Glaucoma Classification

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Problem Statement

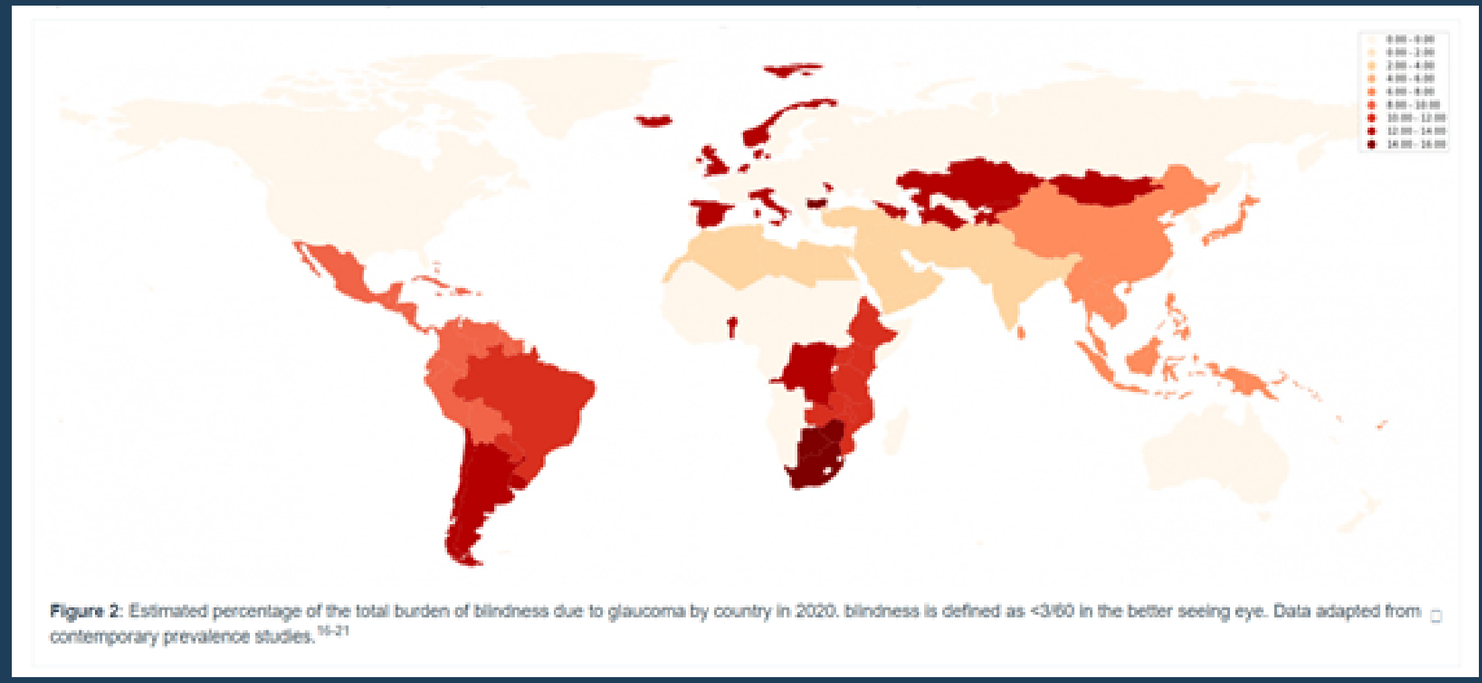


Figure 2: Estimated percentage of the total burden of blindness due to glaucoma by country in 2020. blindness is defined as <3/60 in the better seeing eye. Data adapted from contemporary prevalence studies.[16-21]

ต้อหินเป็นโรคที่พบได้บ่อยในผู้สูงอายุและถือเป็นโรคที่ทำให้เกิดภาวะสูญเสียการมองเห็นชนิดถาวรที่พบได้บ่อยที่สุดทั่วโลกเป็นอันดับ 2 รองจากโรคต้อกระจก โดยในปี ค.ศ. 2020 มีผู้เป็นโรคต้อหินถึง 76 ล้านคน

OBJECTIVES

เพื่อทำการสร้าง AI ที่ใช้ในการตรวจคัดกรองโรค glaucoma, ตาปกติ และ โรคอื่นๆด้วย Machine learning และ deep learning

Machine Learning

Data Collection

ประเภท	จำนวนข้อมูล
Glaucoma	1,363
Normal	2,009
Others	896
Total	4,268

EXPERIMENTAL SETUP

1) **คัดรูป และ ทำการสกัด** Features
2) **ทำการเก็บ** Dataset **จาก** Features **ที่สกัดได้**
3) **นำ** Dataset **ไปทำการ** Train Model **ด้วย** Quadratic Discriminant Analysis Algorithm **และ** RBF SVM Algorithm
4) **ทดสอบ** Model **ที่ได้ ด้วย** Test Data

