

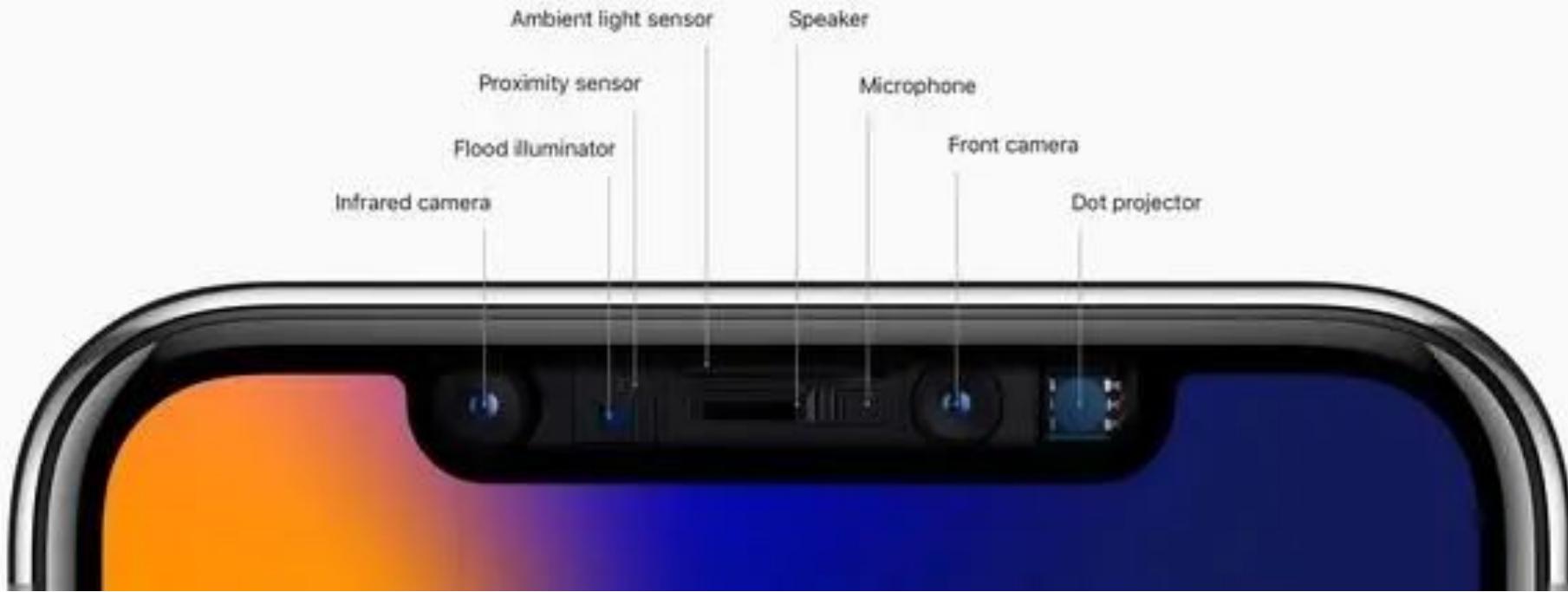
# **3D Face Reconstruction using iPhone's True Depth Camera**

Kesav Ravichandran, Utkarsh Gupta, Yunqian Cheng

# Outline

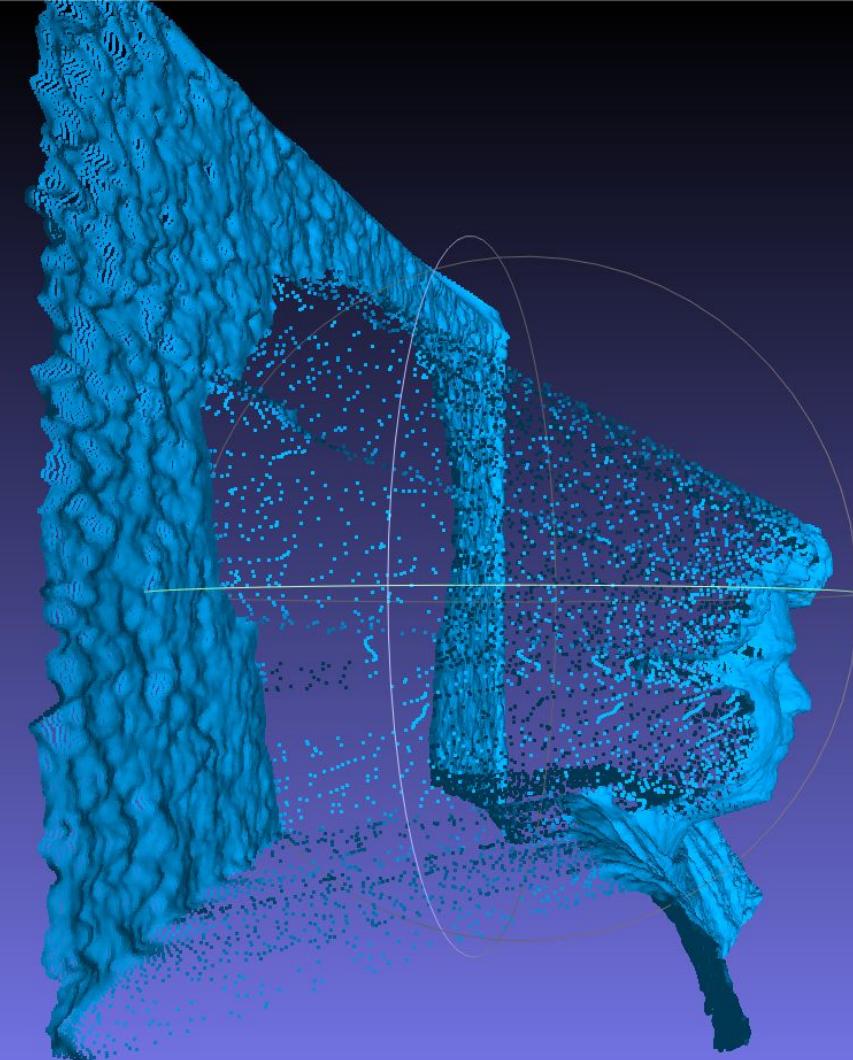
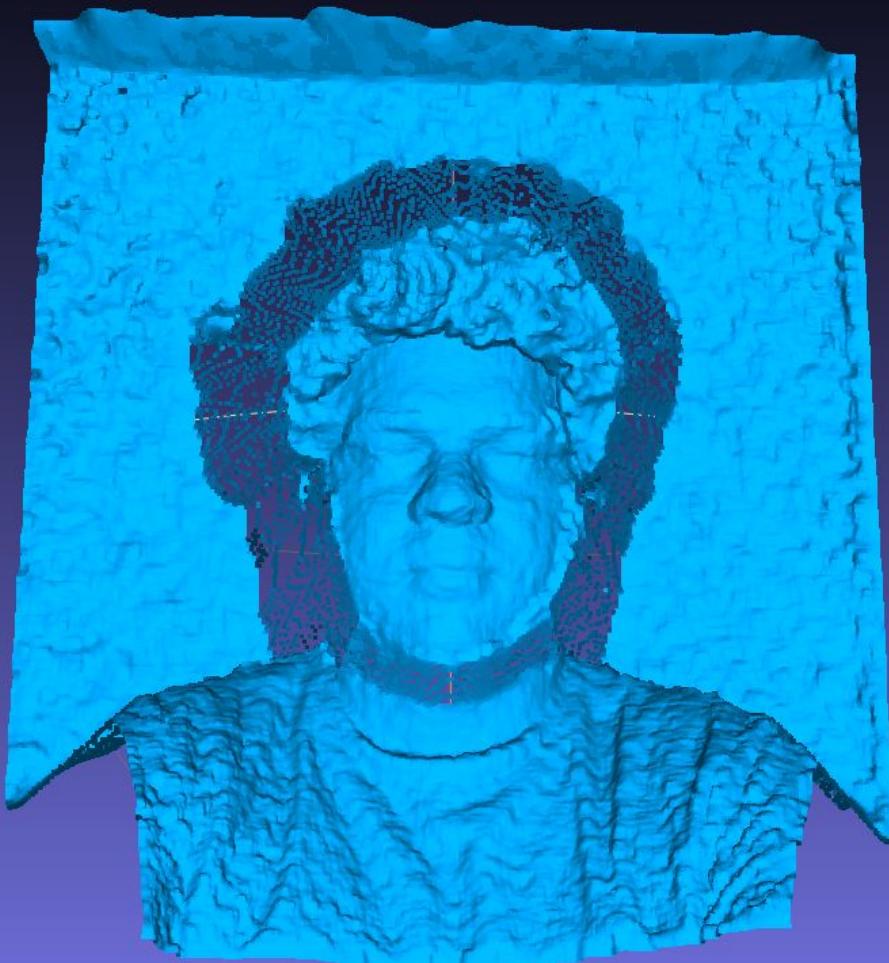
- Develop an iOS application that captures a 3D point cloud of the user's face using the iPhone's front-facing True Depth camera.
- Fit the FLAME model to the data to create a 'complete' face model with only one view.
- Compare results with baseline models that uses RGB images from NoW challenge (MICA and DECA) mainly in low lighting/pitch-dark conditions.

