# exc\_01\_solution

# April 26, 2018

### 0.1 Exercise 01

## 1 Random Forests

```
In [2]: %matplotlib inline
    import numpy as np
    import h5py as h5
    import matplotlib.pyplot as plt
    import scipy.ndimage
    import skimage.color
    import skimage.feature
    import sklearn.ensemble
    from matplotlib.colors import LinearSegmentedColormap
    import ipywidgets as widgets

    from ipywidgets import interact, interactive, fixed
    import ipywidgets as widgets

1.0.1 1.1 Load the data

In [3]: images_02 = []
    labels_02 = []
```

```
images_10 = []
labels_10 = []
f_im_02 = h5.File('images_subject02.h5','r')
f_lb_02 = h5.File('labels_subject02.h5','r')
f_im_10 = h5.File('images_subject10.h5','r')
f_lb_10 = h5.File('labels_subject10.h5','r')
for key in f_im_02:
    images_02.append(f_im_02[key][...])
for key in f_lb_02:
    labels_02.append(f_lb_02[key][...])
for key in f_im_10:
    images_10.append(f_im_10[key][...])
for key in f_lb_10:
    labels_10.append(f_lb_10[key][...])
f_im_02.close()
f_lb_02.close()
```

```
f_im_10.close()
        f_lb_10.close()
In [4]: train_set = [images_02[0],images_02[1],images_02[2]]
        train_label = [labels_02[0], labels_02[1],labels_02[2]]
        test_set = images_10[0]
        test_label = labels_10[0]
  Visualize the training set
In [30]: colors = ['w','r','b','g','y']
         cmap_name = 'my_list'
         cm = LinearSegmentedColormap.from_list(
                 cmap_name, colors, N=5)
         fig,ax = plt.subplots(3,2)
         for i in range(3):
             masked_label = np.ma.array(train_label[i], mask=train_label[i]<0)</pre>
             ax[i,0].imshow(train_set[i],cmap=plt.cm.gray)
             ax[i,1].imshow(train_set[i],cmap=plt.cm.gray)
             ax[i,1].imshow(masked_label,cmap=cm)
                  0
                                                      0
               200
                                                    200
               400
                                                    400
                                                                    500
                  0
                                                      0
               200
                                                    200
```

500

500

### 1.0.2 1.2 Compute Featrues

400

200

400

0

0

400

200

400

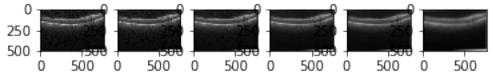
0

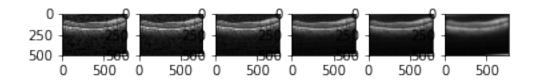
0

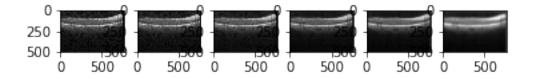
500

500

# for t,train in enumerate(train\_set): f = plt.figure() sigmas = [0.7, 1, 1.6, 3.5, 5, 10] for i, sigma in enumerate(sigmas): res = scipy.ndimage.gaussian\_filter(train, sigma=sigma) ax = f.add\_subplot(1, len(sigmas), i+1) ax.imshow(res, cmap='gray') all\_filters['train\_'+str(t)].append(res)

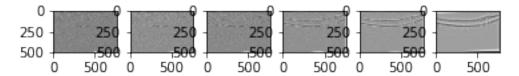


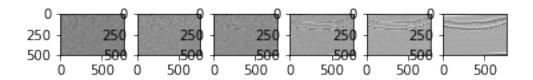


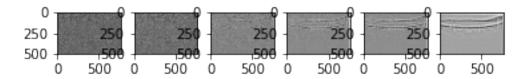


In [32]: for t,train in enumerate(train\_set):

```
f = plt.figure()
for i, sigma in enumerate(sigmas):
    res = scipy.ndimage.gaussian_laplace(train, sigma=sigma)
    ax = f.add_subplot(1, len(sigmas), i+1)
    ax.imshow(res, cmap='gray')
    all_filters['train_'+str(t)].append(res)
```

