naczynia-oka

May 10, 2022

```
[4]: import cv2
import numpy as np
import math
import matplotlib.pyplot as plt
import os
import re
```

1 1. Przetwarzanie obrazów

1.0.1 Wczytanie obrazów

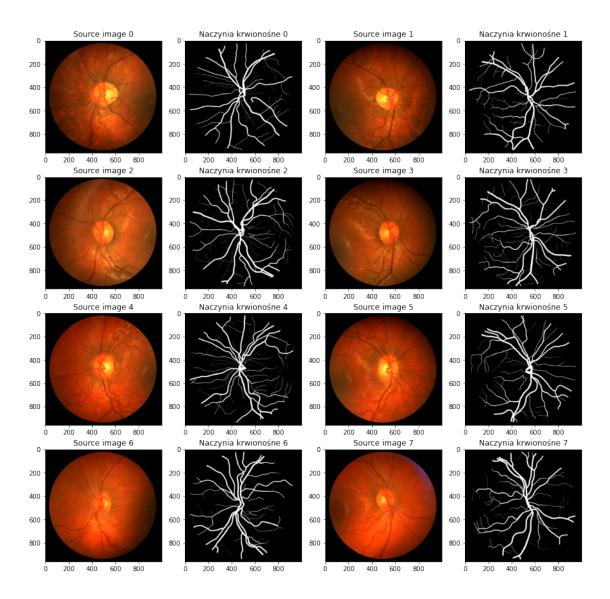
```
[5]: # source: https://blogs.kingston.ac.uk/retinal/chasedb1/
     imageFolder = "images"
     imageNames = os.listdir(imageFolder)
     imageNames.sort()
     images = {}
     for filename in imageNames:
         img = cv2.imread(imageFolder + "/" + filename)
         images[filename] = img
     manualLabeledImageFolder = "manual1"
     labeledImageNames = os.listdir(manualLabeledImageFolder)
     labeledImageNames.sort()
     manualVessels = {}
     for filename in labeledImageNames:
         img = cv2.imread(manualLabeledImageFolder + "/" + filename)
         manualVessels[filename] = img
     print(images.keys())
     print(manualVessels.keys())
```

```
dict_keys(['01_h.jpg', '02_h.jpg', '03_h.jpg', '04_h.jpg', '05_h.jpg',
'06_h.jpg', '07_h.jpg', '08_h.jpg', '09_h.jpg', '10_h.jpg', '11_h.jpg',
'12_h.jpg', '13_h.jpg', '14_h.jpg', '15_h.jpg'])
```

```
dict_keys(['01_h.jpg', '02_h.jpg', '03_h.jpg', '04_h.jpg', '05_h.jpg',
'06_h.jpg', '07_h.jpg', '08_h.jpg', '09_h.jpg', '10_h.jpg', '11_h.jpg',
'12_h.jpg', '13_h.jpg', '14_h.jpg', '15_h.jpg'])
```

1.0.2 Przykładowe obrazy

```
[6]: figure = plt.figure(figsize=(15, 15))
    columns = 4
    rows = 4
    for i in range(0, columns*rows):
        figure.add_subplot(rows, columns, i+1)
        if (i+1)%2:
            img = list(images.values())[int(i/2)]
            plt.title(f"Source image {int(i/2)}")
            plt.imshow(img[:,:,::-1])
        else:
            img = list(manualVessels.values())[int(i/2)]
            plt.title(f"Naczynia krwionośne {int(i/2)}")
            plt.imshow(img, cmap='gray')
        plt.show()
```

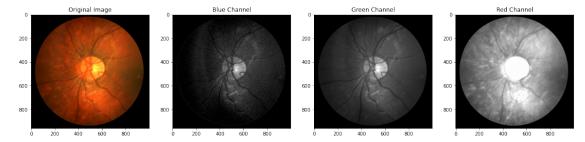


1.0.3 Wybór najlepszego kanału koloru

```
[7]: img = images['01_h.jpg']
  imgBlueChannel = img[:,:,0]
  imgGreenChannel = img[:,:,1]
  imgRedChannel = img[:,:,2]

plt.figure(figsize=(20,12))
  plt.subplot(141)
  plt.imshow(img[:,:,::-1]);
  plt.title("Original Image")
  plt.subplot(142)
  plt.imshow(imgBlueChannel, cmap='gray');
```

```
plt.title("Blue Channel")
plt.subplot(143)
plt.imshow(imgGreenChannel, cmap='gray');
plt.title("Green Channel")
plt.subplot(144)
plt.imshow(imgRedChannel, cmap='gray');
plt.title("Red Channel");
```

