral

Light integral

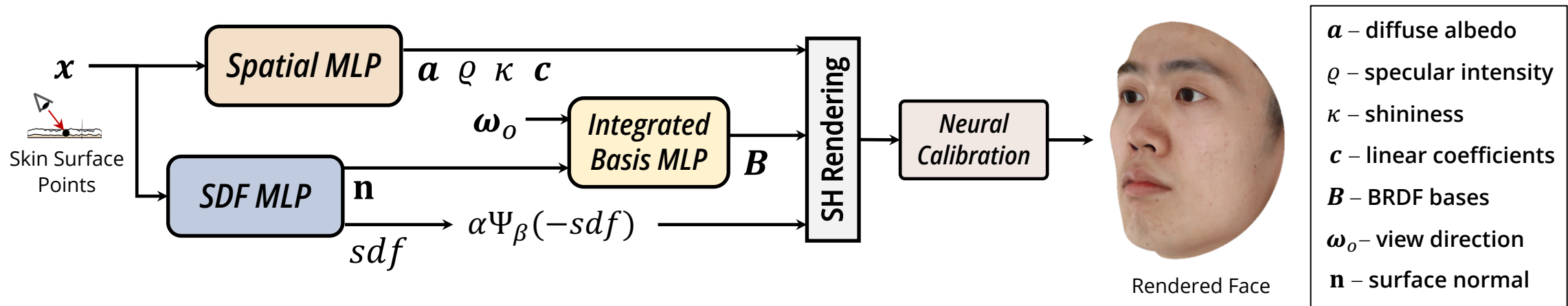
$$= \mathbf{c}(x) \cdot \mathbf{B}(\omega_i, \omega_o, \mathbf{n}) \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$



Low-rank BRDF (rank 3)

A similar specular structure should be low-rank

# Step 2. Material Integral



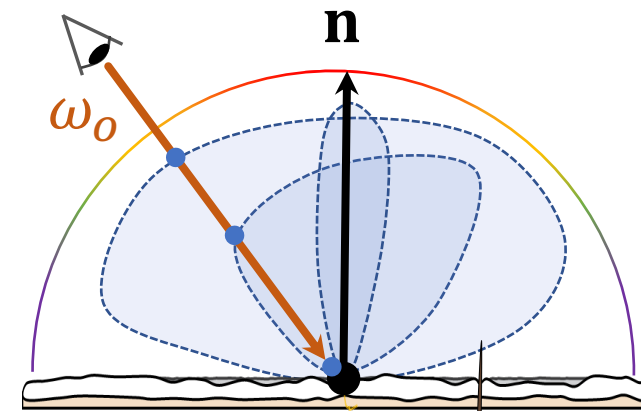
$$\int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot n) d\omega_i$$

$$= \int_{\Omega} f_s(x, \omega_i, \omega_o) (\omega_i \cdot n) d\omega_i \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$

Material integral

Light integral

$$= c(x) \cdot B(\omega_i, \omega_o, n) \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$



Low-rank BRDF (rank 3)

A similar specular structure should be low-rank



# Step 3. Light Integral

$$L_o(x, \omega_o) = \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i + \rho \int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

↑
↓

**Diffuse term  $L_d$** 
**Specular term  $L_s$**

**Specular intensity  $\rho$**

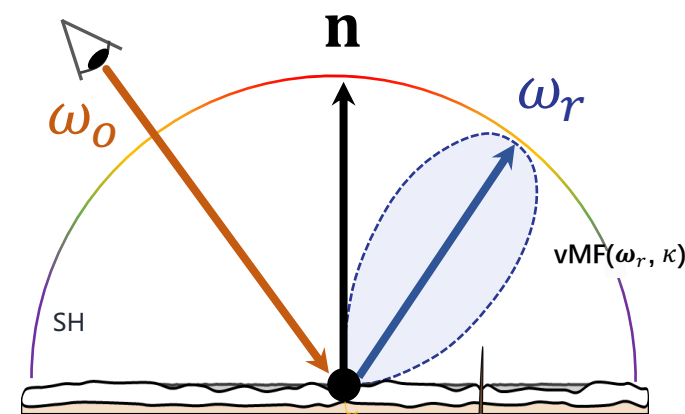
$$\int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

$$= \int_{\Omega} f_s(x, \omega_i, \omega_o) (\omega_i \cdot \mathbf{n}) d\omega_i \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$

