

3D Reconstruction and Appearance Modeling for Human Face

Rebirth in Deep Neural Network

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Chen Li

“3D reconstruction is the process of capturing the **geometry** and **appearance** of real objects.”

--from Wikipedia

Outline

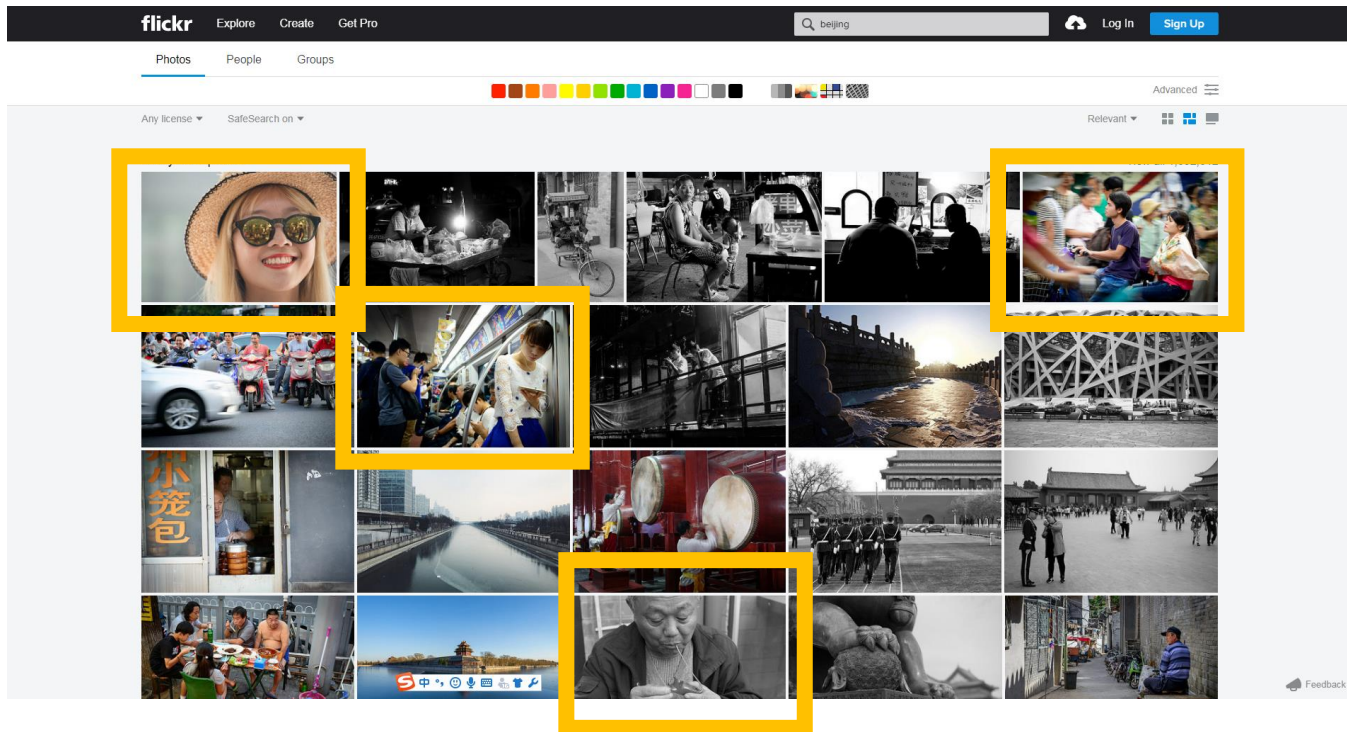
- Why we focus on facial images ?
- Solutions in Industry
- Solutions for single image
 - Geometry
 - Appearance
- Rebirth in Deep Neural Network

Why we focus on facial
images ?



*“Faces exist in more than **60%** of camera phone images.”*

--International Imaging Industry Association



The Curious Case Of
BENJAMIN BUTTON



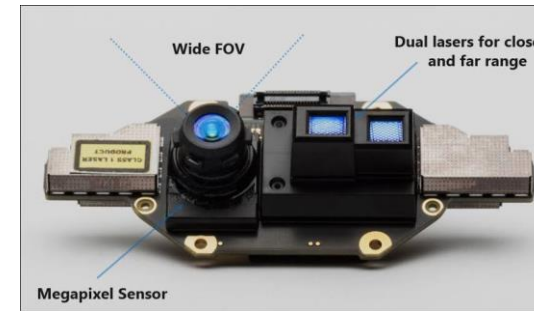
Solutions in Industry

Geometry Reconstruction

- Active methods
 - Structured light (Kinect, iPhone X, *etc.*)
 - Laser range finder
 - Time-of-flight
- Passive methods
 - Multi-view Stereo
 - Shape-from-X (shading, focus/defocus, *etc.*)
 - Data-driven



Kinect 2.0



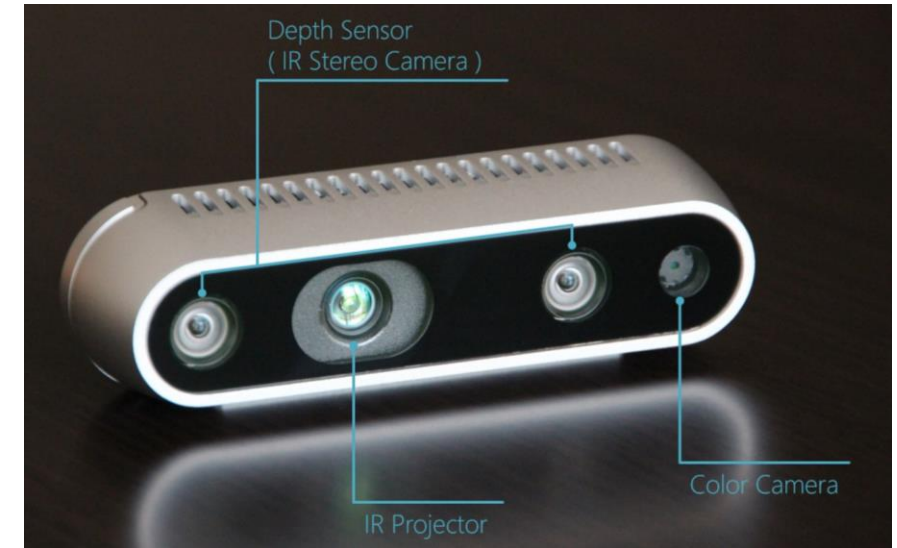
ToF of Microsoft



iPhone X

Geometry Reconstruction

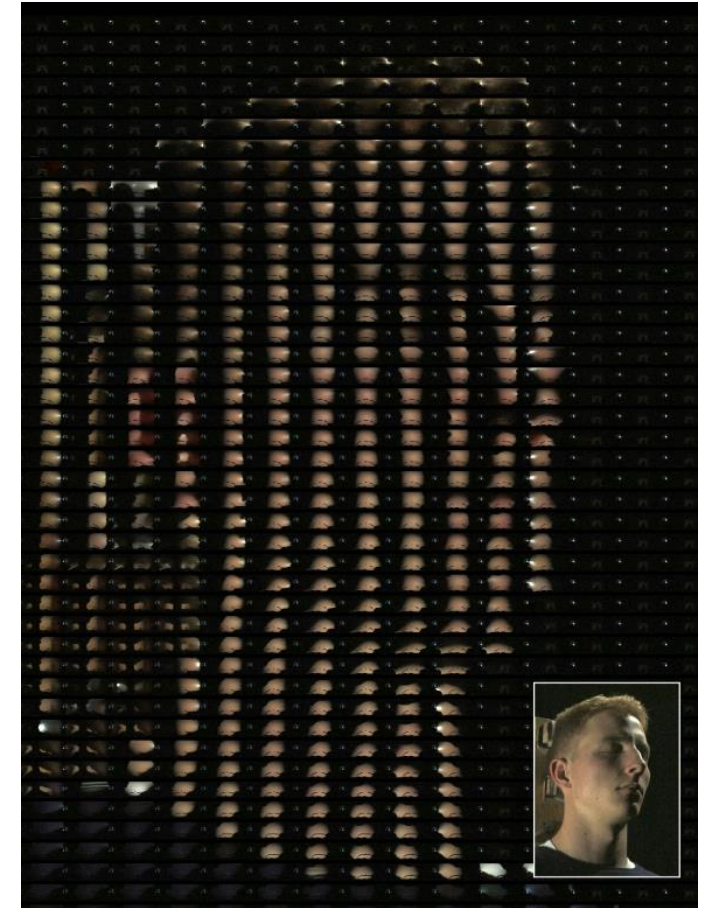
- Active methods
 - Structured light (Kinect, iPhone X, *etc.*)
 - Laser range finder
 - Time-of-flight
- Passive methods
 - Multi-view Stereo
 - Shape-from-X (shading, focus/defocus, silhouettes)
 - Data-driven



