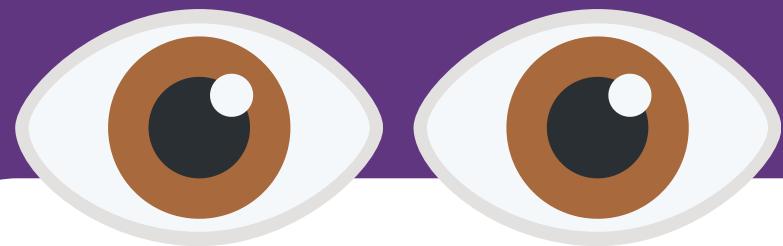
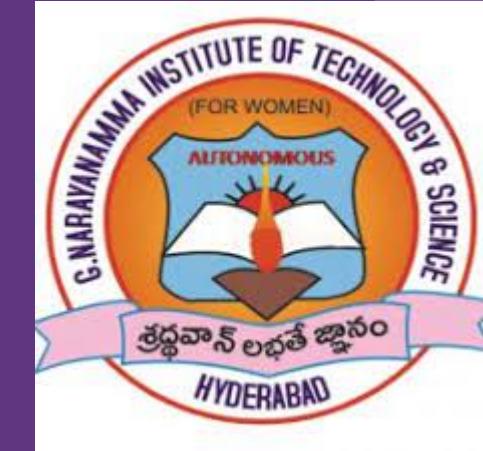


# G. Narayananamma Institute of Technology and Science

## (For Women)

Computer Science and Engineering Department



# A Novel Deep Learning Approach for Classification of Eye Diseases using Fundus Images

Guide:  
**Mrs. Y. Sravani Devi**  
Assistant Professor

- Sneha Boddupalli(21251A0569)
- Charani Gangireddygari(21251A0571)
- Kyama Kalpana(21251A0582)
- Mythili Lenkalapelli(21251A05C6)



# INTRODUCTION



| EYE DISEASE          | NUMBER OF PEOPLE<br>AFFECTED (WORLDWIDE)<br>APPROXIMATELY<br>(IN MILLION) |
|----------------------|---|
| Cataract             | 94  |
| Glaucoma             | 7.7   |
| Diabetic Retinopathy | 3.9   |

- Globally, at least 2.2 billion people have a near or distance vision impairment.
- Vision problems can arise due to various factors
- Such as age, genetics, and environmental influences
- Necessity of in time treatment to preserve eye health and ensure clear sight.

# LITERATURE SURVEY

| Sl.no | Title   | Author  | Algorithm  | Findings  |
|-------|---|---|--|---|
| 1     | Enhancing Ocular Healthcare: Deep Learning-Based Multi-Class Diabetic Eye Disease Segmentation and Classification               | MANEESHA VADDURI AND P.KUPPUSAMY (Member, IEEE) | Deep Convolutional Neural Networks (DCNN), Transfer Learning, image enhancement, segmentation, classification                                | Accuracies for the Cataract(CA): 96.43%, Diabetic Retinopathy (DR): 98.33%, Glaucoma (GL): 97%, and NORMAL: 96%                 |
| 2     | Deep Transfer Learning Strategy to Diagnose Eye-Related Conditions and Diseases: An Approach Based on Low-Quality Fundus Images | Gabriel D. A. Aranha                            | Deep Transfer Learning, Convolutional Neural Networks (CNNs), VGG16 Architecture, Low-Quality Fundus Images, Fine-Tuning Pre-Trained Models. | Accuracies of 87.4%, 90.8%, 87.5%, 79.1% to classify cataract, diabetic retinopathy, excavation and blood vessels, respectively |

# LITERATURE SURVEY

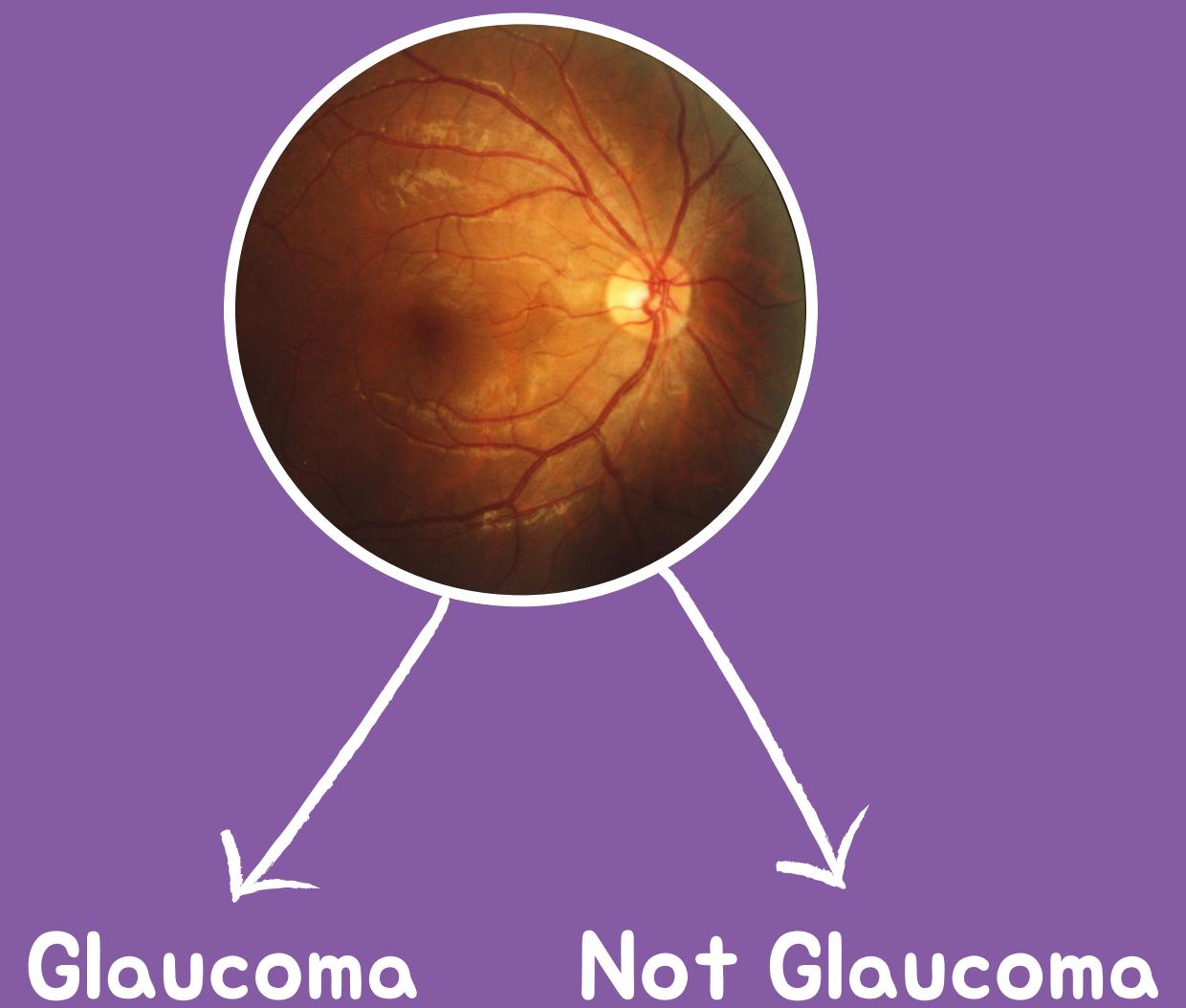
| Sl.no | Title   | Author  | Algorithm  | Findings  |
|-------|---|---|--|---|
| 3     | Artificial Intelligence-Driven Eye Disease Classification Model   | EM. Onyema  | AI-driven classification, multi-class image classification, uncertainty analysis, performance metrics      | Precision: 85%, Specificity: 90%, Sensitivity: 88% for diabetic retinopathy   |
| 4     | The Proposed Convolutional Neural Network Architecture for the Detection and Classification of Eye Diseases | Rahul Singh, Neha Sharma, Rupesh Gupta from Chitkara University Institute of Engineering and Technology | Convolutional Neural Network (CNN), Adam optimizer, multi-class classification, confusion matrix analysis. | The model achieves a 94% overall accuracy in classifying eye conditions, with 87.4% for Cataract, 90.8% for Diabetic Retinopathy, 87.5% for Glaucoma, and 79.1% for blood vessels. This underscores its effectiveness in diagnosing various eye diseases. |

# LITERATURE SURVEY

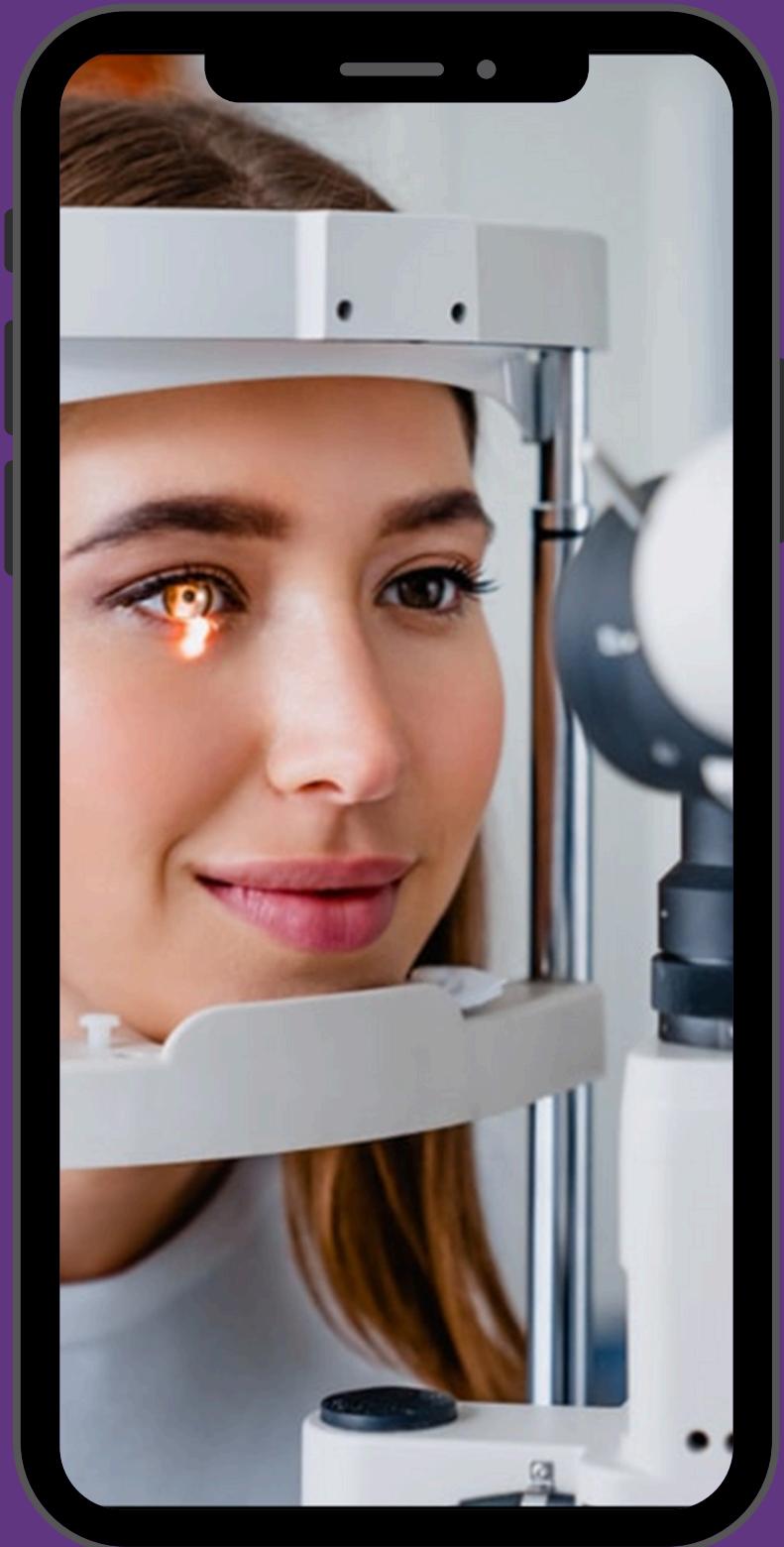
| Sl.no | Title  | Author  | Algorithm   | Findings  |
|-------|--|---|---|---|
| 5     | Glaucoma Retinal Image Synthesis Using the GAN   | N. Kalyani AND G. Sahasra, K. Sathwika, K. Arthi, L. Sharvani<br>(G. Narayananamma Institute of Technology and Science) | Generative Adversarial Network (GAN), Deep Convolutional GAN (DCGAN), Classification(CNN) | Accuracies for DenseNet: 95.83%, MobileNet V2: 92.10%, Inception V3: 90.20%, and Xception: 93.50% in binary classification for Glaucoma screening.                      |
| 6     | A deep transfer learning approach for identification of diabetic retinopathy using data augmentation | Yerrarapu Sravani Devi AND Singam Phani Kumar (GITAM Deemed to be University)   | Deep Learning(ResNet152), Data Augmentation(Gaussian Blur, Circle Crop)                   | Accuracies of 91% using synthetic data and 86% without synthetic data for diabetic retinopathy identification through a deep transfer learning approach with ResNet 50. |