# **Working of TrueDepth Sensor**

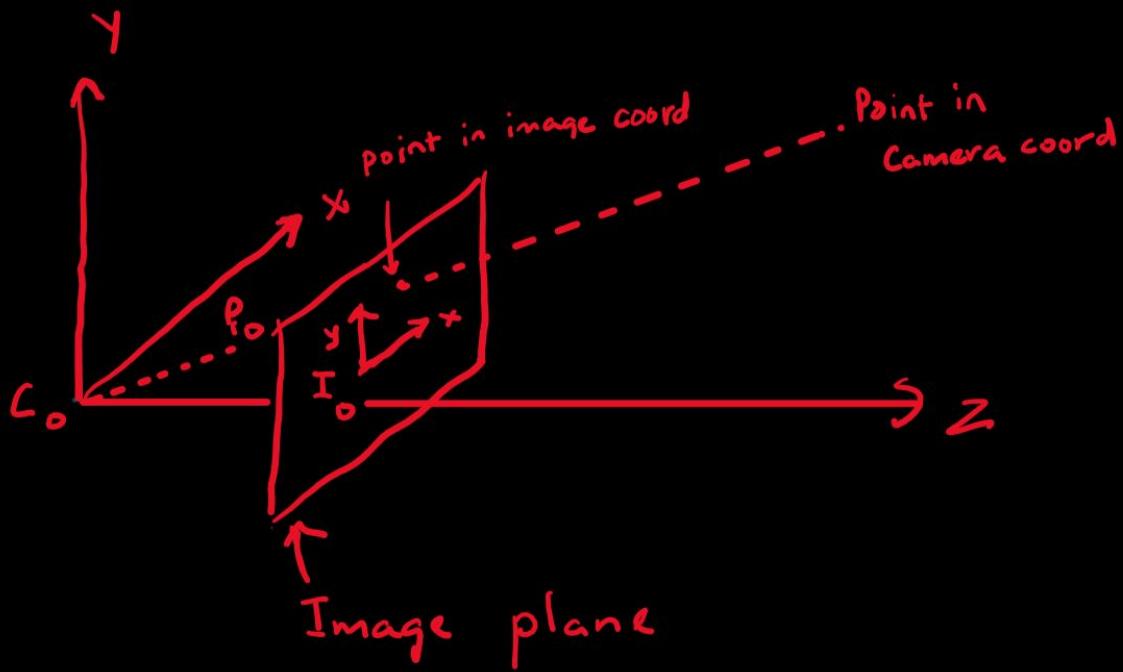


A black and white photograph of a man from the chest up. He is wearing a light-colored, long-sleeved shirt and has short hair. He is holding a dark-colored smartphone in his right hand, which is positioned in front of him at waist level. The background is a dark, textured wall with a faint, large grid or mesh pattern. The lighting is dramatic, coming from the front and slightly above, creating highlights on his face and hands while leaving much of the scene in shadow.

TECH  
INSIDER



# **COORDINATE SYSTEMS**



$C_0$  - Camera Origin.

$I_0$  - Image / Camera Sensor Origin.

$P_0$  - Pixel Origin  $(0,0)$  Top left.

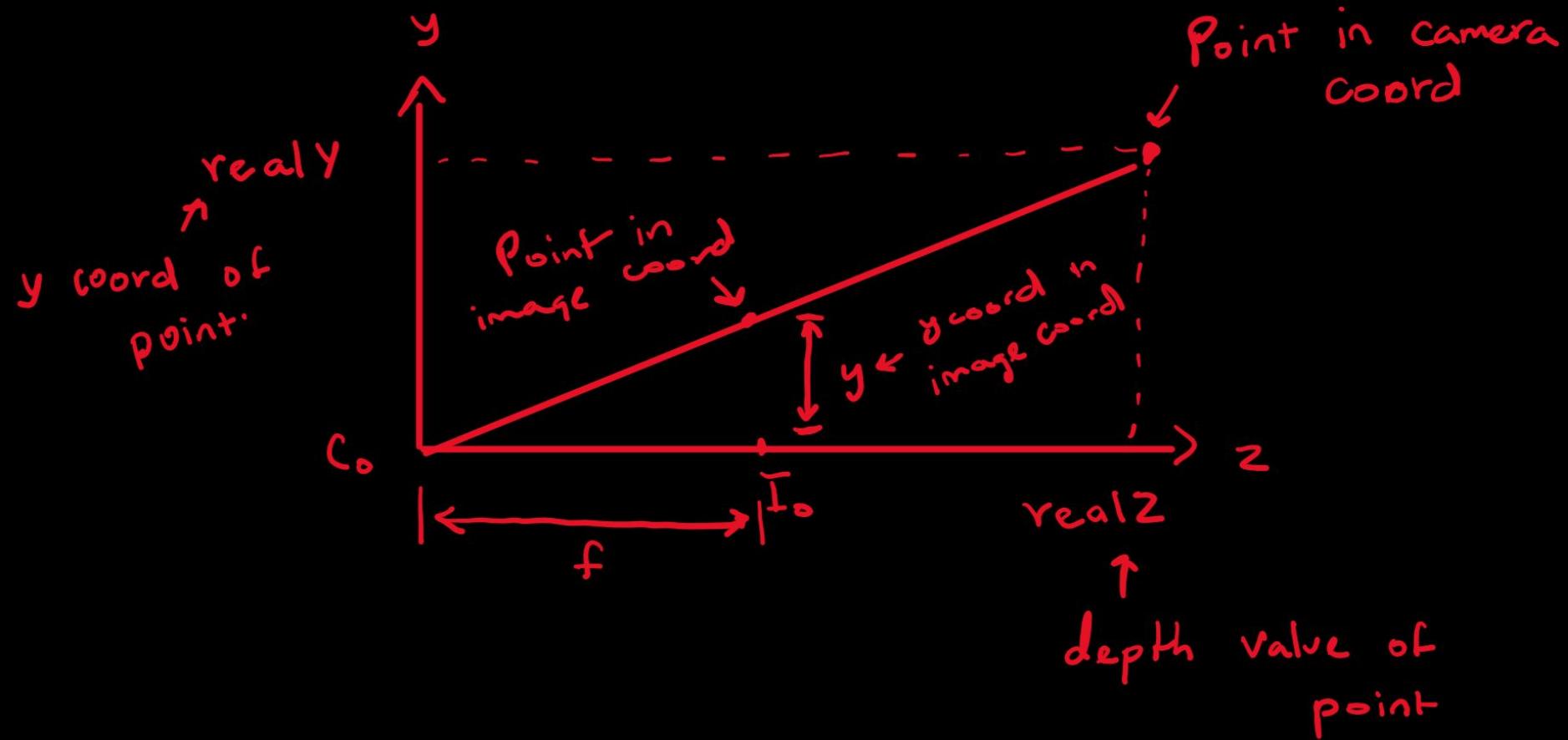
# **CAMERA CALIBRATION (AVFoundation API)**

$$\begin{bmatrix} f_x & 0 & \delta_x \\ 0 & f_y & \delta_y \\ 0 & 0 & 1 \end{bmatrix}$$

$\delta_x$  → offset of center in image coordinate ( $I_o$ )  
 $\delta_y$  → from center in pixel coordinate

$f_x$   
 $f_y$  → Pixel focal length of camera.