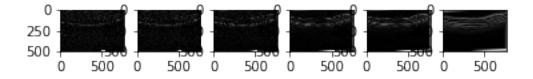


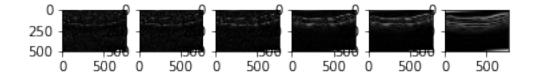


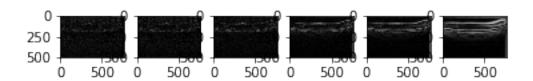


In [33]: for t,train in enumerate(train\_set):

```
f = plt.figure()
for i, sigma in enumerate(sigmas):
    res = scipy.ndimage.gaussian_gradient_magnitude(train, sigma=sigma)
    ax = f.add_subplot(1, len(sigmas), i+1)
    ax.imshow(res, cmap='gray')
    all_filters['train_'+str(t)].append(res)
```



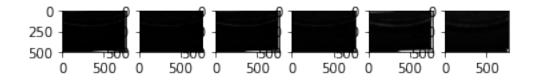




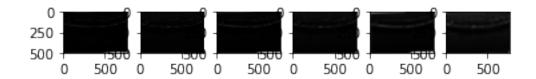
### In [34]: for t,train in enumerate(train\_set):

```
f = plt.figure()
for i, sigma in enumerate(sigmas):
    a,b = skimage.feature.structure_tensor_eigvals(*skimage.feature.structure_tensor(train
    ax0 = f.add_subplot(2, len(sigmas), 2*i+1)
    ax1 = f.add_subplot(2, len(sigmas), 2*i+2)
    ax0.imshow(a, cmap='gray')
    ax1.imshow(b, cmap='gray')
    all_filters['train_'+str(t)].append(a)
    all_filters['train_'+str(t)].append(b)
```

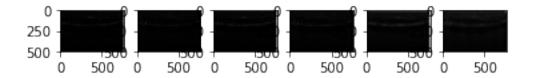












### In [35]: for t,train in enumerate(train\_set):

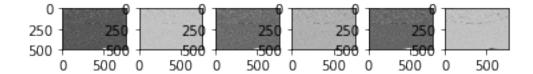
```
f = plt.figure()
for i, sigma in enumerate(sigmas):
```

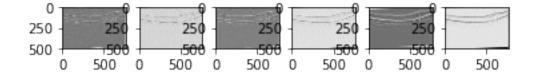
a,b = skimage.feature.hessian\_matrix\_eigvals(\*skimage.feature.hessian\_matrix(train, signal))

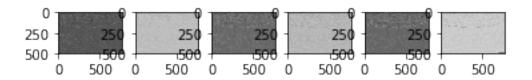
ax0 = f.add\_subplot(2, len(sigmas), 2\*i+1) ax1 = f.add\_subplot(2, len(sigmas), 2\*i+2) ax0.imshow(a, cmap='gray') ax1.imshow(b, cmap='gray')

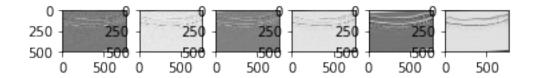
all\_filters['train\_'+str(t)].append(a)

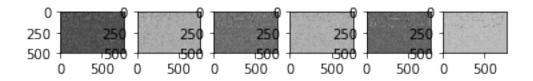
all\_filters['train\_'+str(t)].append(b)

