Cataract Detection using KNN

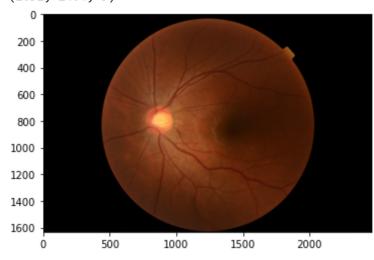
```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import cv2
import os

In [2]: DATADIR = 'Dataset'
# CATEGORIES = ['1_normal','2_cataract','2_glaucoma','3_retina_disease']
CATEGORIES = ['1_normal','2_cataract']
```

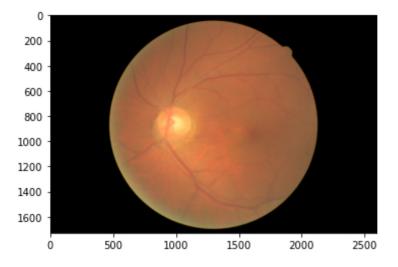
Displaying All Categories of Images

```
image = []
i = 0;
for category in CATEGORIES:
    path=os.path.join(DATADIR, category)
    print(path)
    for img in os.listdir(path):
        img_array=cv2.imread(os.path.join(path,img))
        img_array=cv2.cvtColor(img_array,cv2.COLOR_BGR2RGB)
        print(img_array.shape)
        image.append(img_array)
        plt.imshow(image[i])
        i += 1
        plt.show()
        break
```

Dataset\1_normal (1632, 2464, 3)



Dataset\2_cataract
(1728, 2592, 3)

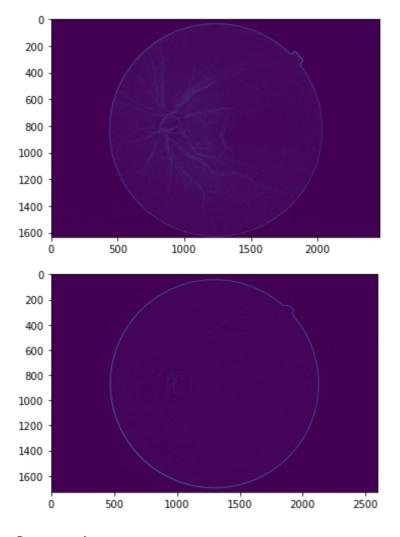


Creating Histogram for Images

```
In [4]:
    def bulid_histogram(image) :
        image_bg = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
        column,row = image_bg.shape
        # Calculate the histogram of the image
        hist, bins = np.histogram(image_bg.flatten(), 256, [0, 256])
        # Calculate the cumulative sum of the histogram
        cdf = hist.cumsum()
        cdf_normalized = cdf * hist.max() / cdf.max()
        # Apply histogram equalization to the image
        img_equalized = np.interp(image_bg.flatten(), bins[:-1], cdf_normalized).reshape(
        # Save the output image
        # cv2.imwrite('output_image.jpg', img_equalized)
        image_eq = img_equalized.copy()
        return image_eq
```

Applying Sobel Operator

```
In [5]: def sobelOperator(image) :
          # Load the image
          img = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
          # Apply the Sobel operator in the x-direction
          sobelx = cv2.Sobel(img, cv2.CV 64F, 1, 0, ksize=3)
          # Apply the Sobel operator in the y-direction
          sobely = cv2.Sobel(img, cv2.CV_64F, 0, 1, ksize=3)
          # Calculate the magnitude of the gradient
          mag = np.sqrt(sobelx**2 + sobely**2)
          # Normalize the magnitude image to range 0-255
          mag = cv2.normalize(mag, None, 0, 255, cv2.NORM_MINMAX, cv2.CV_8U)
          return mag
        for i in range(0,len(image)) :
          im = sobelOperator(image[i])
          plt.imshow(im)
          plt.show()
```