# **COORDINATE SYSTEM CONVERSION**



offset of  $I_o$  from pixel (0,0)



$$y_i = y_p \times d_y - O_y$$



dimension of image for scale

offset of  $I_o$  from pixel (0,0)



$$x_i = x_p \times d_x - O_x$$



dimension of image for scale



Pixel Coordinates  $(x_p, y_p)$   $\rightarrow$  Image Coordinates  $(x_i, y_i)$

Image Coordinates  $(x_i, y_i)$   $\rightarrow$  Camera Coordinates  $(\text{real}X, \text{real}Y, \text{real}Z)$

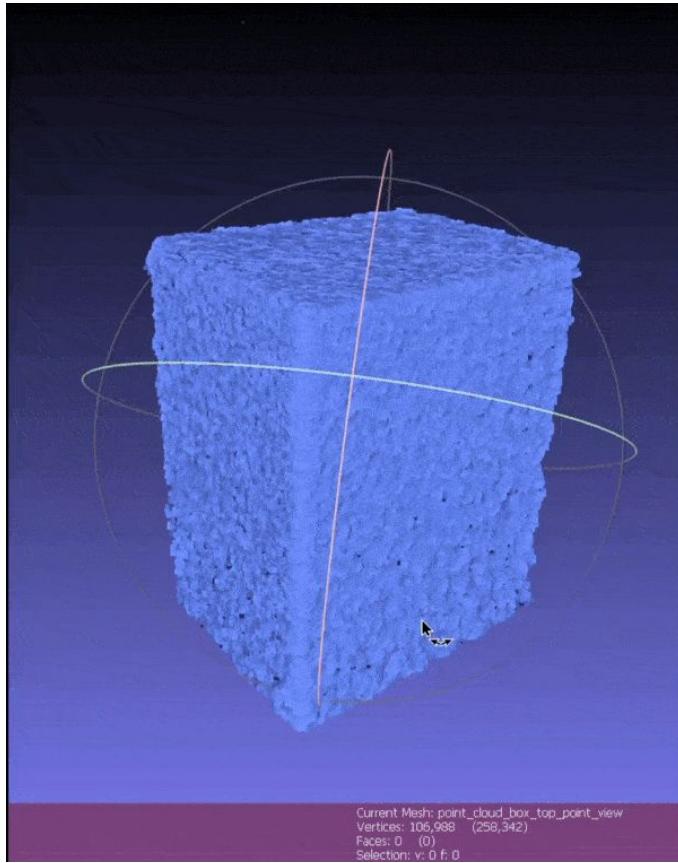
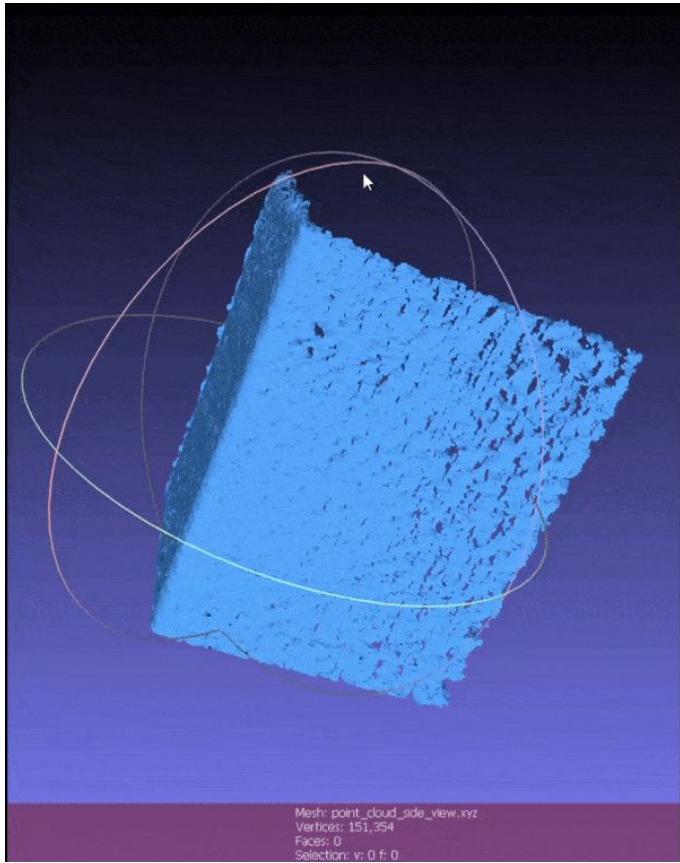


$$\frac{y_i}{f} = \frac{\text{real}Y}{\text{real}Z} \therefore \text{real}Y = \frac{y_i}{f} \times \text{real}Z$$

Similarly,

$$\frac{x_i}{f} = \frac{\text{real}X}{\text{real}Z} \therefore \text{real}X = \frac{x_i}{f} \times \text{real}Z$$