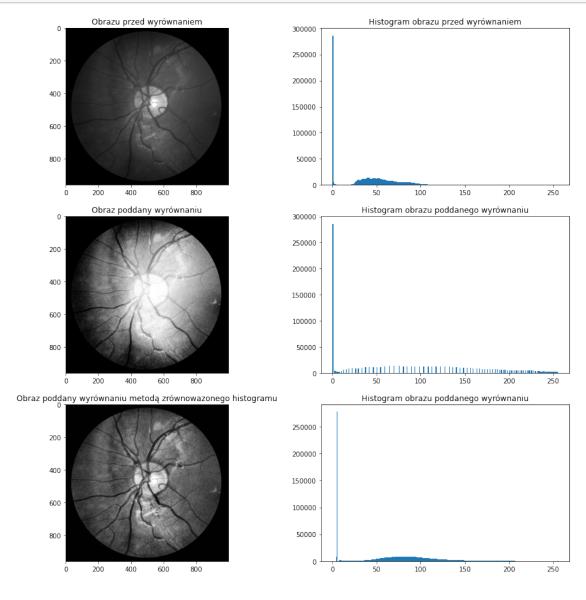


1.0.4 Zwiększenie kontrastu metodą wyrównania histogramu

• od tego momentu przetwarzamy już tylko kanał zielony

```
[8]: imgEqualized = cv2.equalizeHist(imgGreenChannel)
     clahe = cv2.createCLAHE(clipLimit=4.0, tileGridSize=(8,8))
     imgEqualizedCLAHE = clahe.apply(imgGreenChannel)
     fig = plt.figure(figsize=(15, 15))
     plt.subplot(321)
     plt.imshow(imgGreenChannel, cmap='gray')
     plt.title("Obrazu przed wyrównaniem")
     plt.subplot(322)
     plt.hist(imgGreenChannel.ravel(),256,[0,256])
     plt.title("Histogram obrazu przed wyrównaniem")
     plt.subplot(323)
     plt.imshow(imgEqualized, cmap='gray')
     plt.title("Obraz poddany wyrównaniu")
     plt.subplot(324)
     plt.hist(imgEqualized.ravel(),256,[0,256])
     plt.title("Histogram obrazu poddanego wyrównaniu")
     plt.subplot(325)
     plt.imshow(imgEqualizedCLAHE, cmap='gray')
     plt.title("Obraz poddany wyrównaniu metodą zrównowazonego histogramu")
     plt.subplot(326)
```

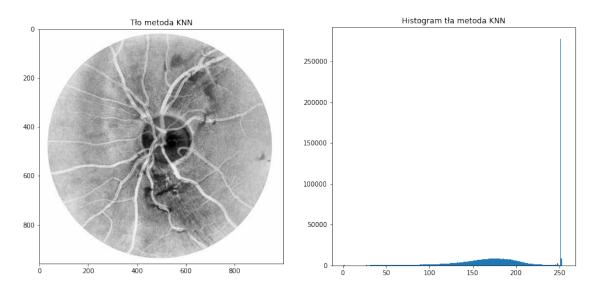
```
plt.hist(imgEqualizedCLAHE.ravel(),256,[0,256])
plt.title("Histogram obrazu poddanego wyrównaniu")
plt.show()
```



```
imgBackgroundKNN = backgroundSubstractorKNN.apply(imgEqualizedCLAHE.copy())
imgKNNSubtracted = imgEqualizedCLAHE * imgBackgroundKNN
fig = plt.figure(figsize=(15, 15))

plt.subplot(221)
plt.imshow(imgKNNSubtracted, cmap='gray')
plt.title("Tło metoda KNN")
plt.subplot(222)
plt.hist(imgKNNSubtracted.ravel(),256,[0,256])
plt.title("Histogram tła metoda KNN")
```

[9]: Text(0.5, 1.0, 'Histogram tła metoda KNN')



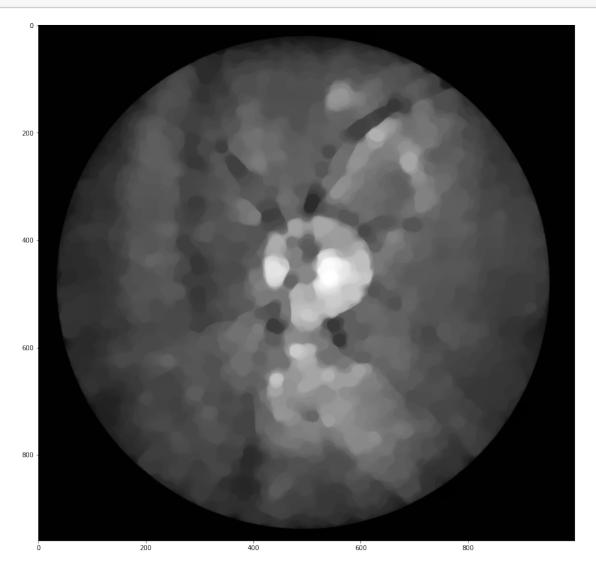
```
kernelSizes = [7, 11, 13, 21]
kernels = [cv2.getStructuringElement(cv2.MORPH_ELLIPSE,(size, size)) for size__
in kernelSizes]

morphology_steps = [imgEqualizedCLAHE.copy()]
for kernel in kernels:
    imageOpened = cv2.morphologyEx(morphology_steps[-1], cv2.MORPH_OPEN,__
kernel, iterations = 1)
    imageClosed = cv2.morphologyEx(imageOpened, cv2.MORPH_CLOSE, kernel,__
iterations = 1)
    morphology_steps.extend([imageOpened, imageClosed])

img = morphology_steps[-1]

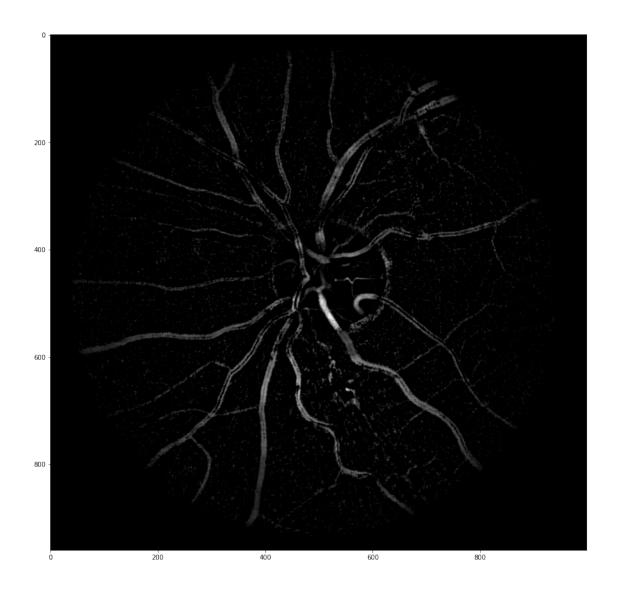
fig = plt.figure(figsize=(15, 15))
plt.imshow(img, cmap='gray')
```