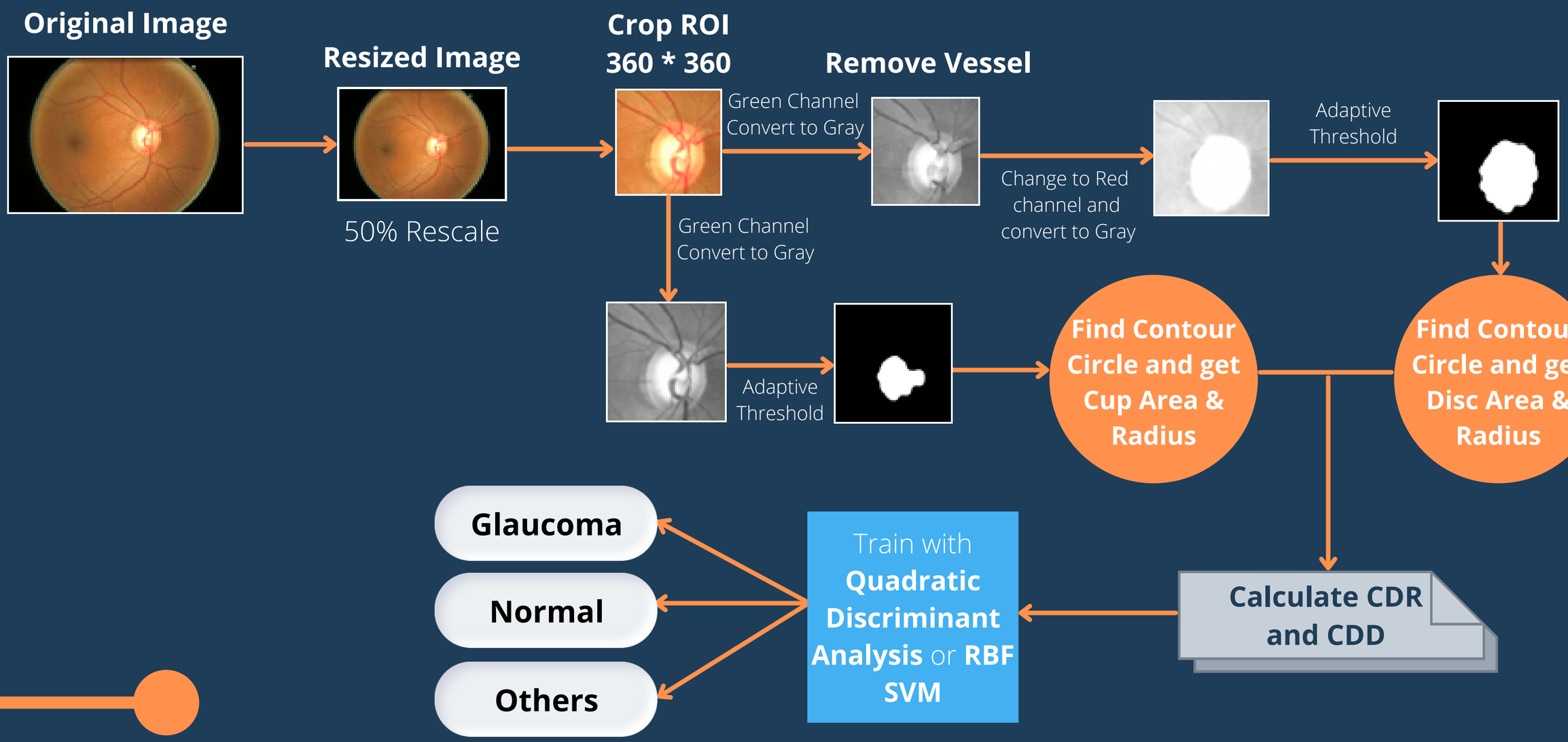
EXPERIMENTAL SETUP

5-Fold Cross Validation

	Data Train					Test Data
Model 1	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Test Data
Model 2	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Test Data
Model 3	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Test Data
Model 4	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Test Data
Model 5	Fold 1	Fold 2	Fold 3	Fold 4	Fold 5	Test Data

Data Validation Data Train Data Test

Machine Learning Methods



Ashish Issac, M. Parthasarthi, Malay Kishore Dutta, Flowchart for the segmentation of optic disc and cup, [Online]. Available: https://ieeexplore.ieee.org/document/7095384. (access Mar. 24th 2021)

ARCHITECTURE

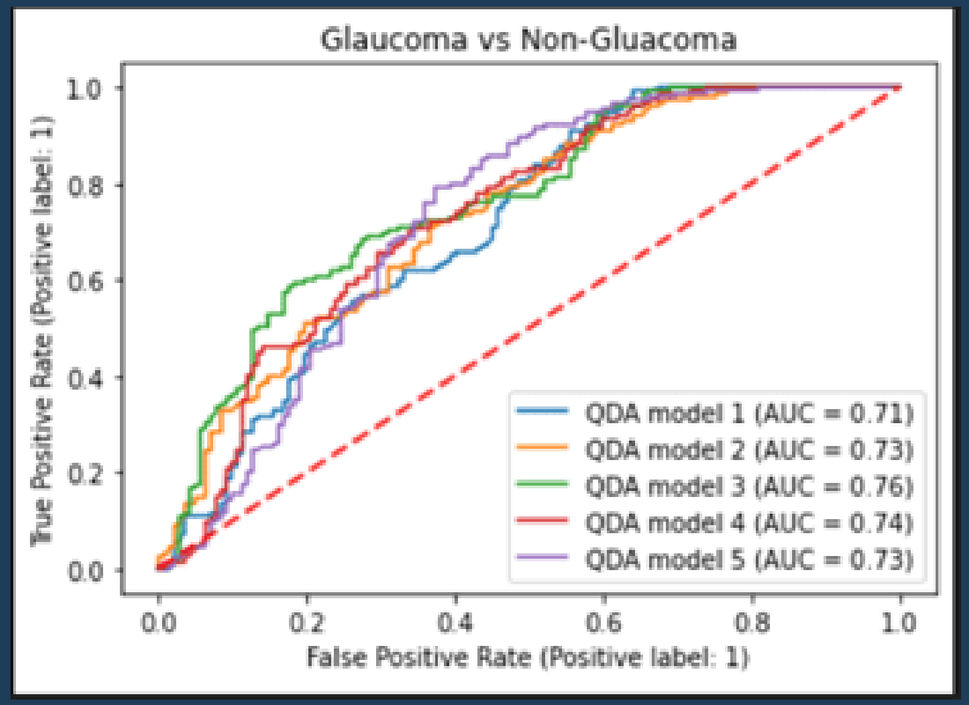
Quadratic Discriminant Analysis

RBF SVM

Experiment Result

Quadratic Discriminant Analysis (Model 3)

Glaucoma vs Non- Glaucoma



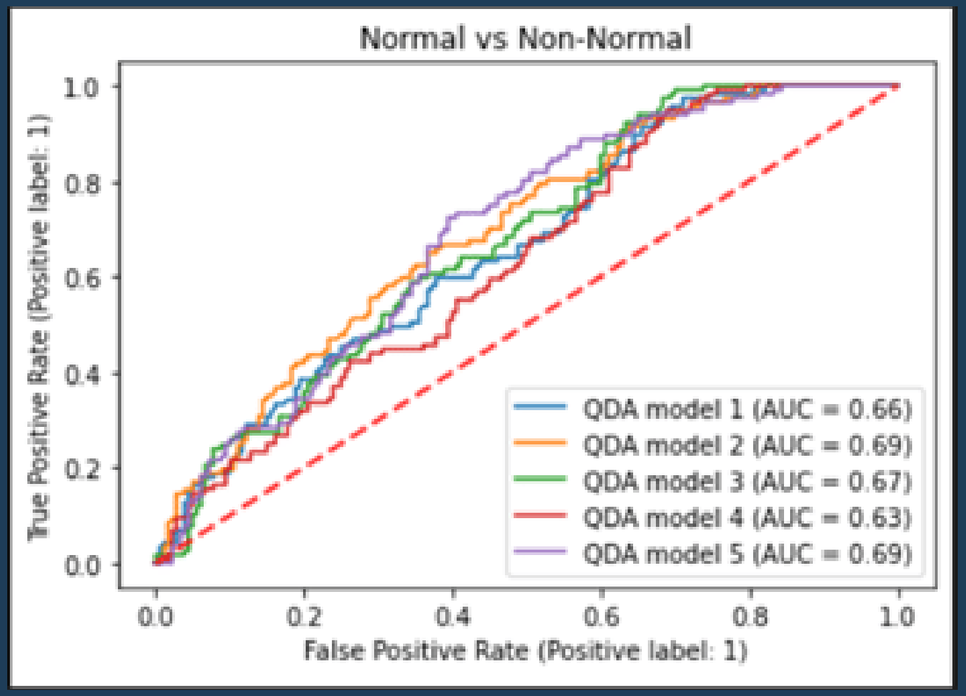
ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	149(80.11%)	37(19.89%)	186
ลบ(Negative)	78(42.16%)	107(57.84%)	185
รวม(Total)	227	144	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.690027	0.578378	0.801075	0.656388	0.690027