**Material integral**
**Light integral**

$$= \mathbf{c}(x) \cdot \mathbf{B}(\omega_i, \omega_o, \mathbf{n}) \int_{\Omega} D(h) L_i(x, \omega_i) d\omega_i$$

$$= \mathbf{c}(x) \cdot \mathbf{B}(\omega_i, \omega_o, \mathbf{n}) \sum_{l=0}^{\infty} \sum_{m=-l}^l e^{-\frac{l(l+1)\rho}{2}} c_{lm}(x, \omega_o) Y_{lm}(\omega_r)$$



**Light Integral**

**Approximate the light term by properties of Spherical Harmonics**

# Step 4. Diffuse Modeling

$$L_o(x, \omega_o) = \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i + \varrho \int_{\Omega} f_s(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i$$

↑
↓  
**Diffuse term  $L_d$** 
**Specular term  $L_s$**

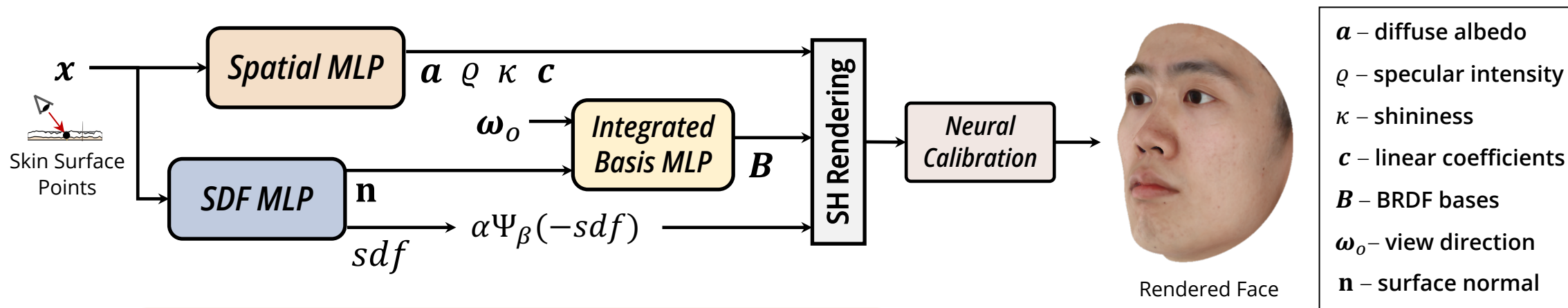
↑
↓  
**Specular intensity  $\varrho$**

$$\begin{aligned}
 L_d(x, \omega_o) &= \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i \\
 &= \frac{\alpha(x)}{\pi} \int_{\Omega} \sum_{l=0}^{\infty} \sum_{m=-l}^l c_{lm} Y_{lm}(\omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i \\
 &= \frac{\alpha(x)}{\pi} \sum_{l=0}^{\infty} \sum_{m=-l}^l c_{lm} \int_{\Omega} Y_{lm}(\omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i \\
 &= \frac{\alpha(x)}{\pi} \sum_{l=0}^{\infty} \sum_{m=-l}^l \Lambda_{lm} c_{lm} Y_{lm}(\mathbf{n})
 \end{aligned}$$

$$\Lambda_{lm} = \begin{cases} \frac{2\pi}{3}, l = 1 \\ \frac{(-1)^{\frac{l}{2}+1} \pi}{2^{l-1} (l-1)(l+2)} \binom{l}{l/2}, l \text{ is even} \\ 0, l \text{ is odd} \end{cases}$$