plt.show()



```
[11]: vessels = cv2.subtract(morphology_steps[-1],imgEqualizedCLAHE)

fig = plt.figure(figsize=(15, 15))
   plt.imshow(vessels, cmap='gray')
   plt.show()
```



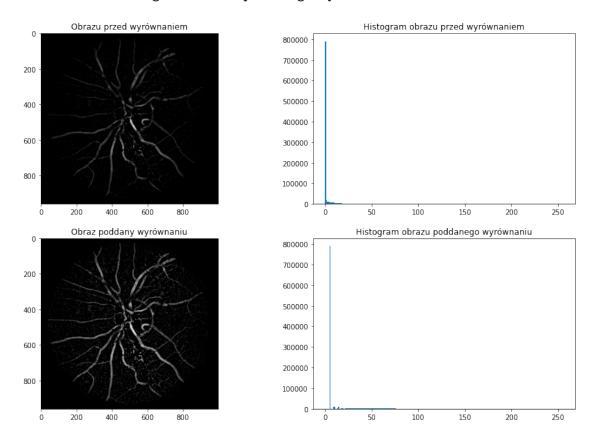
```
[16]: vesselsCLAHE = clahe.apply(vessels)
    fig = plt.figure(figsize=(15, 10))

plt.subplot(221)
    plt.imshow(vessels, cmap='gray')
    plt.title("Obrazu przed wyrównaniem")
    plt.subplot(222)
    plt.hist(vessels.ravel(),256,[0,256])
    plt.title("Histogram obrazu przed wyrównaniem")

plt.subplot(223)
    plt.imshow(vesselsCLAHE, cmap='gray')
    plt.title("Obraz poddany wyrównaniu")
    plt.subplot(224)
```

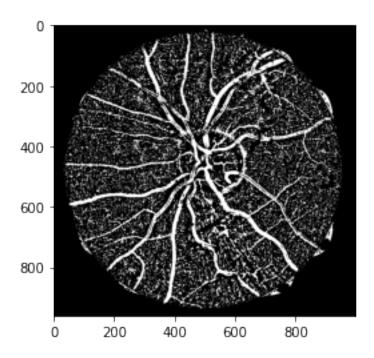
```
plt.hist(vesselsCLAHE.ravel(),256,[0,256])
plt.title("Histogram obrazu poddanego wyrównaniu")
```

[16]: Text(0.5, 1.0, 'Histogram obrazu poddanego wyrównaniu')



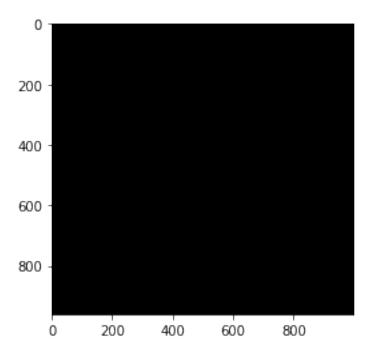
[30]: ret,vesselsThreshold = cv2.threshold(vesselsCLAHE,5,255,cv2.THRESH_BINARY) plt.imshow(vesselsThreshold,cmap='gray')

[30]: <matplotlib.image.AxesImage at 0x7fc956b7d180>



```
[31]: noise_mask = np.ones(vesselsThreshold.shape[:2], dtype="uint8") * 255
plt.imshow(noise_mask,cmap='gray')
print(np.shape(noise_mask))
```

(960, 999)



```
[32]: contours, hierarchy = cv2.findContours(vesselsThreshold.copy(),cv2.

—RETR_LIST,cv2.CHAIN_APPROX_SIMPLE)

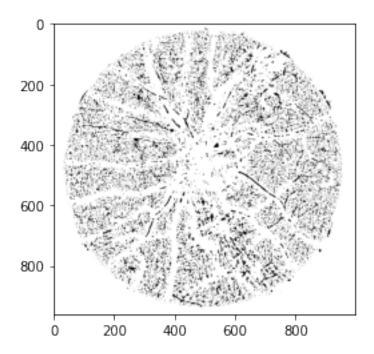
for contour in contours:

    if cv2.contourArea(contour) <= 200:

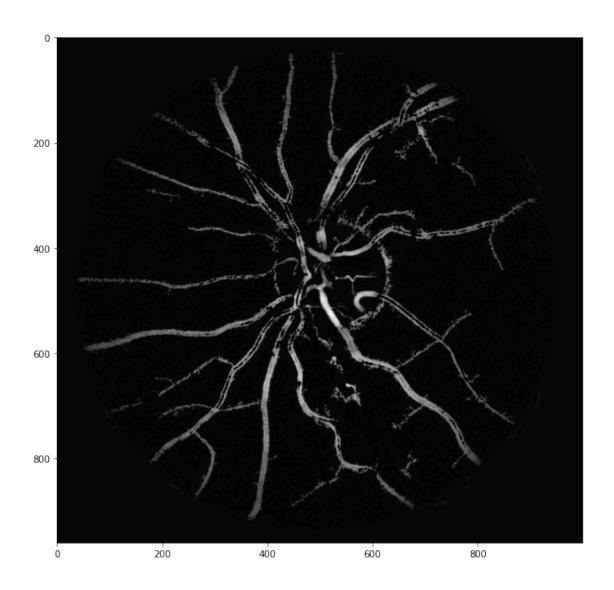
        cv2.drawContours(noise_mask, [contour], -1, 0, -1)

plt.imshow(noise_mask, cmap='gray')
```