Experiment Result

Quadratic Discriminant Analysis (Model 2)

Normal vs Non- Normal



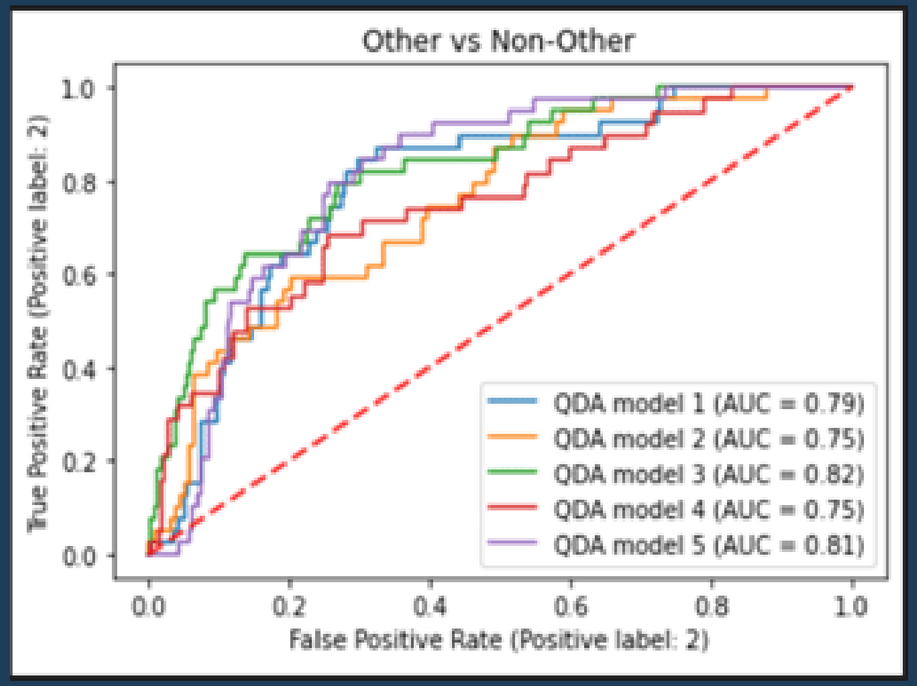
ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	63(45.98%)	74(54.02%)	137
ลบ(Negative)	67(28.63%)	167(71.37%)	234
รวม(Total)	130	241	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.619946	0.713675	0.459854	0.484615	0.619946

Experiment Result

Quadratic Discriminant Analysis (Model 2)

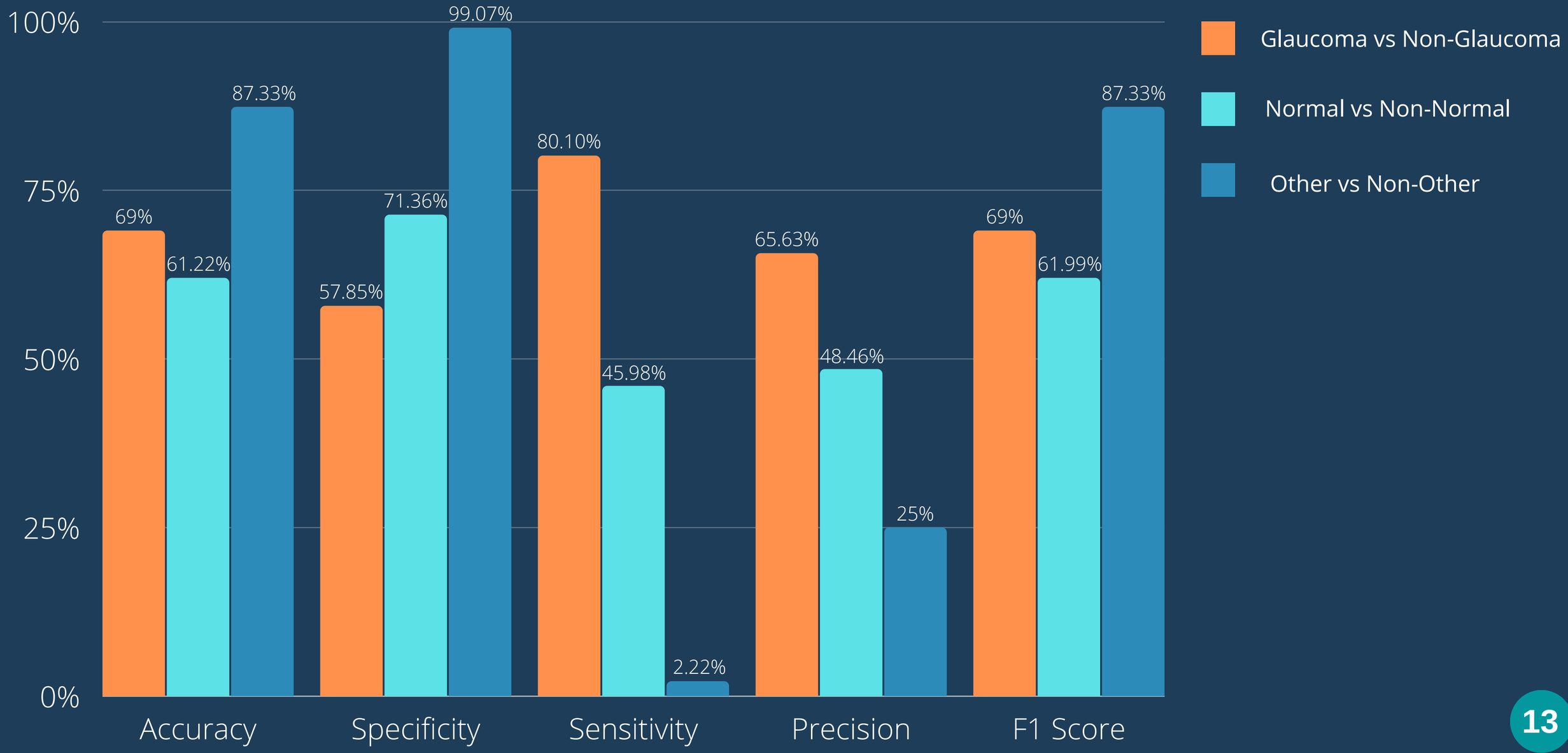
Other vs Non- Other



ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	1(2.22%)	44(97.78%)	45
ลบ(Negative)	3(0.92%)	323(99.08%)	326
รวม(Total)	4	367	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.873315	0.990798	0.022222	0.25	0.873315