Intel RealSense

USC ICT Light Stage Series

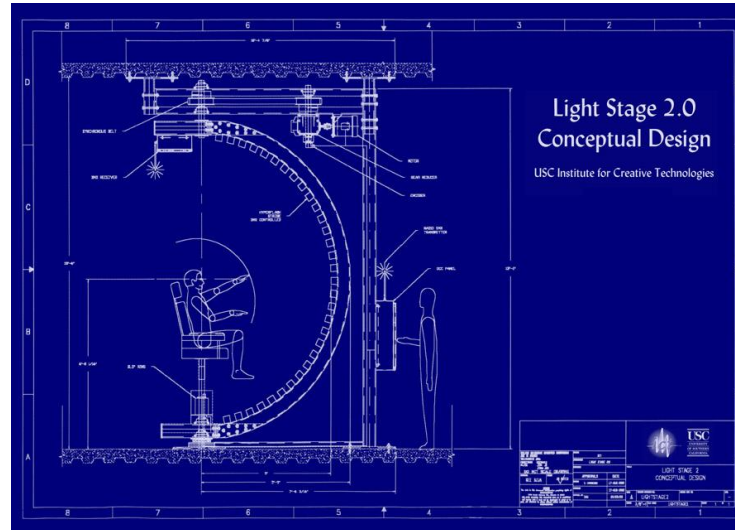
- Light Stage 1
 - Built in 2000
 - One spotlight
 - Takes 1 minute



*Paul Debevec, Tim Hawkins, Chris Tchou, Haarm-Pieter Duiker, Westley Sarokin, and Mark Sagar. 2000. **Acquiring the reflectance field of a human face**. In Proceedings of the 27th annual conference on Computer graphics and interactive techniques (SIGGRAPH '00).*

USC ICT Light Stage Series

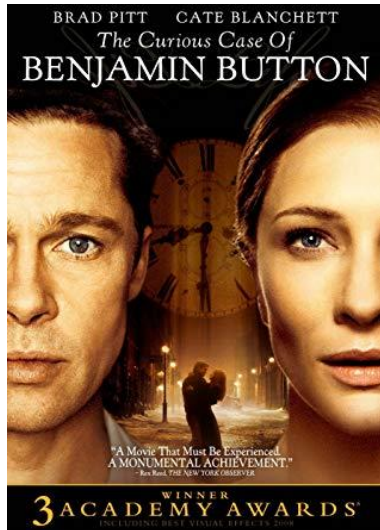
- Light Stage 1
 - Built in 2000
 - One spotlight
 - Takes 1 minute
- Light Stage 2
 - Built in 2002
 - 30 strobe lights
 - 8 seconds



Tim Hawkins, Andreas Wenger, Chris Tchou, Andrew Gardner, Fredrik Göransson, and Paul Debevec. 2004. Animatable facial reflectance fields. In Proceedings of the Fifteenth Eurographics conference on Rendering Techniques (EGSR'04).

USC ICT Light Stage Series

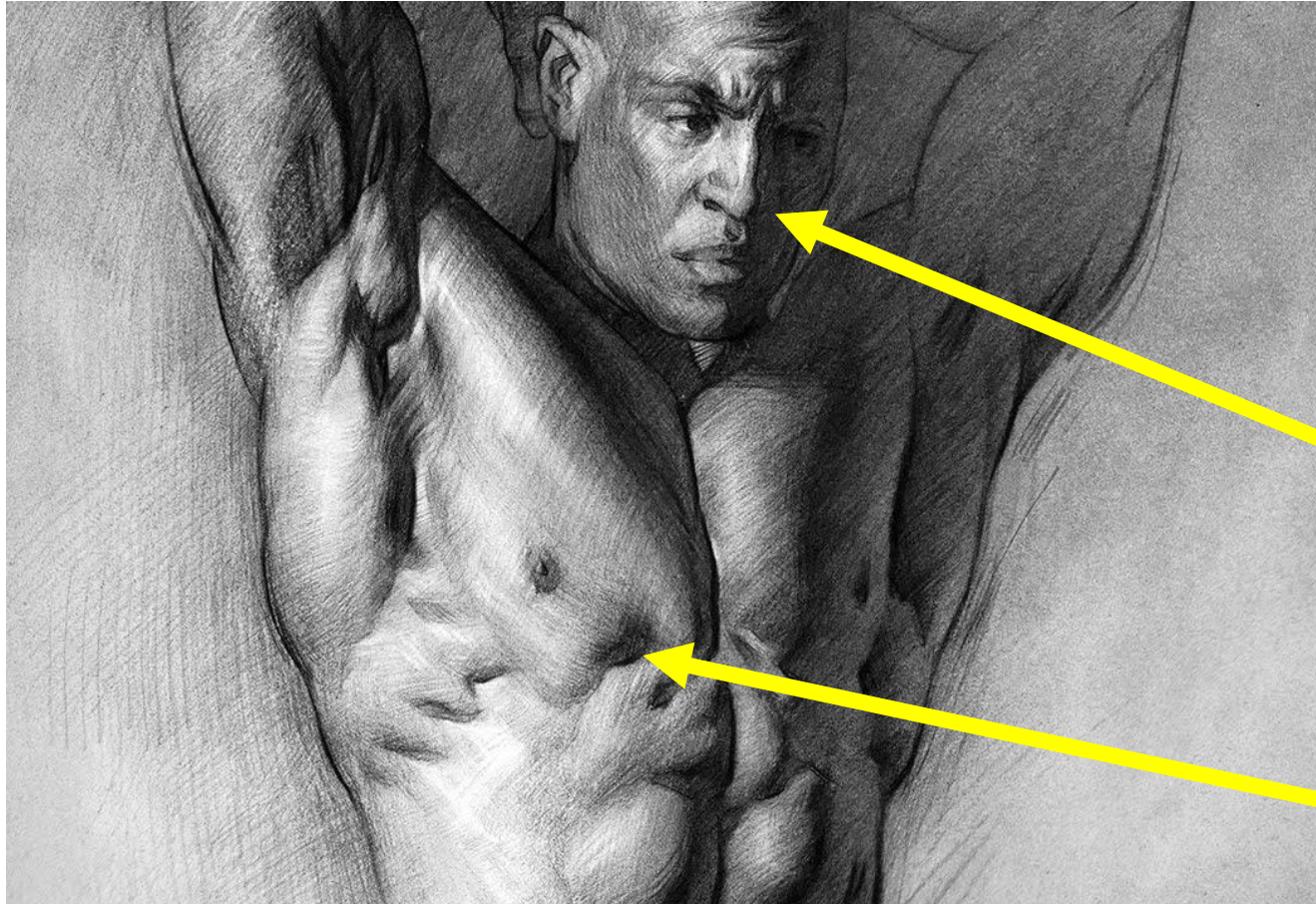
- Light Stage 5
 - Built in 2005
 - 152 lights
 - 24 shot / second



Andreas Wenger, Andrew Gardner, Chris Tchou, Jonas Unger, Tim Hawkins, and Paul Debevec. 2005. Performance relighting and reflectance transformation with time-multiplexed illumination. *ACM Trans. Graph.* 24, 3 (July 2005), 756-764.

Solutions for single image

Analysis by synthesis



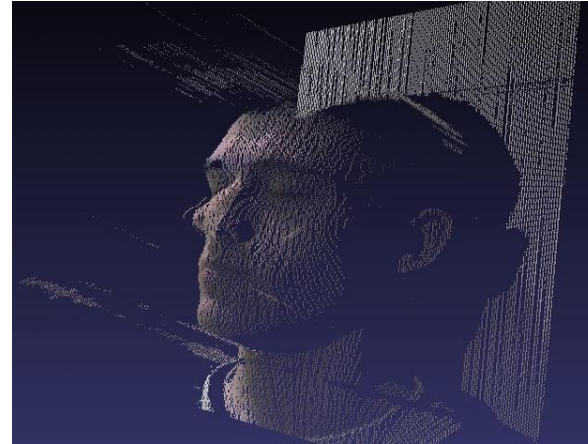
Synthesizing the estimation,
analyzing the accurate.

*Facial
landmarks*

Shading

Problem formulation

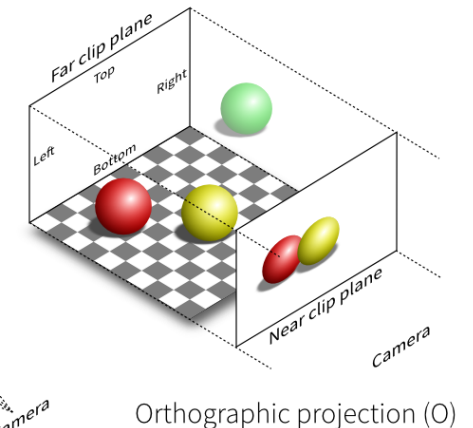
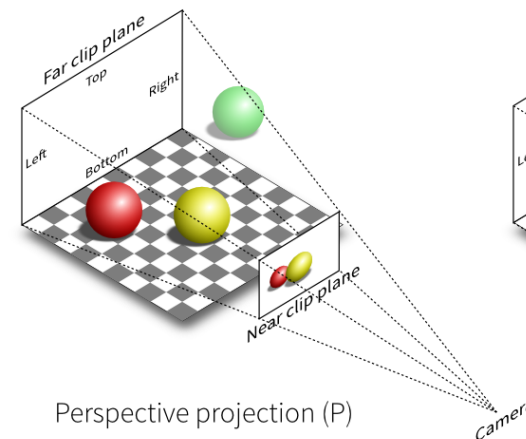
- Geometry representation
 - Point Cloud
 - Depth map/normal map
 - Parametric model
- 3D-to-2D projection
 - Orthographic projection
 - Weakly perspective projection
 - Perspective projection