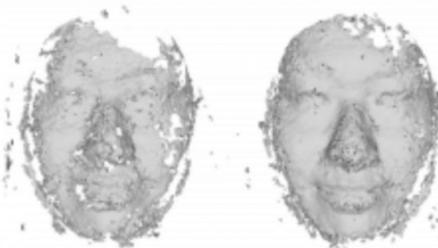
# **RAW DEPTH RESULTS**



# Comparison of results with Stereo



Input images



2-view Stereo



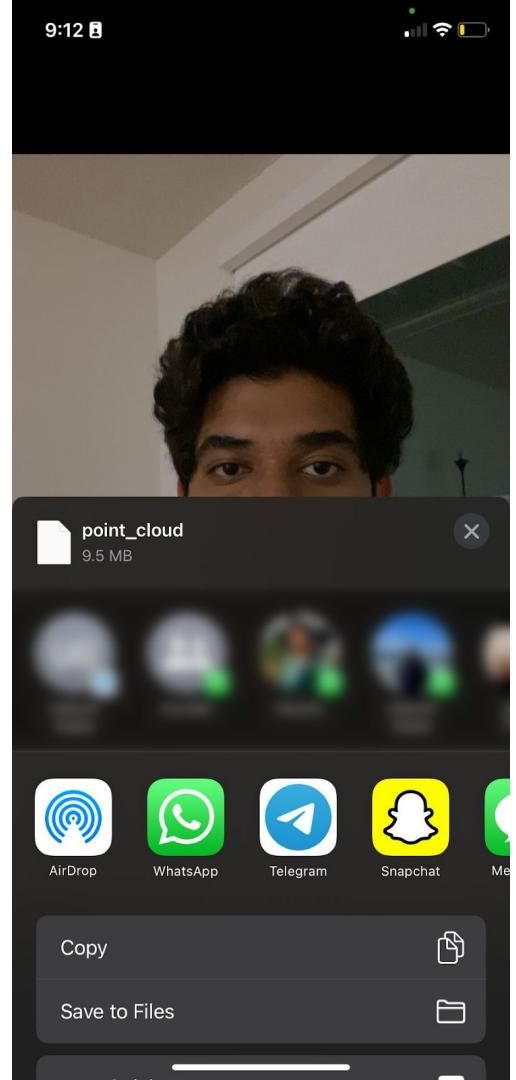
5-view stereo



# APP SCREENSHOTS



← Preview and Capture Scene

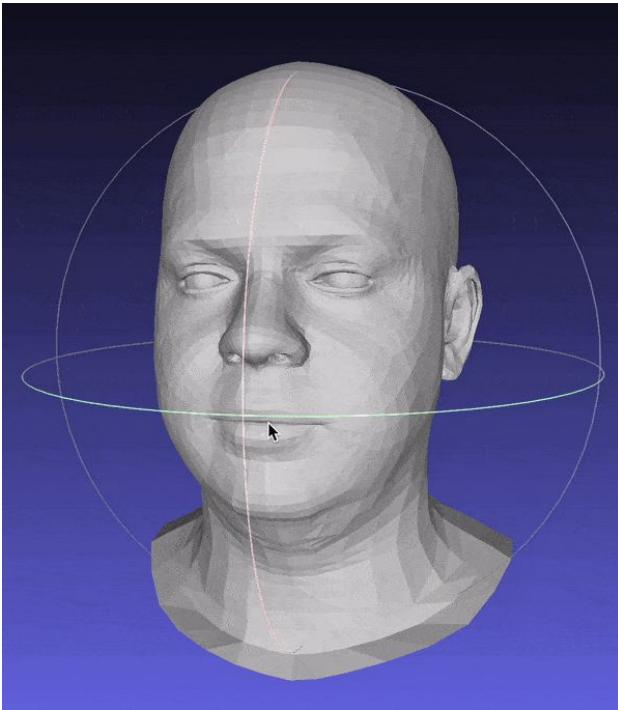


Share Menu for Point Cloud →

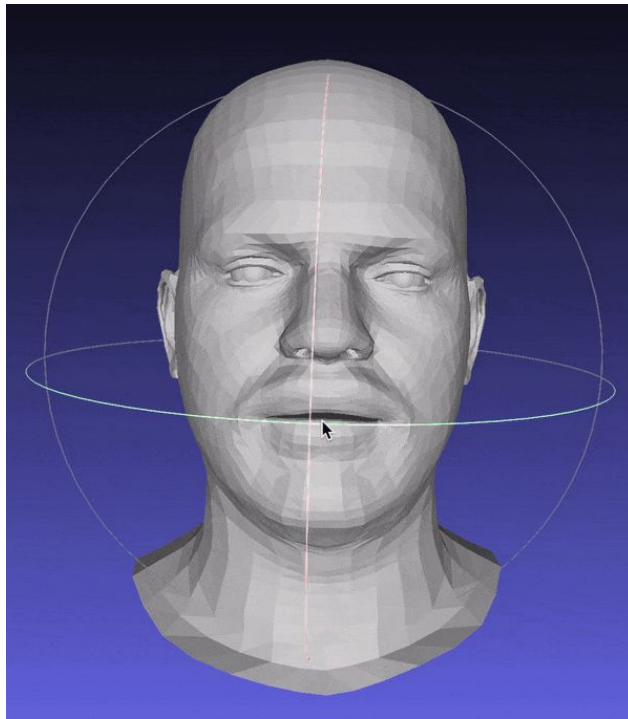


# **FLAME FITTING RESULTS**

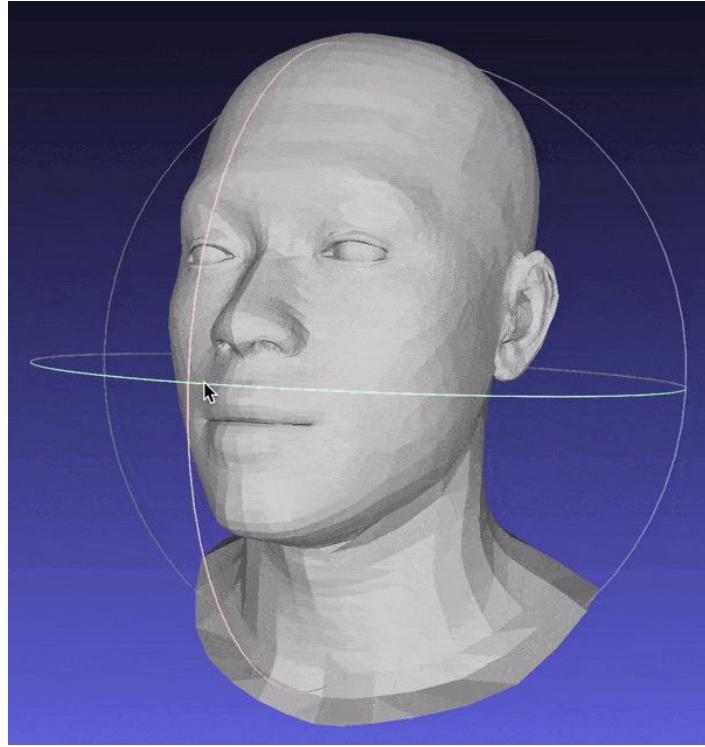
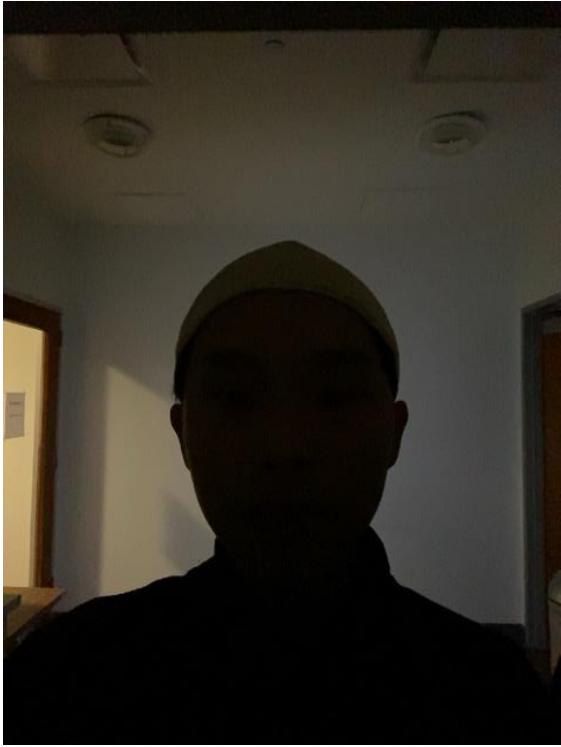
# KESAV



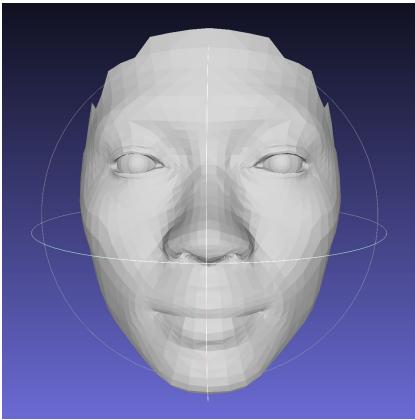
# UTKARSH



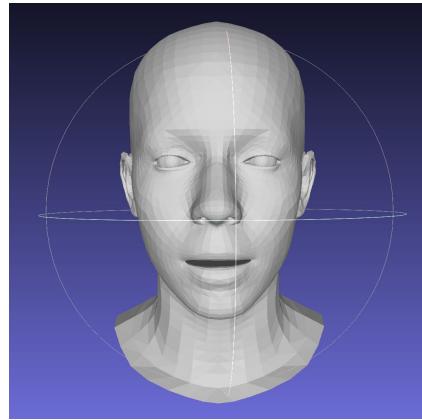
# PATRICK



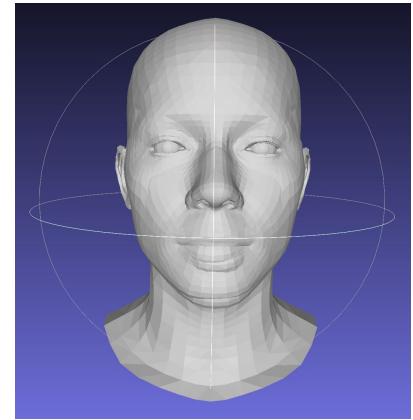
# NoW BENCHMARK



Ground Truth



Predicted  
mesh



Rough  
alignment

1. Ground Truth is generated using MICA.
2. Predicted mesh shown is the DECA output

\*\*\*NOTE: We are only using MICA's ground truth to prove the consistency of our mobile application results.