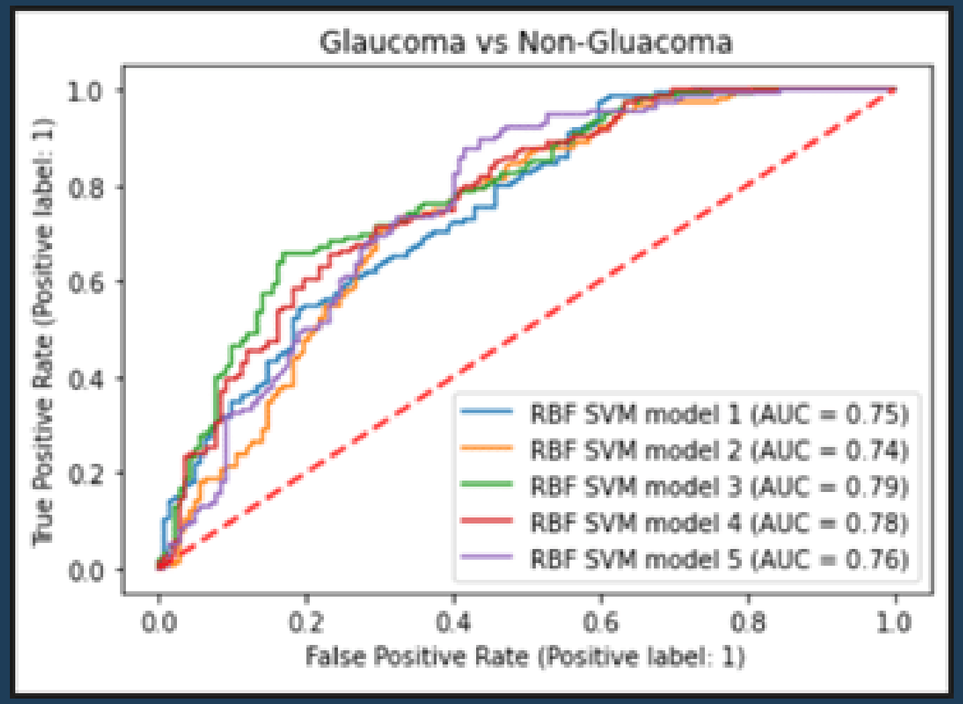
Quadratic Discriminant Analysis



Experiment Result

RBF SVM (Model 3)

Glaucoma vs Non- Glaucoma



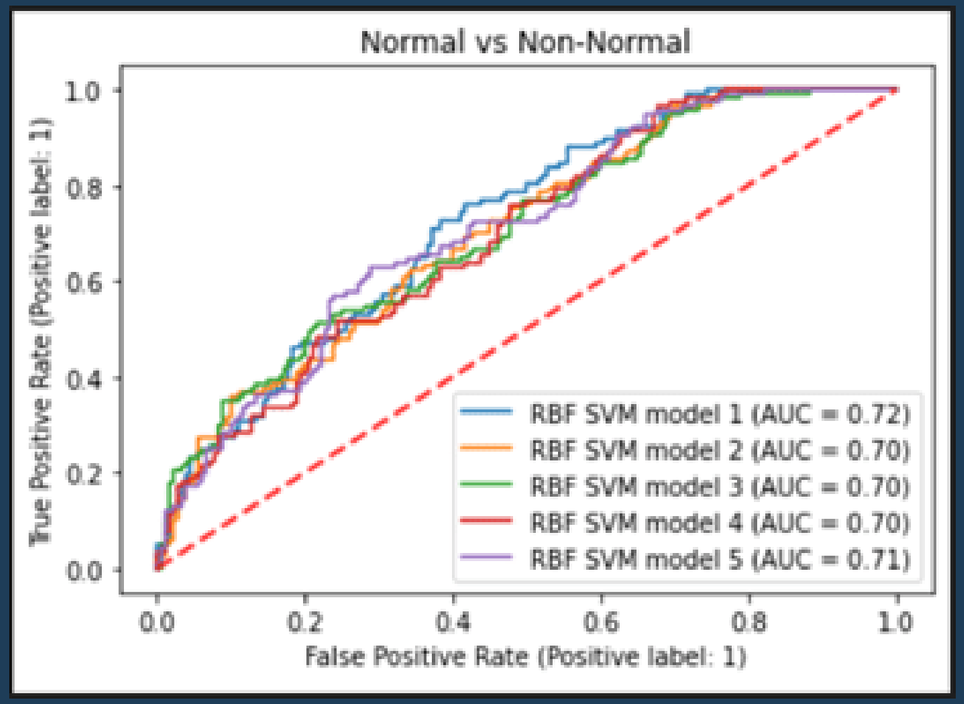
ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	161(86.56%)	25(13.44%)	186
ลบ(Negative)	80(43.24%)	105(56.76%)	185
รวม(Total)	241	130	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.716981	0.567567	0.865591	0.668050	0.716981

Experiment Result

RBF SVM (Model 1)