# EXISTING SYSTEM

- Pre-trained Convolutional Neural Networks (CNNs) Transfer Convolutional Neural Networks (TCNNs)
- Hybrid Neural Networks
- Binary classification
- Severity Check Models



# SMART EYE CARE SYSTEM



Accurate identification of Eye Disease Classification (EDC) retinal fundus images is vital for patients' vision. We need an effective diagnostic model that enhances images and accurately identifies EDC, enabling early diagnosis and personalized treatment using Neural Network.

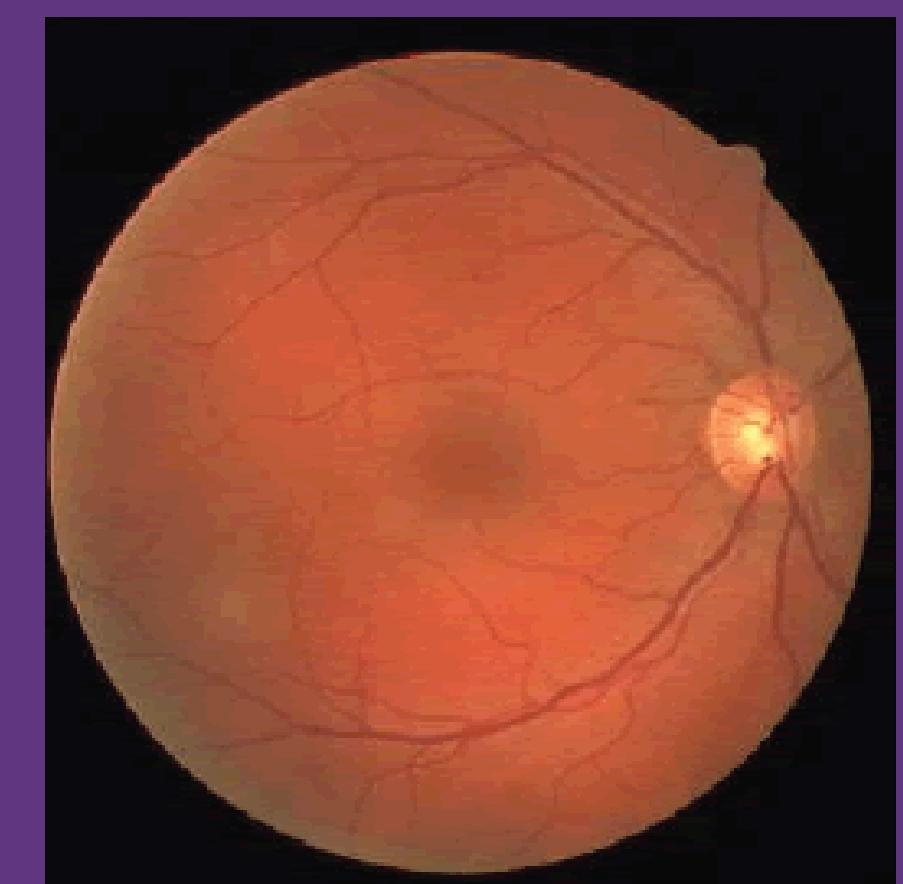
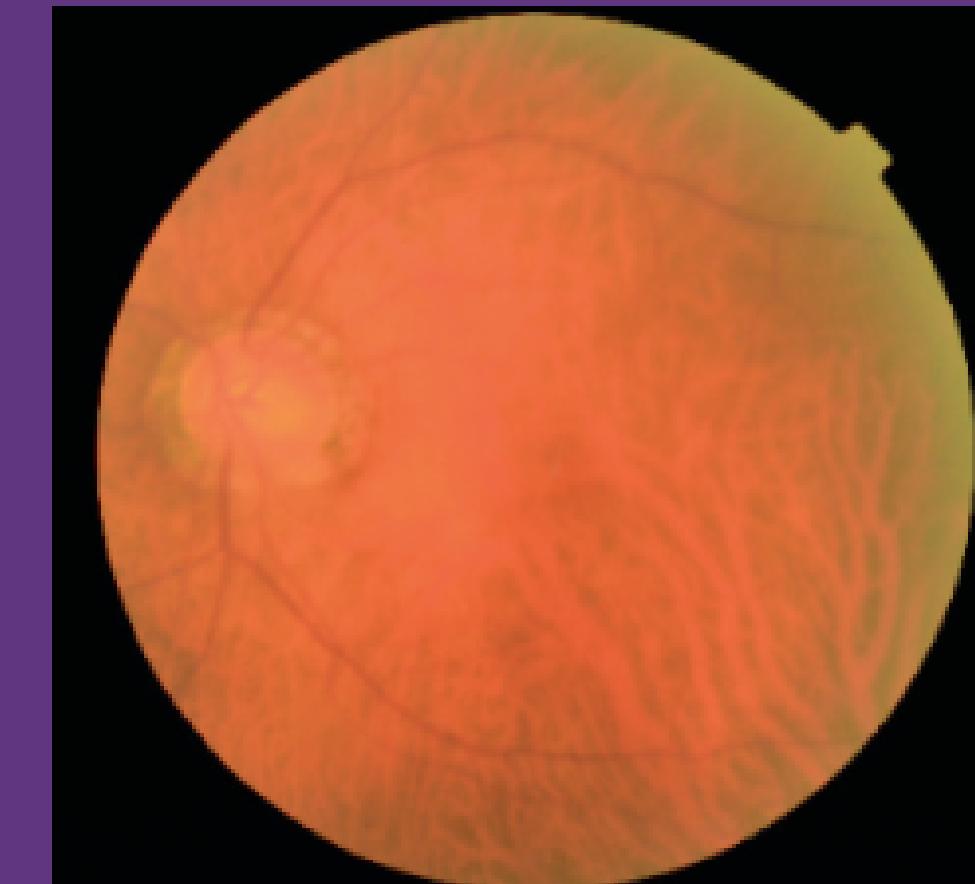
# OBJECTIVES

- 1. To collect data for EDC**
- 2. To adjust scalar pixel and train the Models**
- 3. To compare different architecture models for EDC using Evaluation Metrics**

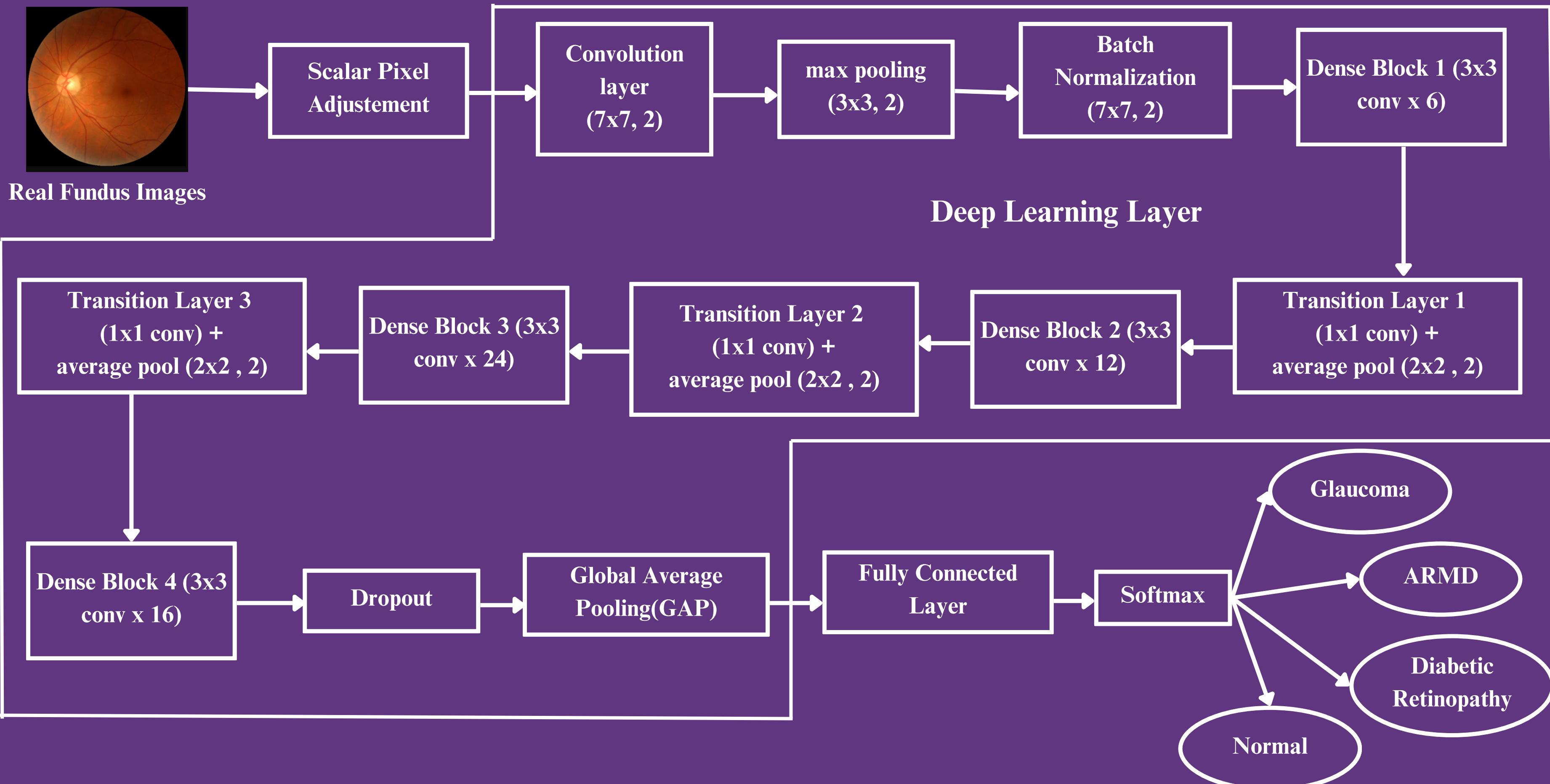
# DATASETS

## Fundus Images

- Publicly available Kaggle datasets used for training  
(4000 images => each 1000 including normal)
- Local dataset from LV Prasad Hospital for testing the model



