

## DERMATOLOGY TEAM 14

Dr. Ricardo Henao, Adrian Lopez, Brian Kim, Kinjal Sanghvi, Meng Xiao, Matt Engelhard





- Problem Statement
- Dataset
- Solution Approach
- Active Learning
- Results
- Accomplishments & Going Forward

# PROBLEM STATEMENT

The dermatology department at the Duke Hospital has **25,509** images of skin lesions. All these images are unlabelled and scale from "a very clear centered lesion" to an image of a full body with tiny lesions.

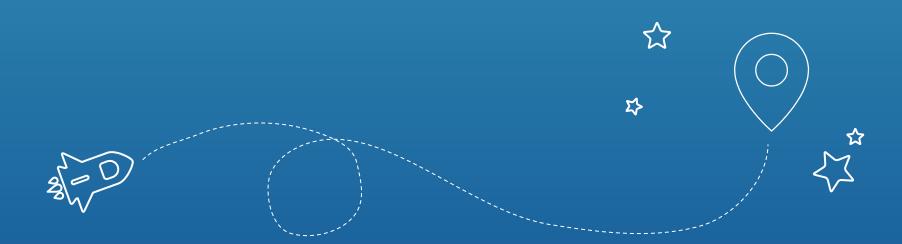
Our task is to train a model to create bounding boxes around each of these lesions and classify them.

### **DATASETS**

- 1. Our Dataset contains Duke Images as well as images from an online dataset (ISIC 2018).
- 2. To train our models, we use a combination of these images.
- We have minimal number of ground truths to the 25,509 images provided by Duke Hospital. Therefore, we utilize Active Learning in training the model to maximize performance gain for each label.





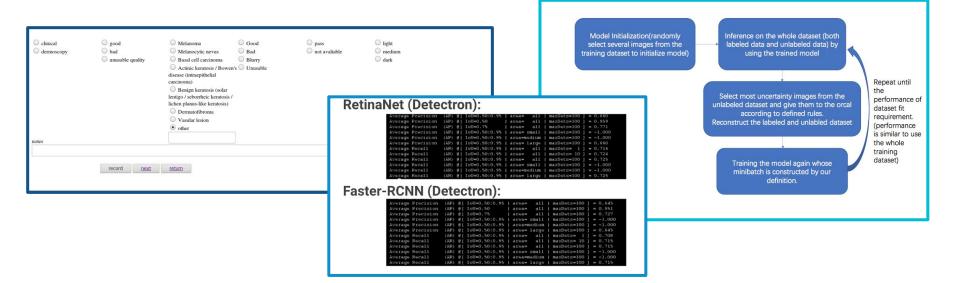


## **SOLUTION APPROACH**

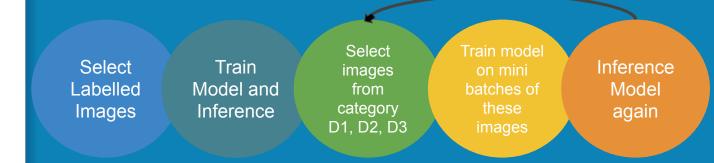


# TEST OBJECT DETECTION MODELS

ACTIVE LEARNING SOLUTION







D1 = 1 Random Labelled Image

D2 = 1 Labelled Image with Score  $\sim 1/(N+1)$ 

D3 = 2 Unlabelled Images with Score  $\sim 1/(N+1)$ 

N = Number of Classes, 1 = Background Class

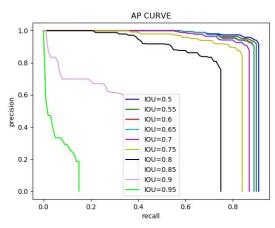


# RESULTS



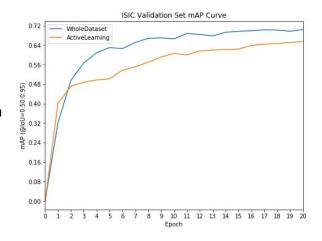
#### ISIC Dataset Results (RetinaNet):

#### Whole Dataset Training:

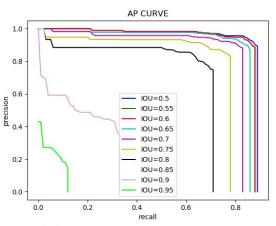


#### **Training Process:**

• Train on the whole dataset (2334 images) at every epoch



#### Active Learning:



#### **Training Process:**

- Train on random subset of 40 labelled images for 5 epochs
- At each following epoch, train on 20 mini-batches of 4 images selected based on D1, D2, D3 criteria
- Total # images used per epoch:80





## ACCOMPLISHMENTS & GOING FORWARD

### **Accomplishments:**

- Successfully set up an AL pipeline using RetinaNet
- 2. Tested and validated its effectiveness and efficiency using the ISIC dataset
- 3. Recognized some key problems with performance on Duke dataset: varying image quality, glare, size of lesions, etc.

### **Going Forward:**

- Incorporate the Duke clinical data into the pipeline
- Further automate the process by integrating the lesion app within the AL pipeline to serve as an oracle
- 3. Continue to improve the performance on the highly varying Duke dataset



# Thank You