Point Cloud



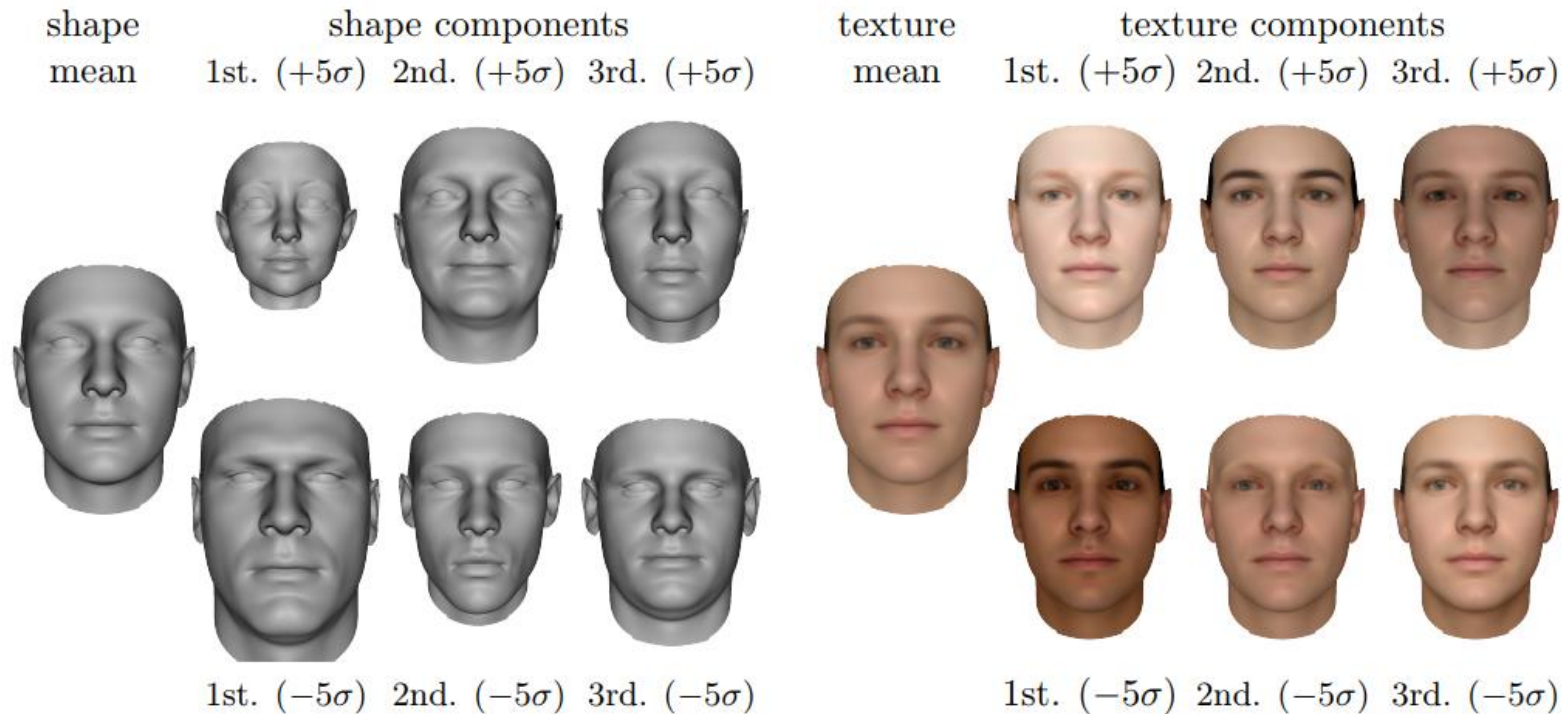
Normal Map



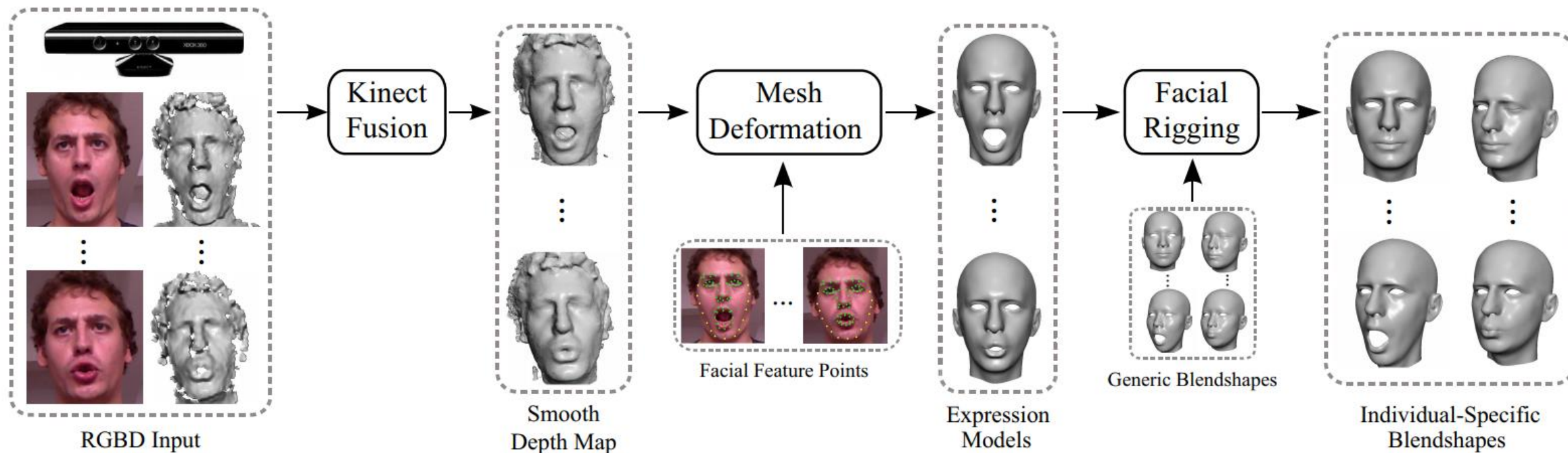
3D Morphable Model

$$G = \bar{G} + U_{id}\alpha_{id} + U_{exp}\alpha_{exp}$$

$$T = \bar{T} + U_{tex}\alpha_{tex}$$



Data Capture



Cao Chen, Yanlin Weng, Shun Zhou, Yiyi Tong, Kun Zhou: "FaceWarehouse: a 3D Facial Expression Database for Visual Computing", *IEEE Transactions on Visualization and Computer Graphics*, 20(3): 413-425, 2014,

Optimization

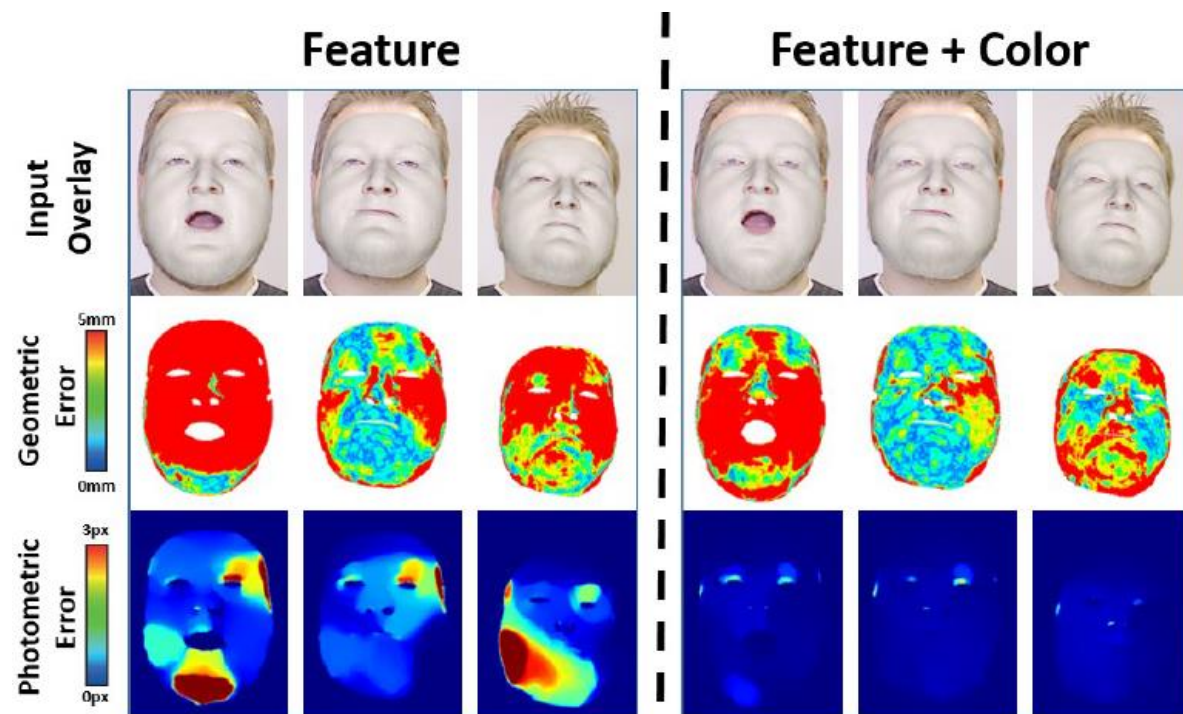
- Target

$$\mathcal{P} = \{\Pi, \alpha_{id}, \alpha_{exp}, \alpha_{tex}\}$$

- Constraints

$$E_{lan} = \sum_f ||\Pi(G(\alpha_{id}, \alpha_{exp})) - f_p||_2^2$$

$$E_{color} = \sum_p ||\mathbf{T}(\alpha_{tex}) - I(\Pi(G(\alpha_{id}, \alpha_{exp})))||_2^2$$

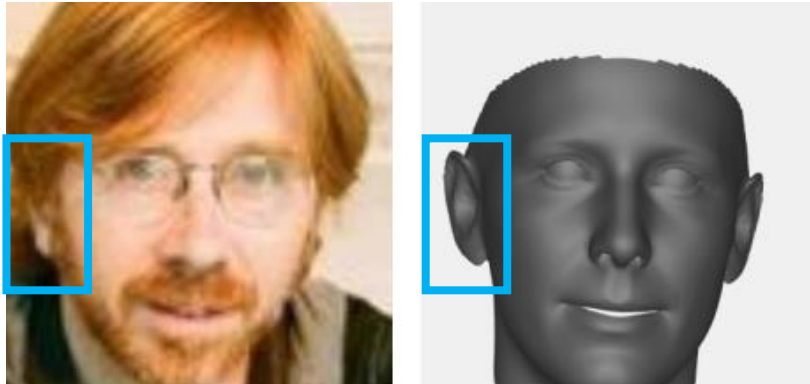


Zollhofer, M. and Thies, J. and Garriod, P. and Bradely, D. and Beeler, T. and Perez, P. and Stamminger, M. and Niesner, M. and Theobalt, C.. 2018. State of the Art on Monocular 3D Face Reconstruction, Tracking, and Applications. Computer Graphics Forum (Eurographics State of the Art Reports 2018).