[32]: <matplotlib.image.AxesImage at 0x7fc956da0f40>



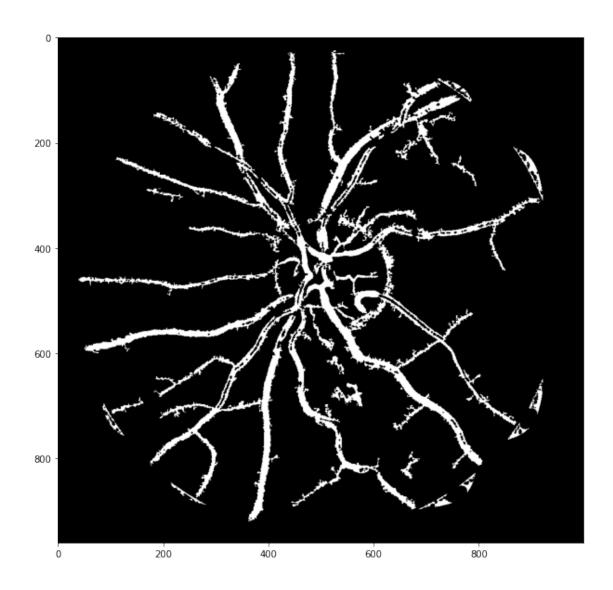
[33]: <matplotlib.image.AxesImage at 0x7fc91459a2c0>



```
[34]: ret,vesselsThresholded = cv2.threshold(vesselsNoiseReducted,5,255,cv2.

OTHRESH_BINARY)
fig = plt.figure(figsize=(15, 10))
plt.imshow(vesselsThresholded, cmap='gray')
```

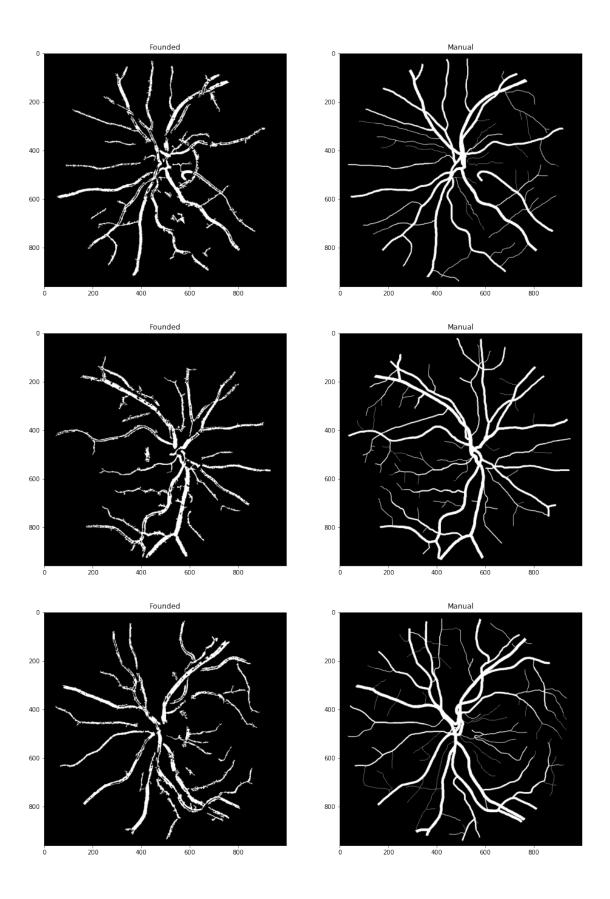
[34]: <matplotlib.image.AxesImage at 0x7fc9145cd330>



