



# Uncovering bias: Analyzing if skin texture intensity illuminates pulse oximetry discrepancies

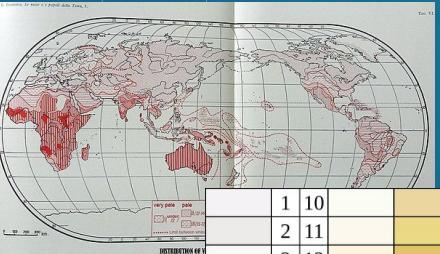
## ENCoDE dataset

# 1

# Introduction

Discrepancy between pulse oximeter and blood gas analysis measurement in patients with **darker skin tones** has been observed in previous studies. Previous analysis of bias in this data set has been limited to tabular data based on self-reported race, pre-existing skin tone scales, and reflectance colorimetry/spectroscopy.

# PRE-EXISTING SKIN TONE SCALES.



1	10		19	28	
2	11		20	29	
3	12		21	30	
4	13		22	31	
5	14		23	32	
6	15		24	33	
7	16		25	34	
8	17		26	35	
9	18		27	36	

**Von Luschan:**  
used in the 1900s to categorize race based on skin color

Fitzpatrick Skin Type Descriptions*	
Skin Type	Description
I	Always burns, never tans
II	Burns easily, tans minimally
III	Burns moderately, tans to light brown
IV	Burns minimally, tans to moderate brown
V	Rarely burns, tans to dark
VI	Never burns, least sensitive to changes

\* Adapted from *The Surgeon General's Call to Action to Prevent Skin Cancer*



**Fitzpatrick:**  
based on skin's reaction to sun exposure



**Monk:**  
introduced by Google to reduce bias in computer models

# Baseline model: Association between skin tone features and pulse oximeter bias ( $\text{SpO}_2$ - $\text{SaO}_2$ )

	coef	std err	z	P> z	[0.025]
Intercept	223.5947	56.062	3.988	0.000	113.0
ADMINISTERED_VISUAL_SCALES_CARD_FITZPATRICKSCALE	-0.5874	1.192	-0.493	0.622	-2.0
ADMINISTERED_VISUAL_SCALES_CARD_MONSKINTONSCALE	-0.9933	0.909	-1.092	0.275	-2.0
ADMINISTERED_VISUAL_SCALES_CARD_VONLUSCHANCHROMATICSSCALE	0.3179	0.333	0.953	0.340	-0.0
DELFIN_SKINCOLORCATCH_CIE_ASTAR	-0.3205	0.610	-0.525	0.599	-1.1
DELFIN_SKINCOLORCATCH_CIE_BSTAR	0.2003	0.077	2.610	0.009	0.0
DELFIN_SKINCOLORCATCH_CIE_LSTAR	-1.4518	0.636	-2.283	0.022	-2.698
DELFIN_SKINCOLORCATCH_ERYTHEMAINDEX	-0.0565	0.057	-0.991	0.322	-0.168
DELFIN_SKINCOLORCATCH_ITA	0.1229	0.109	1.127	0.260	-0.091
DELFIN_SKINCOLORCATCH_MELANININDEX	-0.0815	0.032	-2.564	0.010	-0.144
KONICAMINOLTA_CM700D_CIE_ASTAR	-6.6234	5.526	-1.199	0.231	-17.454
KONICAMINOLTA_CM700D_CIE_BSTAR	1.1688	6.078	0.192	0.847	-10.742
KONICAMINOLTA_CM700D_CIE_CSTAR	-7.8344	6.288	-1.246	0.213	-20.1
KONICAMINOLTA_CM700D_CIE_H	0.3428	0.383	0.895	0.371	-0.0
KONICAMINOLTA_CM700D_CIE_LSTAR	-6.8191	17.416	-0.392	0.695	-40.0
KONICAMINOLTA_CM700D_CIE_X	-70.2849	33.466	-2.100	0.036	-135.0
KONICAMINOLTA_CM700D_CIE_Y	56.8722	47.548	1.196	0.232	-36.0
KONICAMINOLTA_CM700D_CIE_Z	0.7195	5.838	0.123	0.902	-10.0
KONICAMINOLTA_CM700D_HUNTER_ASTAR	20.9237	10.281	2.035	0.042	0.7
KONICAMINOLTA_CM700D_HUNTER_BSTAR	8.2752	9.195	0.900	0.368	-9.747
KONICAMINOLTA_CM700D_HUNTER_L	17.9776	17.169	1.047	0.295	-15.674
KONICAMINOLTA_CM700D_MUNSELL_CSTAR	11.1044	16.124	0.689	0.491	-20.498
KONICAMINOLTA_CM700D_MUNSELL_VALUE	-74.6664	81.354	-0.918	0.359	-234.118
					84.785

$\text{SpO}_2$

$\text{SaO}_2$



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# Hypothesis

We hypothesize that **skin color intensity**, which incorporates visual features such as skin **texture** will provide deeper **insight** into pulse oximeter bias. **Image informatics** related to **intensity** may provide valuable information to extract from the data set.