3D-to-2D Project

Weakly perspective projection

$$p = f * R * v + T_{xy}$$



Tran, et.al. 2018

Perspective projection

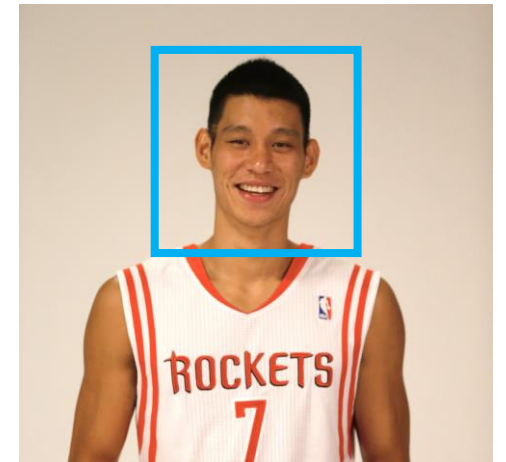
$$v' = R * v + T$$

$$v'' = \left(\frac{v'_x}{v'_z}, \frac{v'_y}{v'_z}, 1 \right)$$

$$p = \begin{bmatrix} f_x & 0 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix} v''$$



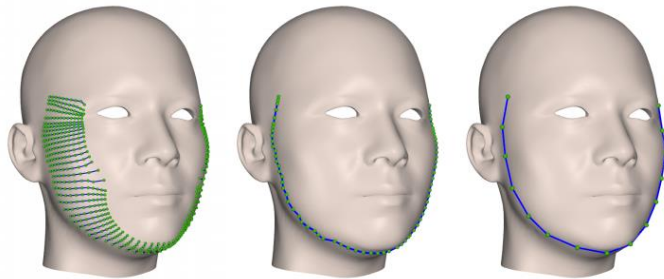
Tewari, et.al. 2018



Face Contour Handling

Dense contour hypothesis

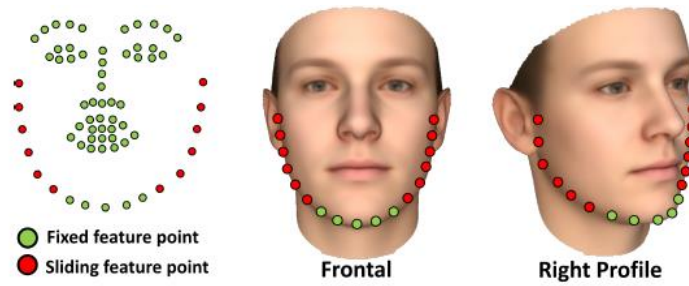
- Pre-defined contour lines
- The computation of visibility is costly



Cao, et.al. 2014

Sliding feature points

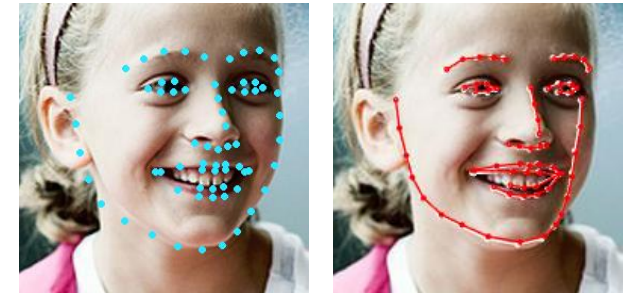
- Iterative updated sliding feature points



Tewari, et.al. 2018

3D landmarks

- Training data is limited
- Hard to capture complex expression



Bulat, et.al. 2017

How about the facial details?

Shape-from-shading

- An intuition

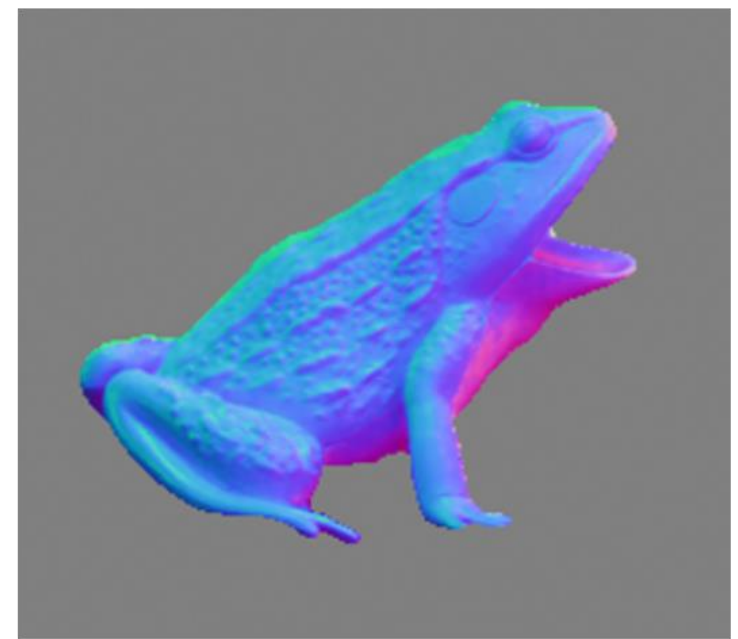
Calibrated illumination



Shading image



Estimated normal map



M. K. Johnson and E. H. Adelson, "Shape estimation in natural illumination," CVPR 2011, Colorado Springs, CO, USA, 2011, pp. 2553-2560.

Formulation

- Distant light

$$I = A \cdot N \cdot L$$

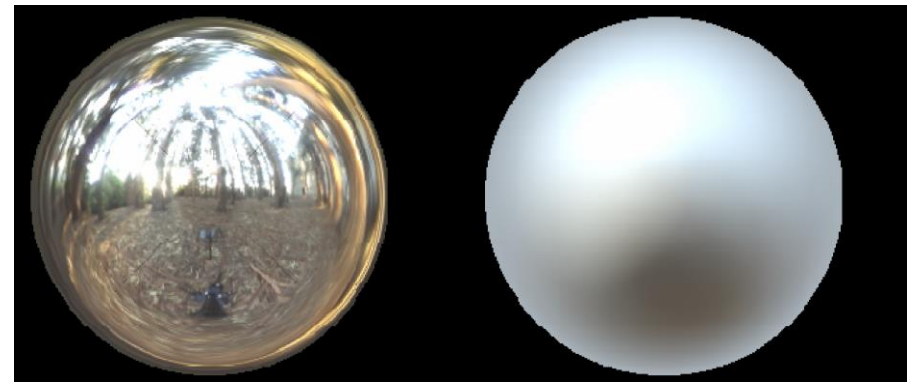
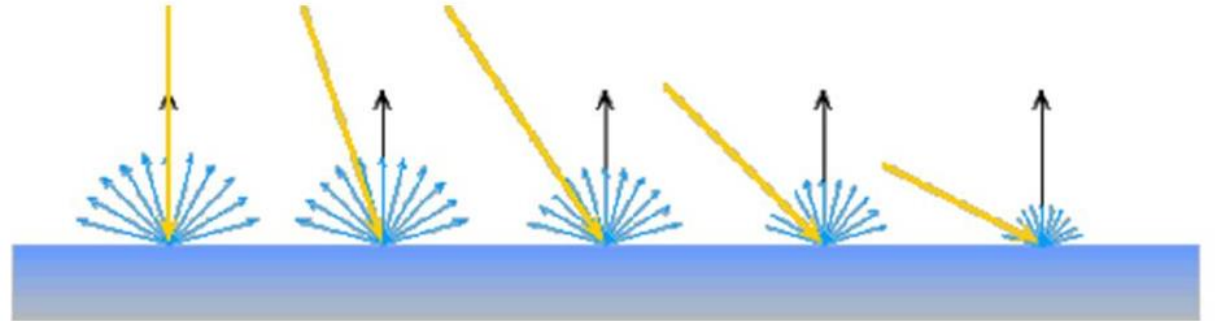
- Parametric illumination