1.1 Now all of this steps for each img

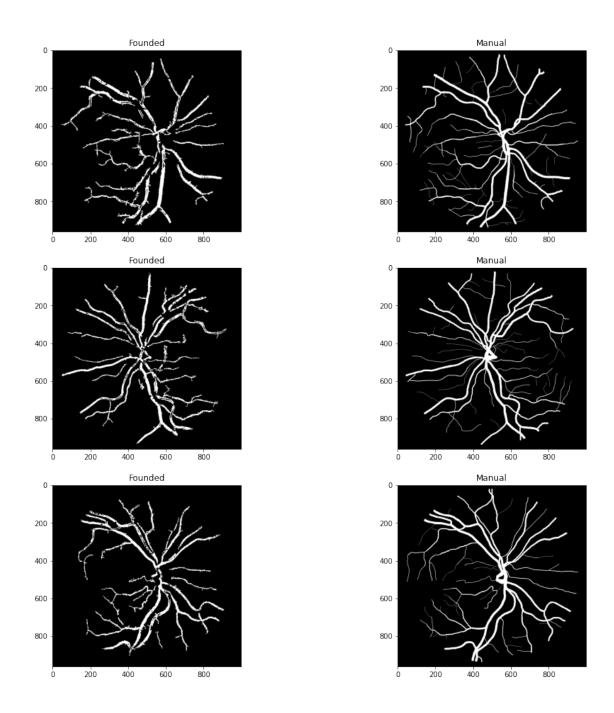
```
morphology_steps = [imageEqualizedCLAHE.copy()]
          for kernel in kernels:
              imageOpened = cv2.morphologyEx(morphology_steps[-1], cv2.MORPH_OPEN,_
       ⇔kernel, iterations = 1)
              imageClosed = cv2.morphologyEx(imageOpened, cv2.MORPH_CLOSE, kernel,_
       \hookrightarrowiterations = 1)
              morphology_steps.extend([imageOpened, imageClosed])
          # subtract vessels from background
          vessels = cv2.subtract(morphology_steps[-1],imageEqualizedCLAHE)
          # first threshold for contours
          _, vesselsThreshold = cv2.threshold(vessels,threshold,255,cv2.THRESH_BINARY)
          # get image noise
          noiseMask = np.ones(vesselsThreshold.shape[:2], dtype="uint8") * 255
          # delete noise in small contours
          contours, _ = cv2.findContours(vesselsThreshold.copy(),cv2.RETR_LIST,cv2.
       →CHAIN_APPROX_SIMPLE)
          for contour in contours:
              if cv2.contourArea(contour) <= noiseContour:</pre>
                  cv2.drawContours(noiseMask, [contour], -1, 0, -1)
          vesselsNoiseReducted = cv2.bitwise_and(vessels, vessels, mask=noiseMask)
          # second threshold for result
          _, vesselsThresholded = cv2.
       sthreshold(vesselsNoiseReducted,threshold,255,cv2.THRESH_BINARY)
          return vesselsThresholded
[36]: imagesVessels = {}
      for name, img in list(images.items())[0:6]:
          print(name)
          # imagesVessels[name] = extractVessels(img, kernelSizes = [37, 49, 55, 73], u
       \rightarrowthreshold = 5, noiseContour = 1000)
          imagesVessels[name] = extractVessels(img, kernelSizes =[7, 11, 13, 21],
       →threshold = 5, noiseContour = 200)
     01_h.jpg
     02_h.jpg
     03_h.jpg
     04_h.jpg
     05_h.jpg
     06_h.jpg
```

```
[37]: figure = plt.figure(figsize=(16,24))
      name = '01_h.jpg'
      plt.subplot(321)
      plt.imshow(imagesVessels[name], cmap='gray')
      print(np.shape(imagesVessels[name]))
      plt.title("Founded")
      plt.subplot(322)
      plt.imshow(manualVessels[name], cmap='gray')
      print(np.shape(manualVessels[name]))
      plt.title("Manual")
      name = '02_h.jpg'
      plt.subplot(323)
      plt.imshow(imagesVessels[name], cmap='gray')
      plt.title("Founded")
      plt.subplot(324)
      plt.imshow(manualVessels[name], cmap='gray')
      plt.title("Manual")
      name = '03_h.jpg'
      plt.subplot(325)
      plt.imshow(imagesVessels[name], cmap='gray')
      plt.title("Founded")
      plt.subplot(326)
      plt.imshow(manualVessels[name], cmap='gray')
      plt.title("Manual")
     (960, 999)
     (960, 999, 3)
[37]: Text(0.5, 1.0, 'Manual')
```



```
[38]: figure = plt.figure(figsize=(16,16))
      name = '04_h.jpg'
      plt.subplot(321)
      plt.imshow(imagesVessels[name], cmap='gray')
      plt.title("Founded")
      plt.subplot(322)
      plt.imshow(manualVessels[name], cmap='gray')
      plt.title("Manual")
      name = '05_h.jpg'
      plt.subplot(323)
      plt.imshow(imagesVessels[name], cmap='gray')
      plt.title("Founded")
      plt.subplot(324)
      plt.imshow(manualVessels[name], cmap='gray')
      plt.title("Manual")
      name = '06_h.jpg'
      plt.subplot(325)
      plt.imshow(imagesVessels[name], cmap='gray')
      plt.title("Founded")
      plt.subplot(326)
      plt.imshow(manualVessels[name], cmap='gray')
      plt.title("Manual")
```

[38]: Text(0.5, 1.0, 'Manual')



[39]: from sklearn.metrics import confusion_matrix

def summary(vesselsManual, vesselsFounded):
 conf = confusion_matrix(vesselsManual.flatten(), vesselsFounded.flatten())
 TrueNeg, FalsePos, FalseNeg, TruePos = conf.ravel()

```
accuracy = (float (TruePos+TrueNeg) / float(TruePos + TrueNeg + FalsePos + FalseNeg))

sensitivity = (TruePos / float(TruePos + FalseNeg))

specificity = (TrueNeg / float(TrueNeg + FalsePos))

geoMean = (sensitivity * specificity)**(1.0/2)

return (accuracy, sensitivity, specificity, geoMean)
```

```
01_h.jpg
accuracy: 0.9582 sensitivity: 0.65767 specificity: 0.98074 geoMean: 0.80312
02_h.jpg
accuracy: 0.9558 sensitivity: 0.60714 specificity: 0.98446 geoMean: 0.77311
03_h.jpg
accuracy: 0.95125 sensitivity: 0.60677 specificity: 0.98167 geoMean: 0.77178
04_h.jpg
accuracy: 0.94936 sensitivity: 0.6229 specificity: 0.97702 geoMean: 0.78012
05_h.jpg
accuracy: 0.95666 sensitivity: 0.68751 specificity: 0.97956 geoMean: 0.82065
06_h.jpg
accuracy: 0.95836 sensitivity: 0.67177 specificity: 0.98177 geoMean: 0.81211
```