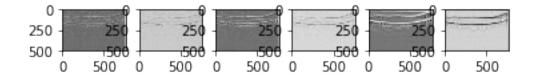




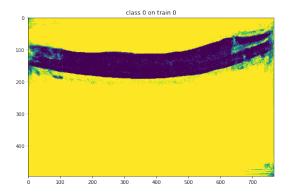


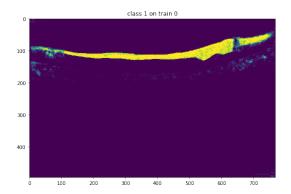


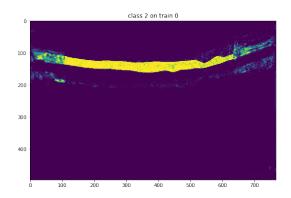


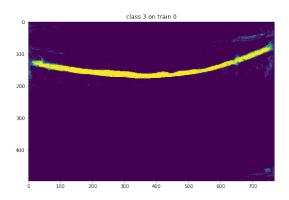


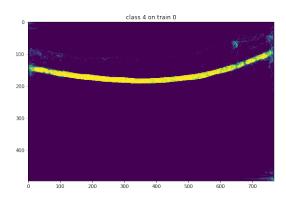
```
print(filters.shape)
             mask = np.logical_not(label < 0)</pre>
             train_X = filters[:,mask].T
             train_Y = label[np.logical_not(label<0)]</pre>
             print(train_X.shape)
             print(train_Y.shape)
             train_X_all.append(train_X)
             train_Y_all.append(train_Y)
         train_X = np.concatenate(train_X_all,axis=0)
         train_Y = np.concatenate(train_Y_all)
         print(train_X.shape)
         print(train_Y.shape)
(42, 496, 768)
(259904, 42)
(259904,)
(42, 496, 768)
(259904, 42)
(259904,)
(42, 496, 768)
(259904, 42)
(259904,)
(779712, 42)
(779712,)
1.0.3 1.3 Random Forest
In [13]: rf = sklearn.ensemble.RandomForestClassifier(n_estimators=10)
         rf.fit(train_X, train_Y)
Out[13]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                     max_depth=None, max_features='auto', max_leaf_nodes=None,
                     min_samples_leaf=1, min_samples_split=2,
                     min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
                     oob_score=False, random_state=None, verbose=0,
                     warm_start=False)
In [36]: def f(t):
             filters = filter_arrays[t]
             train = train_set[t]
             pred0 = rf.predict_proba(filters.reshape((filters.shape[0], -1)).T)
             pred0 = pred0.reshape(train.shape+(-1,))
             #with h5.File('predictions.h5', 'a') as save:
                  save.create_dataset('train_'+str(t), data=pred0)
             f = plt.figure()
             for i_class in range(pred0.shape[-1]):
                 ax = f.add_subplot(4, 2, i_class+1)
                 ax.imshow(pred0[:,:,i_class])
```