$$L(\theta, \phi) = \sum_{l,m} L_{lm} Y_{lm}(\theta, \phi)$$

$$I = A \cdot N^T \cdot M \cdot N$$

Lambert cosine law

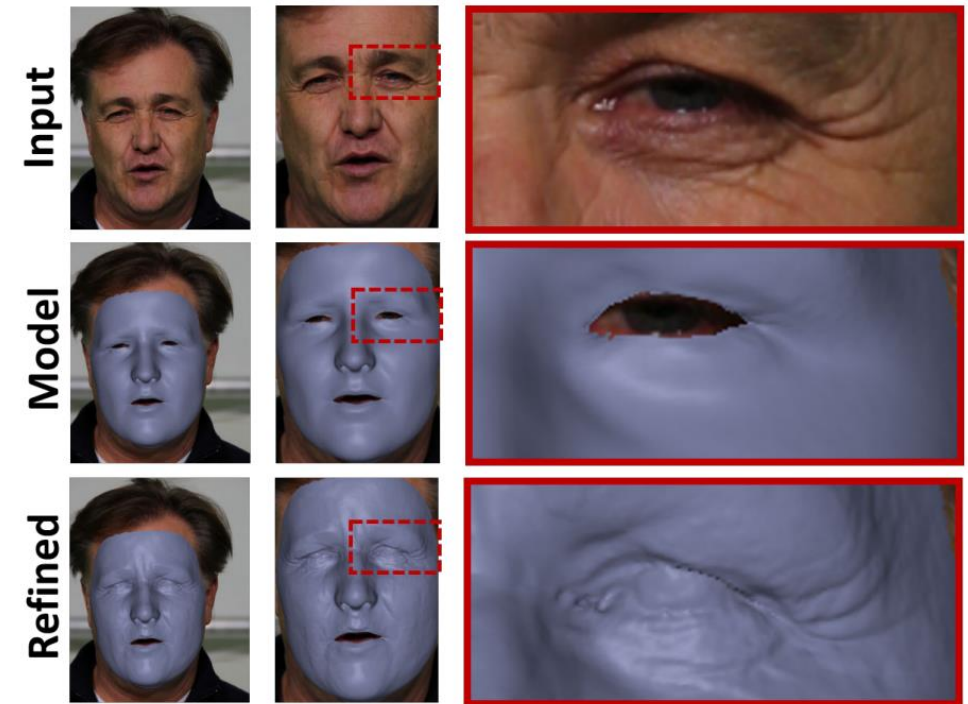
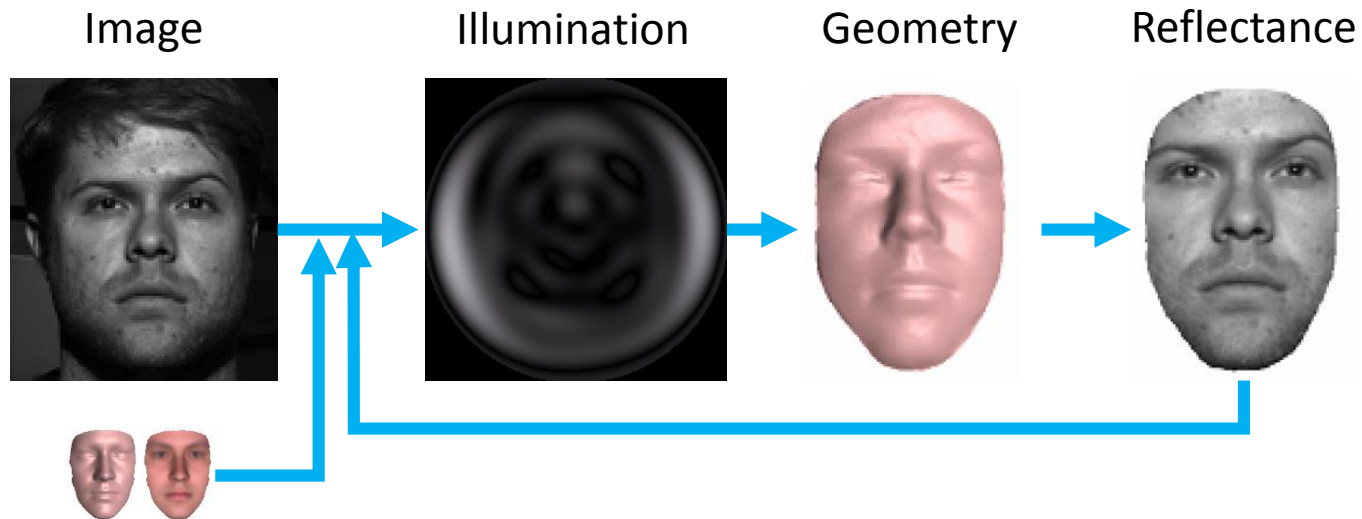
The **radiant intensity** observed from an ideal diffuse diffusely reflecting surface is **directly proportional** to the **cosine** of the angle between the direction of the **incident light** and the **surface normal**.



Ravi Ramamoorthi and Pat Hanrahan. 2001. An efficient representation for irradiance environment maps. In Proceedings of the 28th annual conference on Computer graphics and interactive techniques (SIGGRAPH '01).

A solution for face

$$I = A \cdot N^T \cdot M \cdot N$$

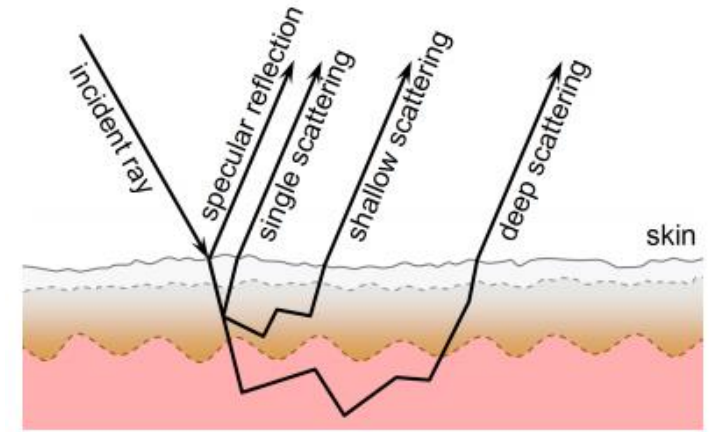


I. Kemelmacher-Shlizerman and R. Basri, "3D Face Reconstruction from a Single Image Using a Single Reference Face Shape," in IEEE Transactions on Pattern Analysis and Machine Intelligence, vol. 33, no. 2, pp. 394-405, Feb. 2011.

Garrido, et.al. 2016

Is face really Lambertian?

Face Appearance Modeling



Specular
Reflection

Single
Scattering

Shallow
Scattering

Deep
Scattering

Rendering

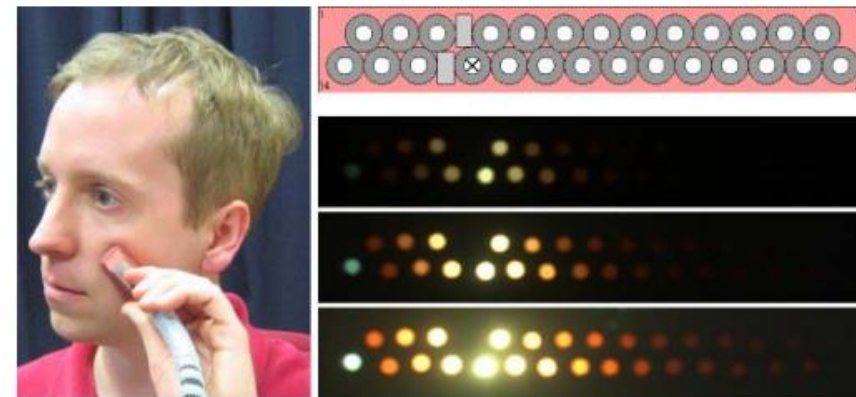
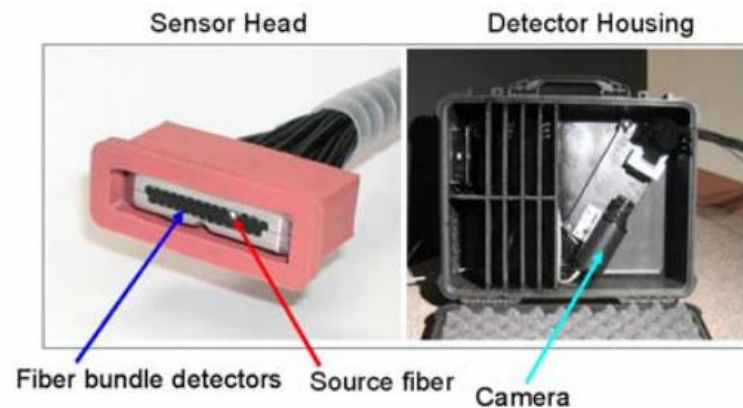
Photograph

Abhijeet Ghosh, Tim Hawkins, Pieter Peers, Sune Frederiksen, and Paul Debevec. 2008. Practical modeling and acquisition of layered facial reflectance. In ACM SIGGRAPH Asia 2008 papers (SIGGRAPH Asia '08).