2 2. Uczenie maszynowe

```
[2]: from skimage.util import view_as_windows
import cv2
import numpy as np
import math
import matplotlib.pyplot as plt
import os
import re
import pandas as pd
from sklearn.ensemble import RandomForestClassifier
from imblearn.under_sampling import RandomUnderSampler
from multiprocessing import Pool
```

```
[111]: imageFolder = "images"
       imageNames = os.listdir(imageFolder)
       imageNames.sort()
       images = {}
       for filename in imageNames:
           img = cv2.imread(imageFolder + "/" + filename)
           images[filename] = img
       manualLabeledImageFolder = "manual1"
       labeledImageNames = os.listdir(manualLabeledImageFolder)
       labeledImageNames.sort()
       manualVessels = {}
       for filename in labeledImageNames:
           img = cv2.imread(manualLabeledImageFolder + "/" + filename)
           manualVessels[filename] = img
[23]: IMAGE\_SHAPE = (999, 960, 3)
       TILE\_SHAPE = (5, 5, 3)
       def imageFeatures(img, img_id):
           return np.array([tileFeatures(tile, img id) for tile in imageToTiles(img)])
       def imageTrainingValues(img):
           return np.array([tileTrainingValue(tile) for tile in imageToTiles(img)])
       def tileFeatures(tile, img_id):
           channels = tile.reshape(-1,3)
           mean = np.mean(channels, axis=0)
           std = np.std(channels, axis=0)
           greenChannel = tile[:,:,1]
           centralMomentsKeys = ['mu20', 'mu11', 'mu02', 'mu30', 'mu21', 'mu12', "mu12']

    'mu03']

           moments = cv2.moments(greenChannel)
           centralMoments = [moments[x] for x in centralMomentsKeys]
           huMoments = cv2.HuMoments(moments)[:,0]
           return np.hstack([img_id, mean, std, centralMoments, huMoments])
       def tileTrainingValue(tile):
           tileMiddlePointIndex = len(tile)//2
           tileMiddlePoint = tile[tileMiddlePointIndex, tileMiddlePointIndex]
           if any(x > 0 for x in tileMiddlePoint):
               return True
```

```
else:
        return False
def imageToTiles(img):
   tiles = []
    for i in range(0, IMAGE_SHAPE[1]-5):
        for j in range(0, IMAGE_SHAPE[0]-5):
            tiles.append(img[i:i+5, j:j+5])
    return tiles
# print(splitImage(images['01_h.jpg'], (5,5,3)))
# print(imageFeatures(images['01_h.jpg'])[0])
# print(imageTrainingValues(manualVessels['01_h.jpg']))
allImagesData = []
# for i, image in enumerate(images.values()):
    allImagesData.append(imageFeatures(image, i))
with Pool() as pool:
    allImagesData = pool.starmap(imageFeatures, zip(images.values(),__
\leftarrow[0,1,2,3,4,5,6,7,8,9,10,11,12,13,14]))
with Pool() as pool:
    allImagesTrainingValues = pool.map(imageTrainingValues, manualVessels.
 ⇔values())
# allImagesTrainingValues = []
# for image in manualVessels.values():
      allImagesTrainingValues.append(imageTrainingValues(image))
print("Done")
```

Done

```
[24]: allImagesData = np.array(allImagesData).reshape(-1, 21)
allImagesTrainingValues = np.array(allImagesTrainingValues).reshape(-1)
print("Done")
```