# Main Objective

To determine if skin intensity features is associated with pulse oximetry discrepancies



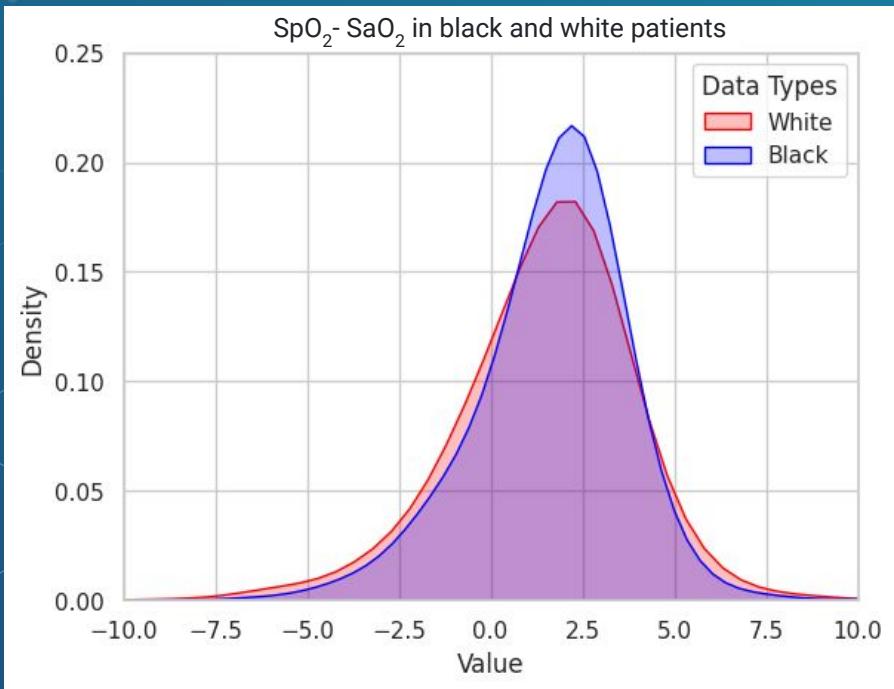
# Population study

Understanding the **demographic distribution** of patients in  
the ENCoDE data set

# Table 1: Racial distribution of patients, subdivided by gender and ethnicity

		Grouped by race				
		Missing	Overall	Black or African American	Other	White
n			127	57	15	55
gender, n (%)	FEMALE		51 (40.2)	23 (40.4)	3 (20.0)	25 (45.5)
	MALE		76 (59.8)	34 (59.6)	12 (80.0)	30 (54.5)
ethnicity, n (%)	Not Hispanic or Latino		120 (94.5)	57 (100.0)	12 (80.0)	51 (92.7)
	Hispanic or Latino		4 (3.1)		1 (6.7)	3 (5.5)
	Unknown		3 (2.4)		2 (13.3)	1 (1.8)
age, median [Q1,Q3]		0 - 58.0 [48.0,67.0]		55.0 [47.0,66.0]	57.0 [49.0,62.5]	61.0 [49.0,68.0]

# Figure 1: Oximeter bias (difference between non-arterial vs arterial oxygen saturation measurements)



Median, [Q1, Q2]:  
- Black: -2.0 [2.9,0.7]  
- White: -1.8 [2.8,0.3]

# Table 2: Oximeter bias was calculated using simultaneous measurements of SpO<sub>2</sub> and SaO<sub>2</sub>

Grouped by race							
		Missing	Overall	Black or African American	Other	White	
n			1581		816	66	699
gender, n (%)	FEMALE		658 (41.6)		367 (45.0)	7 (10.6)	284 (40.6)
	MALE		923 (58.4)		449 (55.0)	59 (89.4)	415 (59.4)
ethnicity, n (%)	Not Hispanic or Latino		1541 (97.5)		816 (100.0)	60 (90.9)	665 (95.1)
	Hispanic or Latino		23 (1.5)		2 (3.0)	21 (3.0)	
	Unknown		17 (1.1)		4 (6.1)	13 (1.9)	
SPO2 - SAO2, median [Q1,Q3]							
		0	1.9 [0.5,2.8]		2.0 [0.7,2.9]	1.9 [0.9,3.0]	1.8 [0.3,2.8]





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# Experimental Design

# Dataset

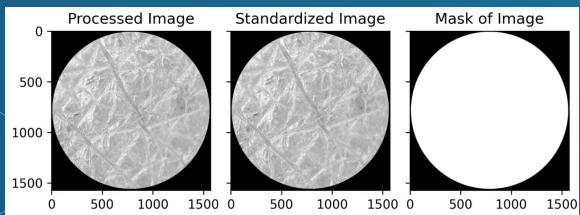
person_id	gender	race	ethnicity	SpO2 - SaO2	original_firstorder_10Percentile	original_firstorder_90Percentile
108	FEMALE	White	Not Hispanic or Latino	-3.9	0.674003	0.855117
111	MALE	Black or African American	Not Hispanic or Latino	2.2	0.764251	0.868282
73	MALE	Asian	Not Hispanic or Latino	2.3	0.724418	0.846694
46	MALE	Black or African American	Not Hispanic or Latino	2.0	0.786955	0.872143
78	FEMALE	White	Not Hispanic or Latino	3.0	0.834364	0.895427
23	MALE	Black or African American	Not Hispanic or Latino	3.2	0.687732	0.814108
70	FEMALE	White	Not Hispanic or Latino	1.7	0.709244	0.850005
13	MALE	White	Not Hispanic or Latino	1.3	0.714899	0.859553
2	FEMALE	White	Not Hispanic or Latino	2.3	0.757256	0.863764
106	MALE	White	Not Hispanic or Latino	0.1	0.684629	0.827562

# Data Engineering

1. Converting to grayscale

2. Standardization

3. Masking of the images



original_fistorder_10Percentile
0.674003
0.764251
0.724418
0.786955

person_id	gender	race	SpO2 - SaO2	original_fistorder_10Percentile	original_fistorder_90Percentile
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# Process

## 1. Extract radiomics

Process the patient images to extract radiomic data, specifically relate to **image intensity**.

## 2. Perform PCA

Perform principal component analysis (PCA) on the data gathered in step 1

## 3. Integrate EHR data with Image features

Integrate EHR data with image-based features

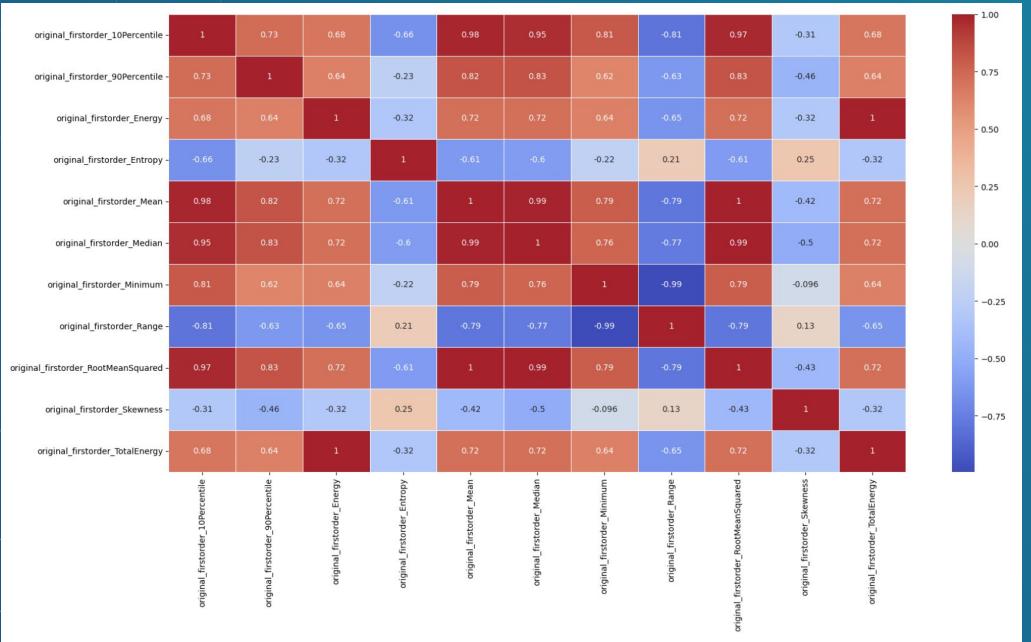
## 4. Adjust for environmental factors

Adjust the blood pressure and body temperature

## 5. Association Study

Interpret the association between the skin intensity features and arterial oxygen saturation ( $\text{SaO}_2$ ), using multivariate regression model