Face Appearance Capture



<https://docs.sharktacos.com/photography/xpol.html>



Weyrich, et.al. 2006

Specular separation

$$I = A \cdot N^T \cdot M \cdot N$$

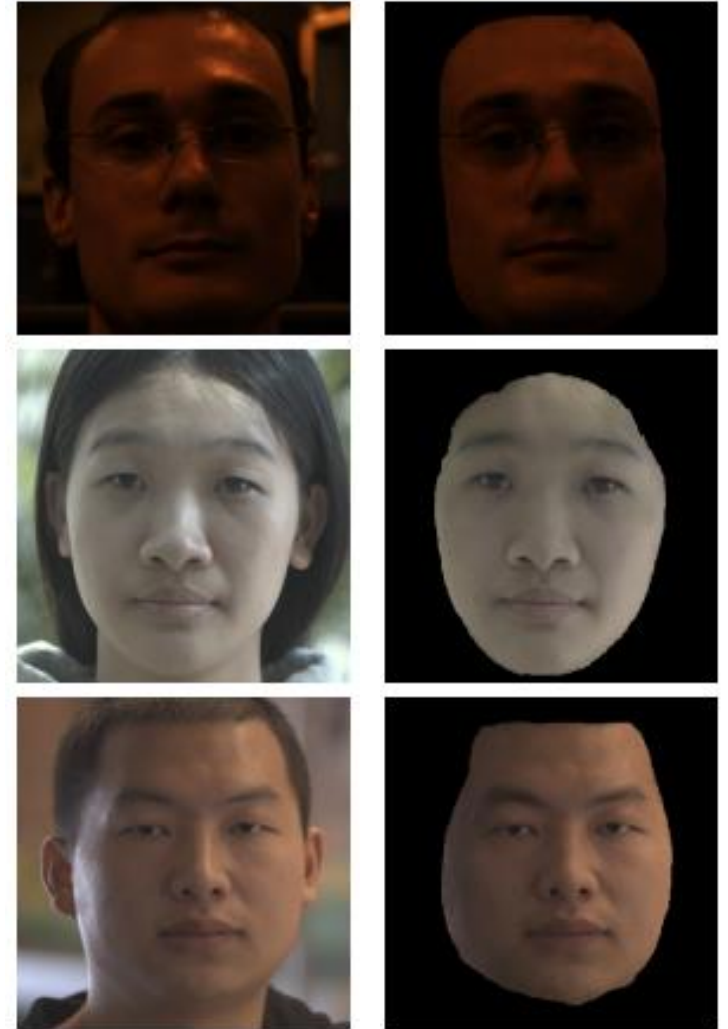
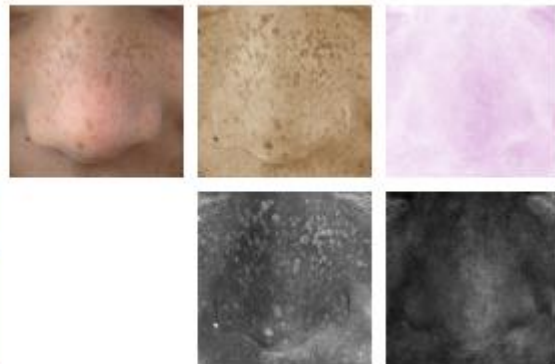
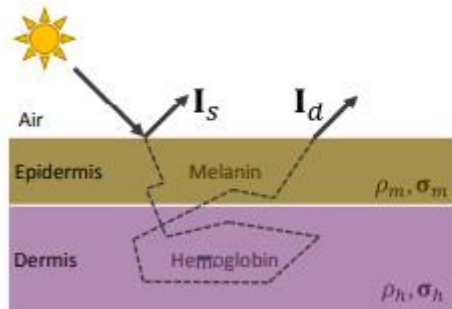


$$I = A \cdot N^T \cdot M \cdot N + S$$

$$A = \sigma_m^{\rho_m} \sigma_h^{\rho_h}$$

Low-frequency
illumination

High-frequency
illumination



C. Li, S. Lin, K. Zhou and K. Ikeuchi, "Specular Highlight Removal in Facial Images," 2017 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), Honolulu, HI, 2017, pp. 2780-2789.

Subsurface scattering

$$I = A \cdot N^T \cdot M \cdot N + S$$



$$I = BSSRDF + S$$

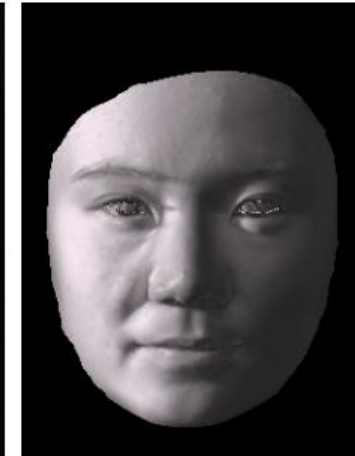
Lambertian

BSSRDF



Lambertian

BSSRDF



Li C., Zhou K., Lin S. (2014) *Intrinsic Face Image Decomposition with Human Face Priors*. In: Fleet D., Pajdla T., Schiele B., Tuytelaars T. (eds) *Computer Vision – ECCV 2014*. ECCV 2014. Lecture Notes in Computer Science, vol 8693. Springer, Cham

Subsurface scattering

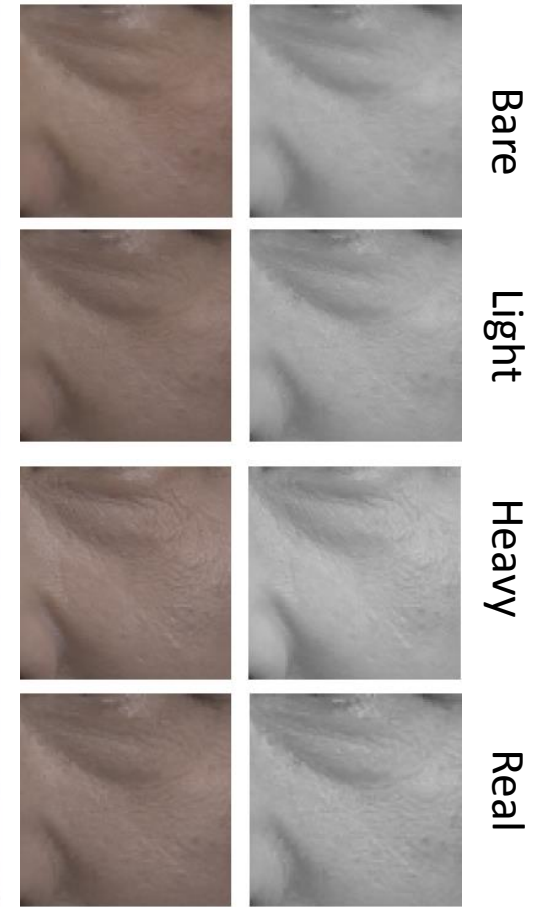
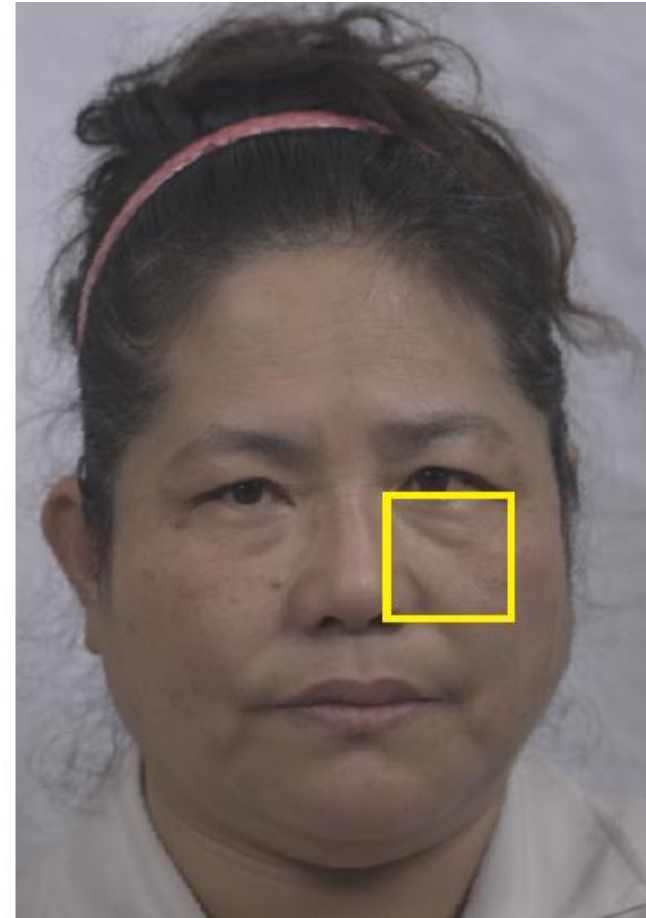
$$I = A \cdot N^T \cdot M \cdot N + S$$



$$I = BSSRDF + S$$

Lambertian

BSSRDF



C. Li, K. Zhou and S. Lin, "Simulating makeup through physics-based manipulation of intrinsic image layers," 2015 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), Boston, MA, 2015, pp. 4621-4629.

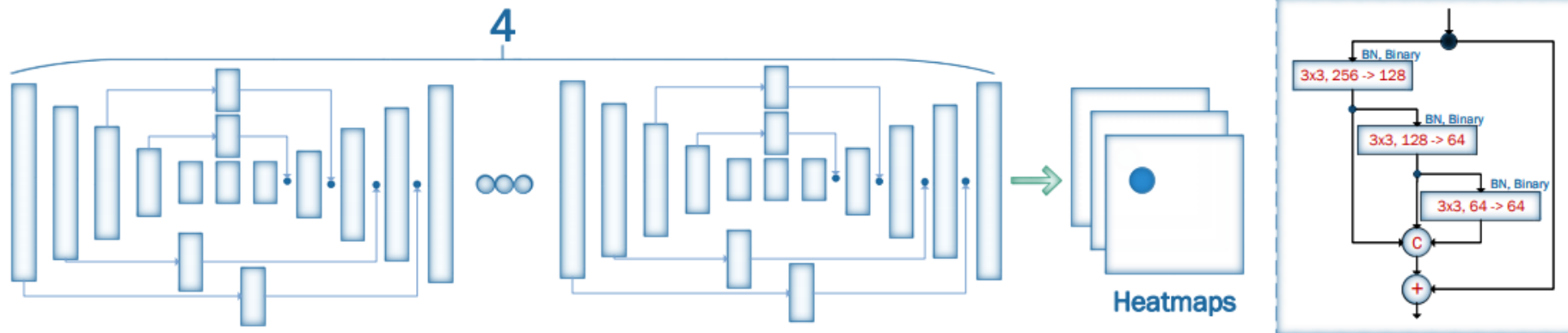
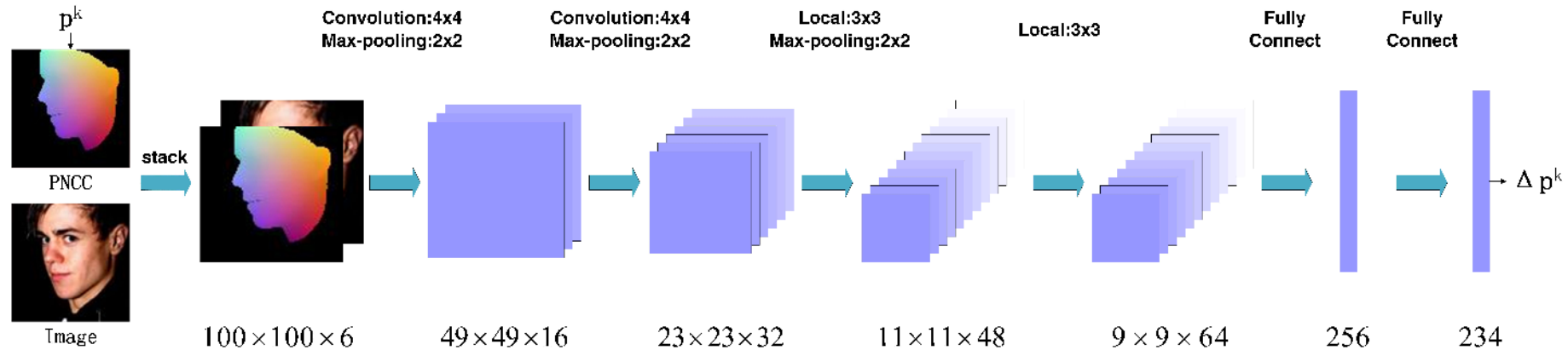
Rebirth in Deep Neural Network

Rebirth in Deep Neural Network

- Frontal face to arbitrary poses
- Occlusion handling
- Separation of training and testing
- Synthetic data + unsupervised real images

Frontal face to arbitrary poses

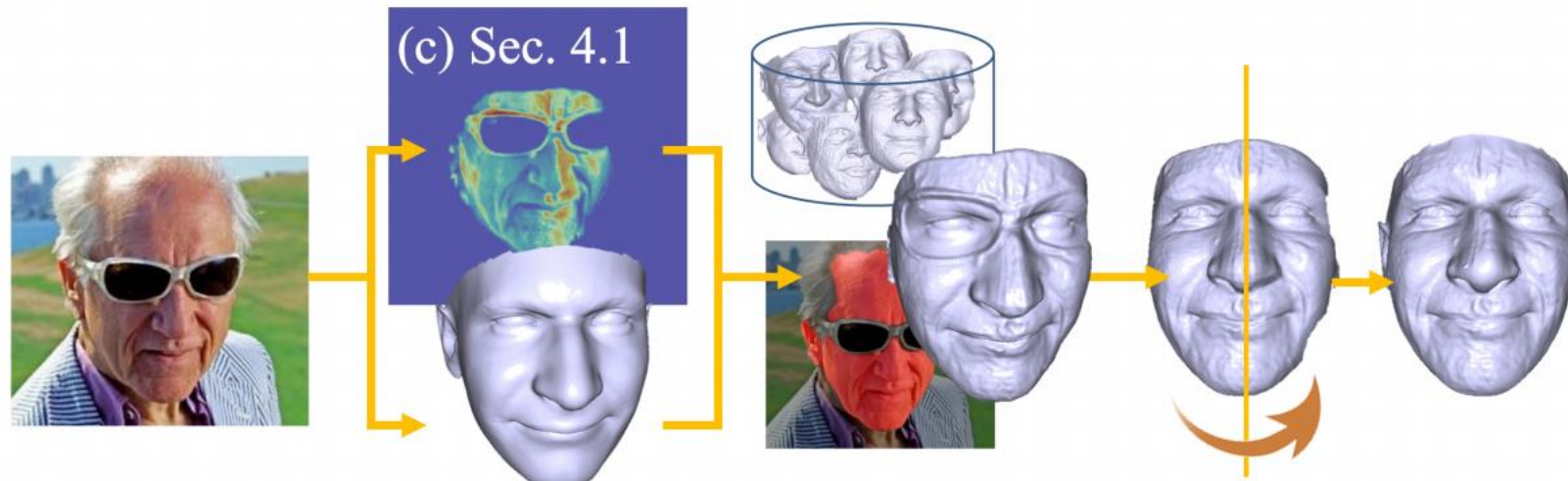
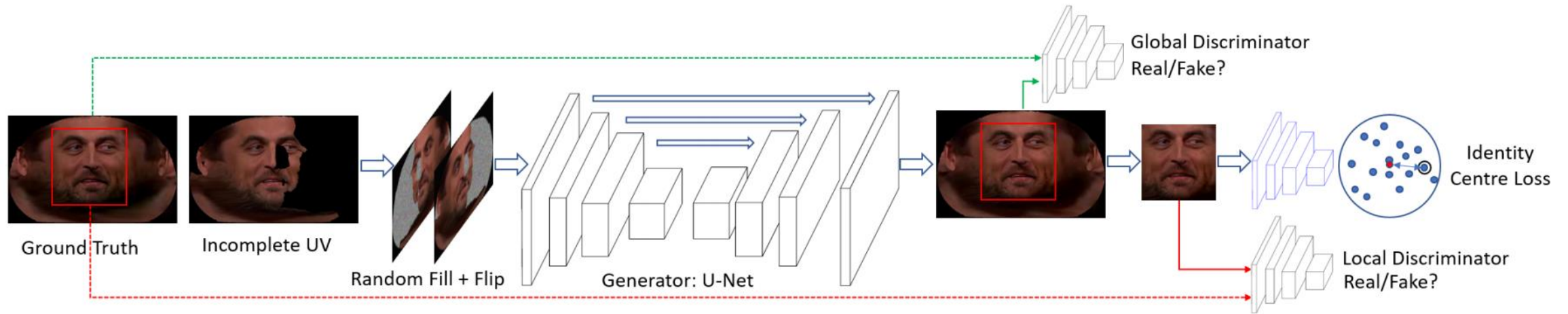
X. Zhu, Z. Lei, X. Liu, H. Shi and S. Z. Li, "Face Alignment Across Large Poses: A 3D Solution," 2016 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), Las Vegas, NV, 2016, pp. 146-155.



A. Bulat and G. Tzimiropoulos, "How Far are We from Solving the 2D & 3D Face Alignment Problem? (and a Dataset of 230,000 3D Facial Landmarks)," 2017 IEEE International Conference on Computer Vision (ICCV), Venice, 2017, pp. 1021-1030.

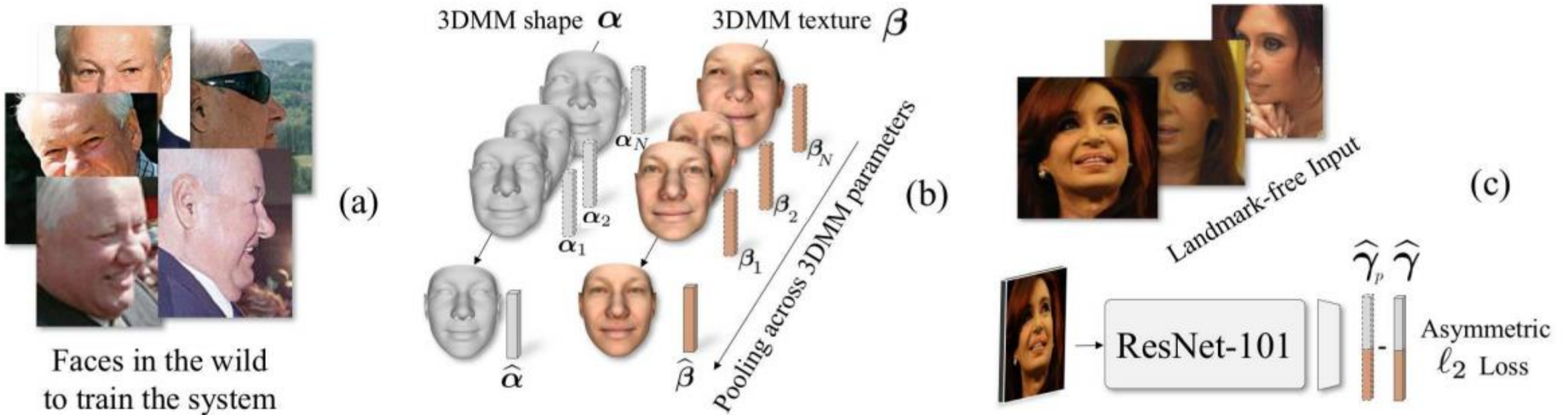
Occlusion handling

Deng, Jiankang and Cheng, Shiyang and Xue, Niannan and Zhou, Yuxiang and Zafeiriou, Stefanos, "UV-GAN: Adversarial Facial UV Map Completion for Pose-Invariant Face Recognition," 2018 IEEE Conference on Computer Vision and Pattern Recognition (CVPR).



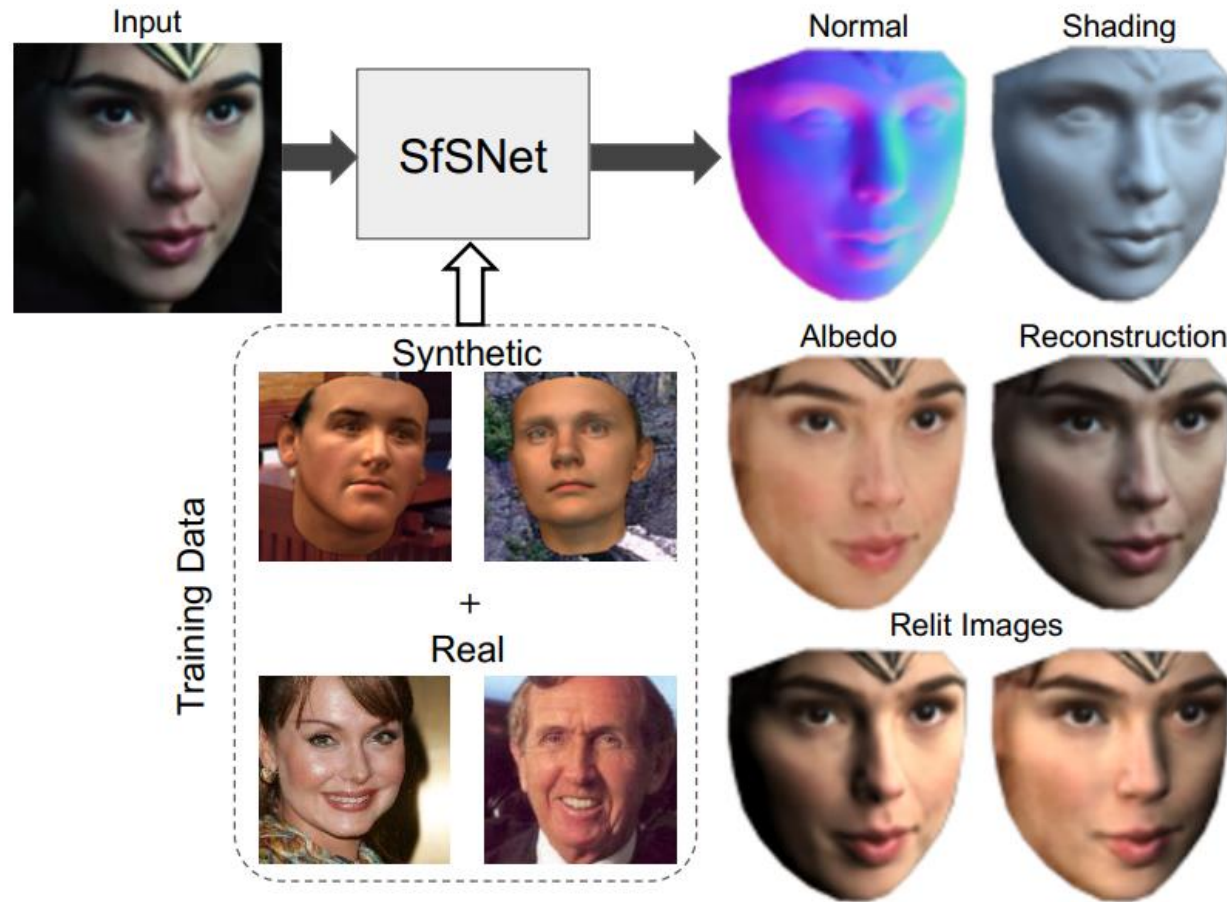
Anh Tuấn Trần, Tal Hassner, Iacopo Masi, Eran Paz, Yuval Nirkin, Gérard Medioni, "Extreme 3D Face Reconstruction: Seeing Through Occlusions," 2018 IEEE Conference on Computer Vision and Pattern Recognition (CVPR).

Separation of training and testing



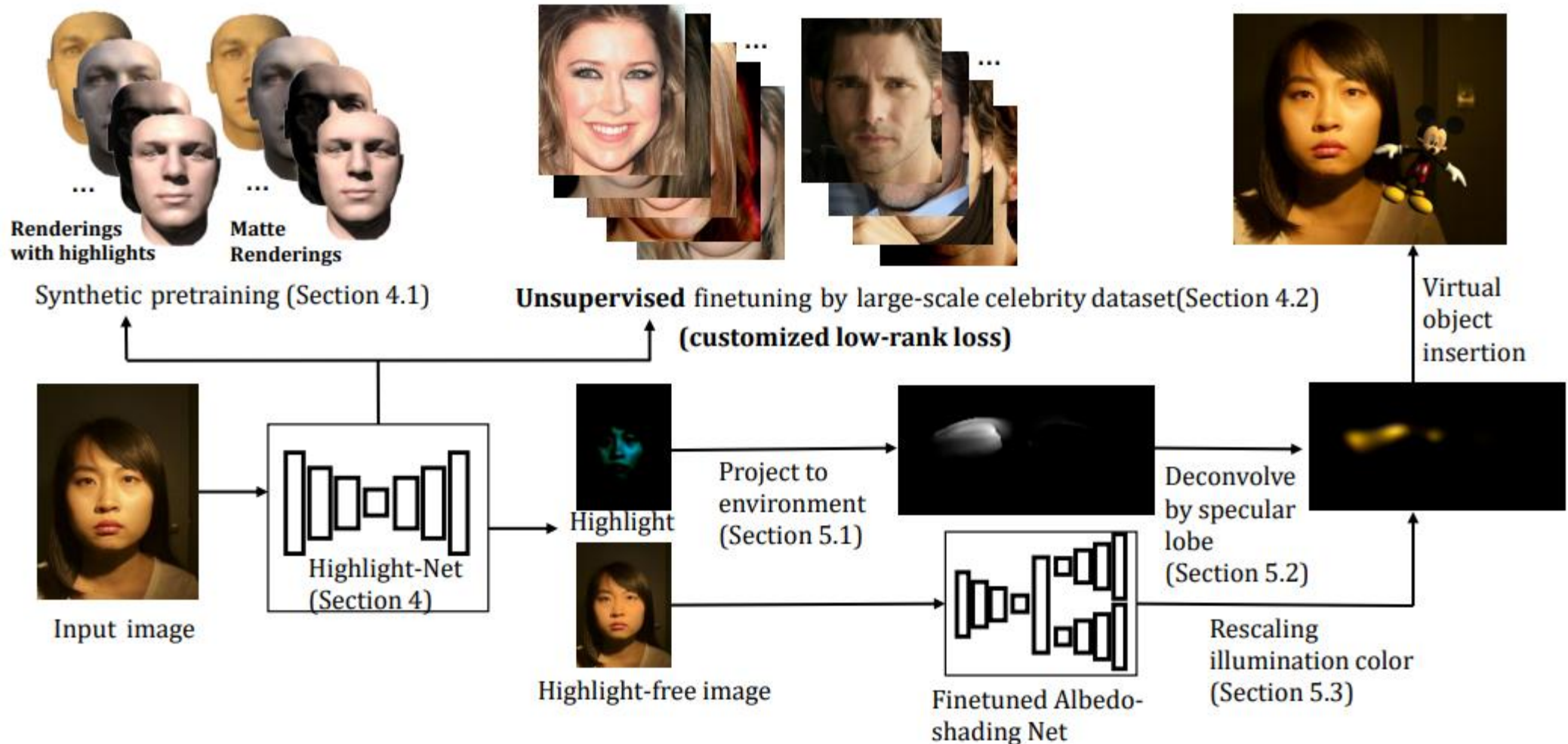
A. T. Tran, T. Hassner, I. Masi and G. Medioni, "Regressing Robust and Discriminative 3D Morphable Models with a Very Deep Neural Network," 2017 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), Honolulu, HI, 2017, pp. 1493-1502.

Synthetic data + unsupervised real images



Soumyadip Sengupta and Angjoo Kanazawa and Carlos D. Castillo and David W. Jacobs, "SfSNet: Learning Shape, Reflectance and Illuminance of Faces in the Wild," 2018 IEEE Conference on Computer Vision and Pattern Recognition (CVPR).

Synthetic data + unsupervised real images



Yi, Renjiao and Zhu, Chenyang and Tan, Ping and Lin, Stephen, "Faces as Lighting Probes via Unsupervised Deep Highlight Extraction," 2018 The European Conference on Computer Vision (ECCV).

Good Reference

- ZJU: Kun Zhou (<http://kunzhou.net/>)
- USC: Hao Li (<https://www.hao-li.com>)
- MSU: Xiaoming Liu (<http://cvlab.cse.msu.edu/>)
- MPI: GVV (<http://gvv.mpi-inf.mpg.de/>)