Done

```
df = pd.DataFrame(allImagesData, columns = features)
      df['is_vessel'] = allImagesTrainingValues
      df.head(5000000)
[25]:
               img_id avg_r
                                        avg_b
                                                   std r
                                                             std_g
                                                                        std b \
                               avg_g
                                0.00
      0
                   0.0
                         0.00
                                         0.00
                                               0.000000
                                                          0.000000
                                                                    0.000000
      1
                   0.0
                         0.00
                                0.00
                                         0.00
                                               0.000000
                                                          0.000000
                                                                    0.00000
      2
                   0.0
                         0.00
                                0.00
                                         0.00
                                               0.000000
                                                          0.000000
                                                                     0.000000
      3
                   0.0
                         0.00
                                0.00
                                         0.00
                                               0.000000
                                                          0.000000
                                                                     0.00000
      4
                   0.0
                         0.00
                                0.00
                                         0.00
                                               0.000000
                                                          0.000000
                                                                    0.000000
                                 •••
                           •••
      4999995
                   5.0
                         2.52
                               39.24
                                       142.04
                                               1.472956
                                                          1.175755
                                                                     1.708333
      4999996
                   5.0
                         2.96
                               39.40
                                       142.64
                                               1.038460
                                                          1.131371
                                                                     1.162067
      4999997
                   5.0
                         3.00
                               39.56
                                       142.92
                                               0.894427
                                                          1.267438
                                                                     1.197330
      4999998
                   5.0
                         2.88
                               39.88
                                       143.60
                                               0.765245
                                                          1.394848
                                                                     1.697056
                         3.04
                               40.32
                                       144.40
                                               0.823650
      4999999
                   5.0
                                                          1.406272
                                                                    1.939072
                       cm 1
                                   cm 2
                                                cm 3
                                                               cm 6
                                                                           cm 7
      0
                   0.000000
                              0.000000
                                            0.000000
                                                           0.000000
                                                                       0.000000
      1
                   0.000000
                              0.000000
                                            0.000000
                                                           0.000000
                                                                       0.000000
      2
                   0.000000
                              0.000000
                                            0.000000
                                                           0.000000
                                                                       0.000000
      3
                   0.000000
                              0.000000
                                            0.000000
                                                           0.000000
                                                                       0.00000
      4
                   0.000000
                              0.000000
                                            0.000000
                                                           0.000000
                                                                       0.000000
      4999995
               1957.983690 -27.032620
                                         1991.934760
                                                          21.318575 -34.732881
      4999996
               1976.998985 -13.995939
                                         2003.983756
                                                          23.851810
                                                                     -3.585919
      4999997
               1985.801820
                             -5.900910
                                         2018.950455
                                                           3.336084
                                                                     -0.130125
      4999998
               1998.557673
                              6.147442
                                         2038.950853
                                                          -4.861520
                                                                     -0.052847
                                                       ... -12.937213
      4999999
               2017.329365
                             12.128968
                                         2056.975198
                                                                     -4.390127
                    hm 1
                                  hm 2
                                                 hm 3
                                                                hm 4
                                                                               hm 5
      0
               0.000000
                          0.000000e+00
                                         0.000000e+00
                                                        0.000000e+00
                                                                       0.000000e+00
      1
               0.000000
                          0.000000e+00
                                         0.000000e+00
                                                        0.000000e+00
                                                                       0.000000e+00
      2
               0.000000
                          0.000000e+00
                                         0.000000e+00
                                                        0.000000e+00
                                                                       0.000000e+00
      3
               0.000000
                          0.00000e+00
                                         0.000000e+00
                                                        0.000000e+00
                                                                       0.000000e+00
      4
               0.000000
                          0.000000e+00
                                         0.000000e+00
                                                        0.000000e+00
                                                                       0.00000e+00
                          4.400772e-09
                                         9.079326e-12
      4999995
               0.004104
                                                        1.127439e-12 1.513383e-24
      4999996
               0.004103
                          1.605933e-09
                                         6.426963e-12
                                                        3.849430e-13 -6.013833e-25
      4999997
               0.004094
                          1.294124e-09
                                         3.080712e-12
                                                        1.123190e-12
                                                                       1.971613e-24
      4999998
                                                        4.018175e-12
               0.004062
                          1.804328e-09
                                         2.728317e-12
                                                                       1.317953e-23
      4999999
               0.004010
                          2.092473e-09
                                         9.481519e-13
                                                        6.653168e-12
                                                                       1.614154e-23
                                             is vessel
                        hm 6
                                       hm 7
      0
               0.000000e+00 0.000000e+00
                                                 False
```

'hm_6','hm_7']

```
1
               0.000000e+00 0.000000e+00
                                                False
      2
               0.000000e+00 0.000000e+00
                                                False
      3
               0.000000e+00 0.000000e+00
                                                False
               0.000000e+00 0.000000e+00
      4
                                                False
                                                False
      4999995 7.167321e-17 -3.274347e-24
      4999996 -8.774278e-18 -7.028720e-26
                                                False
      4999997 -2.278728e-17 6.913789e-25
                                                False
      4999998 -1.706627e-16 -1.817492e-24
                                                False
      4999999 -2.386435e-16 -4.322219e-24
                                                False
      [5000000 rows x 22 columns]
[26]: from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test = train_test_split(df.drop(['is_vessel'],_

¬axis='columns'),allImagesTrainingValues, test_size=0.2)

[27]: undersampler = RandomUnderSampler(sampling_strategy=1, random_state=1)
      x_train, y_train = undersampler.fit_resample(x_train, y_train)
      classifier = RandomForestClassifier()
      classifier.fit(x_train, y_train)
[27]: RandomForestClassifier()
[28]: classifier.score(x_train,y_train)
[28]: 0.9999841140520287
[161]: predictedImages = {}
      for i, imageKey in enumerate(images.keys()):
           imageDf = df[df.img_id == i]
          x_train, y_train = imageDf[features], df.is_vessel
          Y = classifier.predict_proba(x_train)[:,1] > 0.5
          originalImage = images[imageKey][:,:,::-1]
          manualImage = manualVessels[imageKey]
          mask = np.pad(Y.reshape(955, 994), 2)
          mask = np.pad(mask, ((1,0), (1,0)))
          mask = np.dstack([mask] * 3)
          predictedImage = np.broadcast_to([1.0,1.0,1.0], originalImage.shape)
          predictedImage = mask*(predictedImage + originalImage)
          predictedImages[imageKey] = predictedImage
```

```
fig, (ax1, ax2, ax3) = plt.subplots(1, 3, figsize=(20,100))
ax1.imshow(originalImage)
ax2.imshow(manualImage)
ax3.imshow(predictedImage, cmap='gray')
if i == 4:
    break
```

