

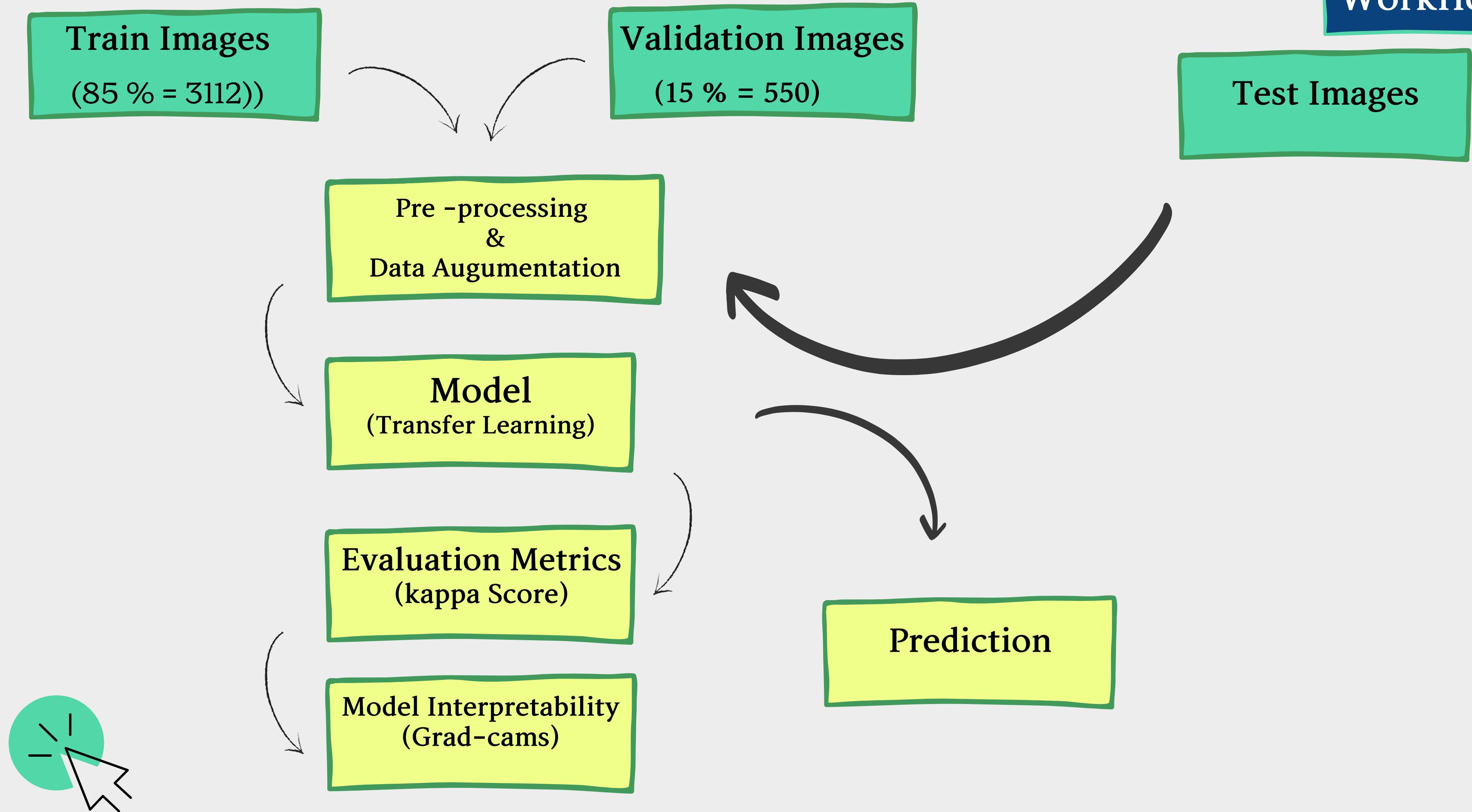
Classifying Diabetic Retinopathy from fundus image Using Deep Learning

JOVITA V
20-PDS-012



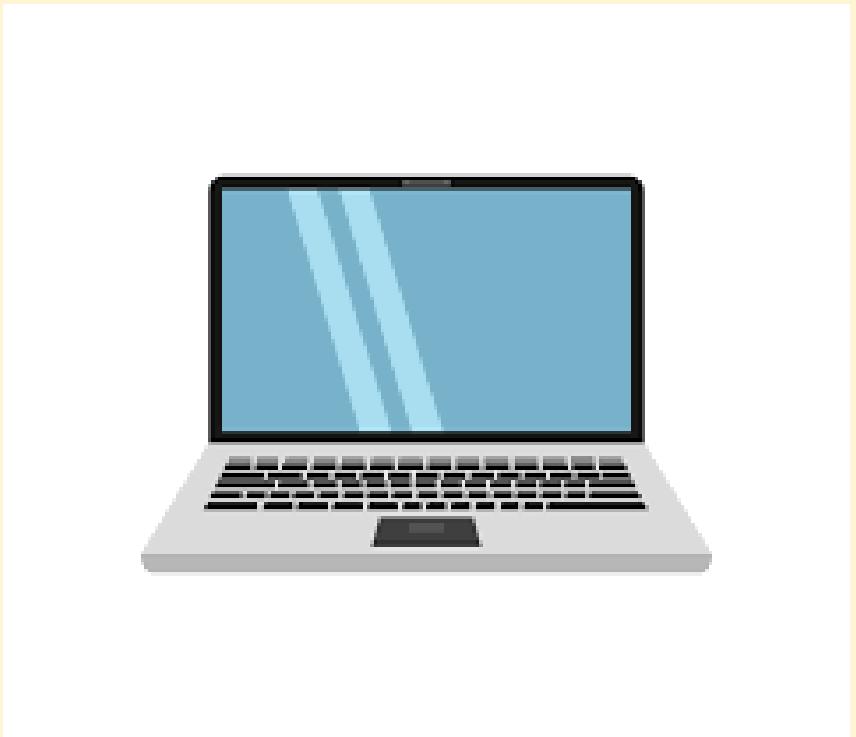
- Description of the workflow
- Hardware and software requirements
- Method of Solution
- Expected Outcome

Workflow



HARDWARE

- RAM : 4GB
- System Type : 64-bit operating system, x64-based processor
- Processor : AMD PRO A4-3350B APU with Radeon R4 Graphics 2.00 GHz
- Device Name : Lenova

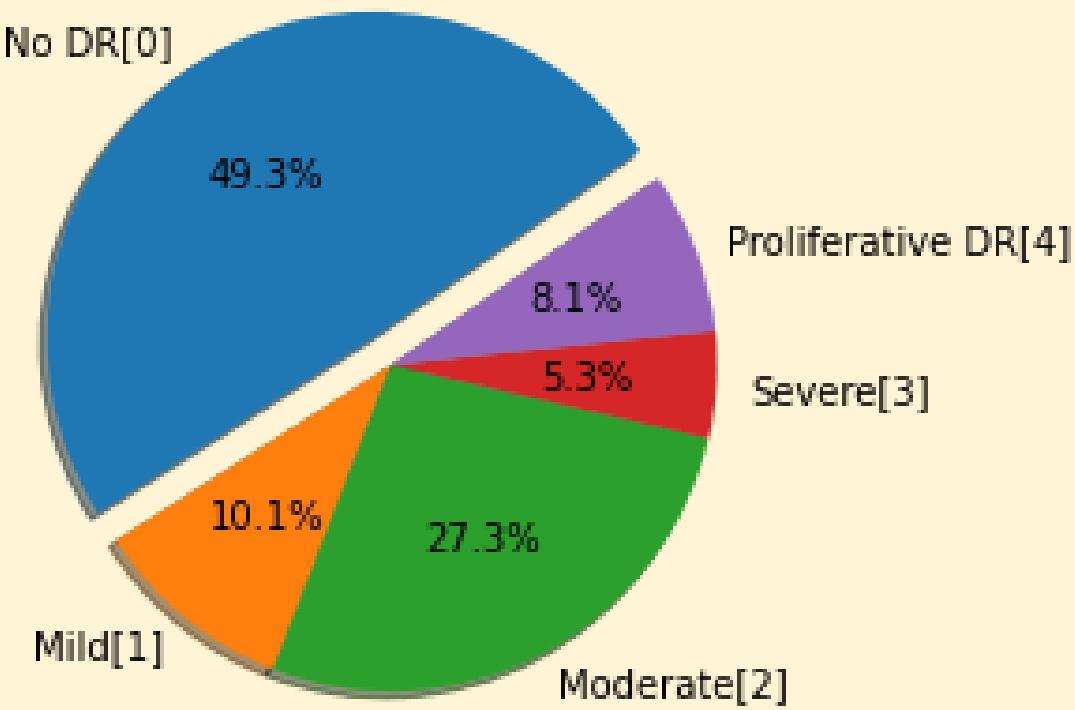


SOFTWARE

- Google Colab
- Programming Language : Python

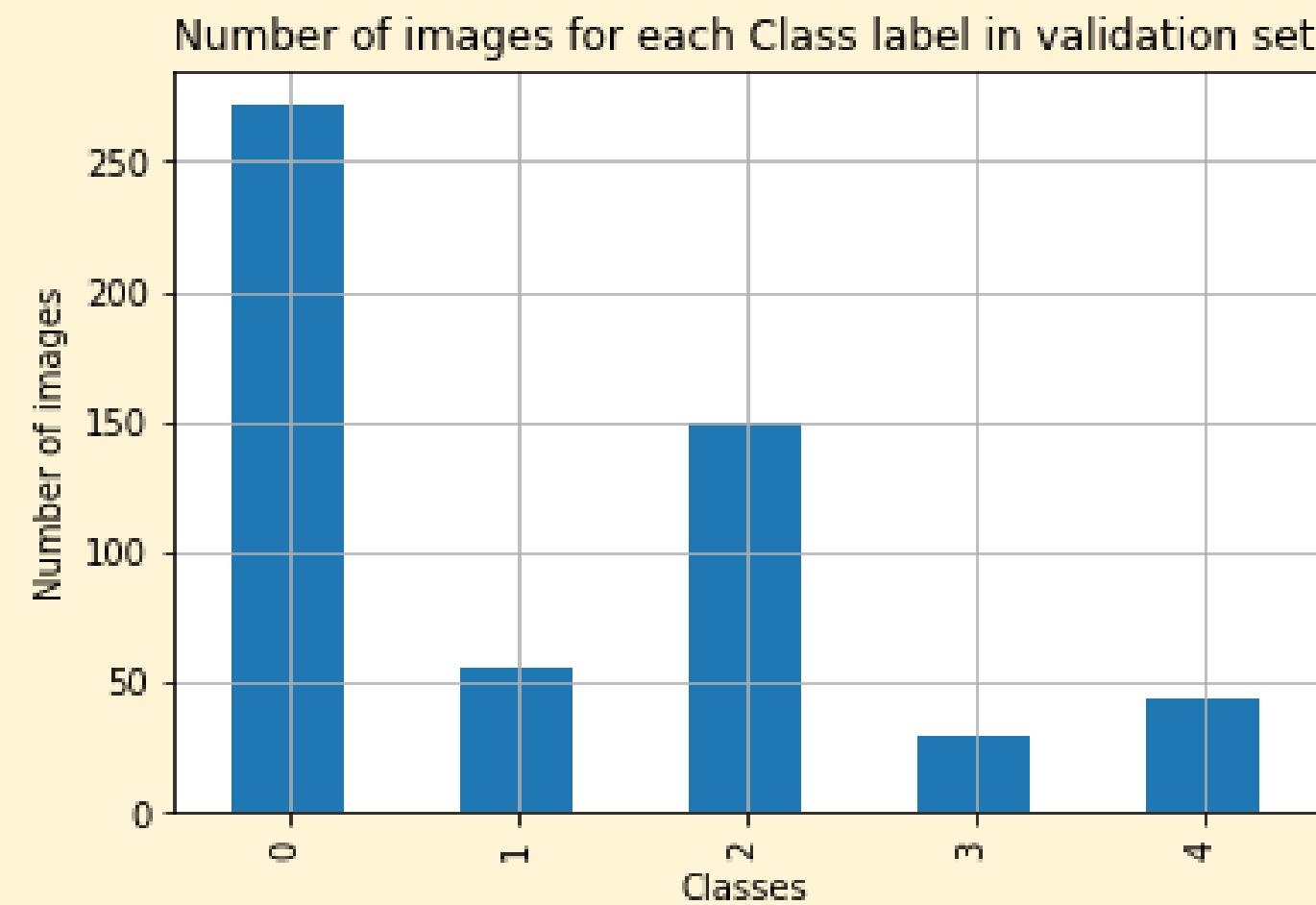
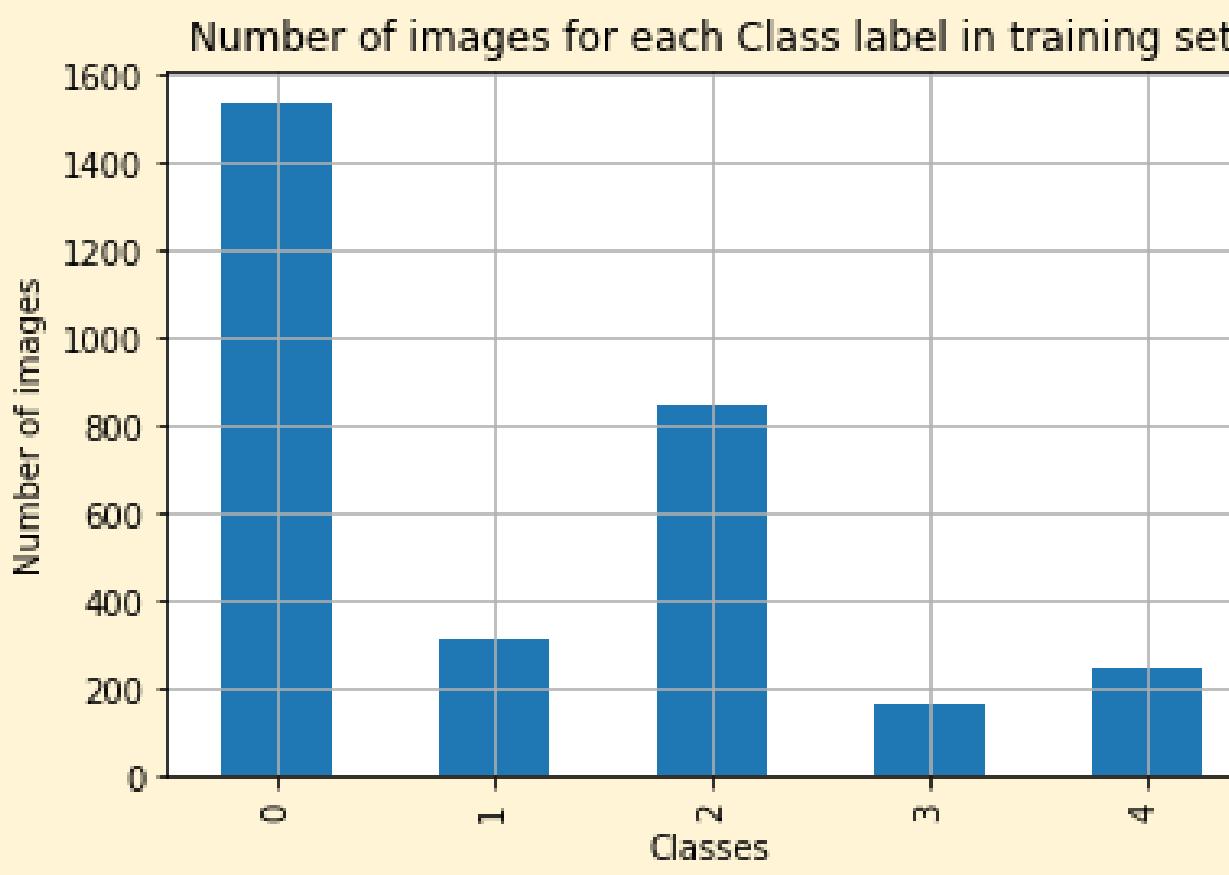


Pie Chart Analysis of Number of Images on each target label:

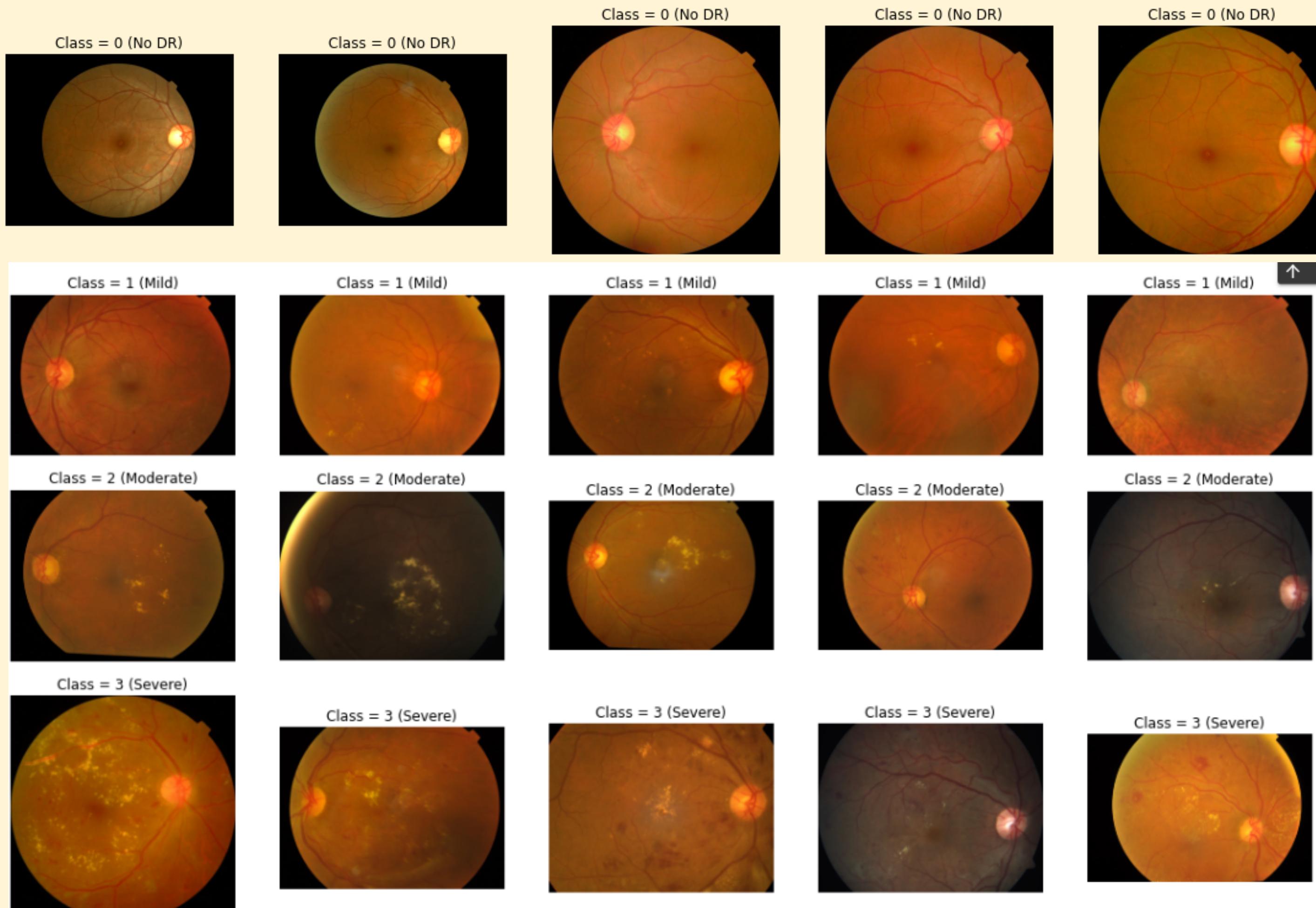


Train Images
(85 % = 3112))

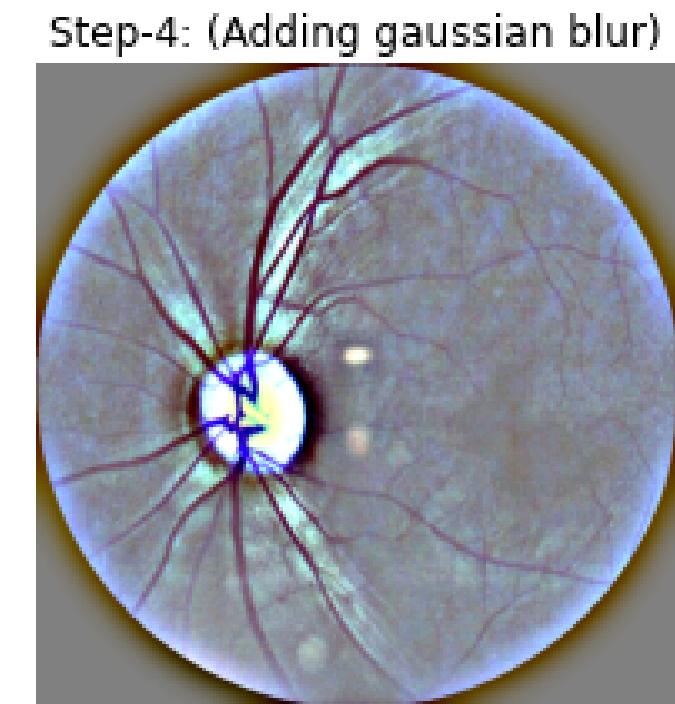
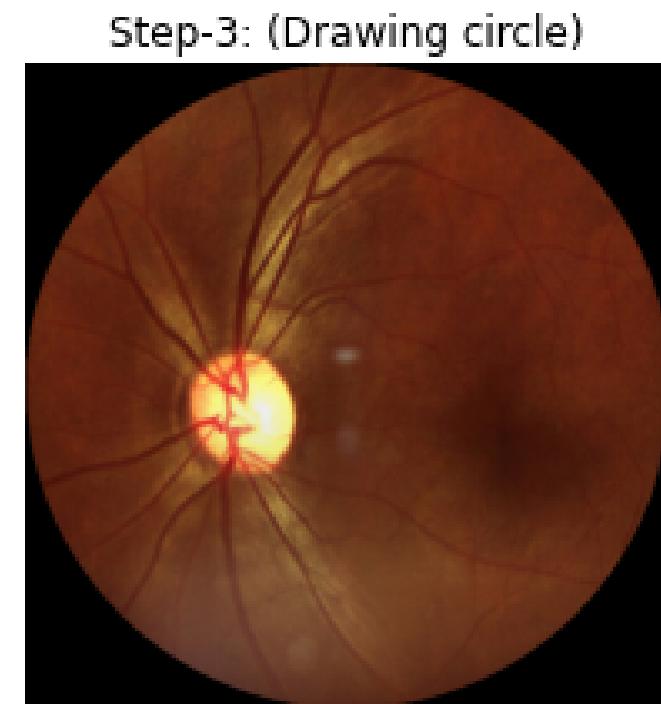
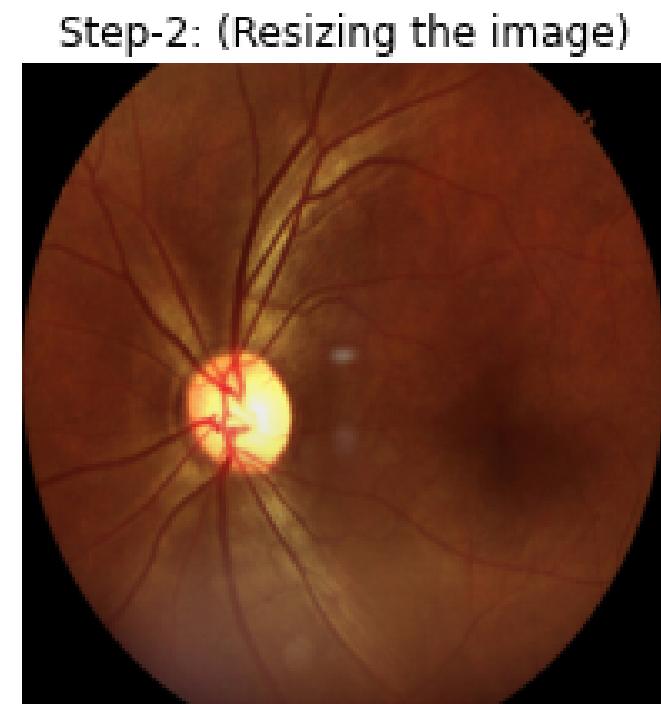
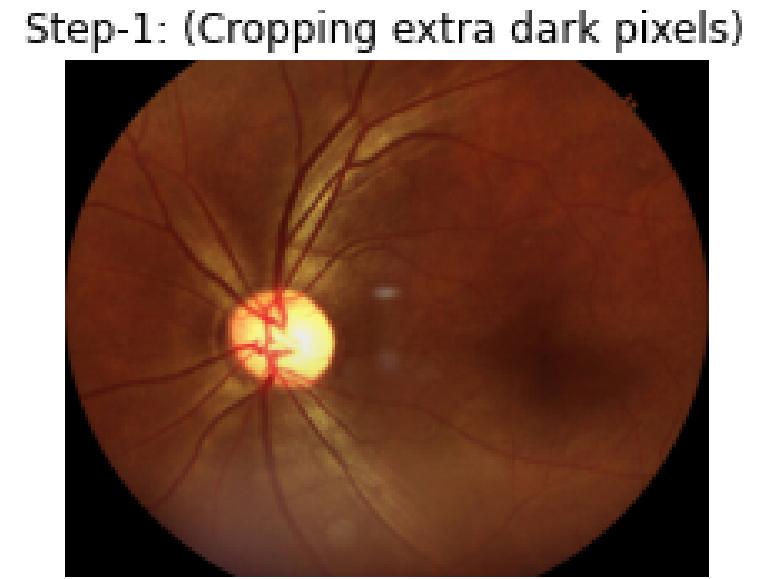
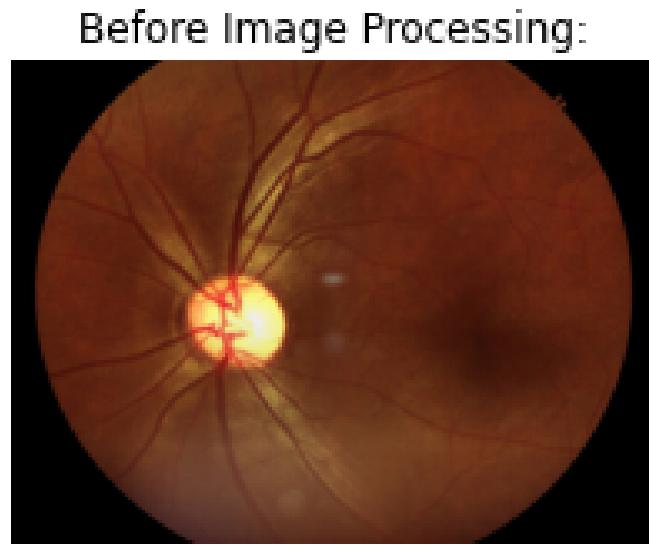
Validation Images
(15 % = 550)



Dataset



Data Pre-processing

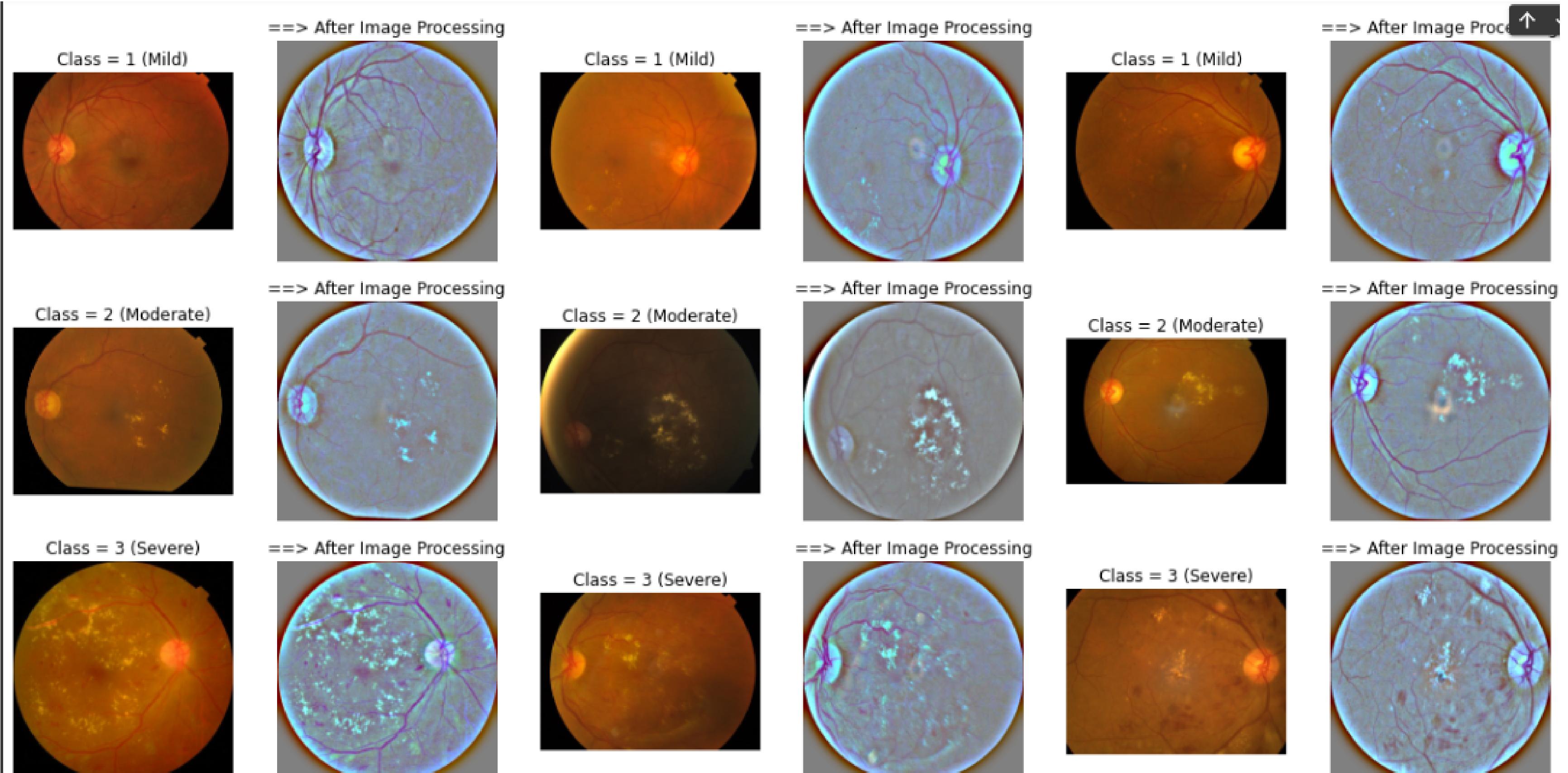


Cropping the
Extra Dark
Images

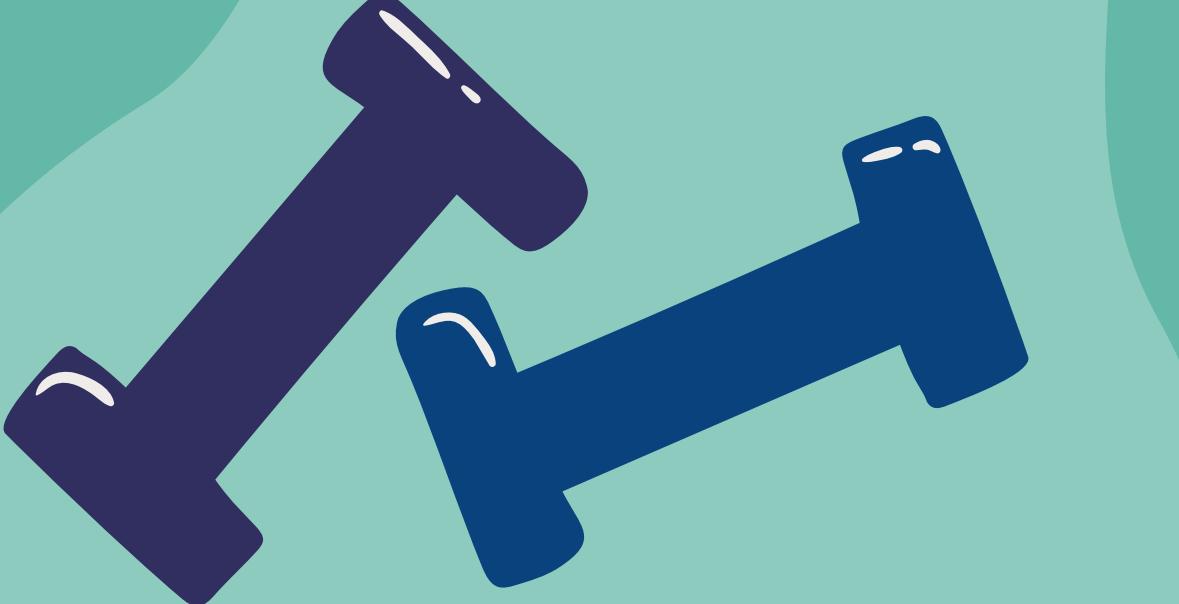
Resizing =
 $(512,512)$

Drawing a circle
from the center
to give a similar
shape to all the
images.

Using Smoothing
technique in
order to remove
the noise in
images



Computing Classweight:



Since the data set is highly imbalanced adding classweight can make it balanced by adding weights to each of the classes.

The class with more weight means they are less in samples. Similarly the class with less weight means they are more in samples.

```
class_weights = class_weight.compute_class_weight(class_weight='balanced',
                                                 classes = [0,1,2,3,4],
                                                 y = train_labels)

print(class_weights)

[0.40573664 1.98216561 0.73309776 3.79512195 2.47968127]
```

Data Augmented

```
def create_datagen():
    return ImageDataGenerator(
        zoom_range=0.2, # set range for random zoom
        rotation_range = 180,
        horizontal_flip=True, # randomly flip images
        vertical_flip=True, # randomly flip images
    )

# Using generator
data_generator = create_datagen().flow(x_train,train_labels, batch_size=BATCH_SIZE, seed=2019)
```

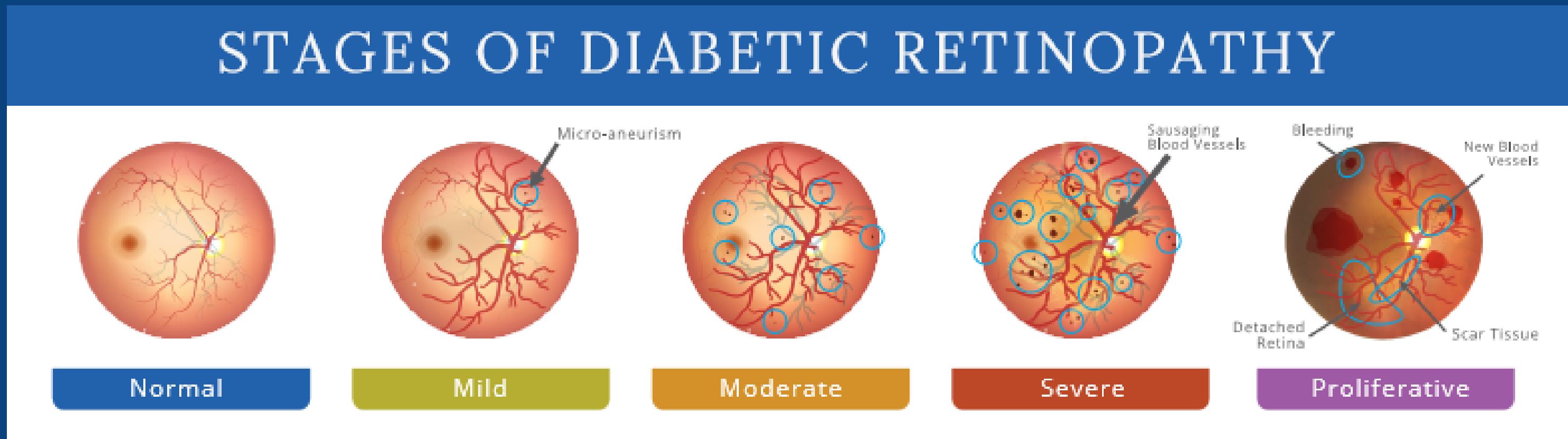
Cohen's Kappa:

$$\kappa \equiv \frac{p_o - p_e}{1 - p_e} = 1 - \frac{1 - p_o}{1 - p_e},$$

where,

Po is observed agreement (same as accuracy).

Pe is chance of disagreement observed between raters.



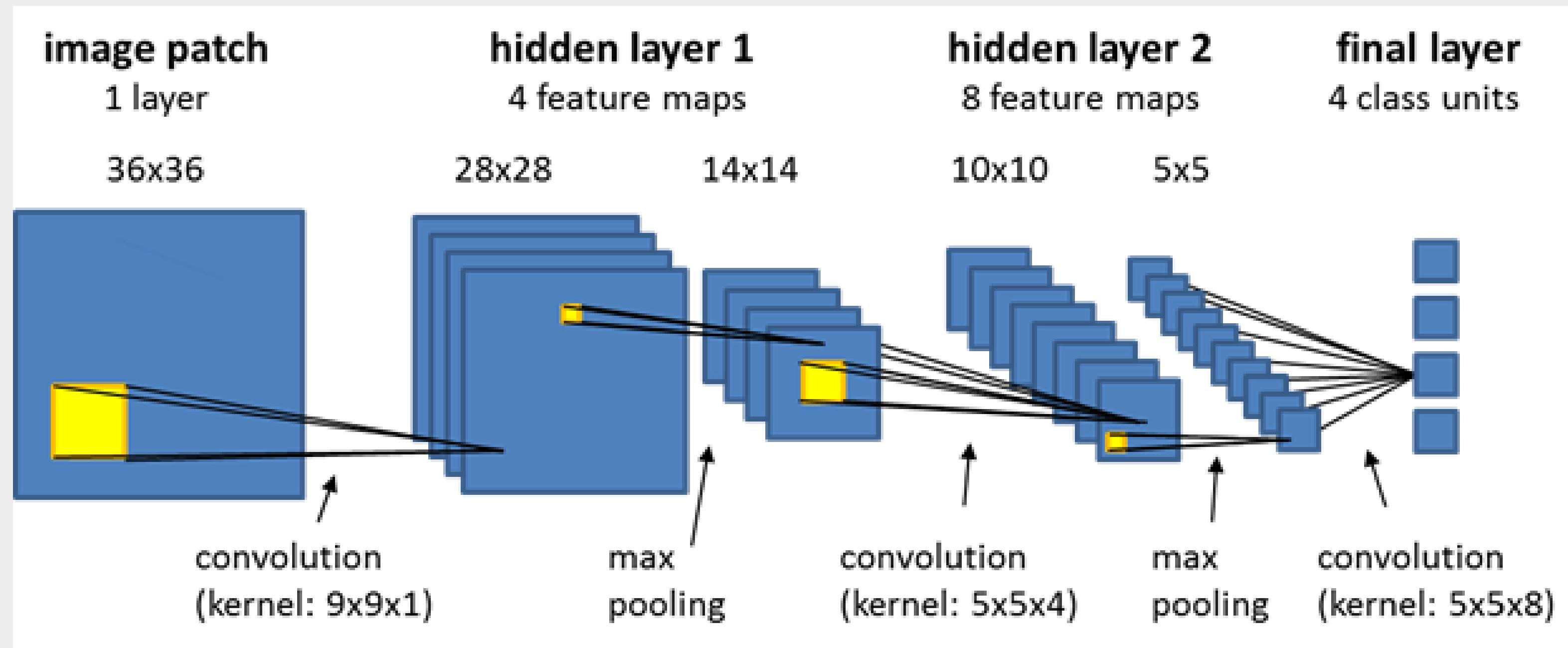
This property will be very helpful because in medical getting higher than expected will be okay because on later diagnosis it can be rectified.



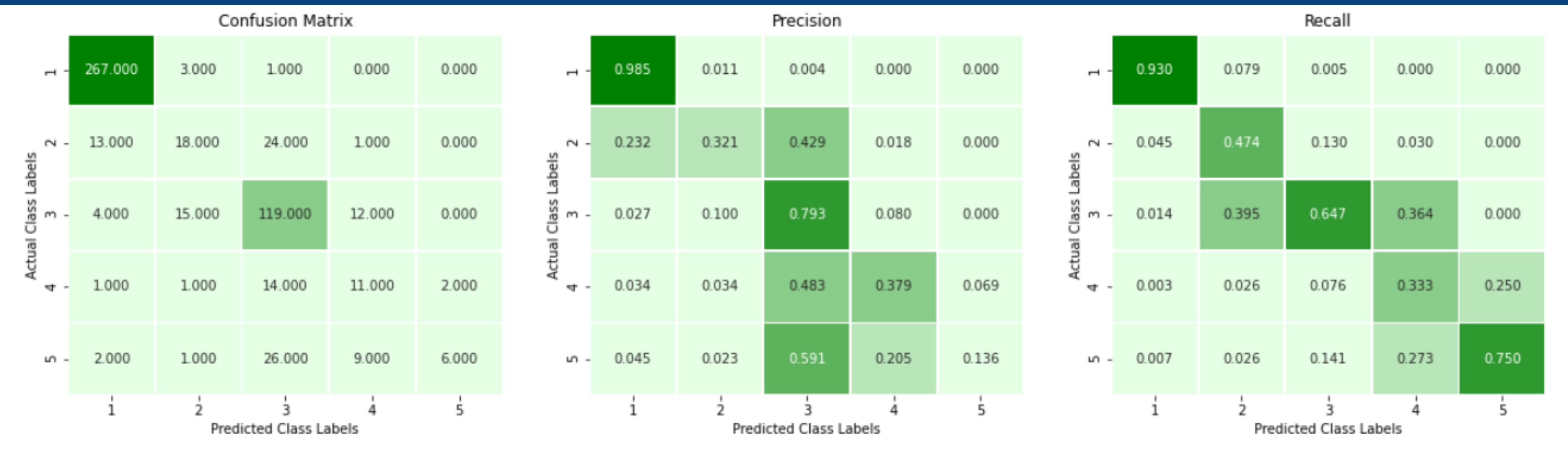
Callbacks

In order to print the kappa metrics
on each epoch on validation data

Training Model

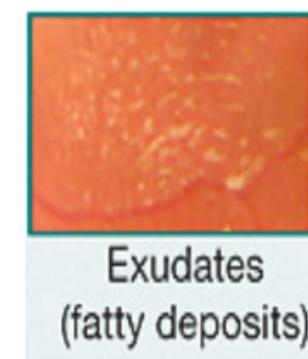
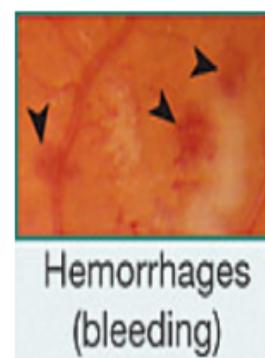
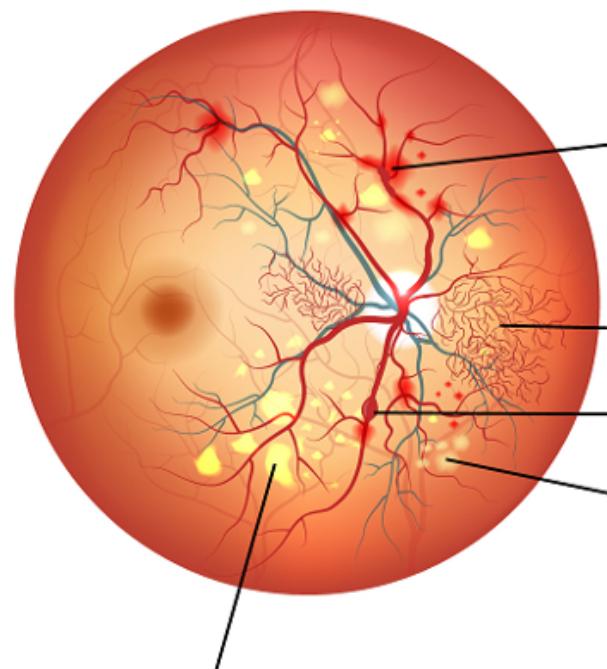


Expected Outcome:

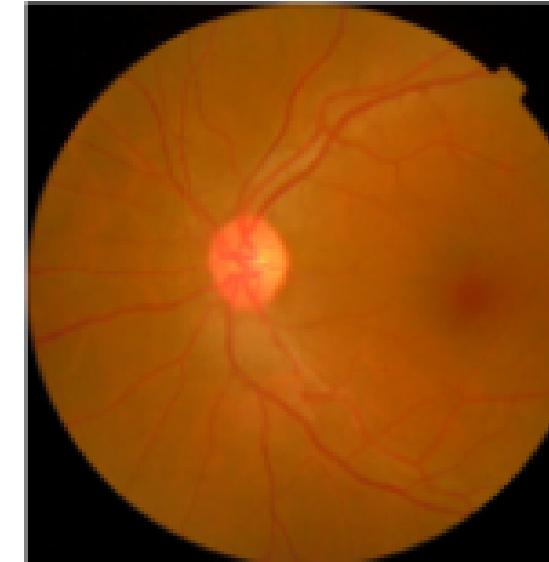


Model Interpretability (Grad-cams)

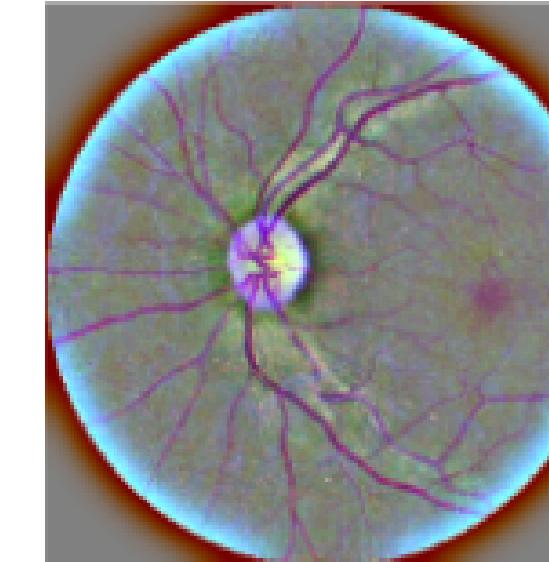
DIABETIC RETINOPATHY



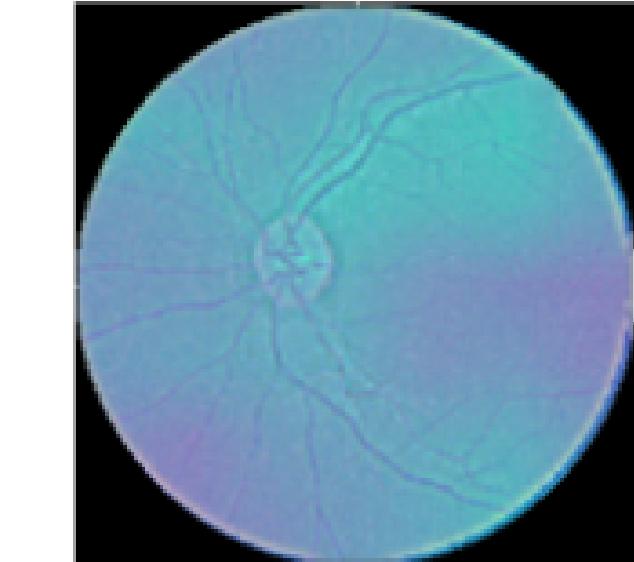
00f6c1be5a33



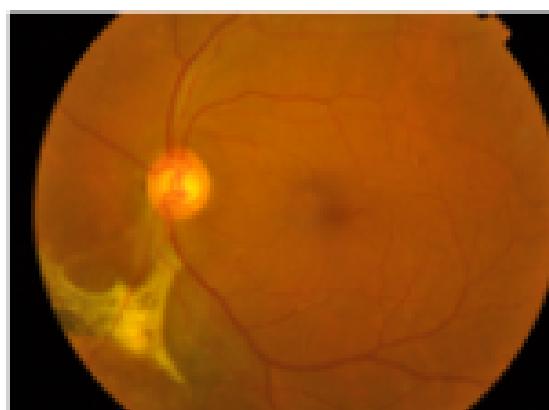
Target Class: 0



Predicted Class: 0



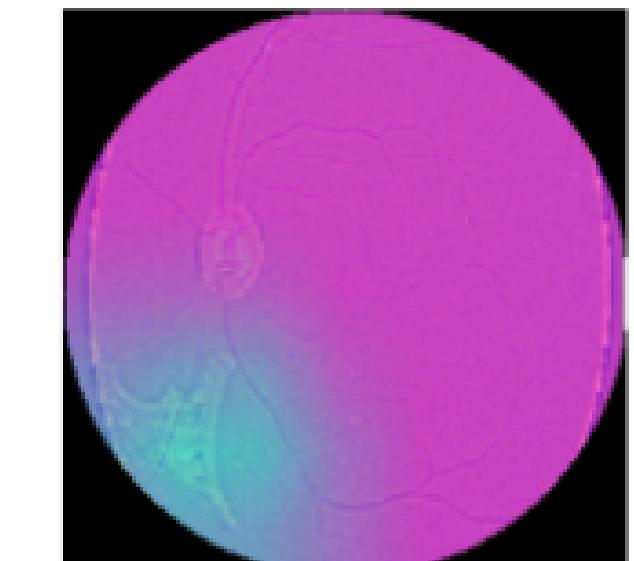
080f66eedfb9



Target Class: 4



Predicted Class: 4



Expected Outcome:



Thank You