Normal vs Non- Normal



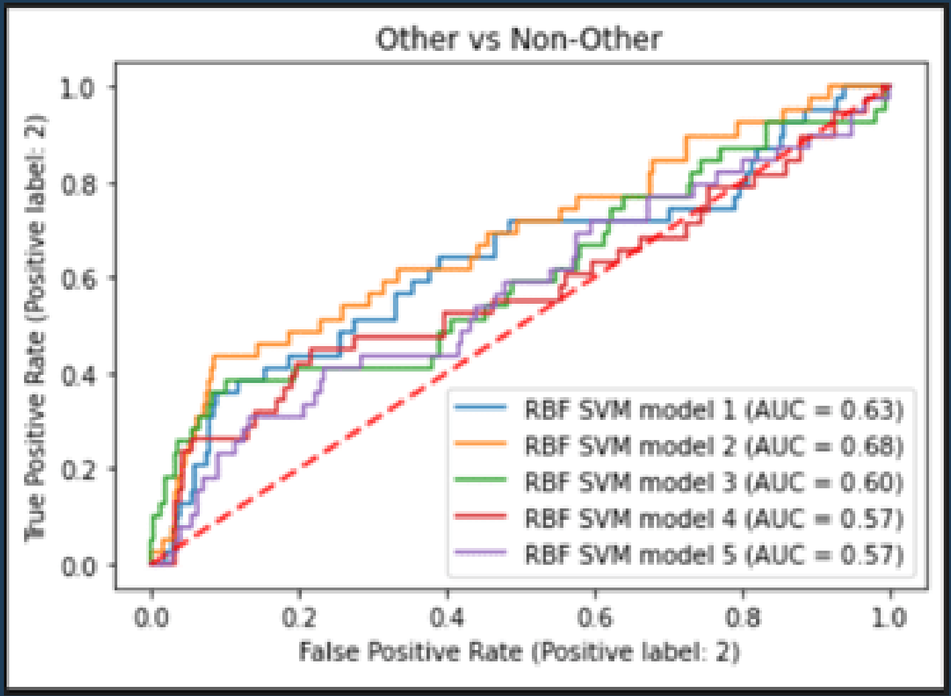
ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	50(36.50%)	87(63.50%)	137
ลบ(Negative)	41(17.52%)	193(82.48%)	234
รวม(Total)	91	280	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.649596	0.829060	0.343066	0.540230	0.617251

Experiment Result

RBF SVM (Model 5)

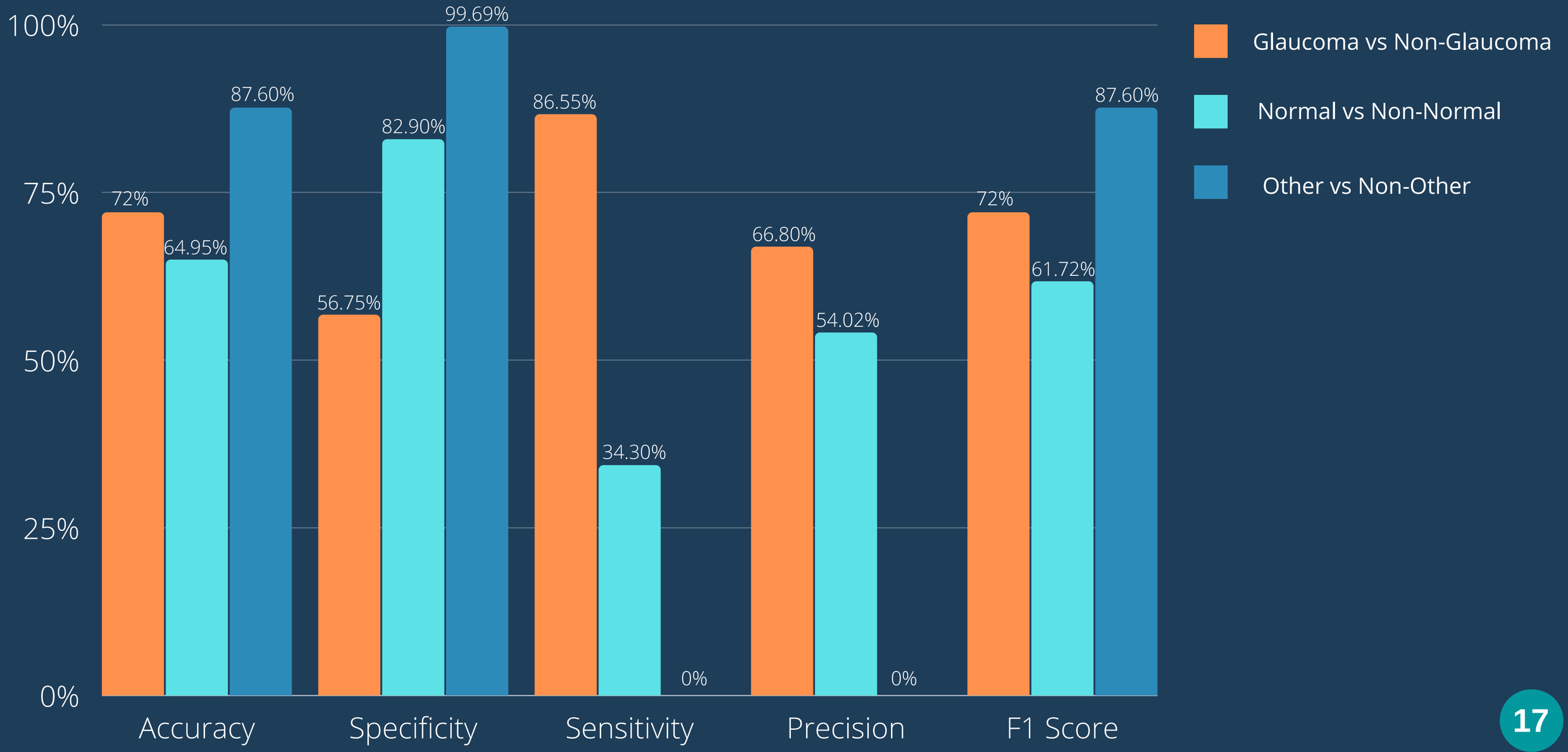
Other vs Non- Other



ประเภทจริง (True class)	ประเภทจริง		รวม (Total)
	บวก(Positive)	ลบ(Negative)	
บวก(Positive)	0(0%)	45(100%)	45
ลบ(Negative)	1(0.30%)	325(99.70%)	326
รวม(Total)	1	370	371

Accuracy	Specificity	Sensitivity	Precision	F1 Score
0.876010	0.996932	0	0	0.876010

RBF SVM


100%
75%
50%
25%
0%
72%
64.95%
87.60%
56.75%
82.90%
99.69%
86.55%
34.30%
0%
66.80%
54.02%
0%
72%
61.72%
87.60%
Accuracy
Specificity
Sensitivity
Precision
F1 Score
Glaucoma vs Non-Glaucoma
Normal vs Non-Normal
Other vs Non-Other

CODE
EXPLANATION

```
def delVessel(image):
    blue,green,red = cv.split(image)
    kernel = cv.getStructuringElement(cv.MORPH_ELLIPSE,(26,26))
    ves = cv.morphologyEx(green, cv.MORPH_BLACKHAT, kernel)

    vessel2 = cv.bitwise_or(ves,green)
    vessel = cv.bitwise_or(ves,red)
    vessel = cv.medianBlur(red,7)
    plt.imshow(vessel2)

    return vessel
```

CODE
EXPLANATION

```
def GetDisc(image):
    M = 60    #filter size

    filter = signal.gaussian(M, std=7) #Gaussian Window
    filter=filter/sum(filter)
    STDf = filter.std()  #It'standard deviation

    image_pre = image-image.mean()-image.std()

    thr = (0.5 * M) - (2*STDf) - image_pre.std()

    r,c = image.shape
    Dd = np.zeros(shape=(r,c))

    for i in range(1,r):
        for j in range(1,c):
            if image_pre[i,j]>thr:
                Dd[i,j]=255
            else:
                Dd[i,j]=0

    Dd = cv.morphologyEx(Dd, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(2,2)), iterations = 1)
    Dd = cv.morphologyEx(Dd, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(7,7)), iterations = 1)
    Dd = cv.morphologyEx(Dd, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(1,21)), iterations = 1)
    Dd = cv.morphologyEx(Dd, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(21,1)), iterations = 1)
    Dd = cv.morphologyEx(Dd, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(23,23)), iterations = 1)
    Dd = cv.morphologyEx(Dd, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(43,43)), iterations = 1)

    Dd = np.uint8(Dd)

    return Dd
```

CODE
EXPLANATION

```
def GetCup(image):
    blue, green, red = cv.split(image)
    green = cv.medianBlur(green,7)

    M = 60    #filter size

    filter = signal.gaussian(M, std=7) #Gaussian Window
    filter = filter/sum(filter)
    STDf = filter.std()  #It'standard deviation

    green_pre = green-green.mean()-green.std()

    thr = (0.5 * M) + (2 * STDf) + (green_pre.std()) + (green_pre.mean())
    r,c = green.shape
    Dc = np.zeros(green.shape[:2])

    for i in range(1,r):
        for j in range(1,c):
            if green_pre[i,j]>thr:
                Dc[i,j]=255
            else:
                Dc[i,j]=0

    Dc = cv.morphologyEx(Dc, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(2,2)), iterations = 1)
    Dc = cv.morphologyEx(Dc, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(7,7)), iterations = 1)
    Dc = cv.morphologyEx(Dc, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(1,21)), iterations = 1)
    Dc = cv.morphologyEx(Dc, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(21,1)), iterations = 1)
    Dc = cv.morphologyEx(Dc, cv.MORPH_CLOSE, cv.getStructuringElement(cv.MORPH_ELLIPSE,(33,33)), iterations = 1)
    Dc = cv.morphologyEx(Dc, cv.MORPH_OPEN, cv.getStructuringElement(cv.MORPH_ELLIPSE,(33,33)), iterations = 1)

    Dc = np.uint8(Dc)
    return Dc
```

CODE
EXPLANATION

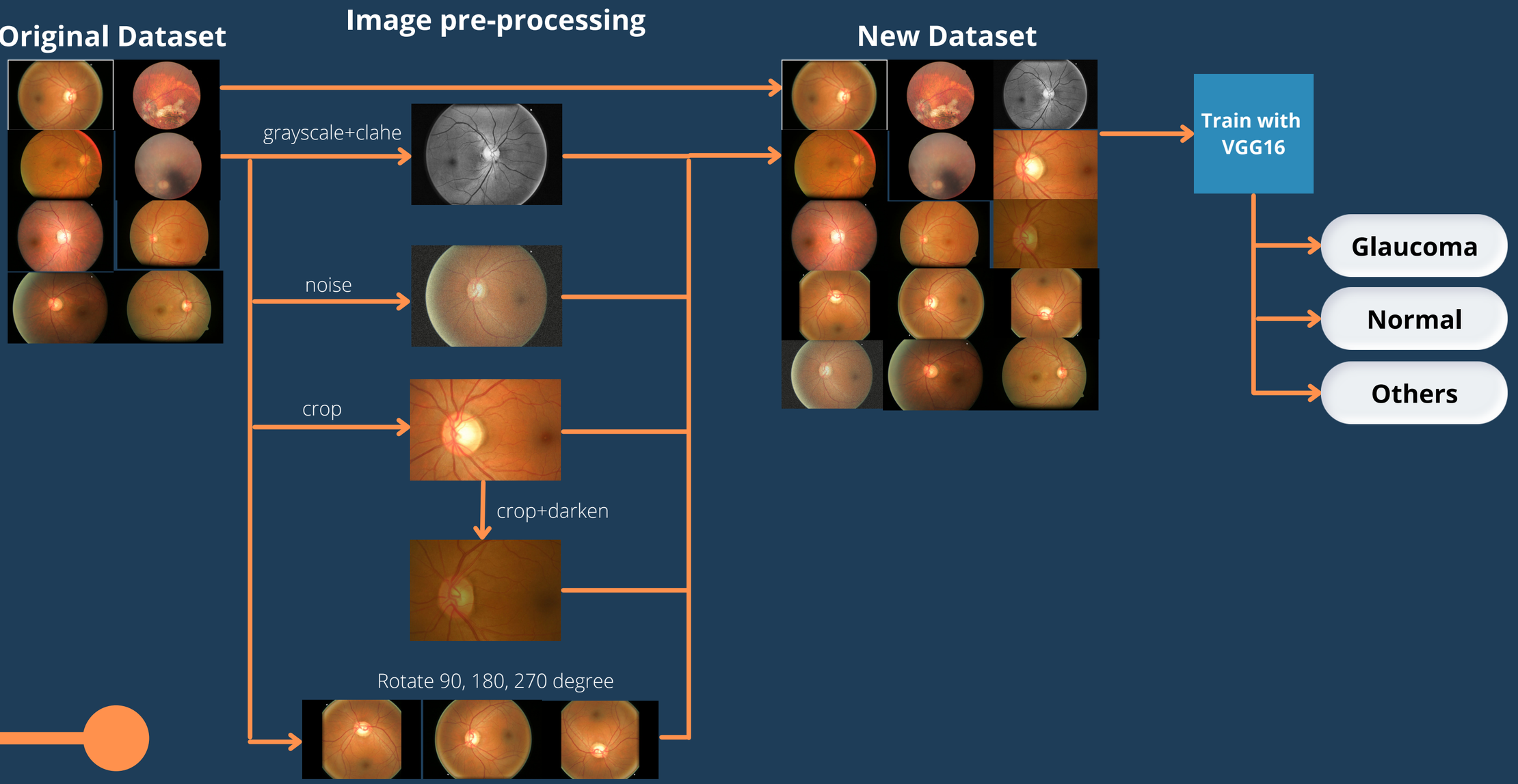
```
trainData01,testData01,trainType01,testType01 = train_test_split(data01, eye01, test_size=0.2 , random_state=1)
skf = StratifiedKFold(n_splits=5)

plt.scatter(data01[:,0],data01[:,1], c=eye01)
plt.xlabel('CDR')
plt.ylabel('y')


# QDA model for Glucoma vs Non-Glaucoma
clf = QuadraticDiscriminantAnalysis()
tprs = []
aucs = []
mean_fpr = np.linspace(0, 1, 100)
fig, ax = plt.subplots()
i = 1

for train_index, test_index in skf.split(trainData01, trainType01):
    x_train, x_test = trainData01[train_index], trainData01[test_index]
    y_train, y_test = trainType01[train_index], trainType01[test_index]
    clf.fit(x_train,y_train)
    filename = 'savemodel/GOQDA_model'+str(i)+'.sav'
    joblib.dump(clf,filename)

    viz = plot_roc_curve(clf,x_test, y_test,name='QDA model {}'.format(i), ax=ax)
    interp_tpr = np.interp(mean_fpr, viz.fpr, viz.tpr)
    interp_tpr[0] = 0.0
    tprs.append(interp_tpr)
    aucs.append(viz.roc_auc)

    i+=1
```

