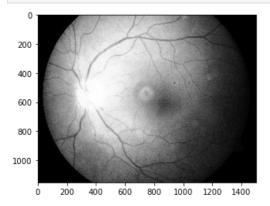
```
In [2]: from scipy import misc
         from PIL import Image
         from skimage import exposure
         from sklearn import svm
         import scipy
         from math import sqrt,pi
         from numpy import exp
         from matplotlib import pyplot as plt
         import numpy as np
         import glob
         import matplotlib.pyplot as pltss
         import cv2
         from matplotlib import cm
         import pandas as pd
         \textbf{from} \ \text{math} \ \textbf{import} \ \text{pi, sqrt}
         import pywt
```

Pre-processing

```
In [103...
         immatrix=[]
         im_unpre = []
         for i in range(1,90):
              img pt = r'C:\Users\Rohan\Desktop\Diabetic Retinopathy\diaretdb1 v 1 1\diaretdb1 v 1 1\resources\images\ddb
             if \bar{i} < 10:
                 img_pt = img_pt + "00" + str(i) + ".png"
             else:
                 img_pt = img_pt + "0" + str(i)+ ".png"
             img = cv2.imread(img_pt)
             img_gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
             equ = cv2.equalizeHist(img_gray)
             immatrix.append(np.array(equ).flatten())
 In [4]: np.shape(np.array(equ).flatten())
```

```
(1728000,)
Out[4]:
```

```
In [111...
         np.shape(immatrix)
         np.shape(equ)
         plt.imshow(immatrix[78].reshape((1152,1500)),cmap='gray')
         plt.show()
```



```
In [6]:
          imm dwt = []
          for equ in immatrix:
               equ = equ.reshape((1152,1500))
               coeffs = pywt.dwt2(equ, 'haar')
equ2 = pywt.idwt2(coeffs, 'haar')
               imm_dwt.append(np.array(equ2).flatten())
```

Visualising a random image

```
In [7]:
        np.shape(imm_dwt)
        np.shape(equ2)
        plt.imshow(imm dwt[78].reshape((1152,1500)),cmap='gray')
        plt.show()
```

```
0
200 -
400 -
800 -
1000 -
0 200 400 600 800 1000 1200 1400
```

Performing K-means Clusttering with PP centers(non random) neighbours on the final image

```
In []: img = equ3
    Z = img.reshape((-1,3))

Z = np.float32(Z)
    k=cv2.KMEANS_PP_CENTERS

criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 10, 1.0)
    K = 2
    ret,label,center=cv2.kmeans(Z,K,None,criteria,10,k)

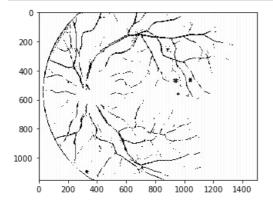
center = np.uint8(center)
    res = center[label.flatten()]
    res2 = res.reshape((img.shape))
```

```
In [10]: imm_kmean = []
for equ3 in imm_gauss2:
    img = equ3.reshape((1152,1500))
    Z = img.reshape((-1,3))

Z = np.float32(Z)
    k=cv2.KMEANS_PP_CENTERS

criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 10, 1.0)
    K = 2
    ret,label,center=cv2.kmeans(Z,K,None,criteria,10,k)

center = np.uint8(center)
    res = center[label.flatten()]
    res2 = res.reshape((img.shape))
    imm_kmean.append(np.array(res2).flatten())
```



Model training

```
In [42]: from sklearn.svm import SVC
         clf = SVC()
In [64]: Y = np.ones(89)
In [65]: Y[1]=Y[5]=Y[7]=Y[17]=Y[6]=0
In [66]: clf.fit(imm_kmean, Y)
Out[66]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
           decision_function_shape=None, degree=3, gamma='auto', kernel='rbf',
           max_iter=-1, probability=False, random_state=None, shrinking=True,
           tol=0.001, verbose=False)
In [72]: y_pred = clf.predict(imm_kmean)
 In [1]: k = [1,3,4,9,10,11,13,14,20,22,24,25,26,27,28,29,35,36,38,42,53,55,57,64,70,79,84,86]
 In [3]: k = k-np.ones(len(k))
In [87]: k
Out[87]: array([ 0.,
                                          9., 10., 12., 13., 19., 21.,
                       2., 3.,
                                   8.,
                 24., 25., 26., 27., 28., 34., 35., 37., 41., 52., 54., 56., 63., 69., 78., 83., 85.])
In [92]: k = [int(x) for x in k]
In [93]: k
Out[93]:
          3,
          8,
          9,
          12,
          13,
          19,
          21,
          23,
          24,
          25,
          26,
          27,
          28,
          34,
          35,
          37,
          41,
          52,
          54,
          63,
          69.
          78,
          83,
          85]
In [98]: imm_train = []
         y train = []
          k.append(5)
          k.append(7)
          for i in k:
             imm_train.append(imm_kmean[i])
              y_train.append(Y[i])
In [99]: y_train
```

```
Out[99]: [1.0,
              1.0,
              1.0,
              1.0,
              1.0.
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              1.0,
              0.0,
              0.0]
   In [100... clf.fit(imm_train, y_train)
  Out[100]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
               decision function shape=None, degree=3, gamma='auto', kernel='rbf',
               max_iter=-1, probability=False, random_state=None, shrinking=True,
               tol=0.001, verbose=False)
   In [101... y_pred = clf.predict(imm kmean)
   In [102... accuracy_score(Y,y_pred)
  Out[102]: 0.9662921348314607
   In [114. from sklearn.neighbors import KNeighborsClassifier
   In [115... neigh = KNeighborsClassifier(n_neighbors=3)
   In [116... neigh.fit(imm_train, y_train)
             KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
  Out[116]:
                         metric_params=None, n_jobs=1, n_neighbors=3, p=2,
                         weights='uniform')
   In [117... y_pred2=neigh.predict(imm_kmean)
   In [119... neigh.score(imm_kmean,Y)
             0.9438202247191011
  Out[119]:
Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js
```