# ARCHITECTURE



# METHODOLOGY

Fundus RGB Images are taking as input and converted into feature maps for image processing

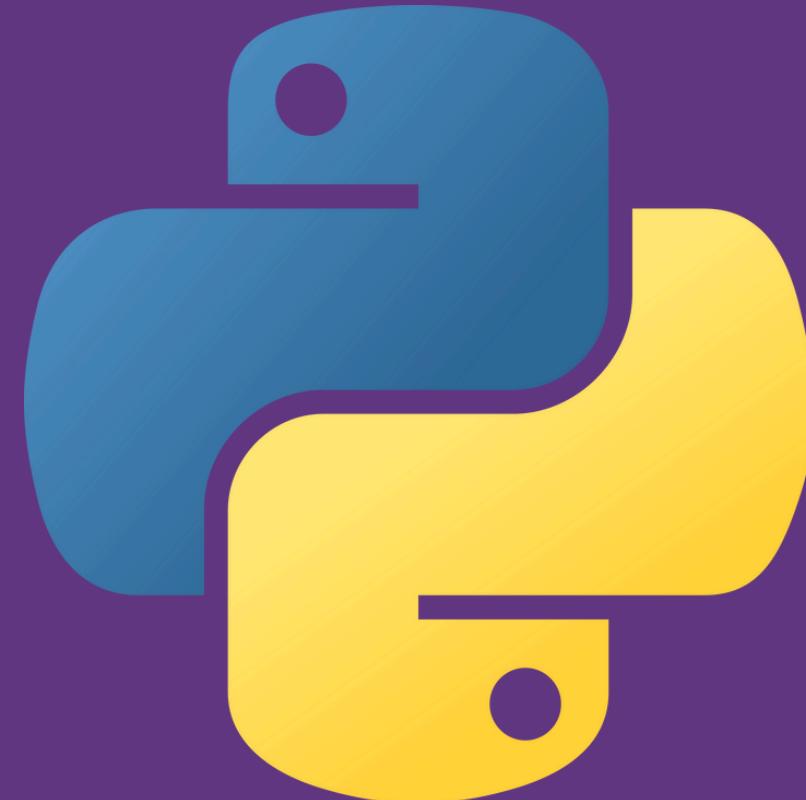
| Sl.no. | Layer                 | Description  |
|--------|-----------------------|--|
| 1      | Convolution Layer     | Uses feature Maps to extract patterns or features from input                 |
| 2      | ReLU                  | A non-linear activation function that increases non-linearity in images      |
| 3      | Max Polling           | Decreases the size of Feature maps taking the largest element from map       |
| 4      | Fully Connected Layer | Connects the neurons among different layers                                  |
| 5      | Softmax               | An activation function used in multi-class classification in neural networks |

# MODULES

- 1) Training using different architecture models on Kaggle dataset**
- 2) Changing the parameters for all models on Kaggle dataset**
- 3) Classifying into different eye diseases or normal eye on real dataset from locality on best model**
- 4) Creating a GUI(Website) out of best model to doing predictions**

# LIBRARIES USED

- TensorFlow and Keras
  - tensorflow & keras
  - layers
  - optimizers
  - metrics
  - regularizers
  - ImageDataGenerator
  - Model, load\_model, Sequential
- NumPy
- Pandas
- shutil, time, os
- cv2
- tqdm
- scikit-learn
  - train\_test\_split
  - confusion\_matrix
  - classification\_report
- matplotlib and seaborn
- PIL
- IPython
- logging
- Flask
- render\_template
- request



# RESULTS

- VGG16  
(LR=0.001, BS=60, HF=T)

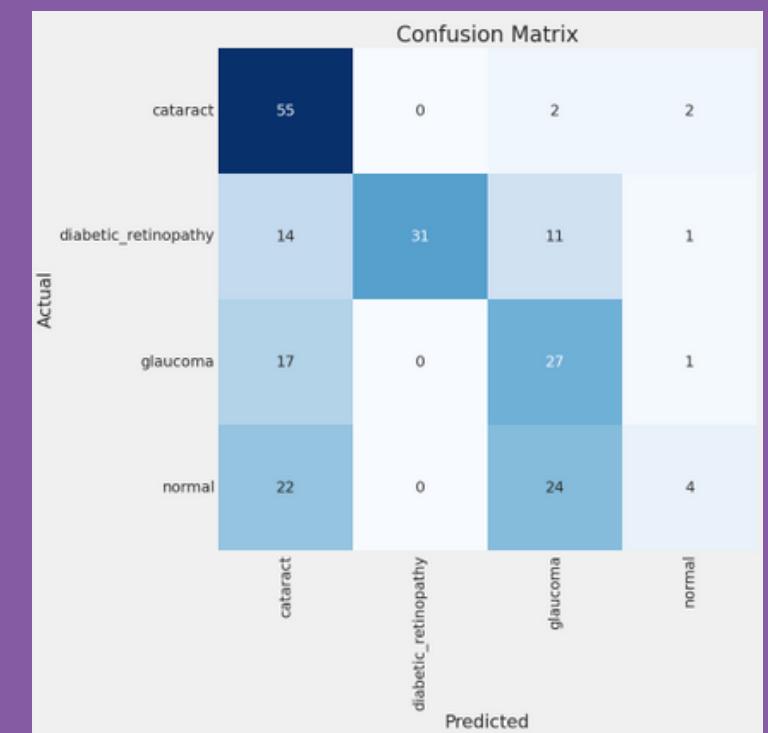
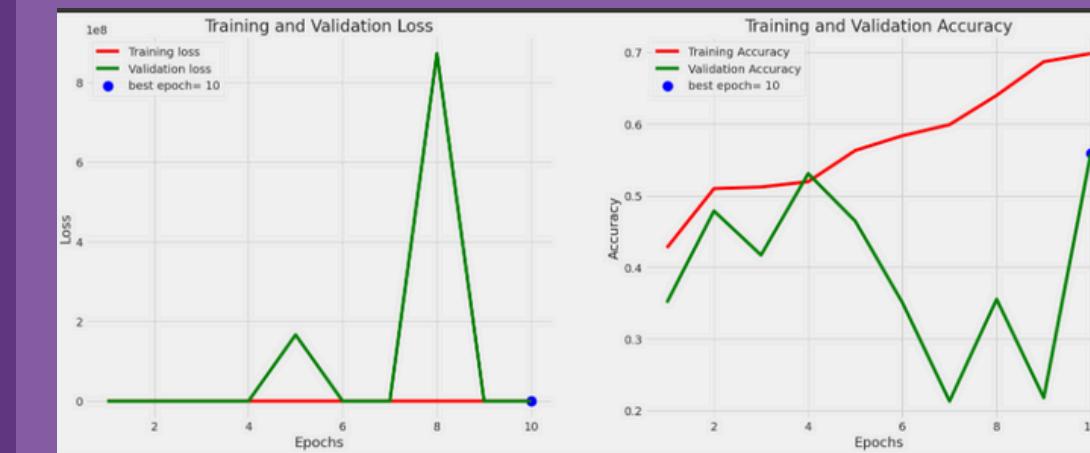
```
211/211 [=====] - 3s 13ms/step - loss: 0.
accuracy on the test set is 91.94 %

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.
    saving_api.save_model(
model was saved as ./VGG16-eye_disease-91.94.h5

class csv file was saved as ./class_dict.csv
```

| Classification Report: |           |        |          |         |
|------------------------|-----------|--------|----------|---------|
|                        | precision | recall | f1-score | support |
| cataract               | 0.97      | 0.95   | 0.96     | 59      |
| diabetic_retinopathy   | 0.97      | 0.98   | 0.97     | 57      |
| glaucoma               | 0.84      | 0.91   | 0.87     | 45      |
| normal                 | 0.89      | 0.82   | 0.85     | 50      |
| accuracy               |           |        | 0.92     | 211     |
| macro avg              | 0.91      | 0.92   | 0.91     | 211     |
| weighted avg           | 0.92      | 0.92   | 0.92     | 211     |

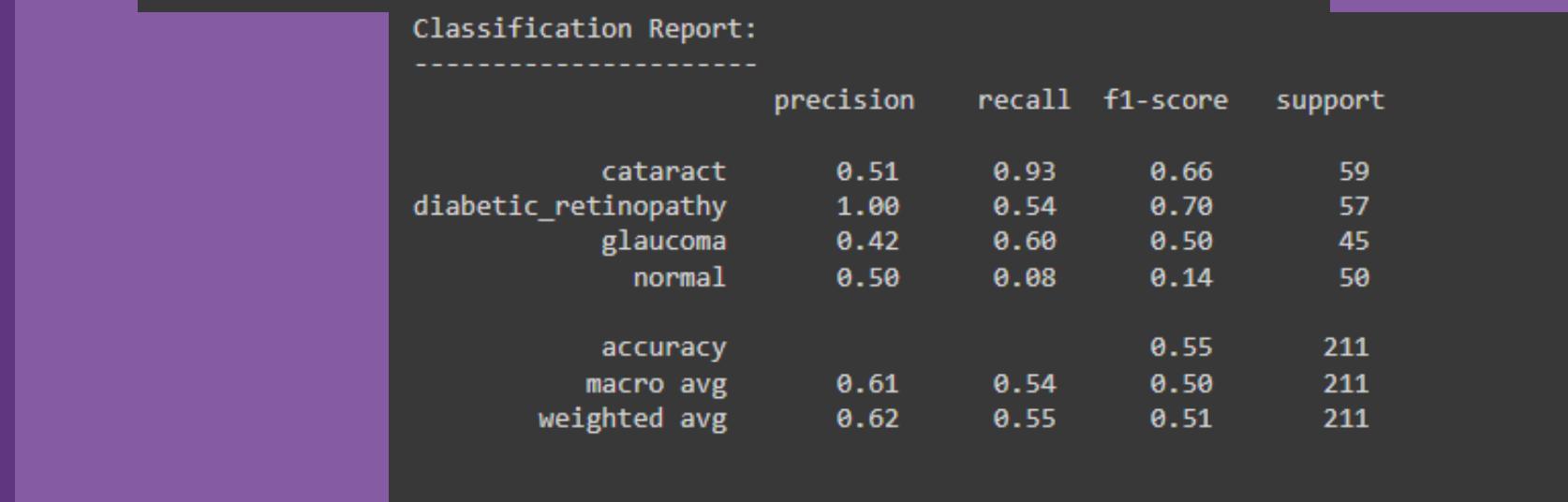
- VGG16  
(LR=0.01, BS=100, HF=T)



```
211/211 [=====] - 4s 11ms/step -
accuracy on the test set is 55.45 %

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.
    saving_api.save_model(
model was saved as ./VGG16-eye_disease-55.45.h5

class csv file was saved as ./class_dict.csv
```



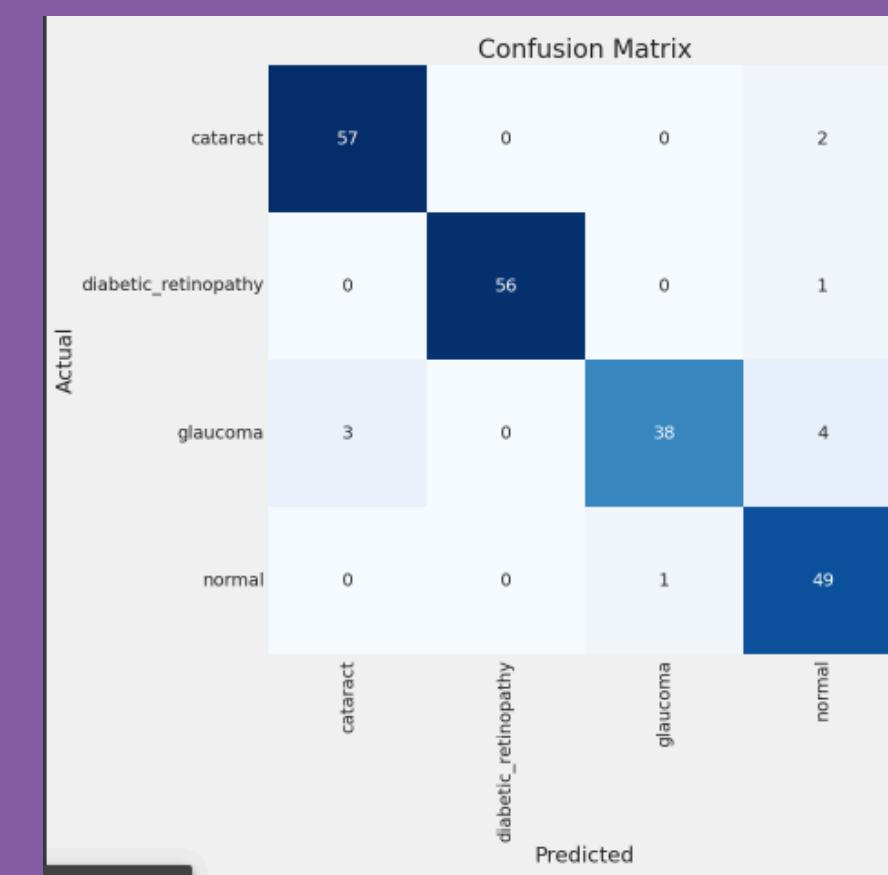
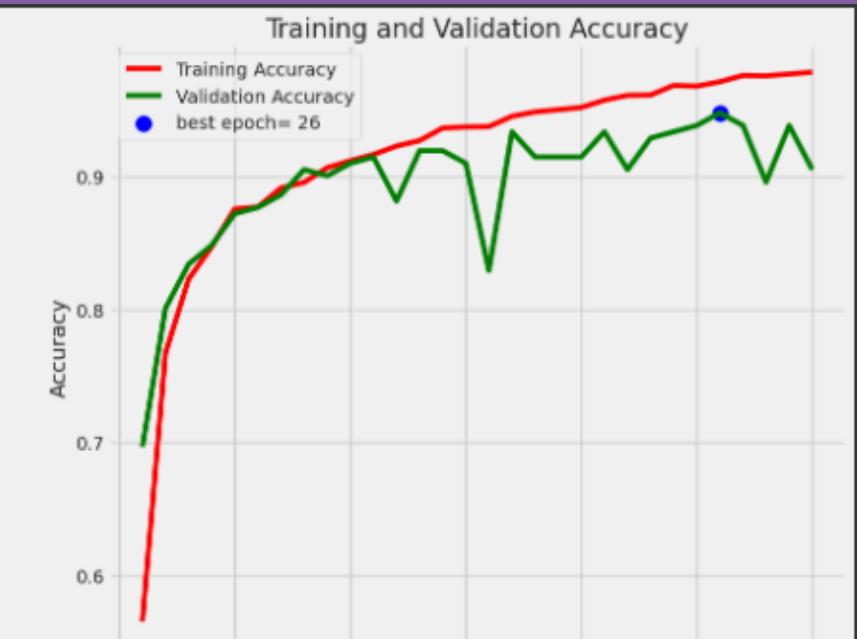
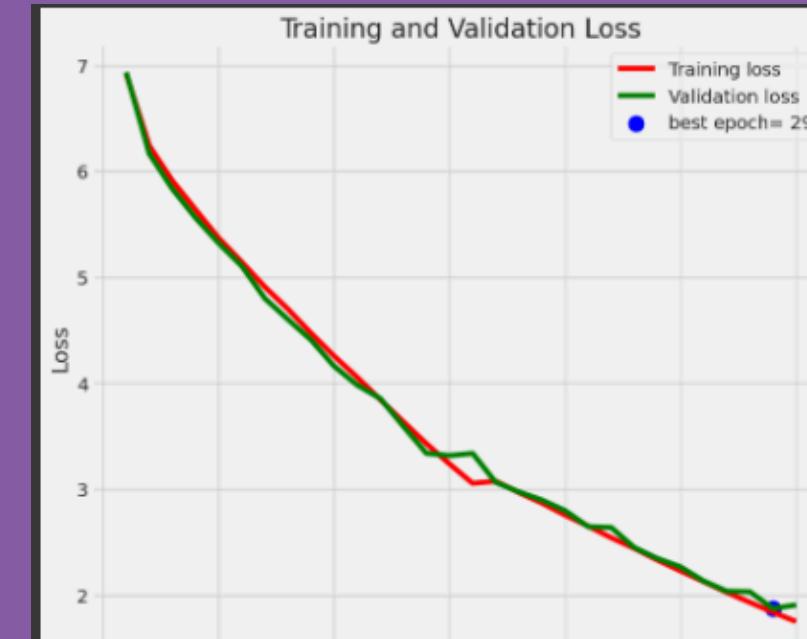
## • VGG16 (LR=0.01, BS=100, HF=T)

```
211/211 [=====] - 3s 12ms/step - loss: 1.8563 -
accuracy on the test set is 94.79 %

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103
    saving_api.save_model(
model was saved as ./VGG16-eye_disease-94.78.h5

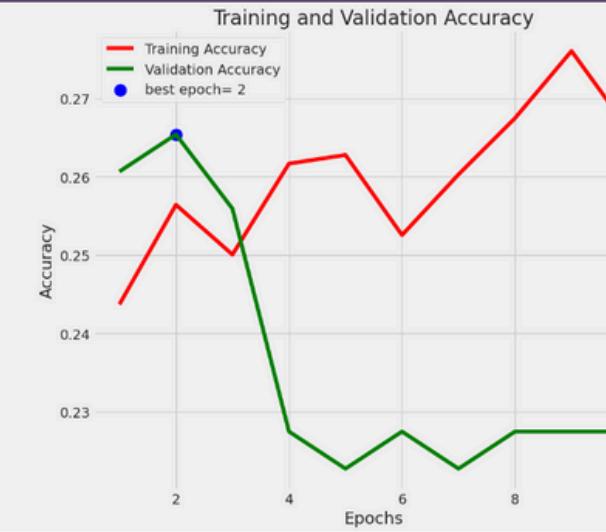
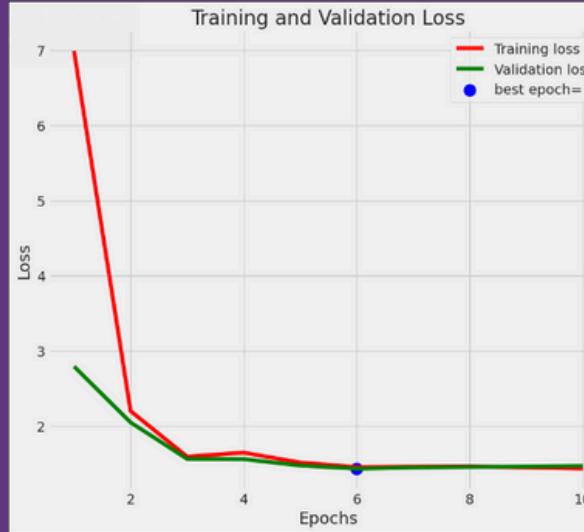
class csv file was saved as ./class_dict.csv
```

| Classification Report: |           |        |          |         |
|------------------------|-----------|--------|----------|---------|
|                        | precision | recall | f1-score | support |
| cataract               | 0.95      | 0.97   | 0.96     | 59      |
| diabetic_retinopathy   | 1.00      | 0.98   | 0.99     | 57      |
| glaucoma               | 0.97      | 0.84   | 0.90     | 45      |
| normal                 | 0.88      | 0.98   | 0.92     | 50      |
| accuracy               |           |        | 0.95     | 211     |
| macro avg              | 0.95      | 0.94   | 0.94     | 211     |
| weighted avg           | 0.95      | 0.95   | 0.95     | 211     |



# • ResNet

(LR=0.0001, BS=60, HF=T)



accuracy on the test set is 29.86 %

```
/usr/local/lib/python3.10/dist-packages/keras/src/engine/trai
    saving_api.save_model(
        model was saved as ./ResNet152-eye_disease-29.85.h5
        class csv file was saved as ./class_dict.csv
```

Classification Report:

|                      | precision | recall | f1-score | support |
|----------------------|-----------|--------|----------|---------|
| cataract             | 0.29      | 0.97   | 0.45     | 59      |
| diabetic_retinopathy | 0.00      | 0.00   | 0.00     | 57      |
| glaucoma             | 0.43      | 0.13   | 0.20     | 45      |
| normal               | 0.00      | 0.00   | 0.00     | 50      |
| accuracy             | 0.18      | 0.27   | 0.30     | 211     |
| macro avg            | 0.17      | 0.30   | 0.17     | 211     |
| weighted avg         | 0.17      | 0.30   | 0.17     | 211     |



# • ResNet

(LR=0.001, BS=100, HF=T)



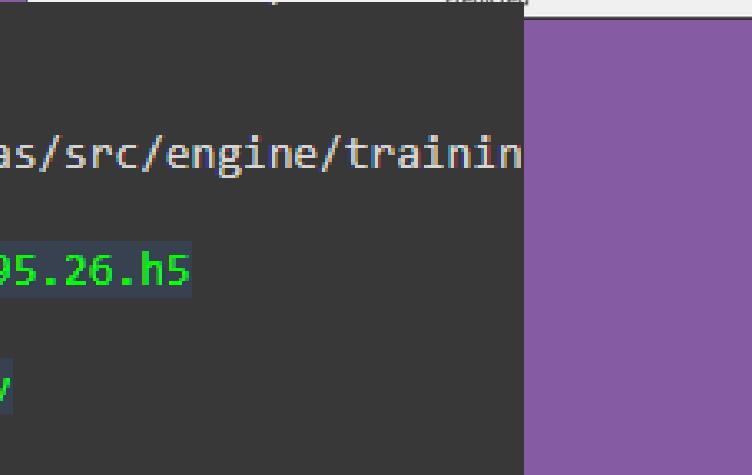
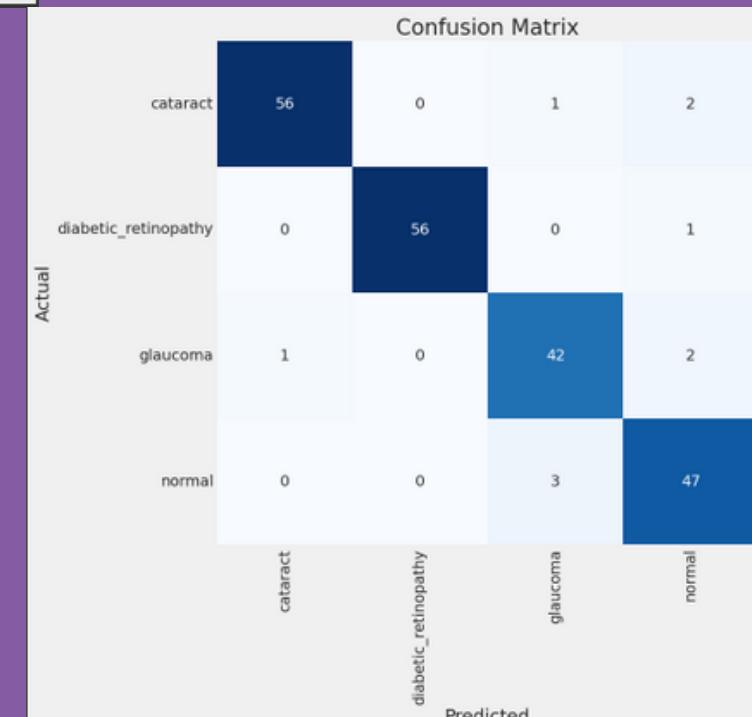
Classification Report:

|                      | precision | recall | f1-score | support |
|----------------------|-----------|--------|----------|---------|
| cataract             | 0.98      | 0.95   | 0.97     | 59      |
| diabetic_retinopathy | 1.00      | 0.98   | 0.99     | 57      |
| glaucoma             | 0.91      | 0.93   | 0.92     | 45      |
| normal               | 0.90      | 0.94   | 0.92     | 50      |
| accuracy             |           |        | 0.95     | 211     |
| macro avg            | 0.95      | 0.95   | 0.95     | 211     |
| weighted avg         | 0.95      | 0.95   | 0.95     | 211     |

accuracy on the test set is 95.26 %

```
/usr/local/lib/python3.10/dist-packages/keras/src/engine/trai
    saving_api.save_model(
        model was saved as ./ResNet152-eye_disease-95.26.h5
        class csv file was saved as ./class_dict.csv
```

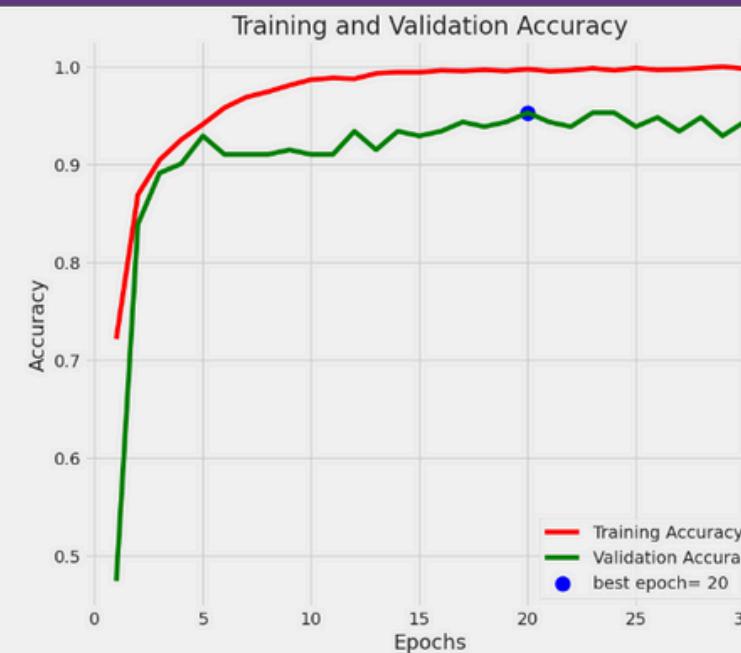
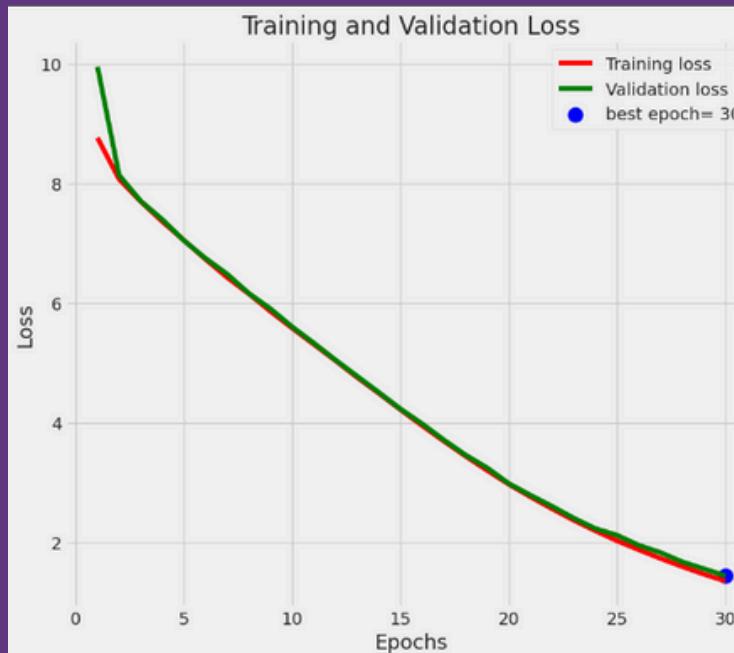
classification report:



## • ResNet

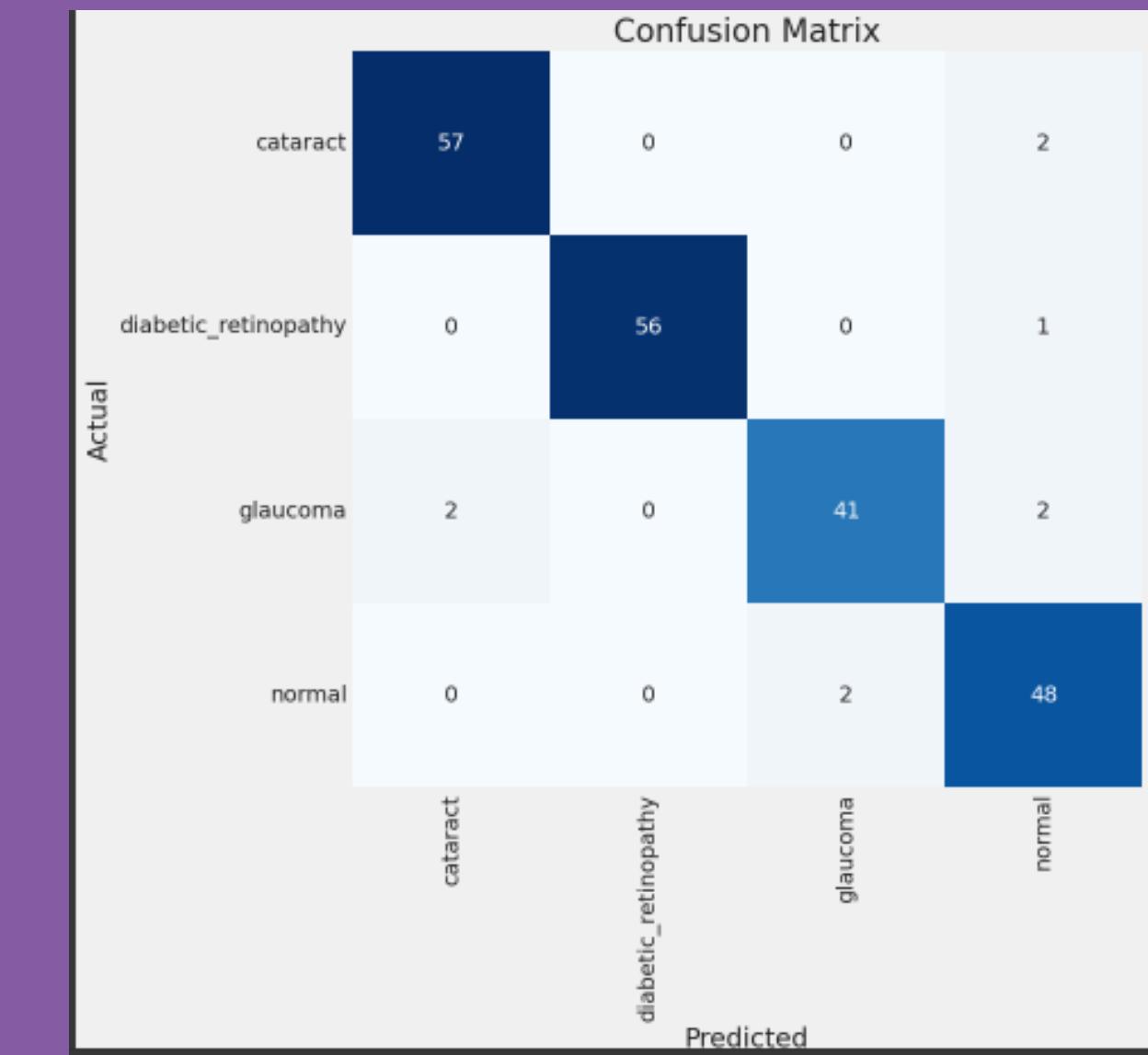
(LR=0.0001, BS=100, HF=F)

```
accuracy on the test set is 95.73 %
/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:364: UserWarning: saving_api.save_model(
model was saved as ./ResNet152-eye_disease-95.73.h5
class csv file was saved as ./class_dict.csv
```



Classification Report:

|                      | precision | recall | f1-score | support |
|----------------------|-----------|--------|----------|---------|
| cataract             | 0.97      | 0.97   | 0.97     | 59      |
| diabetic_retinopathy | 1.00      | 0.98   | 0.99     | 57      |
| glaucoma             | 0.95      | 0.91   | 0.93     | 45      |
| normal               | 0.91      | 0.96   | 0.93     | 50      |
| accuracy             |           |        | 0.96     | 211     |
| macro avg            | 0.96      | 0.95   | 0.96     | 211     |
| weighted avg         | 0.96      | 0.96   | 0.96     | 211     |



- DenseNet  
(LR=0.01, BS=60, HF=T)

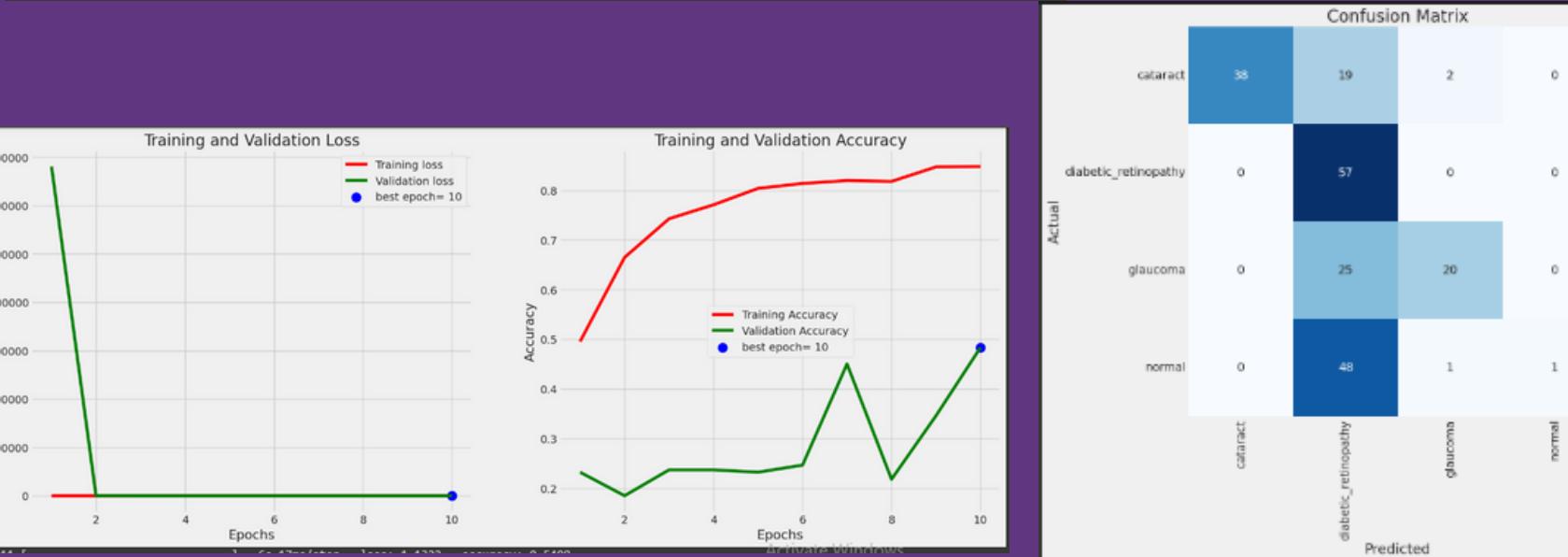
```
211/211 [=====] - 6s 17ms/step - loss: 1.13
accuracy on the test set is 54.98 %

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:366: UserWarning: saving_api.save_model(
model was saved as ./DenseNet121-eye_disease-54.97.h5

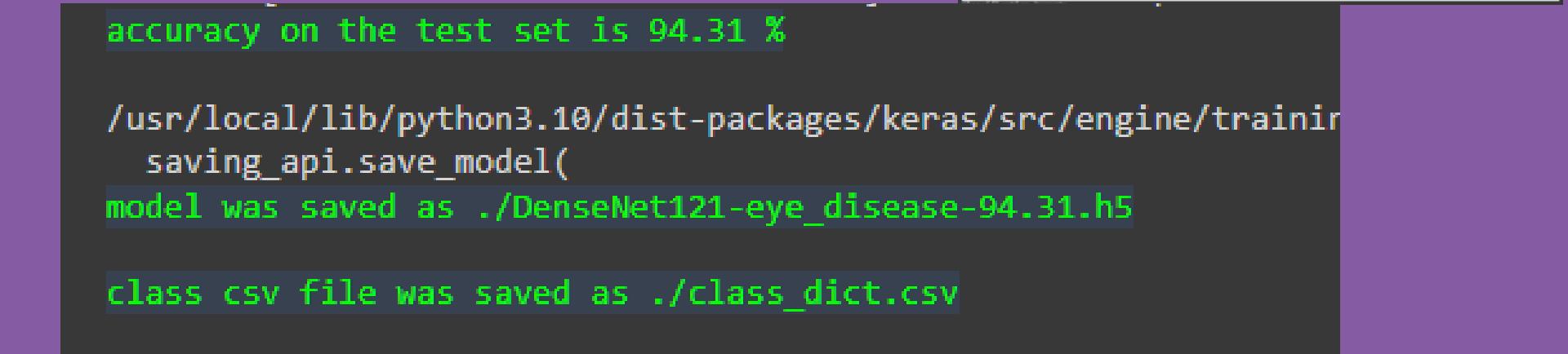
class csv file was saved as ./class_dict.csv
```

| Classification Report: |           |        |          |         |
|------------------------|-----------|--------|----------|---------|
|                        | precision | recall | f1-score | support |
| cataract               | 1.00      | 0.64   | 0.78     | 59      |
| diabetic_retinopathy   | 0.38      | 1.00   | 0.55     | 57      |
| glaucoma               | 0.87      | 0.44   | 0.59     | 45      |
| normal                 | 1.00      | 0.02   | 0.04     | 50      |
| accuracy               |           |        | 0.55     | 211     |
| macro avg              | 0.81      | 0.53   | 0.49     | 211     |
| weighted avg           | 0.81      | 0.55   | 0.50     | 211     |

- DenseNet  
(LR=0.0001, BS=60, HF=F)



| Classification Report: |           |        |          |         |
|------------------------|-----------|--------|----------|---------|
|                        | precision | recall | f1-score | support |
| cataract               | 0.98      | 0.93   | 0.96     | 59      |
| diabetic_retinopathy   | 1.00      | 0.98   | 0.99     | 57      |
| glaucoma               | 0.88      | 0.93   | 0.90     | 45      |
| normal                 | 0.90      | 0.92   | 0.91     | 50      |
| accuracy               |           |        | 0.94     | 211     |
| macro avg              | 0.94      | 0.94   | 0.94     | 211     |
| weighted avg           | 0.95      | 0.94   | 0.94     | 211     |



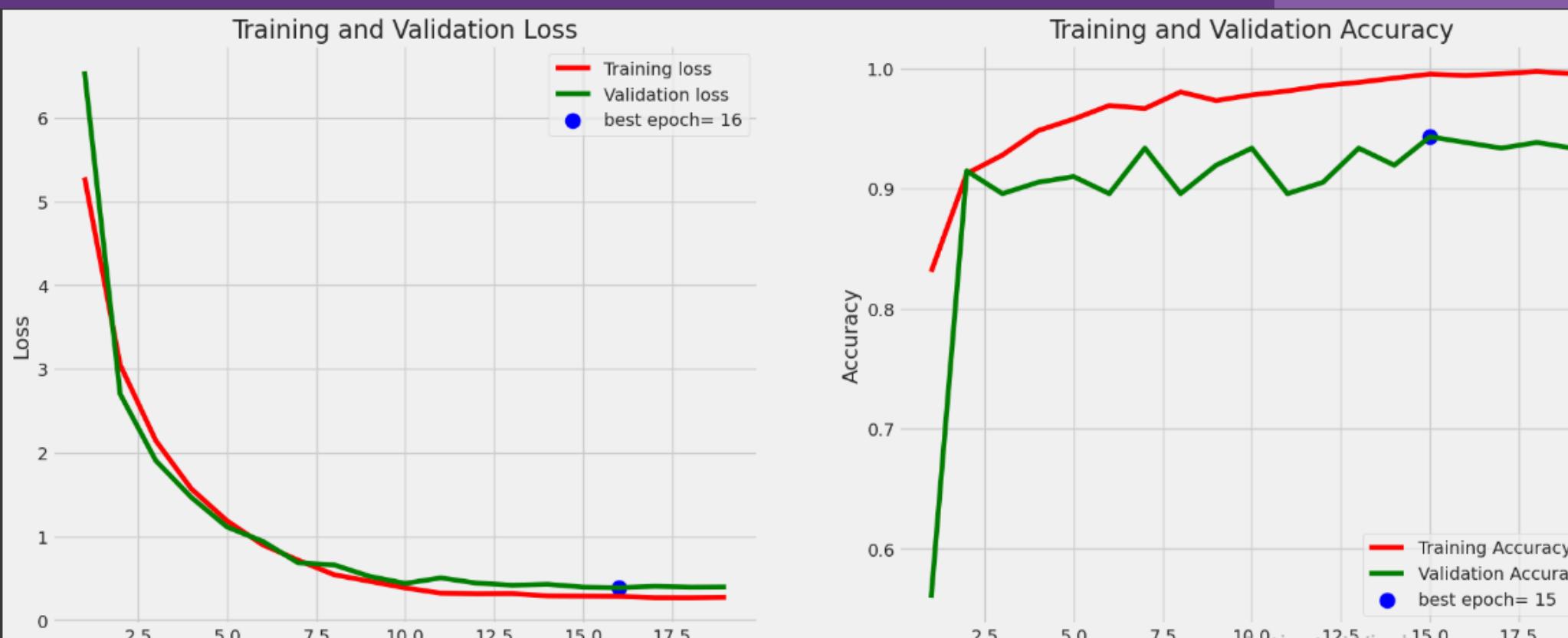
# • DenseNet

(LR=0.001, BS=100, HF=T)

```
211/211 [=====] - 4s 17ms/step - loss: 0.0000 - accuracy: 0.9668
accuracy on the test set is 96.68 %
```

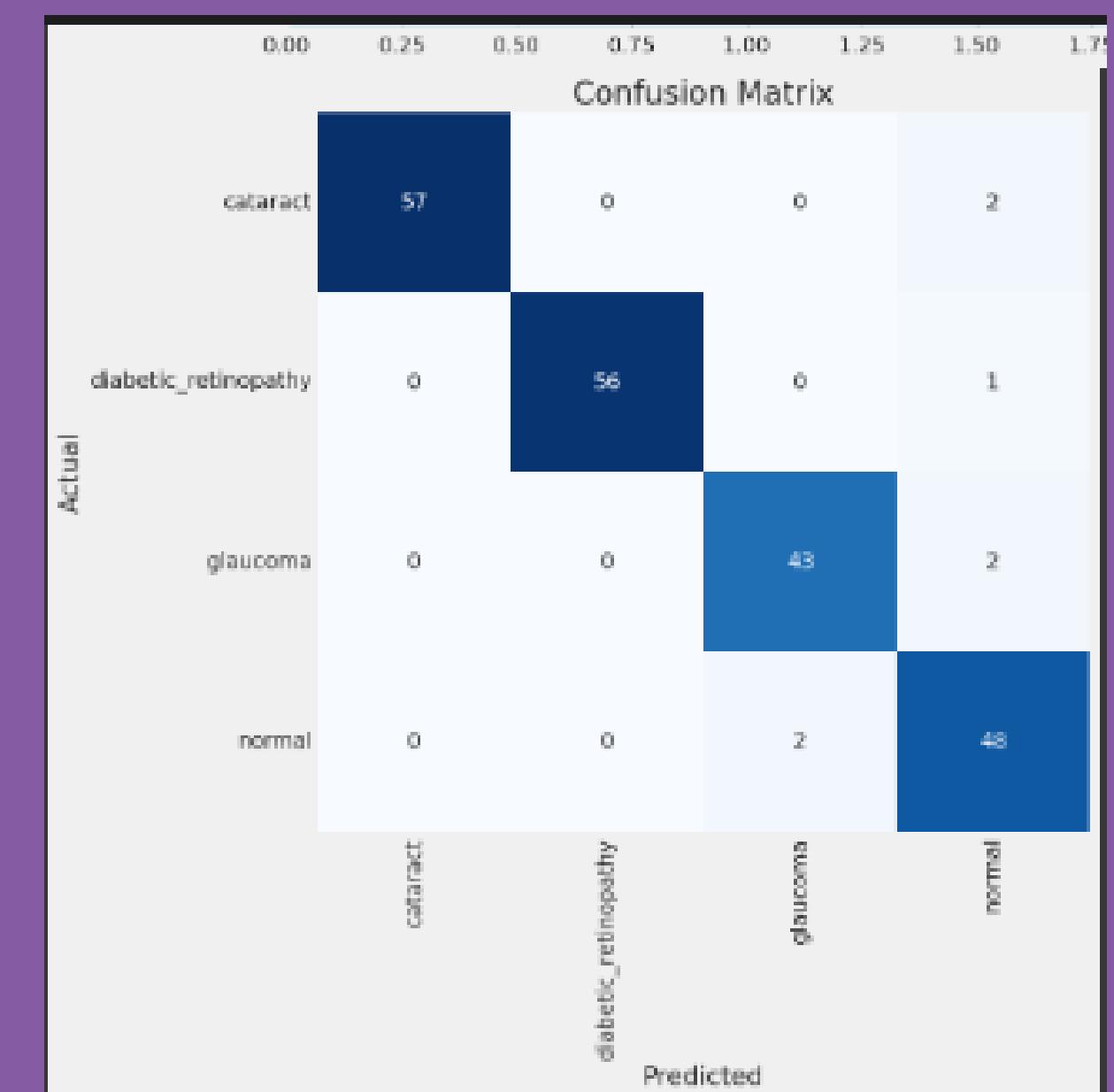
```
/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:147: UserWarning: saving_api.save_model(
  model was saved as ./DenseNet121-eye_disease-96.68.h5

class csv file was saved as ./class_dict.csv
```



Classification Report:

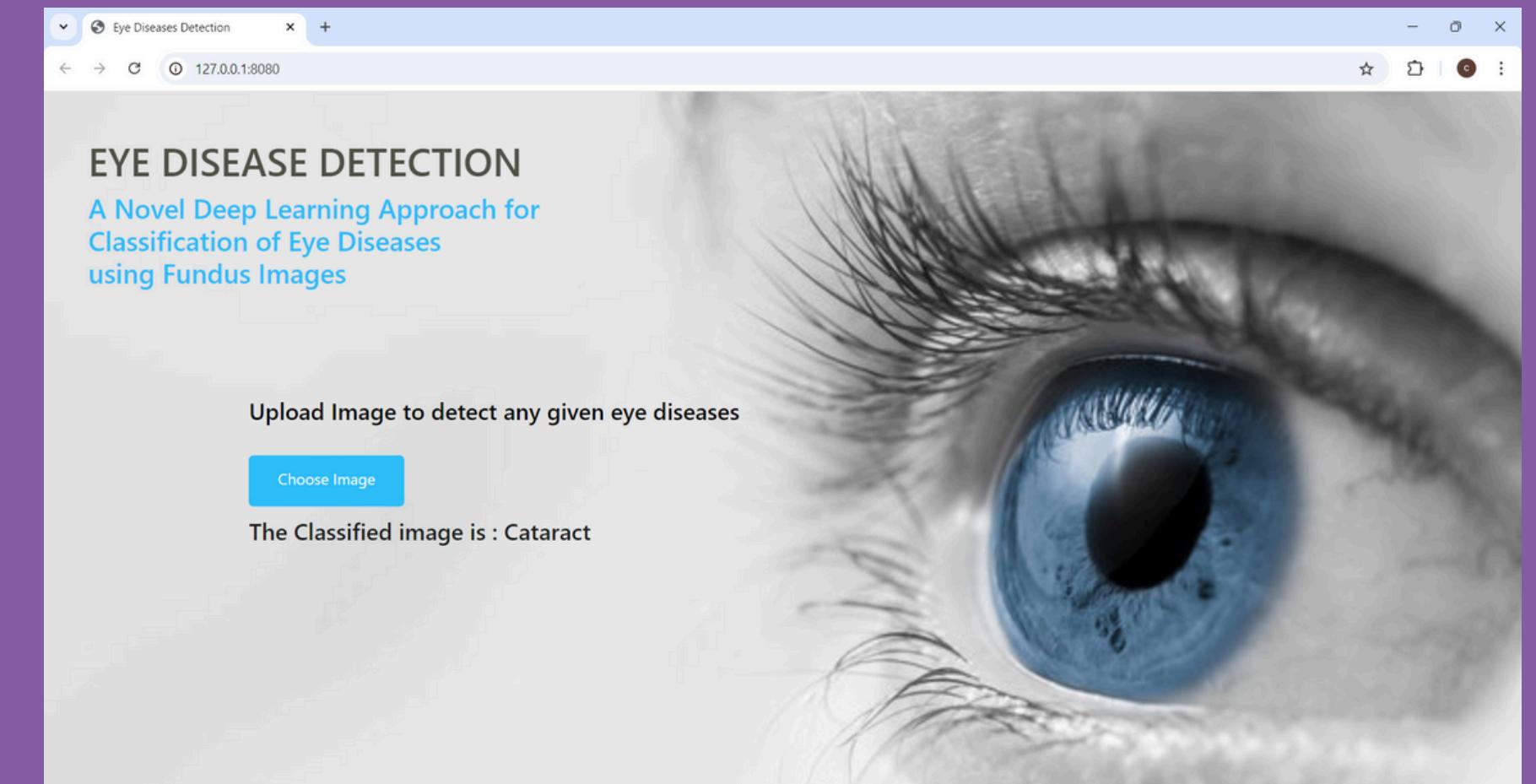
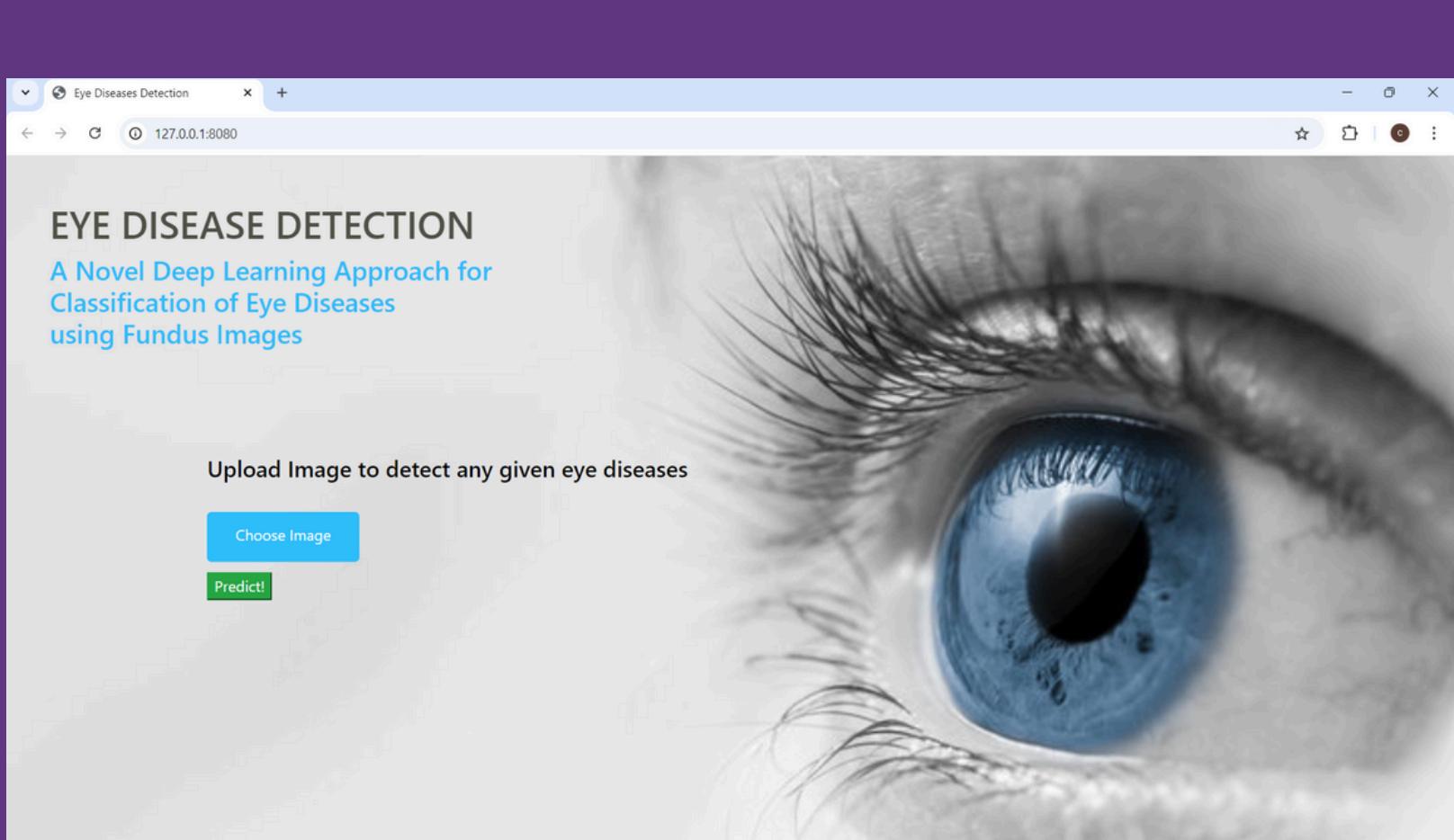
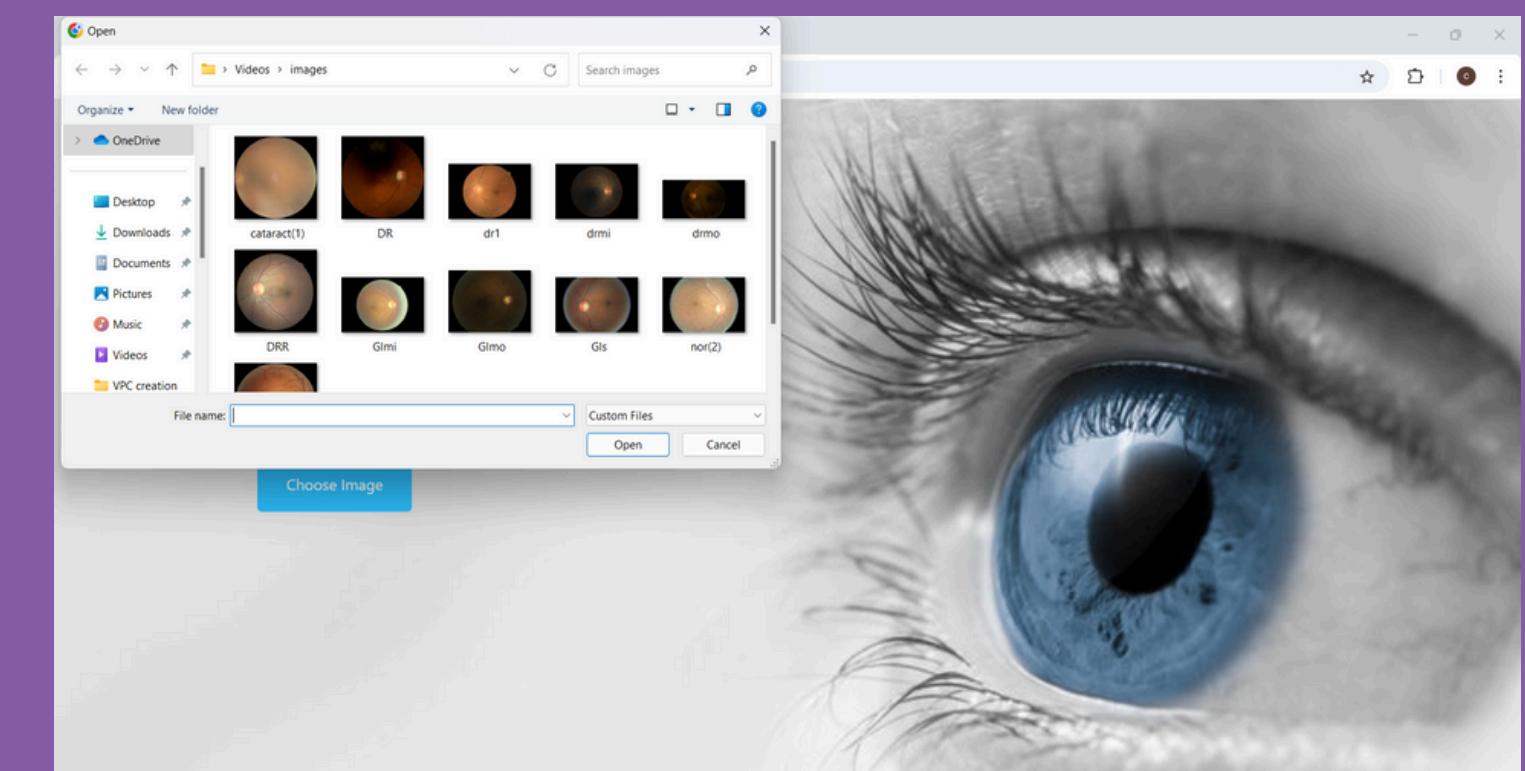
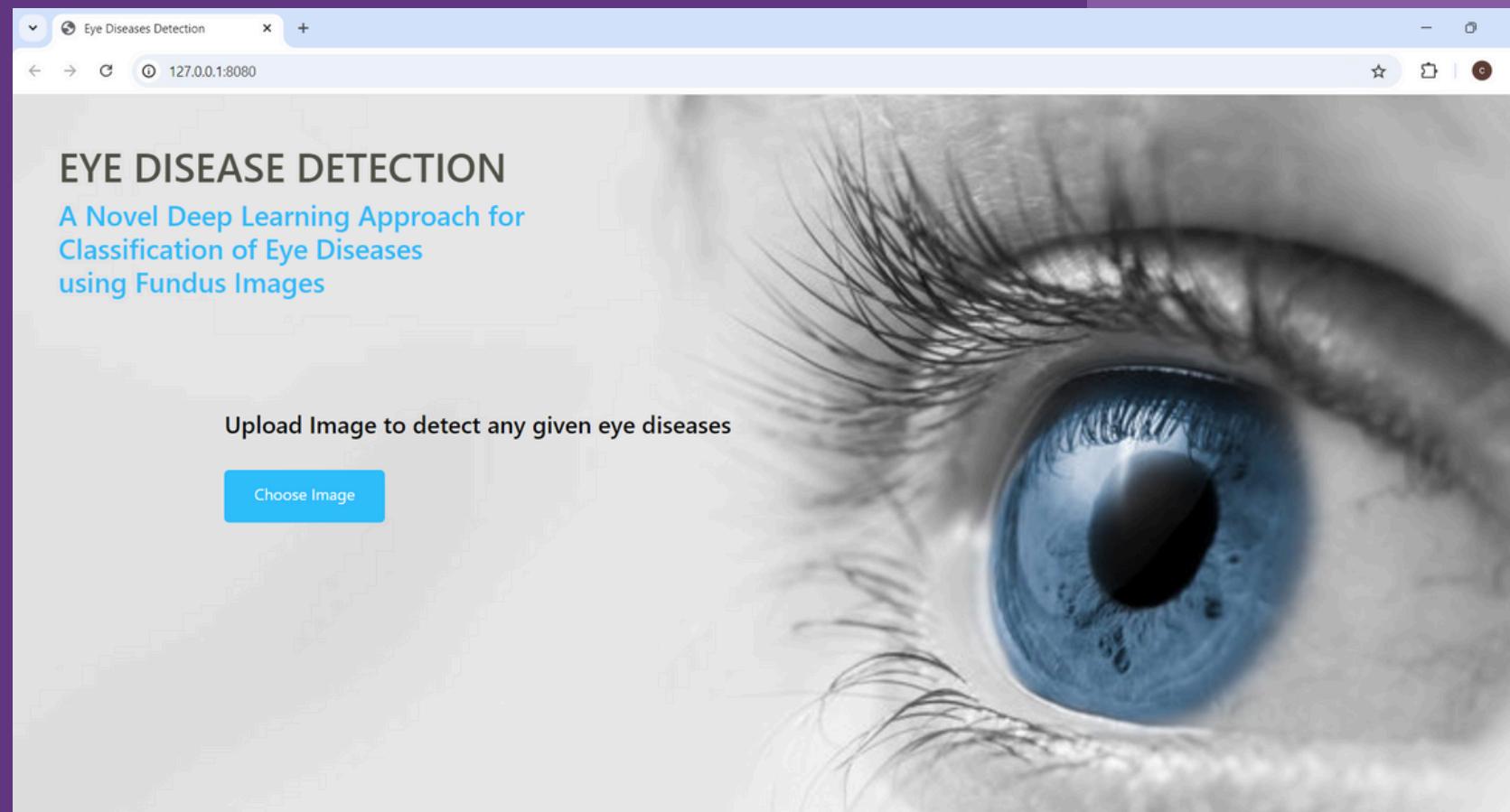
|                      | precision | recall | f1-score | support |
|----------------------|-----------|--------|----------|---------|
| cataract             | 1.00      | 0.97   | 0.98     | 59      |
| diabetic_retinopathy | 1.00      | 0.98   | 0.99     | 57      |
| glaucoma             | 0.96      | 0.96   | 0.96     | 45      |
| normal               | 0.91      | 0.96   | 0.93     | 58      |
| accuracy             |           |        | 0.97     | 211     |
| macro avg            | 0.97      | 0.97   | 0.97     | 211     |
| weighted avg         | 0.97      | 0.97   | 0.97     | 211     |



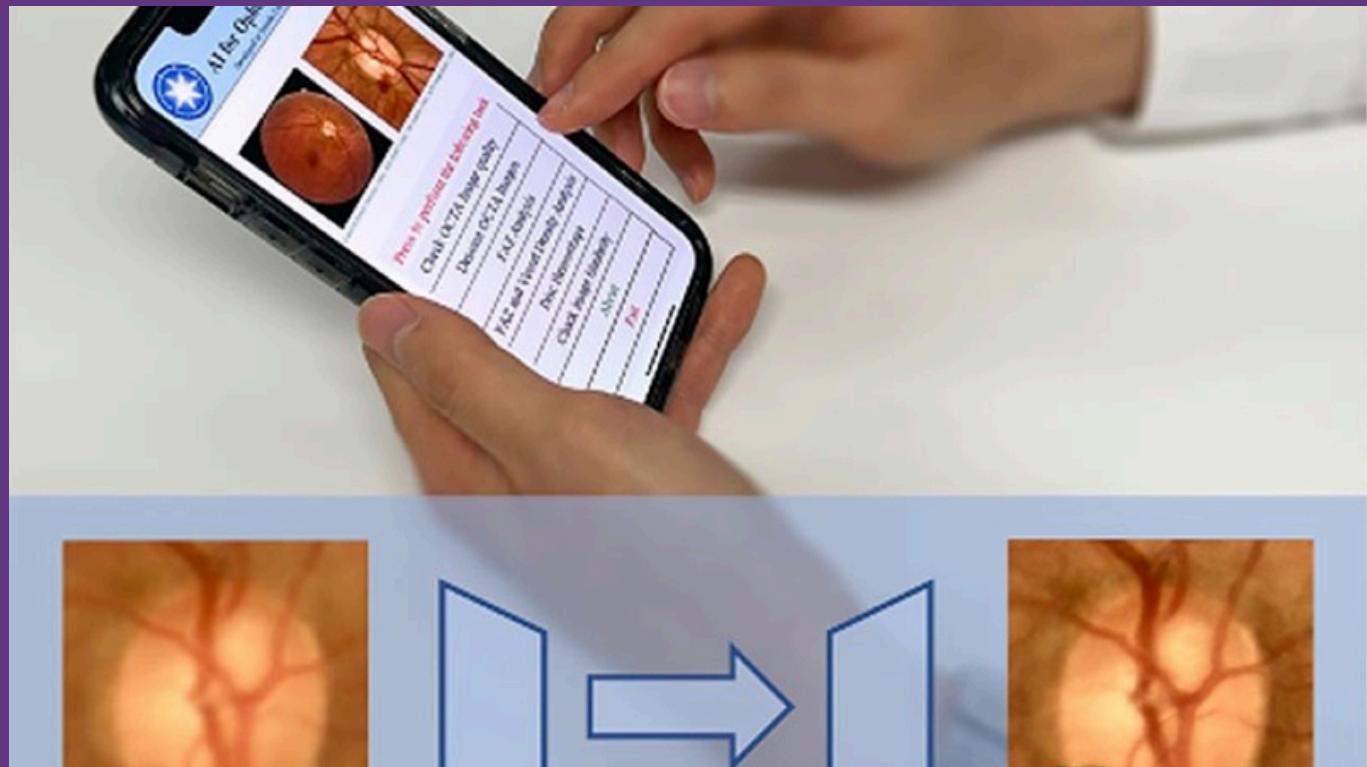
# EVALUATION

| Model    | Learning Rate | Epochs | Accuracy | Precision | Recall |
|----------|---------------|--------|----------|-----------|--------|
| VGG16    | 0.01          | 10     | 55.45    | 62.07     | 51.87  |
| ResNet   | 0.01          | 10     | 29.86    | 17.86     | 27.32  |
| DenseNet | 0.01          | 10     | 54.98    | 81.98     | 52.98  |
| VGG16    | 0.001         | 20     | 91.94    | 91.87     | 92.98  |
| ResNet   | 0.001         | 20     | 95.26    | 95.26     | 95.35  |
| DenseNet | 0.001         | 20     | 96.68    | 97.08     | 96.33  |
| VGG16    | 0.0001        | 30     | 94.79    | 95.08     | 94.09  |
| ResNet   | 0.0001        | 30     | 95.73    | 95.21     | 95.73  |
| DenseNet | 0.0001        | 30     | 94.31    | 94.53     | 93.98  |

# GUI



# CONCLUSION

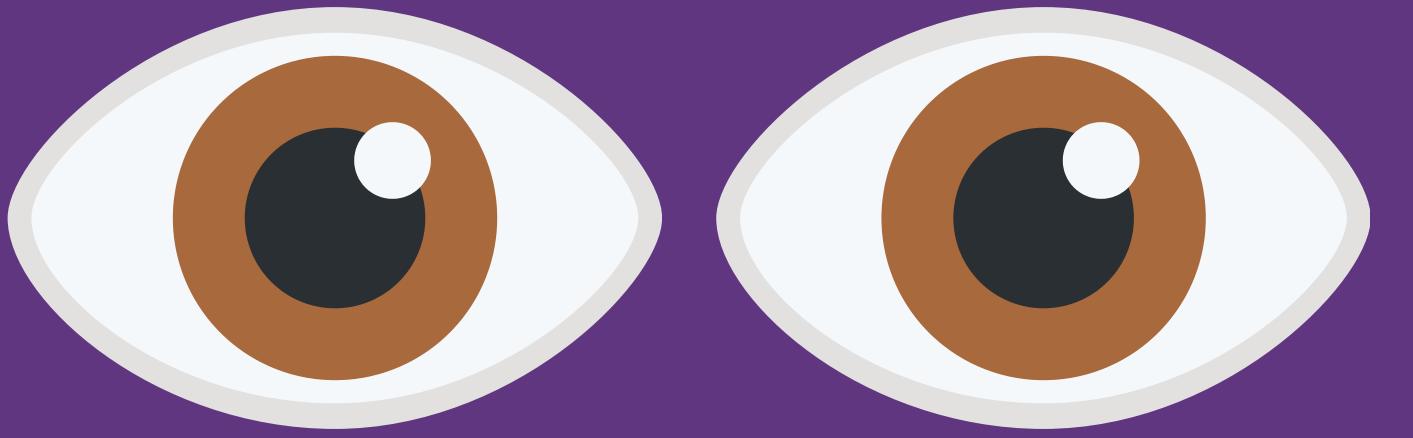


This project introduces a method for identifying EDC using deep learning. It achieves high accuracies for detecting EDC using DenseNet Architecture (96.68% accuracy) with a user friendly GUI.

# FUTURE SCOPE

- Develop a model that identifies diseases and also assesses their severity.
- Improve detection and analysis of other diseases through further research.
- Provide valuable insights for healthcare professionals and improve patient outcomes.





# THANK YOU!