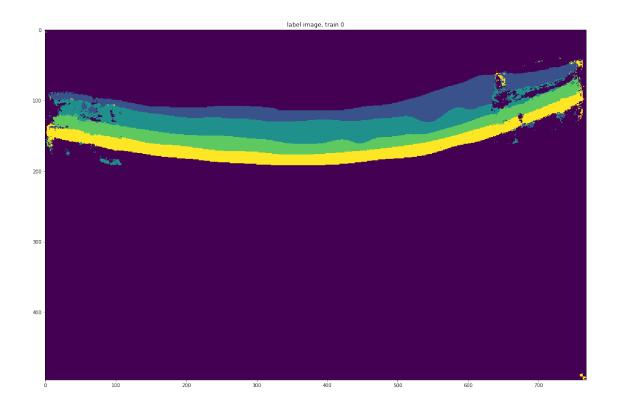








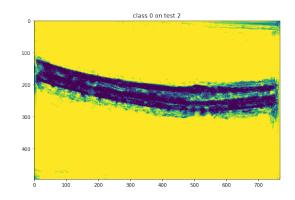


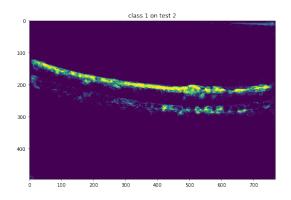


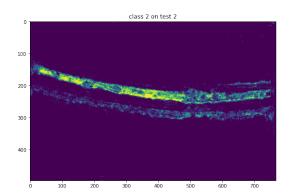
Predict labels on test set

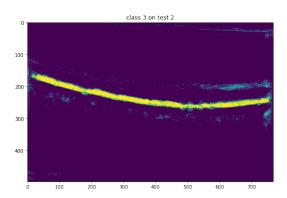
```
In [15]: test_filters = []
         for i, sigma in enumerate(sigmas):
             res = scipy.ndimage.gaussian_filter(test_set, sigma=sigma)
             test_filters.append(res)
         for i, sigma in enumerate(sigmas):
             res = scipy.ndimage.gaussian_laplace(test_set, sigma=sigma)
             test_filters.append(res)
         for i, sigma in enumerate(sigmas):
             res = scipy.ndimage.gaussian_gradient_magnitude(test_set, sigma=sigma)
             test_filters.append(res)
         for i, sigma in enumerate(sigmas):
             a,b = skimage.feature.structure_tensor_eigvals(*skimage.feature.structure_tensor(test_set,
             test_filters.append(a)
             test_filters.append(b)
         for i, sigma in enumerate(sigmas):
             a,b = skimage.feature.hessian_matrix_eigvals(*skimage.feature.hessian_matrix(test_set, signature.hessian_matrix)
             test_filters.append(a)
             test_filters.append(b)
In [16]: test_filters = np.array(test_filters)
         pred = rf.predict_proba(test_filters.reshape((test_filters.shape[0], -1)).T)
         pred = pred.reshape(test_set.shape+(-1,))
         with h5.File('predictions.h5', 'a') as save:
             if not 'test' in save:
                 save.create_dataset('test',data=pred)
```

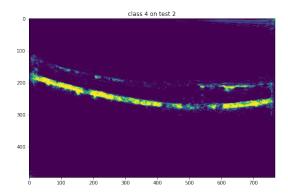
```
In [37]: f = plt.figure()
    for i_class in range(pred.shape[-1]):
        ax = f.add_subplot(4, 2, i_class+1)
        ax.imshow(pred[:,:,i_class])
        ax.set_title("class %i on test %i" %(i_class,t))
    f.set_size_inches(20,40)
```



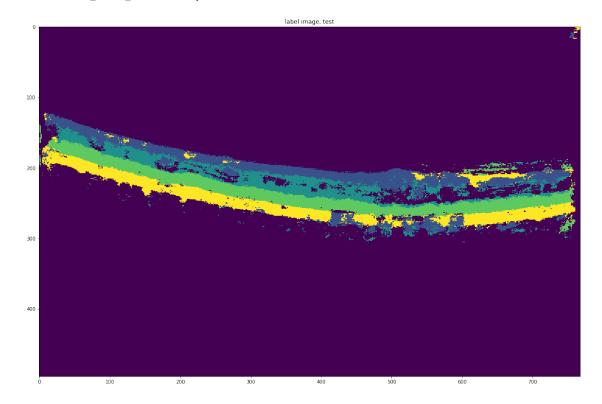




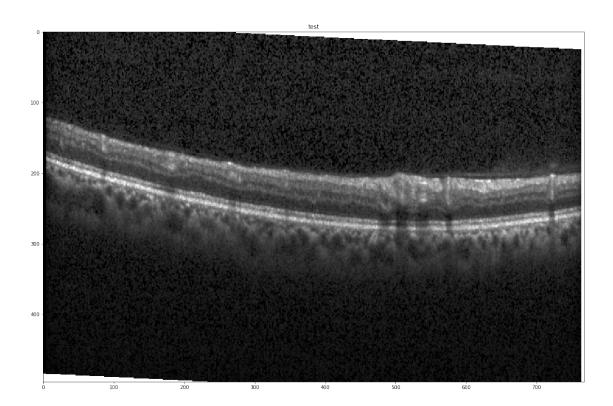




```
In [38]: pred_full = np.argmax(pred, axis=2)
    f = plt.figure()
    ax = f.add_subplot(111)
    ax.imshow(pred_full)
    ax.set_title("label image, test")
    f.set_size_inches(20,40)
```



```
In [39]: f = plt.figure()
    ax = f.add_subplot(111)
    ax.imshow(test_set,cmap=plt.cm.gray)
    ax.set_title("test")
    f.set_size_inches(20,40)
```