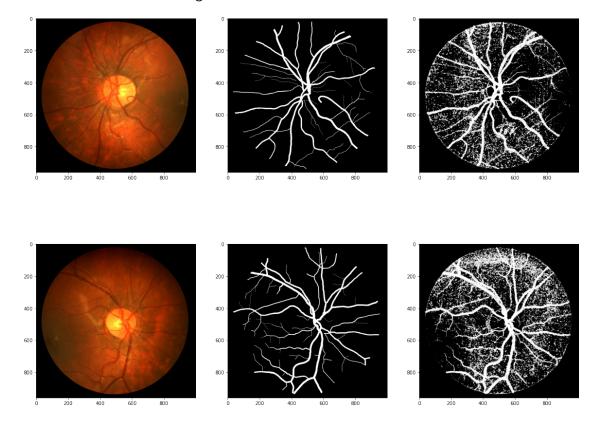
Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

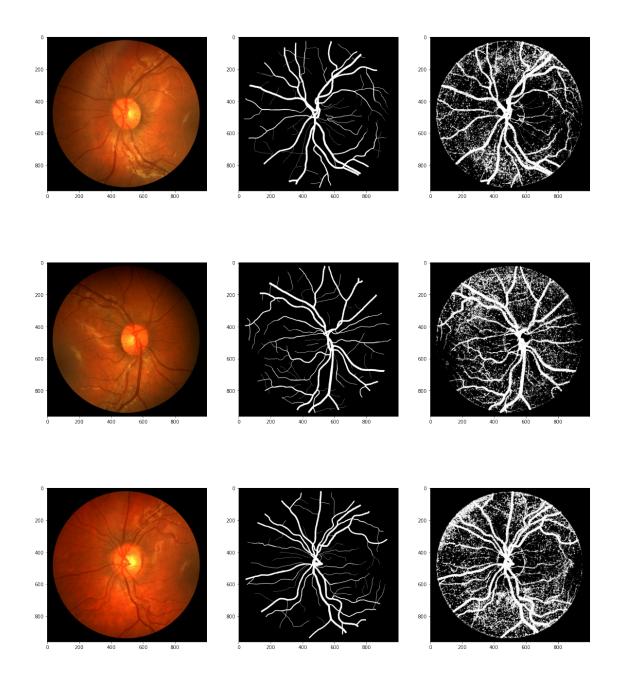
Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).





```
[162]: from sklearn.metrics import confusion_matrix

def summary(vesselsManual, vesselsFound):
    conf = confusion_matrix(vesselsManual.flatten(), vesselsFound.flatten())
    TrueNeg, FalsePos, FalseNeg, TruePos = conf.ravel()

accuracy = (float (TruePos+TrueNeg) / float(TruePos + TrueNeg + FalsePos +
    FalseNeg))
```

```
sensitivity = (TruePos / float(TruePos + FalseNeg))
specificity = (TrueNeg / float(TrueNeg + FalsePos))
geoMean = (sensitivity * specificity)**(1.0/2)
return (accuracy, sensitivity, specificity, geoMean)
```

01_h.jpg

```
ValueError
                                          Traceback (most recent call last)
Input In [163], in <cell line: 1>()
      5 _, vesselsManual = cv2.threshold(vesselsManual,10,255,cv2.THRESH_BINARY
      6 , vesselsFound = cv2.threshold(vesselsFound, 10, 255, cv2.THRESH BINARY)
----> 8 results = summary(vesselsManual, vesselsFound)
      9 print(f"accuracy: {results[0]} sensitivity: {results[1]} specificity:

¬{results[2]} geoMean: {results[3]}")

     10 if i==4:
Input In [162], in summary(vesselsManual, vesselsFound)
      4 def summary(vesselsManual, vesselsFound):
            conf =
  --> 5
 confusion matrix(vesselsManual flatten(), vesselsFound flatten())
            TrueNeg, FalsePos, FalseNeg, TruePos = conf.ravel()
            accuracy = (float (TruePos+TrueNeg) / float(TruePos + TrueNeg +
 →FalsePos + FalseNeg))
File ~/.local/lib/python3.10/site-packages/sklearn/metrics/_classification.py:
 →307, in confusion_matrix(y_true, y_pred, labels, sample_weight, normalize)
    222 def confusion_matrix(
    223
            y_true, y_pred, *, labels=None, sample_weight=None, normalize=None
    224):
            """Compute confusion matrix to evaluate the accuracy of a_{\sqcup}
 ⇔classification.
    226
```

```
By definition a confusion matrix :math: `C` is such that :math: `C_{i __}
    227
 (...)
    305
            (0, 2, 1, 1)
            0.00
    306
            y_type, y_true, y_pred = _check_targets(y_true, y_pred)
--> 307
            if y_type not in ("binary", "multiclass"):
    308
                raise ValueError("%s is not supported" % y_type)
    309
File ~/.local/lib/python3.10/site-packages/sklearn/metrics/_classification.py:

⇔84, in _check_targets(y_true, y_pred)
     57 def _check_targets(y_true, y_pred):
            """Check that y_{true} and y_{pred} belong to the same classification.
 59
            This converts multiclass or binary types to a common shape, and \sqcup
 ⇔raises a
   (...)
     82
            y_pred : array or indicator matrix
     83
            check consistent_length(y_true, y_pred)
---> 84
            type true = type of target(y true)
     85
            type_pred = type_of_target(y_pred)
File ~/.local/lib/python3.10/site-packages/sklearn/utils/validation.py:332, in_
 ⇔check_consistent_length(*arrays)
    330 uniques = np.unique(lengths)
    331 if len(uniques) > 1:
--> 332
            raise ValueError(
    333
                "Found input variables with inconsistent numbers of samples: %r
    334
                % [int(1) for 1 in lengths]
    335
ValueError: Found input variables with inconsistent numbers of samples: [959040]
 ⇒2877120]
```

[]: