



上海交通大学
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人工智能研究院
Artificial Intelligence Institute

Improving Fairness in Facial Albedo Estimation via Visual-Textual Cues

CVPR 2023

TUE-PM-036



highlight

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Preview

- Inferring Albedo from Single Image is ill-posed!

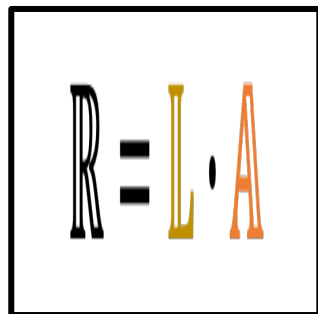


$$\mathbb{R} = \mathbb{L} \cdot \mathbb{A}$$

→ **Illumination Estimation**

→ **Biased Albedo**

- Insight: Semantic Attributes will control Albedo!



→ **Facial Attribute Prior**

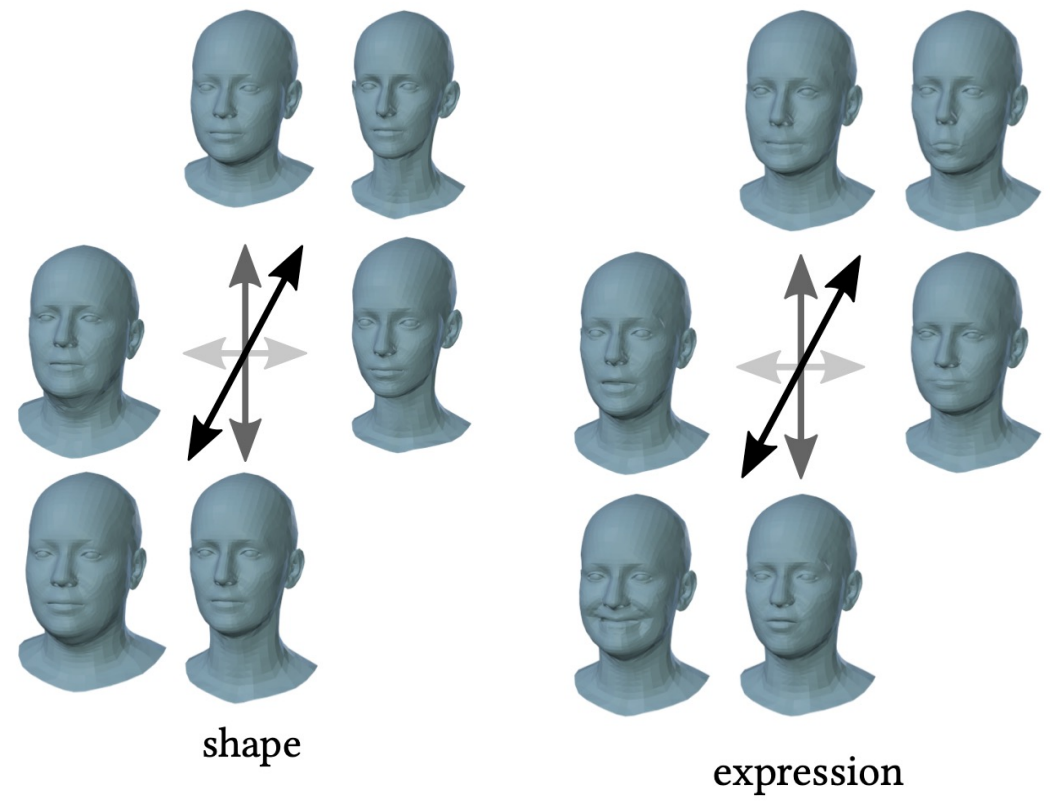
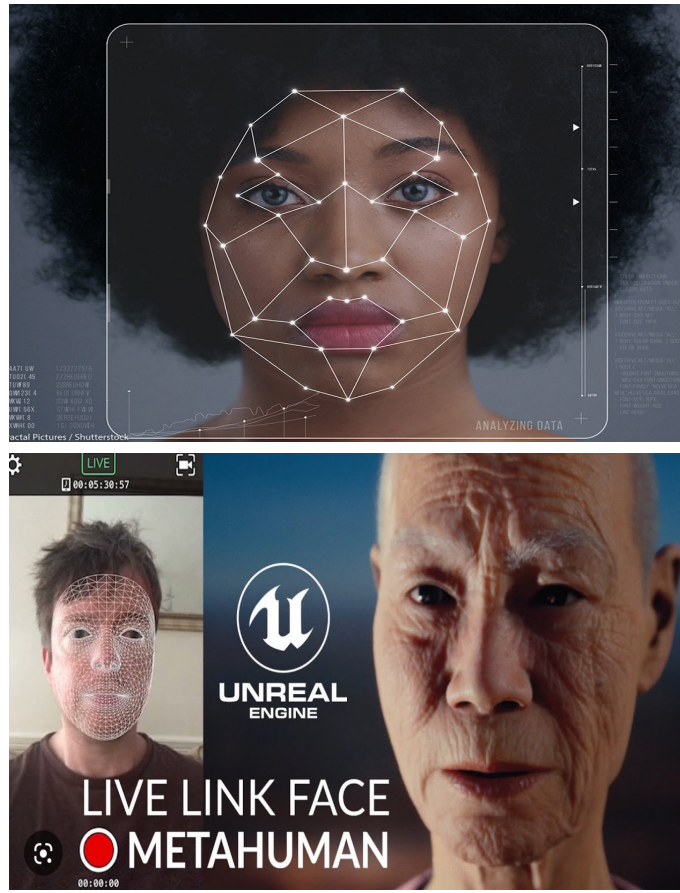
→ **Race: Asian
Age: Young
Skin: Brown
....**

→ **Albedo**





Background: 3D Face Reconstruction



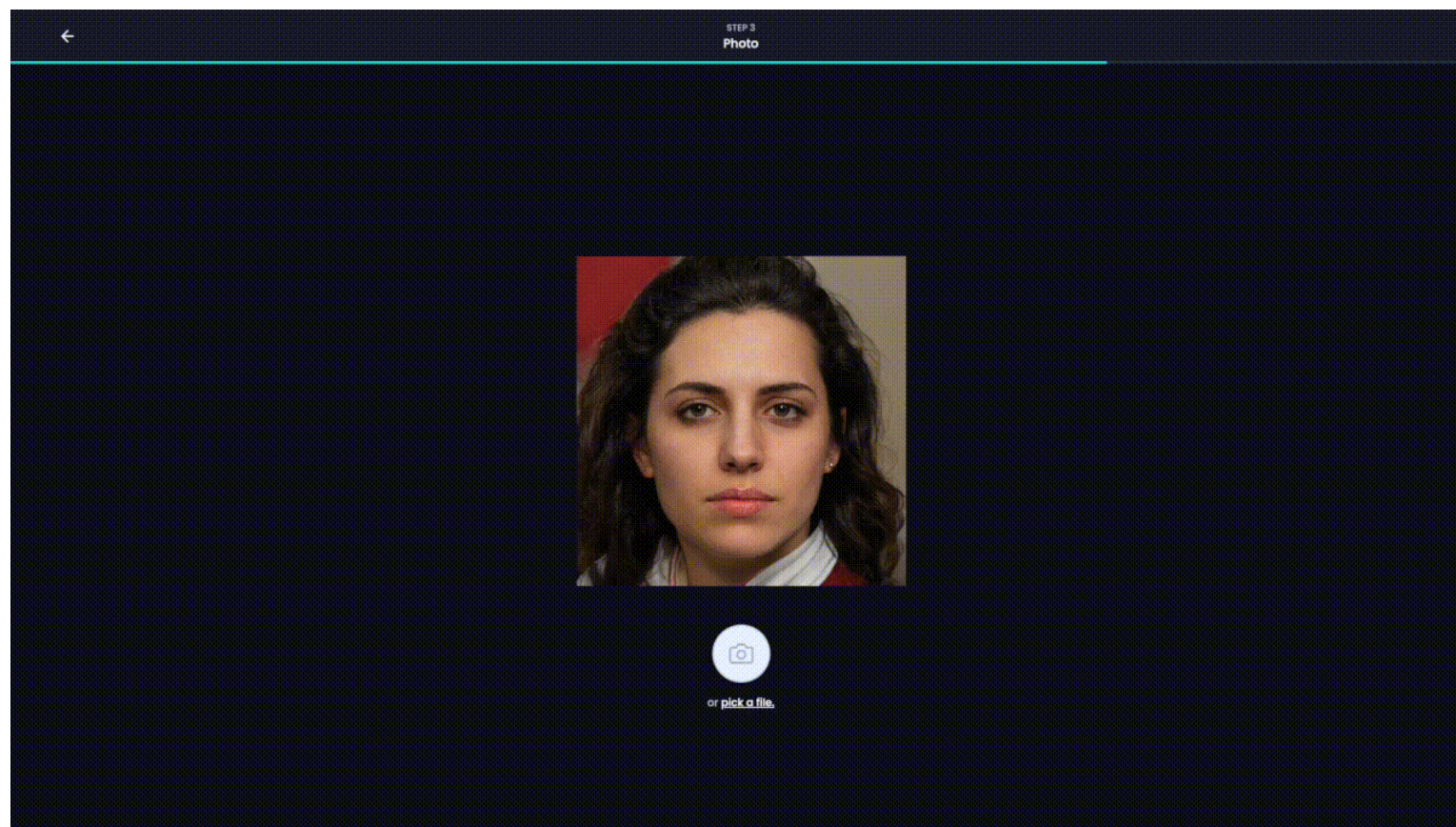
Applications:

- Face Recognition
- Manipulation
- Facial animation
- ...





Background: 3D Face Reconstruction



Applications:

- Avatar Creation
- AR / VR
- ...



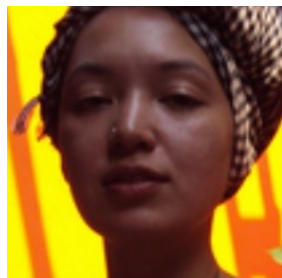


Challenge



- Inferring albedo from pixels is an ill-posed problem

$$\mathcal{R} = \mathcal{A} \odot \mathcal{S}$$



?

?



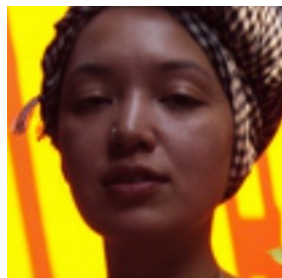


Challenge



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↓ ↓
Albedo Lighting
Model Constrain



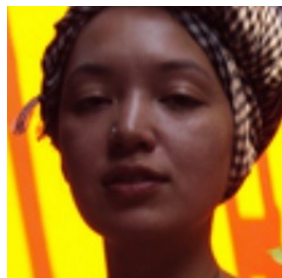


Challenge



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$$\mathcal{R} = \mathcal{A} \odot \mathcal{S}$$



↓ ↓
Albedo Lighting
Model Constrain

Biased! Biased!



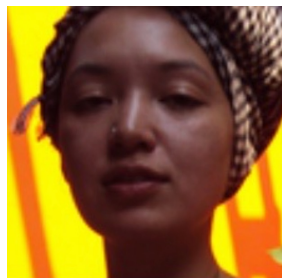


Challenge



- Inferring albedo from pixels is an ill-posed problem

$$\mathcal{R} = \mathcal{A} \odot \mathcal{S}$$



↓ ↓
Unbiased Scene
Albedo Lighting
Model Constrain

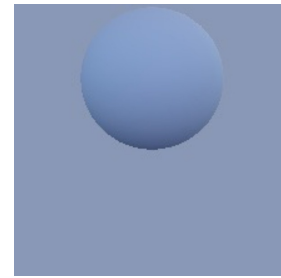
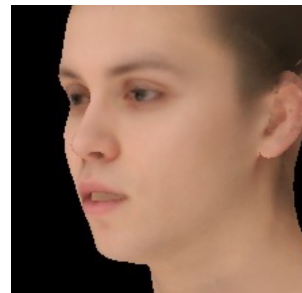
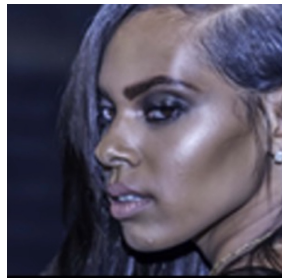
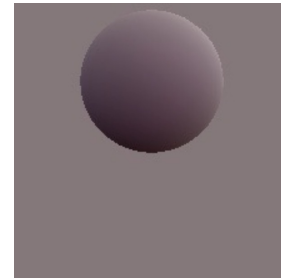
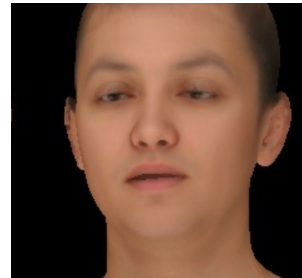




Challenge

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$$\mathcal{R} = \mathcal{A} \odot \mathcal{S}$$



Biased Albedo





Motivation

- How do people disentangle light from a single image?
 - Prior Knowledge
 - Caucasian → White
 - ...
 - African → Dark
 - Facial Attribute → Albedo



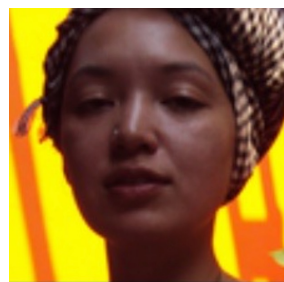


Motivation



- Inferring albedo from pixels is an ill-posed problem

$$\mathcal{R} = \mathcal{A} \odot \mathcal{S}$$



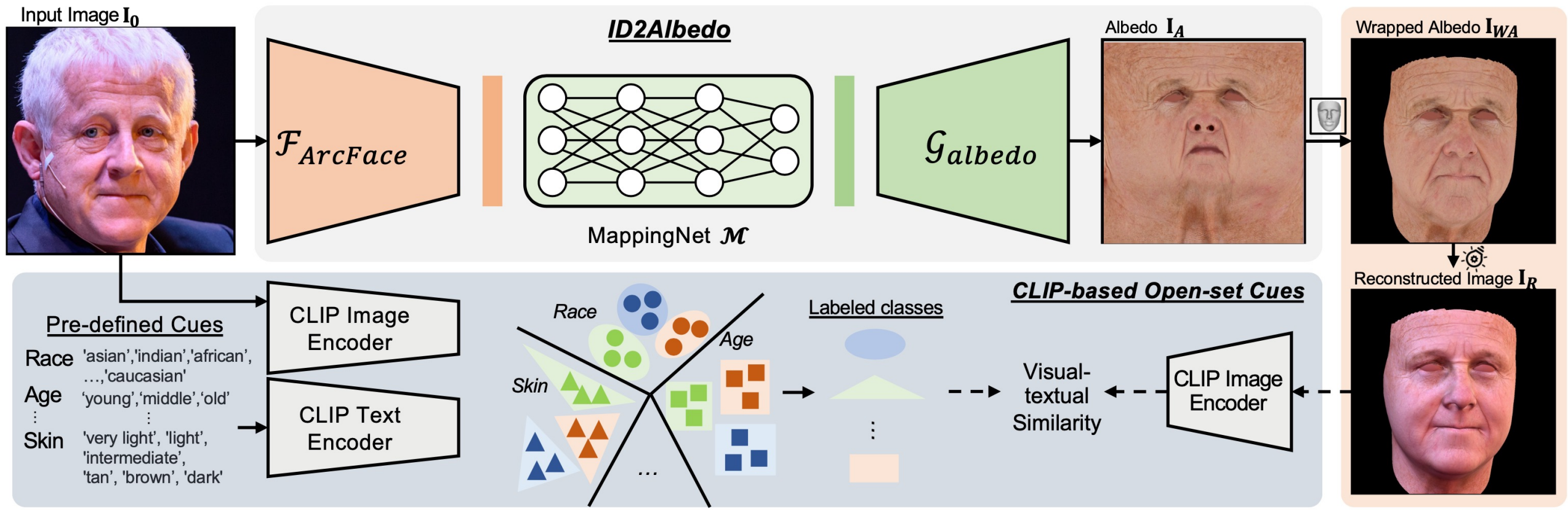
↓
Facial
Attribute

↓ ↓
Unbiased No
Albedo Constrain
Generator



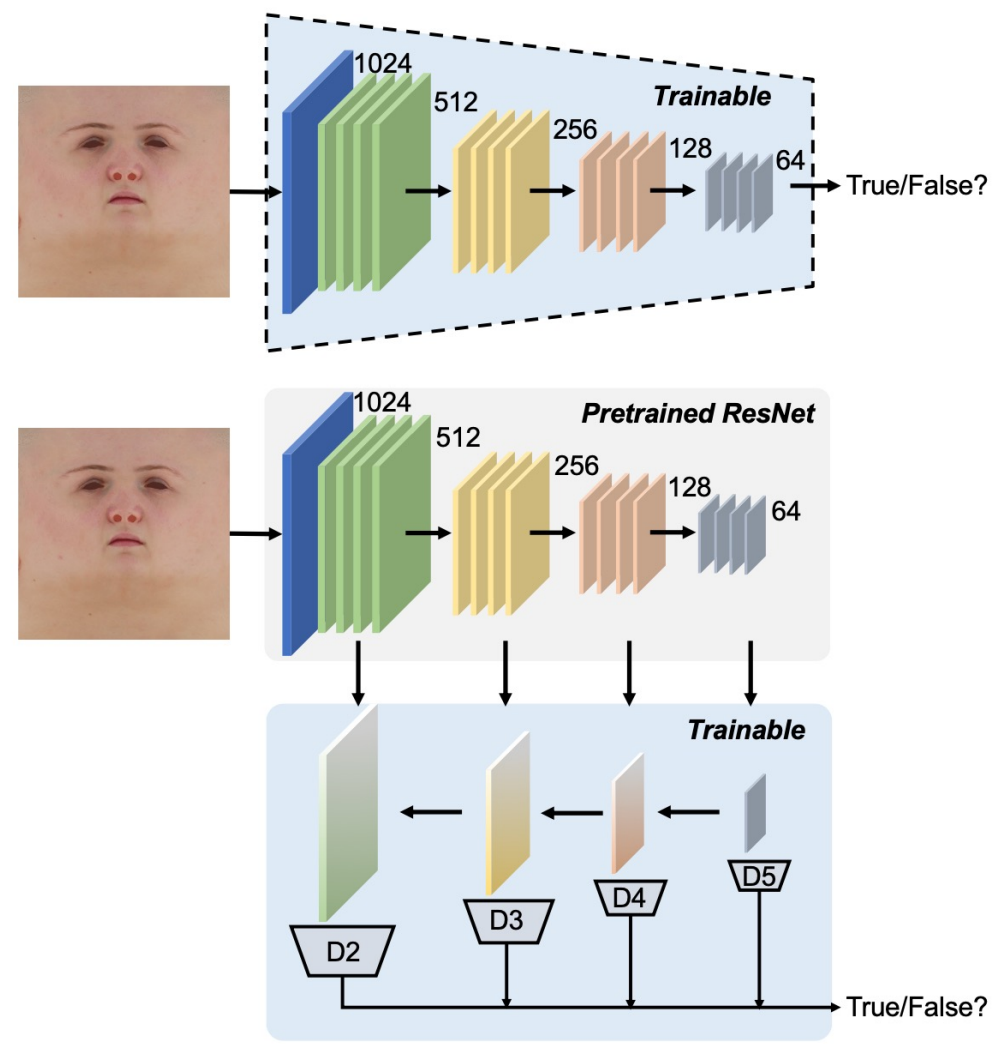


Method: ID2Albedo





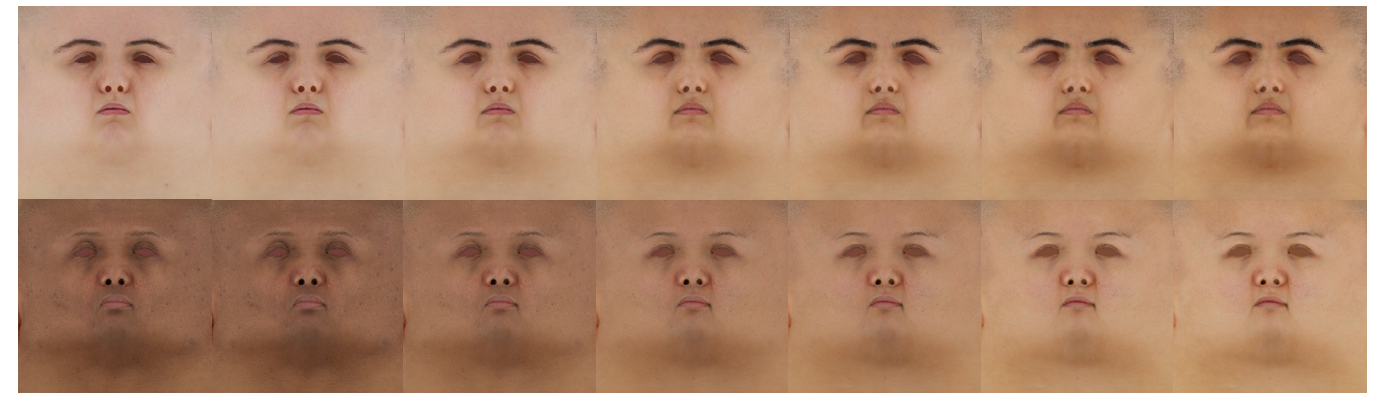
Method: Albedo Generator



Random Seeds



Interpolation





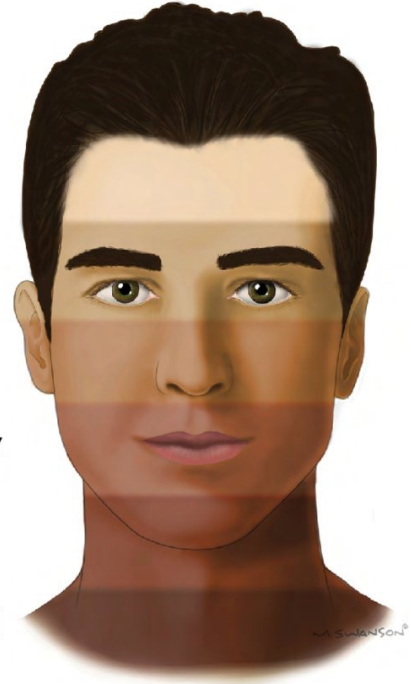
Evaluation

- ITA Score

$$ITA(L^*, b^*) = \frac{180}{\pi} \times \arctan\left(\frac{L^* - 50}{b^*}\right),$$

SKIN TYPE	SKIN COLOR	REACTION TO SUN	
		SUNBURN	TANNING
I	Light, pale white	Always burns	Never Tans
II	White, fair	Usually burns	Tans with difficulty
III	Medium, white to olive	Sometimes mild burns	Gradually tans to olive
IV	Beige olive, moderate brown	Rarely burns	Easy tan to moderate brown
V	Brown, dark brown	Very rarely burns	Tans very easily
VI	Very dark brown to black	Never burns	Always tans

Type I
Type II
Type III
Type IV
Type V
Type VI





Results: FAIR Benchmark

Method	Avg. ITA ↓	Bias ↓	Score ↓	MAE ↓	ITA per skin type ↓					
					I	II	III	IV	V	VI
Deep3D [10]	22.57	24.44	47.02	27.98	8.92	9.08	8.15	10.90	28.48	69.90
GANFIT [19]	62.29	31.81	94.11	63.31	94.80	87.83	76.25	65.05	38.24	11.59
MGCNet [46]	21.41	17.58	38.99	25.17	19.98	12.76	8.53	9.21	22.66	55.34
DECA [15]	28.74	29.24	57.98	38.17	9.34	11.66	11.58	16.69	39.10	84.06
INORig [2]	27.68	28.18	55.86	33.20	23.25	11.88	4.86	9.75	35.78	80.54
CEST [55]	35.18	12.14	47.32	29.92	50.98	38.77	29.22	23.62	21.92	46.57
TRUST [14] (BFM)	16.19	15.33	31.52	21.82	12.44	6.48	5.69	9.47	16.67	46.37
TRUST [14] (AlbedoMM)	17.72	15.28	33.00	19.48	15.50	10.48	8.42	7.86	15.96	48.11
TRUST [14] (BalancedAlb)	13.87	2.79	16.67	18.41	11.90	11.87	11.20	13.92	16.15	18.21
Ours (ID2Albedo)	12.07	4.91	16.98	23.33	18.30	9.13	5.83	9.46	19.09	10.59





Results: In-the-wild Images



Methods	M-SSIM \uparrow	LPIPS \downarrow	FID \downarrow	ID \uparrow
Deep3D [10]	0.73	0.1933	74.41	0.712
DECA [15]	0.61	0.2089	98.13	0.585
TRUST [14]	0.64	0.2112	97.37	0.603
Ours	0.87	0.1549	45.56	0.867





Ablations



Albedo Encoder	Avg. ITA ↓	Bias ↓	Score ↓
ResNet-100 [23] (Scratch)	58.46	32.59	91.05
ResNet-100 [23] (ImageNet)	31.63	15.48	47.11
ArcFace [9] (fully trainable)	41.63	19.81	61.44
ArcFace [9] (L2 + L3 + L4)	28.75	11.87	40.62
ArcFace [9] (L3 + L4)	19.52	9.46	28.98
ArcFace [9] (L4)	14.58	6.79	21.37
ArcFace [9] (Frozen)(Ours)	13.46	5.86	19.32

Configs	Avg. ITA ↓	Bias ↓	Score ↓
w/o any cues	25.66	23.51	49.17
Manual labeled races	18.13	10.46	28.59
CLIP [40] cues (only races)	16.21	7.44	23.65
CLIP cues all (ours)	13.46	5.86	19.32





Improving **Fairness** in Facial Albedo Estimation via **Visual-Textual Cues**