# **LOW LIGHTING EXPERIMENT**

# Low Lighting Experiment



25%



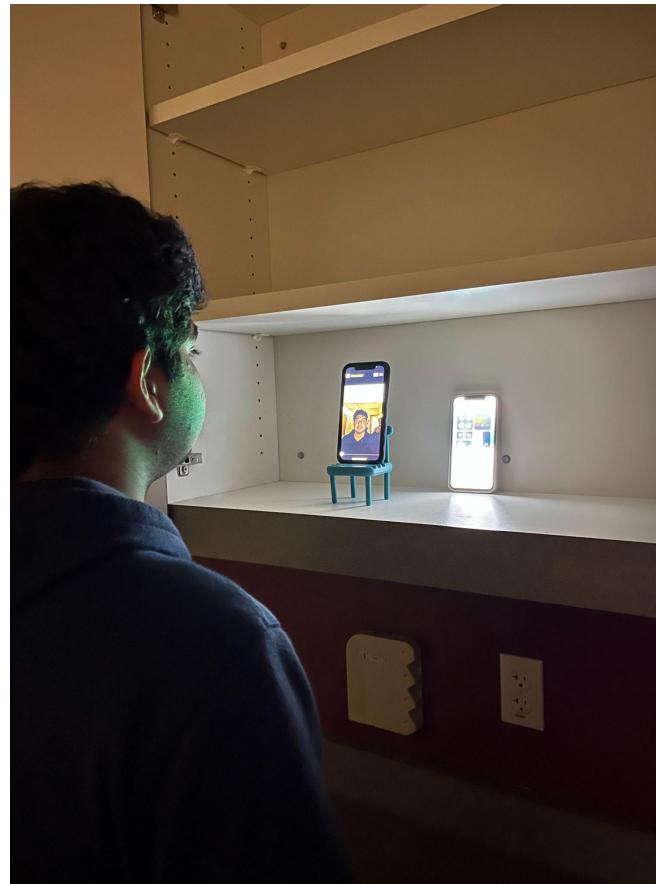
50%



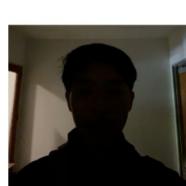
75%



100%



# Low Lighting Experiment



(A: 25%)



(B: 50%)



(C: 75%)



(D: 100%)



(E: natural)



(F: 25%)



(G: 50%)



(H: 75%)



(I: 100%)



(J: natural)



(K: 25%)



(L: 50%)



(M: 75%)



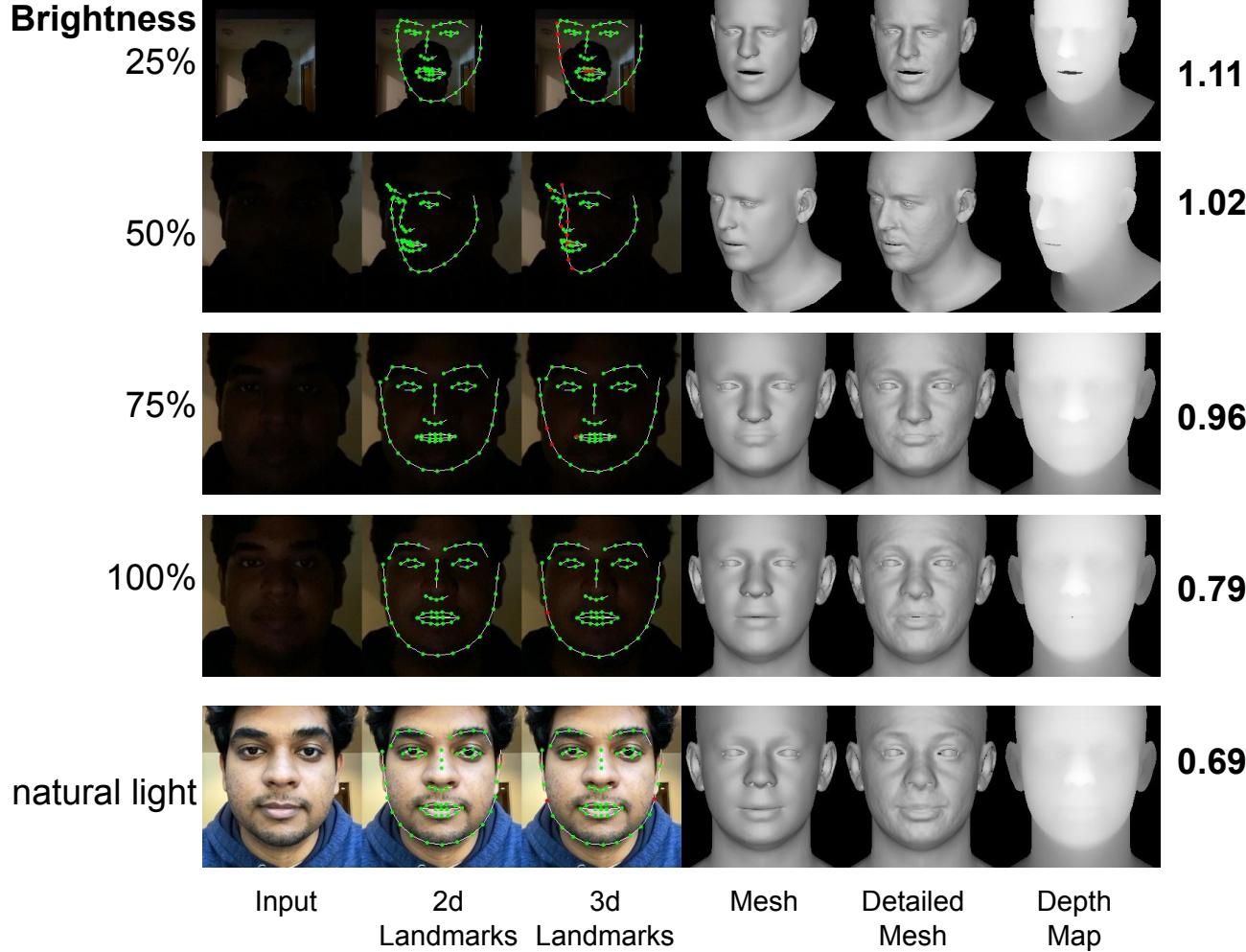
(N: 100%)



(O: natural)