Preprocessing

```
IMG_SIZE = 400
In [6]:
        training_data=[]
        def create_training_data():
          for category in CATEGORIES:
            path=os.path.join(DATADIR, category)
            class_num=CATEGORIES.index(category)
            for img in os.listdir(path):
              try:
                 img_array=cv2.imread(os.path.join(path,img))
                 img_array = sobelOperator(img_array)
                new_array=cv2.resize(img_array,(IMG_SIZE,IMG_SIZE))
                training_data.append([new_array,class_num])
              except Exception as e:
                 pass
        create_training_data()
In [7]:
        print(len(training_data))
        400
In [8]: lenofimage = len(training_data)
```

```
In [9]: X=[]
         y=[]
         for categories, label in training_data:
           X.append(categories)
           y.append(label)
         X= np.array(X).reshape(lenofimage,-1)
In [10]: X.shape
         Χ
         array([[2, 0, 0, ..., 0, 0, 1],
Out[10]:
                [0, 2, 3, \ldots, 0, 1, 0],
                [2, 0, 0, \ldots, 0, 1, 1],
                [1, 0, 2, \ldots, 2, 2, 1],
                [0, 1, 1, \ldots, 0, 1, 0],
                [1, 1, 2, ..., 1, 1, 2]], dtype=uint8)
In [11]: X = X/255
         Χ
Out[11]: array([[0.00784314, 0. , 0.
                                                   , ..., 0.
                                                                   , 0.
                 0.00392157],
                           , 0.00784314, 0.01176471, ..., 0.
                                                                    , 0.00392157,
                [0.
                           ],
                                                                     , 0.00392157,
                [0.00784314, 0.
                                       , 0.
                                                  , ..., 0.
                 0.00392157],
                                     , 0.00784314, ..., 0.00784314, 0.00784314,
                [0.00392157, 0.
                 0.00392157],
                           , 0.00392157, 0.00392157, ..., 0. , 0.00392157,
                [0.00392157, 0.00392157, 0.00784314, ..., 0.00392157, 0.00392157,
                 0.00784314]])
In [12]: X[1]
         array([0.
                          , 0.00784314, 0.01176471, ..., 0.
                                                                   , 0.00392157,
Out[12]:
                          ])
                0.
In [13]: y=np.array(y)
In [14]: y.shape
         (400,)
Out[14]:
         Creating Model Using KNN
In [15]: from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2)
```

```
from sklearn.neighbors import KNeighborsClassifier
In [16]:
         # model = KNeighborsClassifier(n_neighbors=10,weights='distance',metric='minkowski
         model = KNeighborsClassifier(n_neighbors=6)
         model.fit(X_train, y_train)
```

KNeighborsClassifier(n_neighbors=6) Out[16]:

```
In [17]: y2 = model.predict(X_test)
```

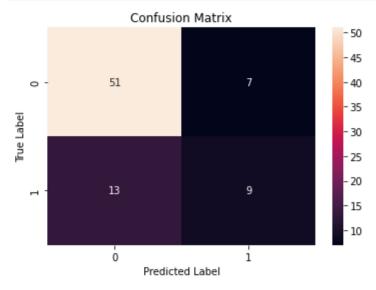
In [24]: | from sklearn.metrics import accuracy_score,classification_report print("Accuracy on unknown data is",accuracy_score(y_test,y2))

Accuracy on unknown data is 0.75

In [25]: | print("Accuracy on unknown data is",classification_report(y_test,y2))

Accuracy on unknown data is				precision		recall	f1-score	support
	0 1	0.80 0.56	0.88 0.41	0.84 0.47	58 22			
accura macro a weighted a	vg	0.68 0.73	0.64 0.75	0.75 0.65 0.74	80 80 80			

```
In [19]:
         from sklearn.metrics import confusion_matrix
          import seaborn as sns
          cm = confusion_matrix(y_test,y2)
          sns.heatmap(cm, annot=True, fmt="d")
          plt.title("Confusion Matrix")
         plt.ylabel("True Label")
         plt.xlabel("Predicted Label")
          plt.show()
```



Elbow Method for Choosing Reasonable K values

```
In [20]: test_error_rates = []
```

In []:

```
In [21]: for k in range(1,30):
              knn_model = KNeighborsClassifier(n_neighbors=k)
              knn_model.fit(X_train,y_train)
              y_pred_test = knn_model.predict(X_test)
              test_error = 1 - accuracy_score(y_test,y_pred_test)
              test_error_rates.append(test_error)
          plt.figure(figsize=(10,6))
In [29]:
          plt.plot(range(1,30),test_error_rates,label='Test Error')
          plt.legend()
          plt.ylabel('Error Rate')
          plt.xlabel('K Value')
          Text(0.5, 0, 'K Value')
Out[29]:
                                                                                           Test Error
            0.31
            0.30
            0.29
          Error Rate
            0.28
            0.27
            0.26
            0.25
                               5
                                                                                    25
                                           10
                                                         15
                                                                      20
                                                                                                 30
                                                       K Value
```