**Analytical solution of the diffuse term using the Funk-Hecke theorem**

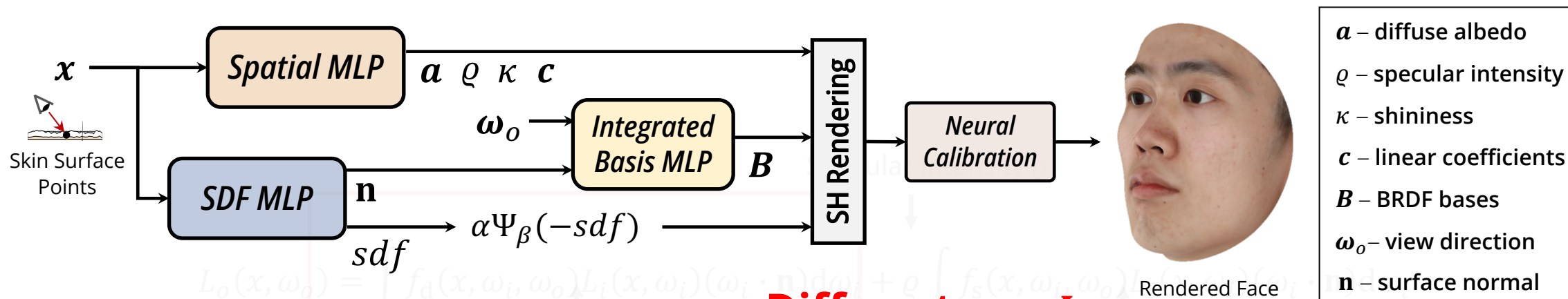
# Step 4. Diffuse Modeling



$$\begin{aligned}
 L_d(x, \omega_o) &= \int_{\Omega} f_d(x, \omega_i, \omega_o) L_i(x, \omega_i) (\omega_i \cdot \mathbf{n}) d\omega_i \\
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**Analytical solution of the diffuse term using the Funk-Hecke theorem**

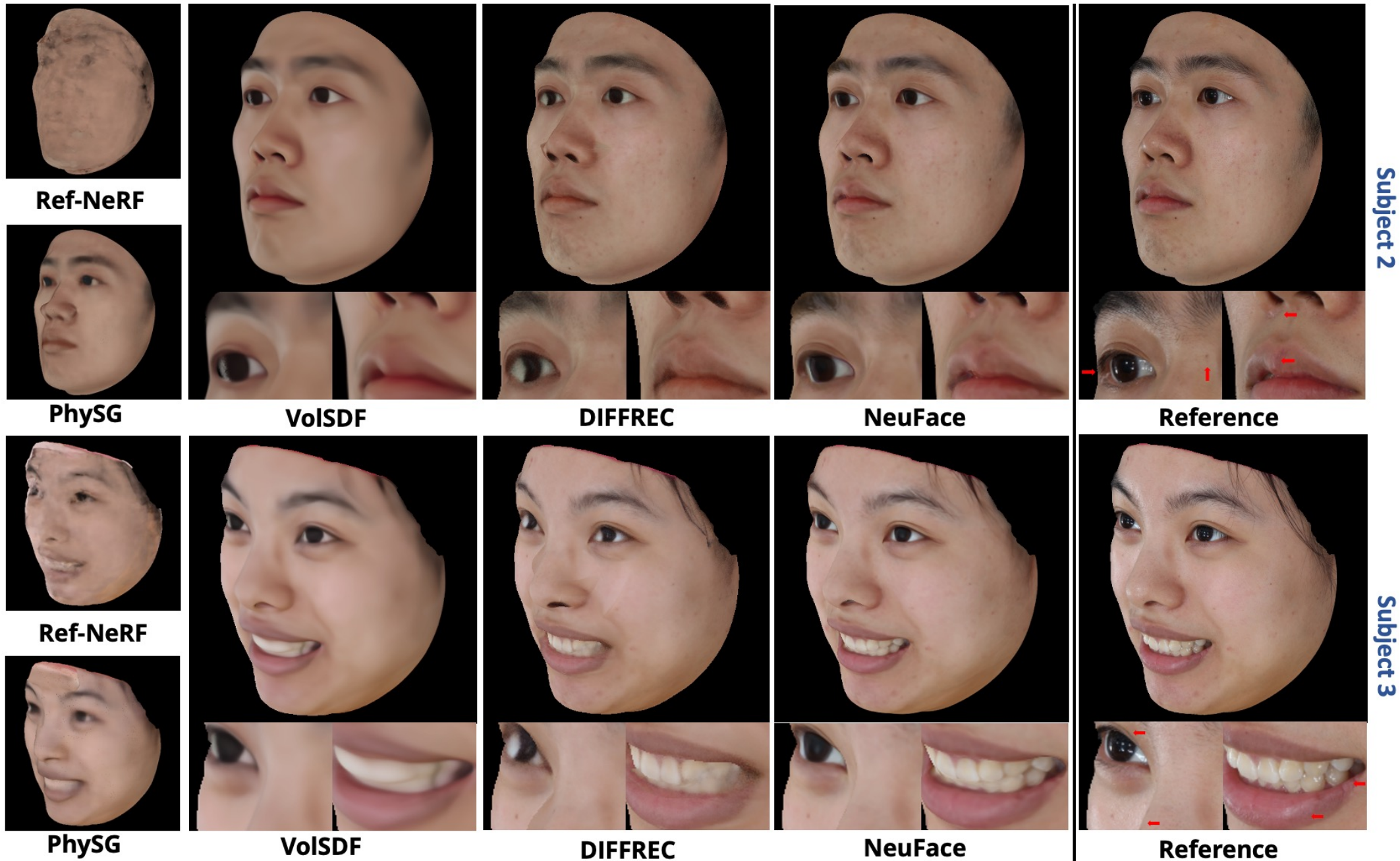


**Diffuse term  $L_d$**

$$L_d(x, \omega_o) = \frac{\alpha(x)}{\pi} \sum_{l=0}^{\infty} \sum_{m=-l}^l \Lambda_{lm} c_{lm} Y_{lm}(\mathbf{n}) + \rho \cdot c(x) \cdot B(\omega_o, \mathbf{n}) \sum_{l=0}^{\infty} \sum_{m=-l}^l e^{-\frac{l(l+1)\rho}{2}} c_{lm}(x, \omega_o) Y_{lm}(\omega_r)$$

**Specular term  $L_s$**

# Comparison with Other Neural Rendering Methods





# Comparison with Other Neural Rendering Methods

NeuFace **Ours**



DIFFREC



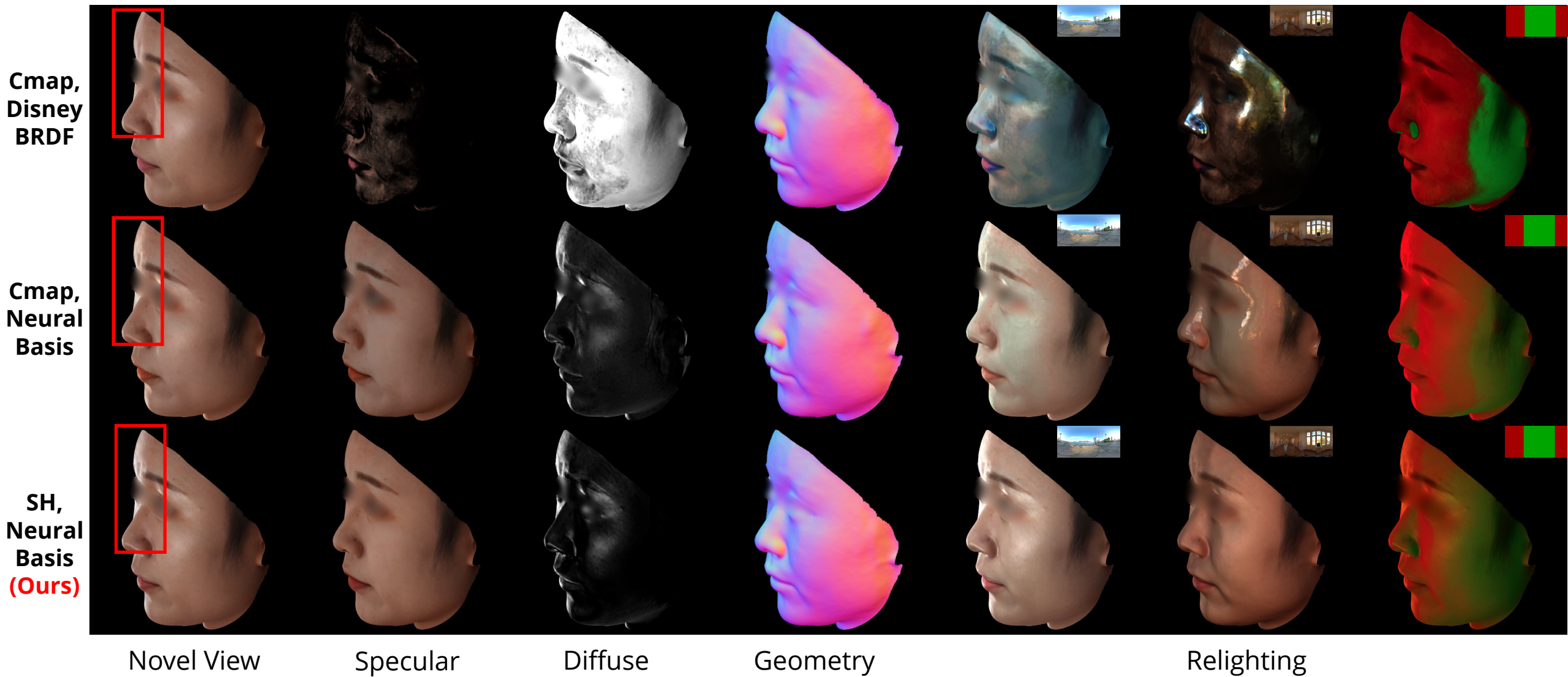
PhySG



VoISDF



# Ablation Study



# Extension to Common Objects

DIFFREC:Diffuse



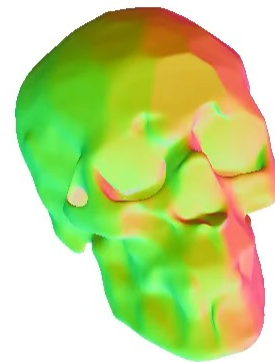
DIFFREC:Specular



DIFFREC:Render



DIFFREC:Normal



DIFFREC:Relighting



NeuFace:Diffuse



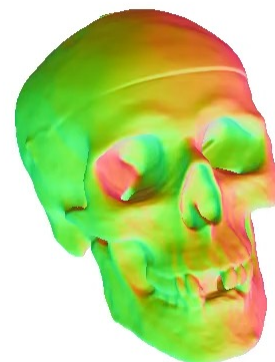
NeuFace:Specular



NeuFace:Render



NeuFace:Normal



NeuFace:Relighting

**Ours**

# Extension to Common Objects

DIFFREC:Diffuse



DIFFREC:Specular



DIFFREC:Render



DIFFREC:Normal



DIFFREC:Relighting



NeuFace:Diffuse



NeuFace:Specular



NeuFace:Render



NeuFace:Normal



NeuFace:Relighting

**Ours**



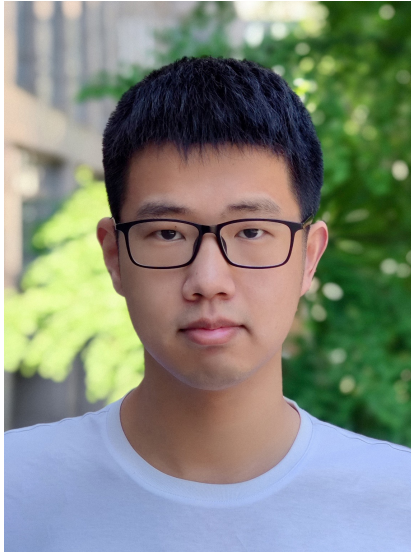


IRIP Laboratory  
<https://irip.buaa.edu.cn>

Code and video are available at  
<https://github.com/aejion/NeuFace>.



# Thank you !



Mingwu Zheng



Haiyu Zhang



Hongyu Yang



Di Huang