# NoW RESULTS

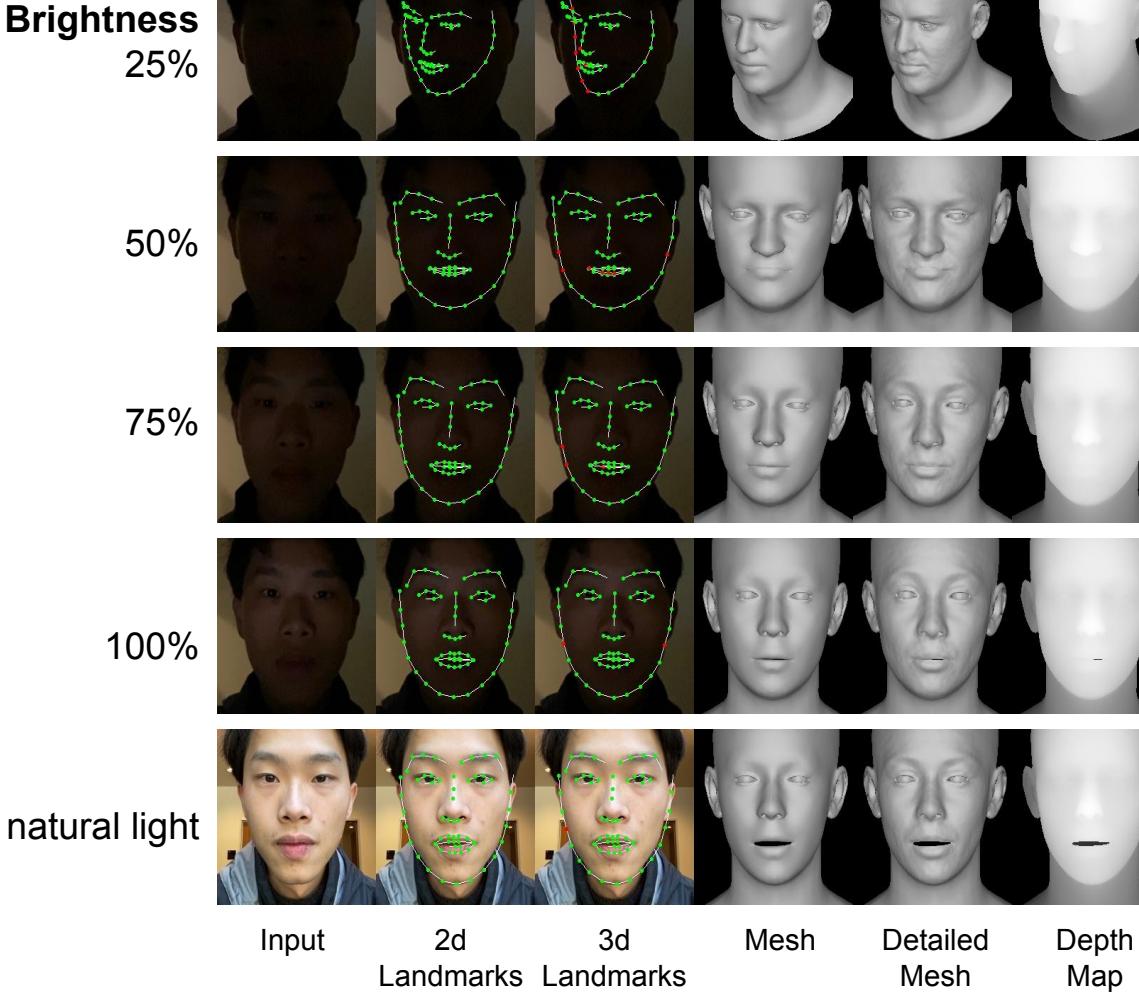
## DECA Results



Mean Reconstruction Error ↓

## Brightness

### DECA Results



1.67  
1.65  
1.24  
1.22  
1.33

Mean Reconstruction Error

Brightness Level	Mean Reconstruction Error			
	DECA		iPhone's depth data + FLAME	
	Kesav	Patrick	Kesav	Patrick
25%	1.11	1.67	1.17	0.75
50%	1.02	1.65	1.13	0.72
75%	0.96	1.24	1.07	0.69
100%	0.79	1.22	1.06	0.78
natural	0.69	1.33	1.13	0.68

The diagram illustrates the consistency of results across different brightness levels. Vertical arrows connect the Mean Reconstruction Error values for each brightness level between the two models. Arrows point downwards from the DECA model (Kesav and Patrick) to the iPhone's depth data + FLAME model (Kesav and Patrick), indicating that the error values are lower for the iPhone's depth data + FLAME model at all brightness levels. The word "CONSISTENT" is written vertically along these downward arrows. Additionally, there are upward-pointing arrows from the iPhone's depth data + FLAME model (Kesav and Patrick) to the DECA model (Kesav and Patrick), also labeled with "CONSISTENT".

\*\*\*NOTE: Do not compare the error across models.  
The point here is to prove the consistency of results

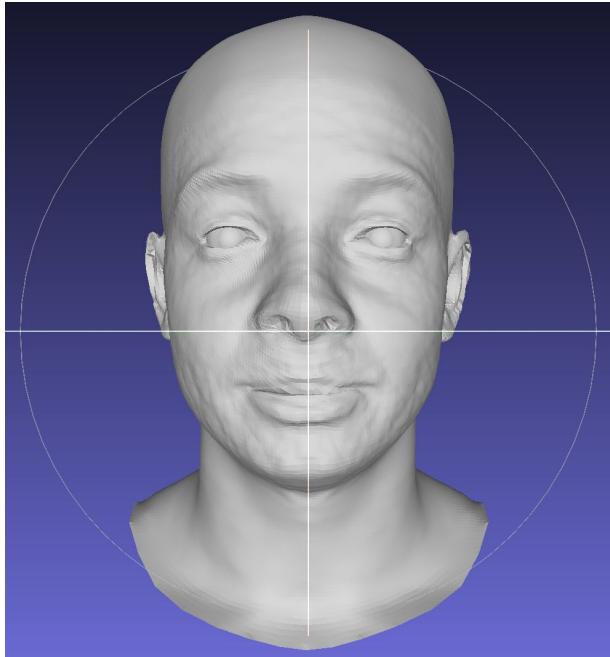
# VISUAL COMPARISON

Image

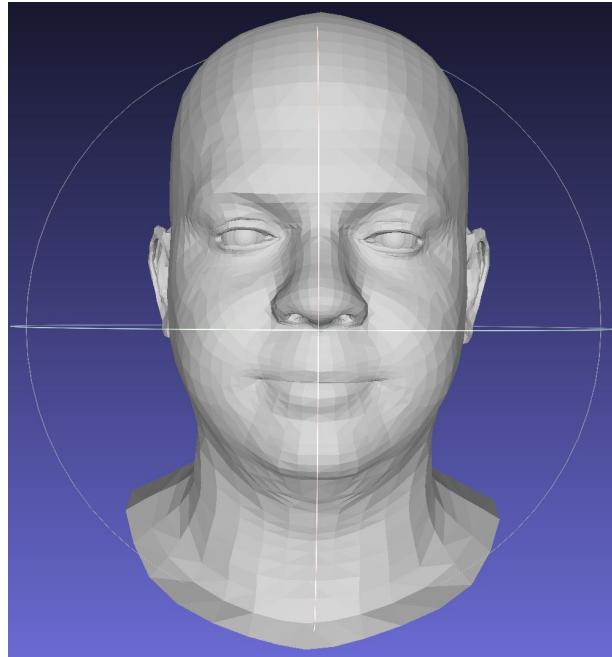


natural

DECA



FLAME

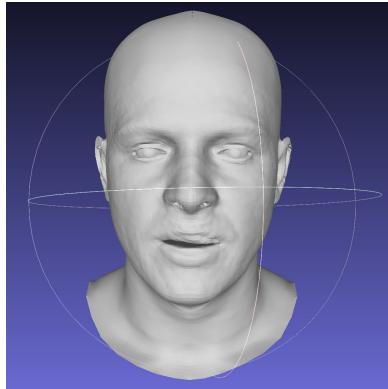


Image

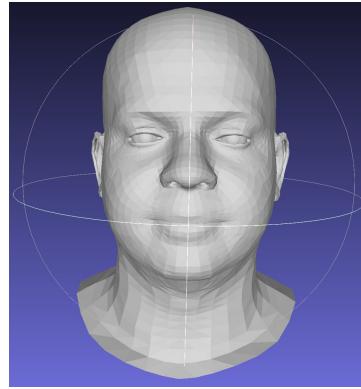


25%

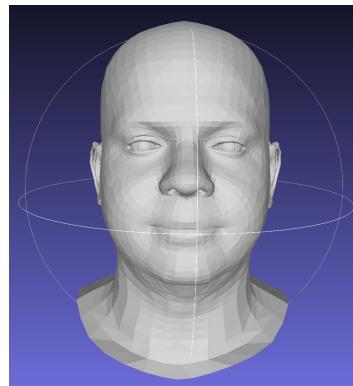
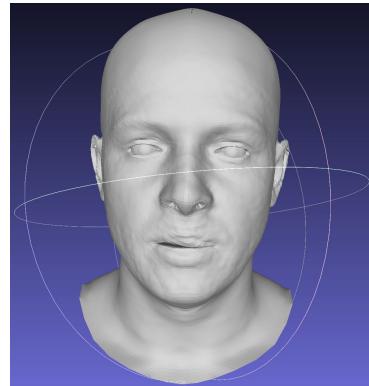
DECA



FLAME



50%

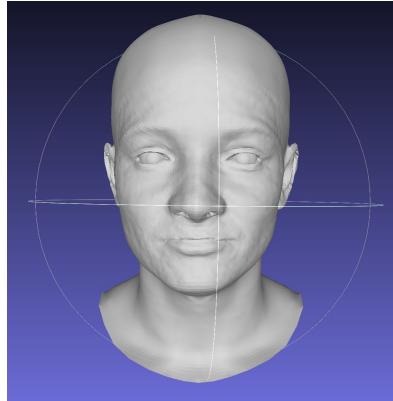


Image



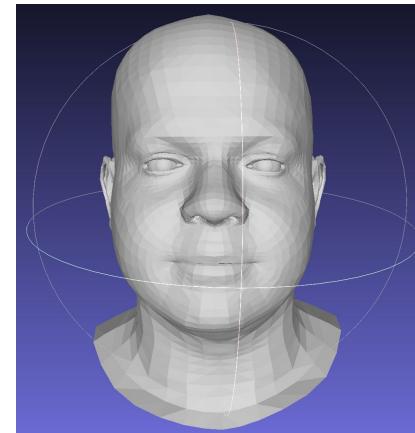
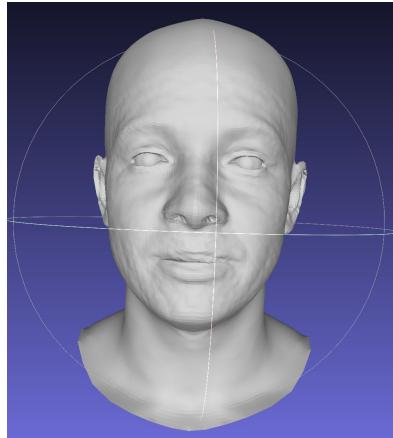
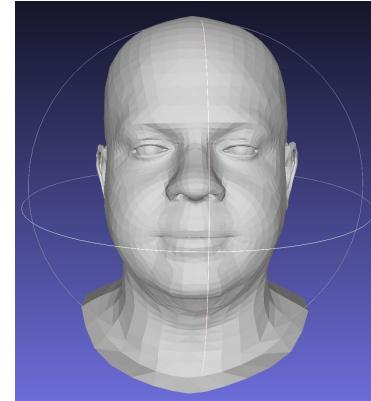
75%

DECA



100%

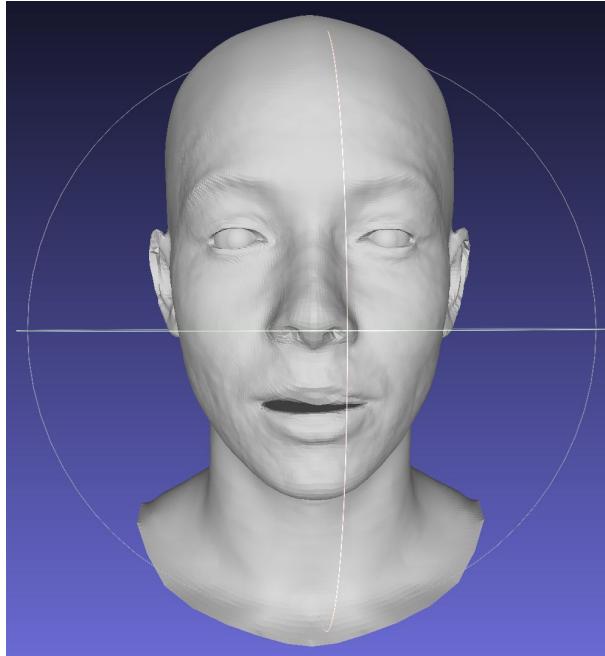
FLAME



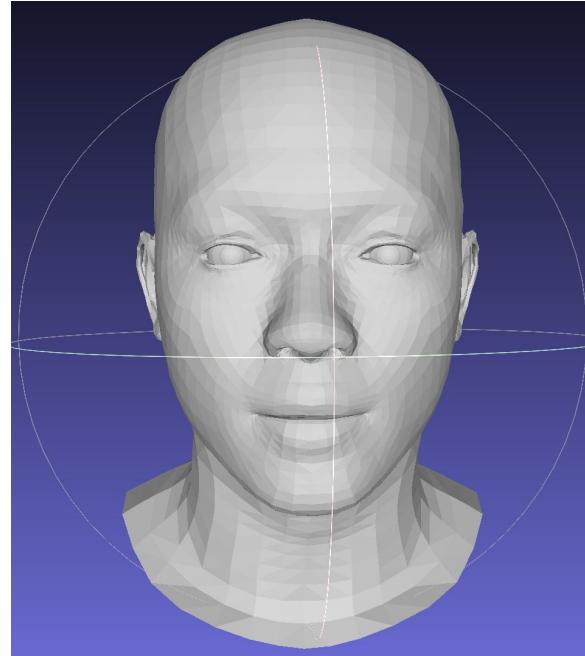
Image



DECA



FLAME



Image

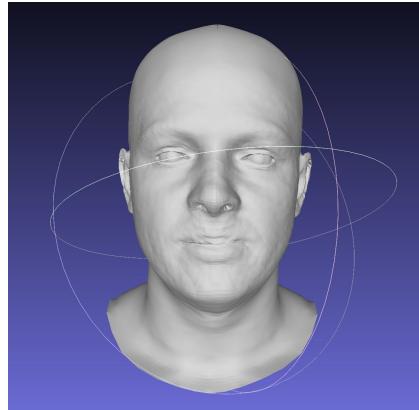


25%

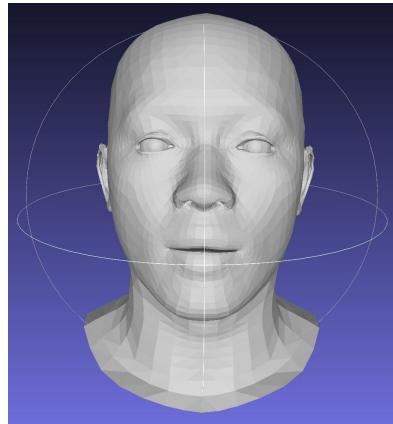
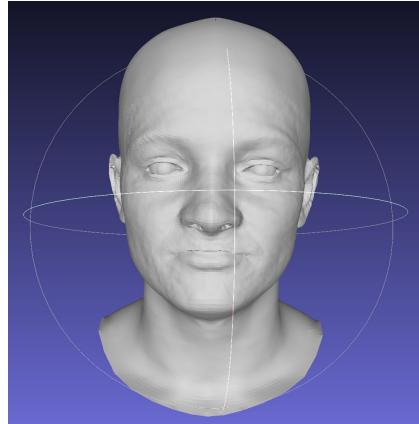
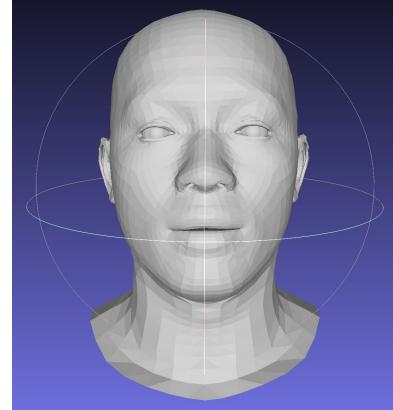


50%

DECA



FLAME



Image

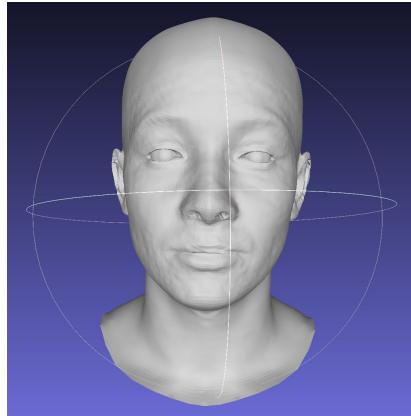


75%

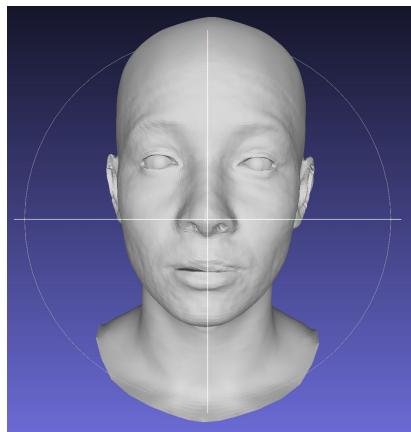
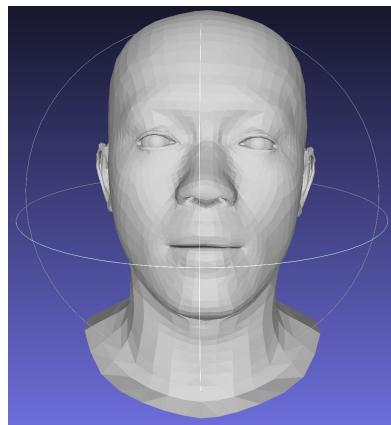
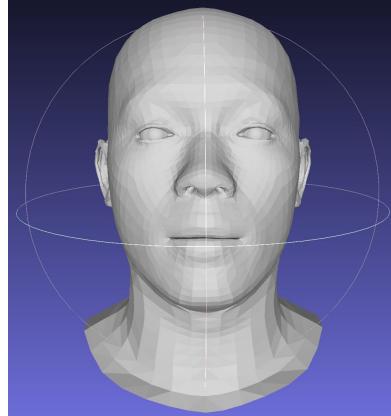


100%

DECA

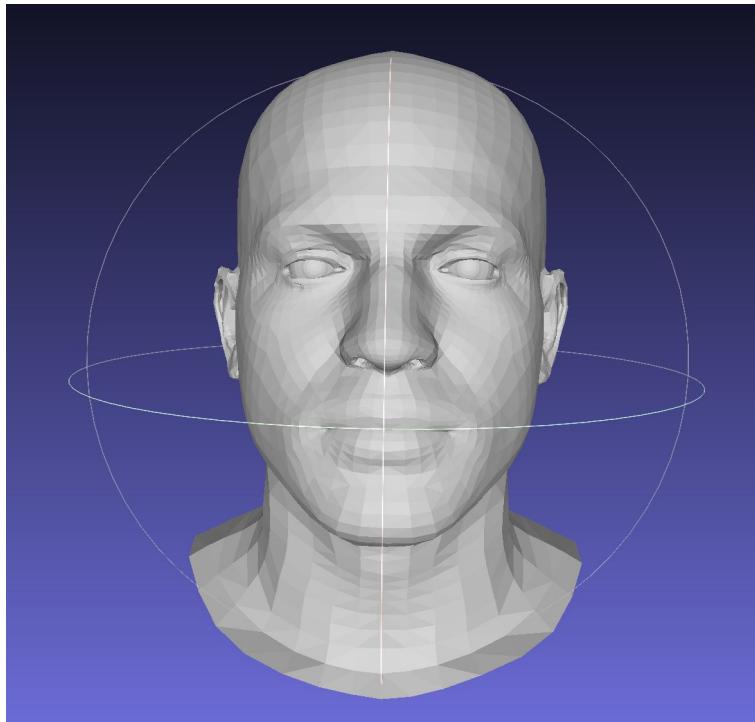


FLAME

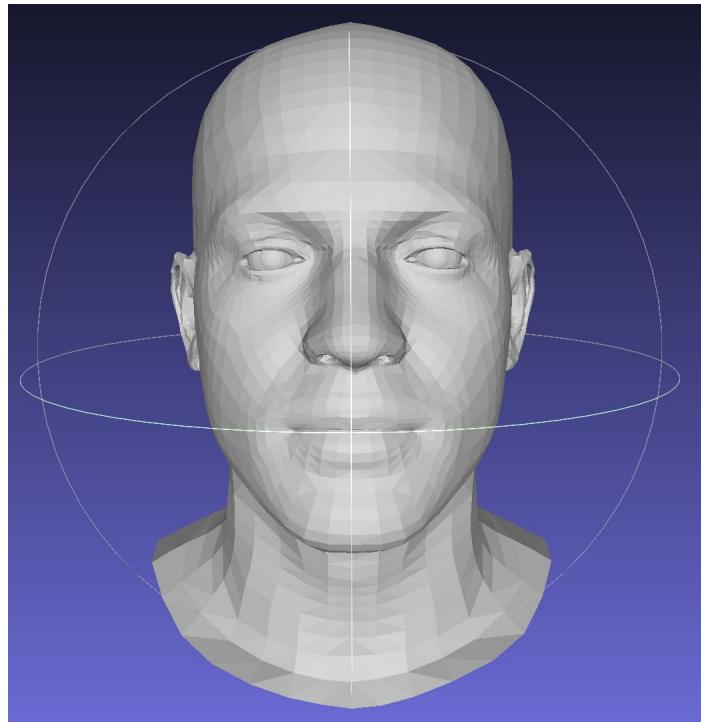


# **CHALLENGES**

25% brightness



Kesav



Utkarsh

# Conclusion

- Made an iPhone mobile application for facial 3D point cloud capturing
- By comparing results with baseline models from NoW challenge (MICA and DECA) we found that:
  - Our method (iPhone capture point cloud + FLAME) produces visually better results than MICA and DECA
  - Image-based Machine Learning algorithms will either not work at all or will perform suboptimally as compared to the 3d face reconstruction done using iPhone's depth data.

## **Discussion & Broader Impact**

Our solution uses an iPhone as the data capturing device and only require one capture to generate a good 3D face model. Based on these characteristics, our method can provide a faster, more accessible, and consistent solution for 3D face model capturing up to a certain degree of accuracy. We hope that this work can help researchers capture more data and potentially lead to better solutions to 3D face model capturing.

# Questions?