# 1.0.4 1.4 Acuracy Score

R = 0.796000F1 = 0.655000

Class 2:

```
In [20]: pred_comp = pred_full[np.logical_not(test_label < 0)]</pre>
                                    labels_comp = test_label[np.logical_not(test_label < 0)]</pre>
                                    print(pred_comp.shape)
                                    print(labels_comp.shape)
(259904,)
(259904,)
In [28]: for i_class in range(pred.shape[-1]):
                                                    TP = np.sum(np.equal(pred_comp[np.where(pred_comp==i_class)],labels_comp[np.where(pred_com
                                                    FP = np.sum(np.not_equal(pred_comp[np.where(pred_comp==i_class)],labels_comp[np.where(pred_comp==i_class)]
                                                    FN = np.sum(np.not_equal(pred_comp[np.where(labels_comp==i_class)],labels_comp[np.where(labels_comp==i_class)]
                                                   P = TP / (TP + FP)
                                                   R = TP / (TP + FN)
                                                   F_1 = 2*P*R/(P+R)
                                                     print("Class \%i: \n\tP = \%f\n\tR = \%f \n\tF1 = \%f"\%(i\_class,np.round\_(P,3),np.round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,3),round\_(R,
Class 0:
                               P = 0.980000
                               R = 0.986000
                               F1 = 0.983000
Class 1:
                               P = 0.557000
```

```
P = 0.839000
                                                   R = 0.563000
                                                   F1 = 0.674000
Class 3:
                                                   P = 0.893000
                                                   R = 0.920000
                                                   F1 = 0.906000
Class 4:
                                                    P = 0.778000
                                                   R = 0.728000
                                                   F1 = 0.752000
In [29]: for i_class in range(pred.shape[-1]):
                                                                                    TP = np.sum(np.equal(pred_comp[np.where(pred_comp==i_class)],labels_comp[np.where(pred_comp
                                                                                    FP = np.sum(np.not_equal(pred_comp[np.where(pred_comp==i_class)],labels_comp[np.where(pred_comp==i_class)]
                                                                                     \texttt{FN} = \texttt{np.sum(np.not\_equal(pred\_comp[np.where(labels\_comp==i\_class)],labels\_comp[np.where(labels\_comp==i\_class)],labels\_comp[np.where(labels\_comp==i\_class)], } \\ \texttt{abels\_comp[np.where(labels\_comp==i\_class)],labels\_comp[np.where(labels\_comp==i\_class)], } \\ \texttt{basels\_comp[np.where(labels\_comp==i\_class)],labels\_comp[np.where(labels\_comp==i\_class)], } \\ \texttt{basels\_comp[np.where(labels\_comp==i\_class)],labels\_comp==i\_class), } \\ \texttt{basels\_comp} \\ \texttt{
                                                                                    P = TP / (TP + FP)
                                                                                    R = TP / (TP + FN)
                                                                                   F_1 = P*R/(P+R)
                                                                                    print("Class \%i: \n\tP = \%f \n\tF1 = \%f"\%(i_class,np.round_(P,3),np.round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R,3),round_(R
Class 0:
                                                    P = 0.980000
                                                   R = 0.986000
                                                   F1 = 0.492000
Class 1:
                                                    P = 0.557000
                                                   R = 0.796000
                                                    F1 = 0.328000
Class 2:
                                                    P = 0.839000
                                                   R = 0.563000
                                                   F1 = 0.337000
Class 3:
                                                    P = 0.893000
                                                    R = 0.920000
                                                   F1 = 0.453000
Class 4:
                                                   P = 0.778000
                                                   R = 0.728000
                                                   F1 = 0.376000
In []:
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