# Extraction of 19 radiomic features



- 10th percentile
- 90th percentile
- Energy
- Entropy
- Interquartile range
- Kurtosis
- Maximum
- Mean absolute deviation
- Mean
- Median
- Minimum
- Range
- Robust mean absolute deviation
- Root mean squared
- Skewness
- Total energy
- Uniformity
- Variance



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# Results from Principal Component Analysis

	Pulse ox bias ( $\text{SpO}_2 - \text{SaO}_2$ )
Skin texture feature-1	Negative
Skin texture feature-2	Positive
Skin texture feature-3	Positive

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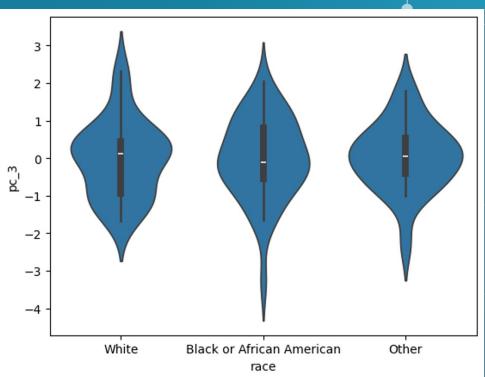
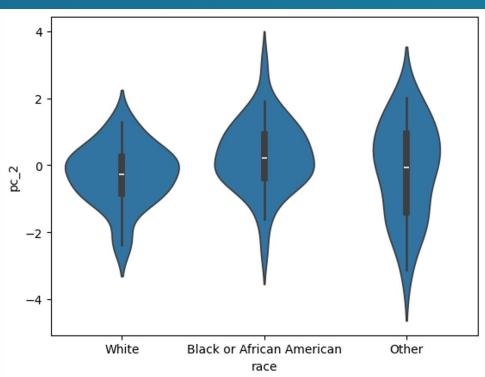
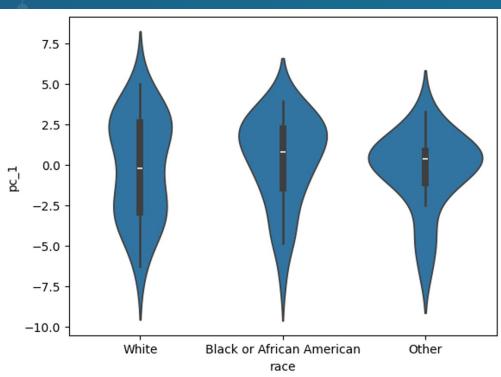
## 5. Association Study

Interpret the association between the skin intensity features and pulse oxy bias, using multivariate regression model

# Results from Generalized Estimating Equation Regression

	Pulse Ox bias ( $\text{SpO}_2$ - $\text{SaO}_2$ )
<b>Skin texture feature-1</b>	<b>Negative</b>
<b>Skin texture feature-2</b>	<b>Positive</b>
<b>Skin texture feature-3</b>	<b>Positive</b>

# Skin Texture Features visualized based on race and ethnicity





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# Conclusion/Relevance



# Novelty

Our proposed method took a in-depth view on patients' **skin texture** from **images**, and found associations between features representing **skin texture** and pulse oximetry **bias**.



# Limitations



- Limitation of scope: Families of data related to different image features (ex. GLCM) would provide more insight into tissue texture
- Sample size too small due to primary dataset

# THANK YOU!

**Team 2:**

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Kwak, Balowa Musa, Marit UyHam



# Image credits:

[https://en.wikipedia.org/wiki/Von\\_Luschan%27s\\_chromatic\\_scale](https://en.wikipedia.org/wiki/Von_Luschan%27s_chromatic_scale)

Sharma AN, Patel BC. Laser Fitzpatrick Skin Type Recommendations. In: StatPearls [Internet]. Treasure Island (FL): StatPearls Publishing; 2024 Jan-. Available from: <https://www.ncbi.nlm.nih.gov/books/NBK557626/>

<https://suttonderm.com/whats-your-type/>

<https://qz.com/2176270/googles-more-inclusive-skin-tone-scale-could-minimise-internet-racism/>

<https://statnews.com/2022/07/05/pulse-oximeters-accuracy-skin-color/>