Deep Learning

Data Collection

ประเภท	จำนวนข้อมูล
Glaucoma	12,516
Normal	9,050
Others	5,049
Total	26,615

Deep Learning Methods



ARCHITECTURE

Model 1

```
Model: "sequential"
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
conv2d (Conv2D)              (None, 224, 224, 64)      1792
_________________________________________________________________
conv2d_1 (Conv2D)            (None, 222, 222, 64)      36928
_________________________________________________________________
max_pooling2d (MaxPooling2D) (None, 111, 111, 64)      0
_________________________________________________________________
dropout (Dropout)            (None, 111, 111, 64)      0
_________________________________________________________________
conv2d_2 (Conv2D)            (None, 111, 111, 32)      18464
_________________________________________________________________
conv2d_3 (Conv2D)            (None, 109, 109, 32)      9248
_________________________________________________________________
max_pooling2d_1 (MaxPooling2 (None, 54, 54, 32)        0
_________________________________________________________________
dropout_1 (Dropout)          (None, 54, 54, 32)        0
_________________________________________________________________
conv2d_4 (Conv2D)            (None, 54, 54, 16)        4624
_________________________________________________________________
conv2d_5 (Conv2D)            (None, 52, 52, 16)        2320
_________________________________________________________________
max_pooling2d_2 (MaxPooling2 (None, 26, 26, 16)        0
_________________________________________________________________
dropout_2 (Dropout)          (None, 26, 26, 16)        0
_________________________________________________________________
flatten (Flatten)            (None, 10816)             0
_________________________________________________________________
dense (Dense)                (None, 64)                692288
_________________________________________________________________
dropout_3 (Dropout)          (None, 64)                0
_________________________________________________________________
dense_1 (Dense)              (None, 32)                2080
_________________________________________________________________
dropout_4 (Dropout)          (None, 32)                0
_________________________________________________________________
dense_2 (Dense)              (None, 16)                528
_________________________________________________________________
dropout_5 (Dropout)          (None, 16)                0
_________________________________________________________________
dense_3 (Dense)              (None, 3)                 51
=================================================================
Total params: 768,323
Trainable params: 768,323
Non-trainable params: 0
```

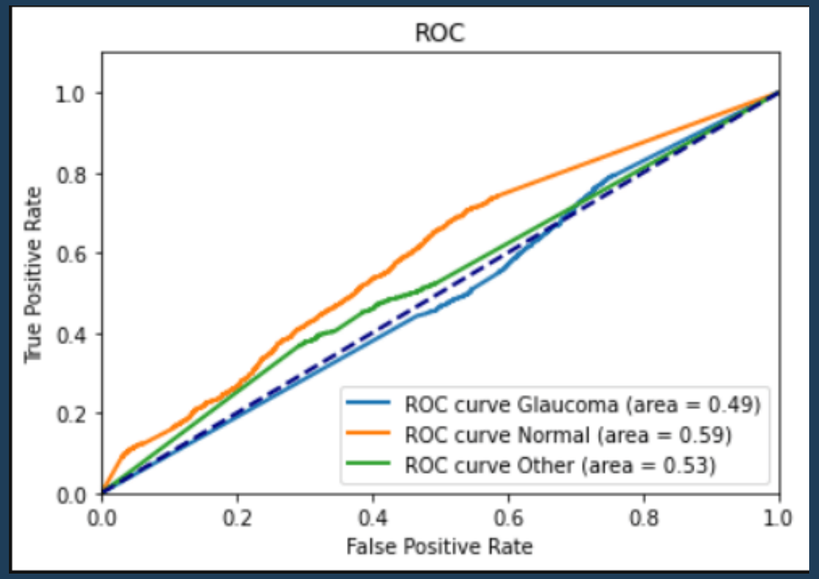
Model 2

VGG16

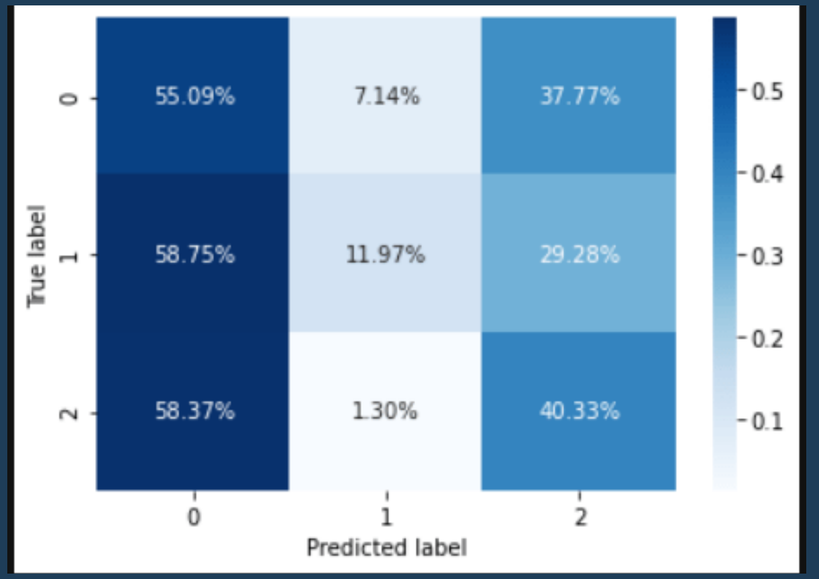
```
Model: "sequential"
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
vgg16 (Functional)           (None, 7, 7, 512)         14714688
_________________________________________________________________
global_average_pooling2d (Gl (None, 512)               0
_________________________________________________________________
dense (Dense)                (None, 3)                 1539
=================================================================
Total params: 14,716,227
Trainable params: 1,539
Non-trainable params: 14,714,688
_________________________________________________________________
```

Model 1

ROC



confusion table

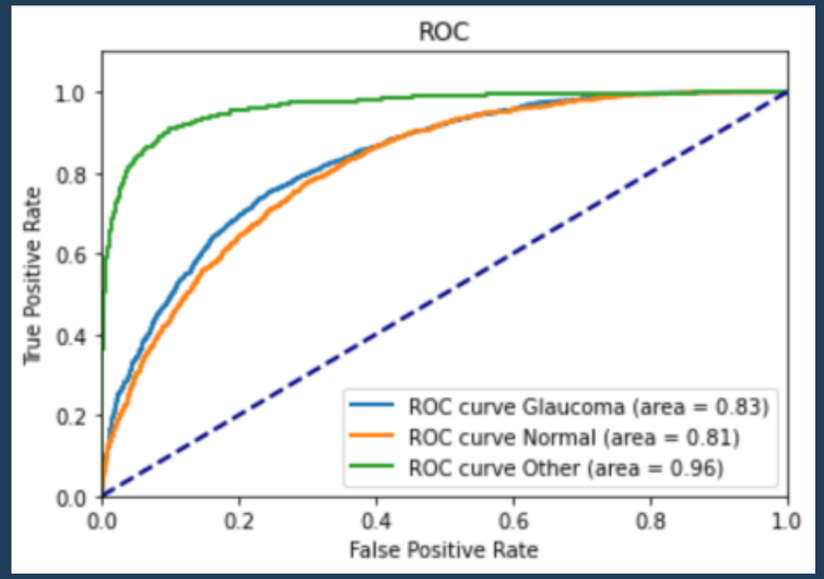


Type	Precision	Recall	F1 Score
Glaucoma	0.45483	0.55090	0.49828
Normal	0.53061	0.11974	0.19538
Other	0.21632	0.40330	0.28160

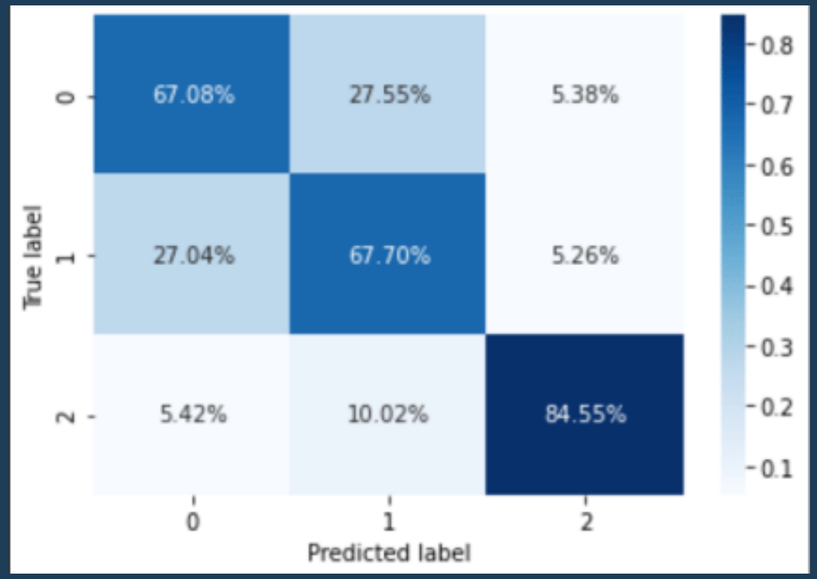
Accuracy = 0.37629

Model 2

ROC



confusion table



Type	Precision	Recall	F1 Score
Glaucoma	0.75522	0.67079	0.71051
Normal	0.60780	0.67697	0.64052
Other	0.78791	0.84552	0.81570

Accuracy = 0.70604

CODE
EXPLANATION

```
import tensorflow as tf
import numpy as np
from tensorflow.keras.preprocessing import image
from sklearn.metrics import accuracy_score, f1_score, precision_score, confusion_matrix, roc_curve, auc, plot_roc_curve
```

```
path_train = "dataset/train"

img_train = tf.keras.preprocessing.image_dataset_from_directory(
    path_train,
    validation_split=0.2,
    subset = "training",
    seed = 125,
    image_size = (224,224),
    batch_size = 32
)

img_validation = tf.keras.preprocessing.image_dataset_from_directory(
    path_train,
    validation_split=0.2,
    subset = "validation",
    seed = 125,
    image_size = (224,224),
    batch_size = 32
)
```

CODE
EXPLANATION

```
VGG16_MODEL=tf.keras.applications.VGG16(input_shape=(224,224,3),
                                        include_top=False,
                                        weights='imagenet')
VGG16_MODEL.trainable=False
global_average_layer = tf.keras.layers.GlobalAveragePooling2D()
prediction_layer = tf.keras.layers.Dense(3,activation='softmax')
```

```
model = tf.keras.Sequential([
  VGG16_MODEL,
  global_average_layer,
  prediction_layer
])

model.compile(optimizer="adam",
              loss=tf.keras.losses.sparse_categorical_crossentropy,
              metrics=["accuracy"])
```

```
model.summary()
```

CODE
EXPLANATION

```
history = model.fit(img_train,
                    epochs=18,
                    steps_per_epoch=30,
                    validation_steps=2,
                    validation_data=img_validation)
model.save("deeplearningVGG16.h5")
```

```
#import Test img

path_test = "dataset/test"

img_test = tf.keras.preprocessing.image_dataset_from_directory(
    path_test,
    image_size = (224,224)
)


label = ["glaucoma","normal","other"]

loadmodel = tf.keras.models.load_model("deeplearningVGG16.h5")

predict = loadmodel.predict(img_test)
prediction = np.argmax(predict)
```

CODE
EXPLANATION

```
i = 0
glau = 0
norm = 0
oth = 0
for i in predict:
    index = np.argmax(i)
    if index == 0:
        glau+=1
    elif index == 1:
        norm+=1
    elif index==2 :
        oth+=1

print(glau)
print(norm)
print(oth)
```

```
# change file type to jpeg

import imghdr
import cv2
import os
import glob


for file in glob.glob('dataset/test/other/*.jpg'):
    image = cv2.imread(file)
    file_type = imghdr.what(file)
    if file_type != 'jpeg':
        print(file + " - invalid - " + str(file_type))
        cv2.imwrite(file, image)
    print("finish jpg",image)
```

PROBLEM

GPU Error.

Some images can't crop.

The feature can't extract other images.

Reference

Machine Learning

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