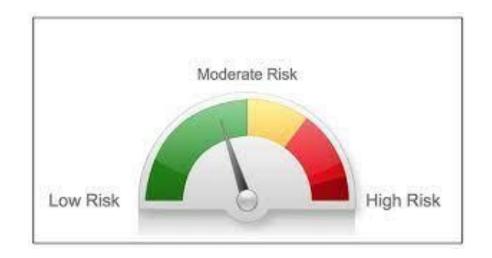
Skin Disease Prediction through Image Analysis on Diverse Skin Tones

Amanda Walsh

#### Introduction

 Objective: To offer a triage solution through predictive modeling to a hospital executive board that is experiencing long wait times for appointment scheduling in Dermatology

- Goal: To show how AI can be one solution for triaging patients and making sure prompt care is provided to those who need it.
  - Educate the board on the importance of using ethical models that are representative of diverse skin tones, to ensure equality in care across all groups.



## Bias in Dermatology

- Lack of diversity in diagnostic images
- Doctor training across the board
- Limited, dated medical research

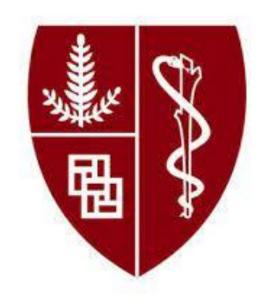


## Warning- Medical Images



#### **Data Collection**

- Stanford University "Diverse Dermatology Images" Dataset
- Selected at Stanford Clinics from 2010-2020
- 656 Images of biopsy proven skin diseases
- Skin tone labeled and cross-referenced using Fitzpatrick
   Scale



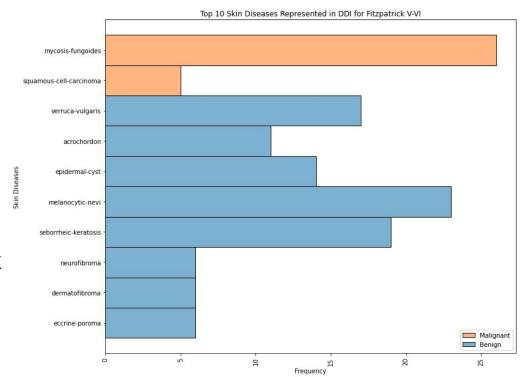
## What is the Fitzpatrick Scale?

Score	Description	Female	Male	
0-6	Pale white skin			
Туре І	Extremely sensitive skin, always burns, never tans  Example: red hair with freckles			
7-13	White skin			
Туре ІІ	Very sensitive skin, burns easily, tans minimally  Example: fair skinned, fair haired Caucasians, northern  Asians			
14-20	Light brown skin			
Type III	Sensitive skin, sometimes burns, slowly tans to light brown  Example: darker Caucasians, some Asians			
21-27	Moderate brown skin  Mildly sensitive, burns minimally, always tans to moderate			
Type IV	brown  Example: Mediterranean and Middle Eastern Caucasians, southern Asians			
28-34	Dark brown skin			
Туре V	Resistant skin, rarely burns, tans well Example: some Hisponics, some Africans			
35+	Deeply pigmented dark brown			
Type VI	to black skin  Very resistant skin, never burns, deeply pigmented  Example: darker Africans, Indigenous Australians			

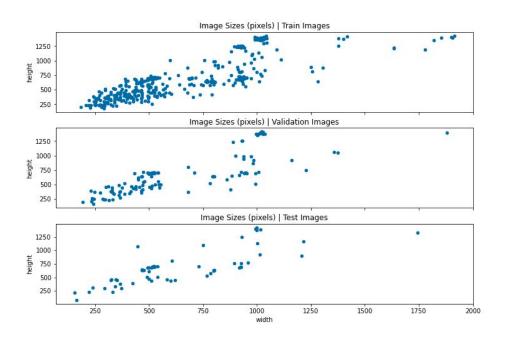
- Fitzpatrick (FST) I-VI
- Score 1-35
- Describes skin's reaction to sunlight exposure commonly used by medical providers to describe patient skin color
- Limitations

## Image Distribution in Data

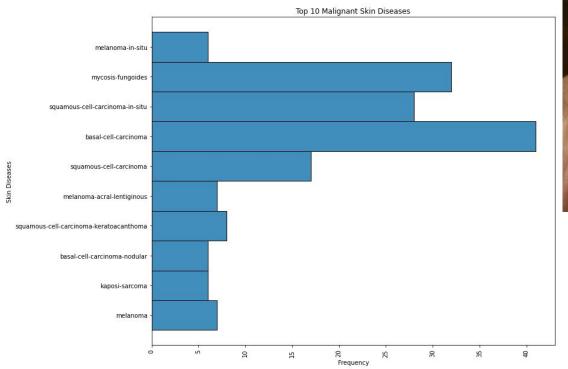
- 656 images
- 208 FST I-II Images (Fair Skin Tones)
  - o 159 benign, 49 malignant
- 241 FST III-IV Images (Medium Skin Tones)
  - o 167 benign, 74 malignant
- 207 FST V-VI Images (Dark Skin Tones)
  - o 159 benign, 48 malignant



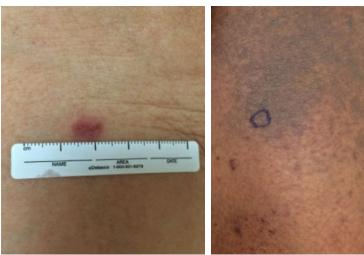
### **Data Science Workflow**

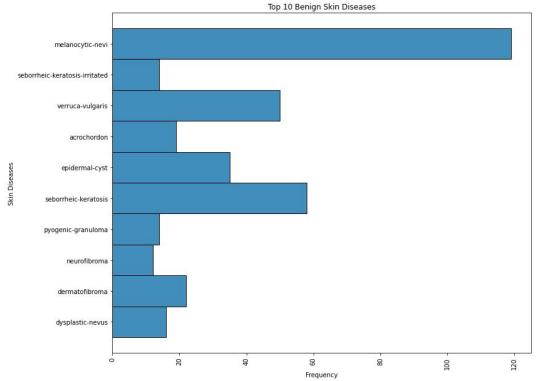


- Cleaned Data
- Exploratory Data Analysis













### Data Science Workflow

- Preprocessed images
- Multiple Convolutional Neural Network Models
  - Recall Optimization (Minimizing False Negatives)
- Best Model Results Report:

	precision	recall	fl-score	support
benign	0.89	0.88	0.88	48
malignant	0.67	0.71	0.69	17
accuracy			0.83	65
macro avg	0.78	0.79	0.78	65
weighted avg	0.83	0.83	0.83	65



benign 100% (benign)



malignant 100% (malignant)



benign 89% (benign)



malignant 100% (malignant)



benign 100% (benign)



malignant 100% (malignant)







benign 100% (malignant)

benign 64% (malignant)

benign 100% (malignant)







malignant 100% (benign)

malignant 100% (benign)

malignant 100% (benign)



#### Conclusions and Recommendations

- Training models on diverse populations
- Knowing model limitations
- Making sure models are up to date and ethical

#### **Final Thoughts:**

- The Diverse Dermatology Images Dataset needs more images, class imbalance should be addressed
- Model and application improvement

# Thank you!