Problem4_DrewHill

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CE263N - Problem 4
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In [1]: import numpy as np
import pickle
from sklearn import svm
from mpl_toolkits.mplot3d import Axes3D

from matplotlib import pyplot as plt
from pylab import *

# run matplotlib inline
%matplotlib inline
```

Step 1 Run clicker.py at appropriate path (in current directory) and identify cars (left click) and non-cars (right-click). Close window, and np arrays automatically saved to the current directory. Because the "clicker.py" script flattens the information captured (4 points each [transparent, red, blue, green] for each pixel in a 20x20 pixel box centered at the point we clicked– total of 1600 points per click), we need to unflatten it. This applies to every X-y pair in the numpy array (note the shape of each X array is 1600).

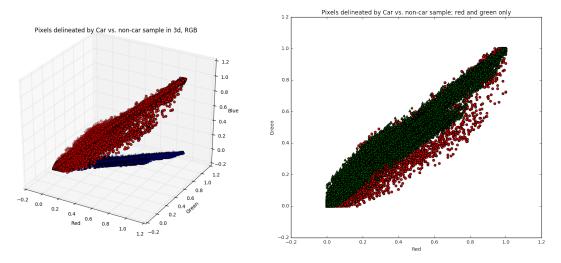
Turns out some only were assigned 1360 or even 0 RGBT sets. So, adjust "unflattening" accordingly:

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In [4]: # unflatten / reshape each "X_trn" element
        # take random sample of each sample (Virtual Sample)
        X_{trn} = np.empty((0,4), float)
        Y_{trn} = np.empty((0,), float)
        for i in xrange(X_trn_og.shape[0]):
            # produce [R,G,B,T,car,i] * [pixel number per sample] array for each sa
                # determine length of new array
            row_num_newarray = X_trn_og[i].shape[0]/ 4
            a = X_tn_oq[i]
            a = a.reshape(row_num_newarray, 4)
                # remove transparency element (4th element)
            b = np.delete(a, 3, 1)
                # create new element for "car" status
            car = np.array([Y_trn_og[i]] * row_num_newarray)
                # append
            c = np.insert(b, 3, car, 1)
                  # random sample of 25% of each sample
                      # create index
                  idx = np.random.randint(10, size= row num newarray/4)
                  e = d i dx,:
            # append to final array set
            X_{trn} = np.append(X_{trn}, c, 0)
                                                   # stack onto X_trn array
            Y_trn = np.append(Y_trn, car, 0)
In [5]: ## Plot
        def plot_RGB_3d():
            X_{cars} = X_{trn}[X_{trn}[:,3] == 1]
            X_{notcars} = X_{trn}[X_{trn}[:,3] == 0]
            fig = plt.figure(figsize=(20,8))
            ax = fig.add_subplot(121, projection='3d')
            ax.scatter(X_cars[:,0], X_cars[:,1], X_cars[:,1], c='r', marker='o')
            ax.scatter(X_notcars[:,0], X_notcars[:,1], X_notcars[:,3], c='b', marke
            ax.set xlabel('Red')
            ax.set_ylabel('Green')
            ax.set zlabel('Blue')
            ax.set_title("Pixels delineated by Car vs. non-car sample in 3d, RGB")
            # Project data to X/Y plane - RG
            ax2d = fig.add\_subplot(122)
            ax2d.scatter(X_cars[:,0], X_cars[:,1], c='r', marker='o')
            ax2d.scatter(X_notcars[:,0], X_notcars[:,1], c='g', marker='^')
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ax2d.set_xlabel('Red')
ax2d.set_ylabel('Green')
ax2d.set_title("Pixels delineated by Car vs. non-car sample; red and gr
  # Project data to X/Y plane - RB
 ax3d = fig.add_subplot(122)
 ax3d.scatter(X_cars[:,0], X_cars[:,2], c='r', marker='o')
 ax3d.scatter(X_notcars[:,0], X_notcars[:,2], c='b', marker='^')
 ax3d.set_xlabel('Red')
 ax3d.set_ylabel('Blue')
 ax3d.set_title("Pixels delineated by Car vs. non-car sample; red and
 # Project data to X/Y plane - BG
 ax2d = fig.add\_subplot(122)
 ax2d.scatter(X_cars[:,2], X_cars[:,1], c='b', marker='o')
 ax2d.scatter(X_notcars[:,2], X_notcars[:,1], c='g', marker='^')
 ax2d.set xlabel('Blue')
 ax2d.set_ylabel('Green')
 ax2d.set_title("Pixels delineated by Car vs. non-car sample; blue and
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plt.show()

In [6]: plot_RGB_3d()



Step 2 Create a support vector machine classifier, using different classifiers.

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In [18]: # kernel = 'rbf'
         # clf_rbf = svm.SVC(kernel = kernel, C=100.0)
         # clf_rbf.fit(X_trn[:,:-1], Y_trn)
In [10]: kernel = 'poly'
         clf p = svm.SVC(kernel = kernel, C=100.0)
         clf_p.fit(X_trn[:,:-1], Y_trn)
Out[10]: SVC(C=100.0, cache_size=200, class_weight=None, coef0=0.0,
           decision_function_shape=None, degree=3, gamma='auto', kernel='poly',
           max_iter=-1, probability=False, random_state=None, shrinking=True,
           tol=0.001, verbose=False)
  Step 3 Pickle my classifier.
In [ ]: with open('classifier_DrewHill_linear_C10000.pickle','wb') as f:
            pickle.dump(clf l, f)
  Step 4 My Submission (as tested on the 'preview' data)
In [ ]: import numpy as np
        import pickle
        from matplotlib import pyplot as plt
        from pylab import *
        ### this will be replaced with the real test image ###
        im_test = plt.imread('parking_test_preview.png')
        ###
        # This function MUST take locations (loc) and an image (im)
        # as input parameters and return the feature vector
        def my_feature_vector(loc, im, size = 10):
          w = size
          # a patch of the size w cenetered at loc is extracted as a feature vector
          patch = im[loc[1]-w:loc[1]+w, loc[0]-w:loc[0]+w]
          p = np.array(patch).flatten()
          return p
        ## 10 preview test locations
        ### these will be replaced with the real set of 100 test locations ###
        test_locs_labs = np.load('test_locations_and_labels_preview.np')
        test_locs = test_locs_labs[:,0:2]
        test_labels = test_locs_labs[:,2]
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X_{test} = []
for loc in test_locs:
  X_test.append( my_feature_vector(loc, im_test) )
######
# Feature selection
## load classifier
my_classifier = pickle.load(open('classifier_DrewHill_linear_truncated.pickle.pickle.goal)
def drew_predict(X):
    X_i = \text{np.empty}((0,3), \text{float})
    for i in xrange(np.array(X).shape[0]):
        row_num_newarray = np.array(X).shape[0]/ 4
        a = np.array(X)
        a = a.reshape(row_num_newarray, 4)
        b = np.delete(a, 3, 1)
         car = np.array([test_locs_labs[i,2]] * row_num_newarray)
         c = np.insert(b, 3, car, 1)
        x_i = np.empty((0,), int)
        for i in xrange(b.shape[0]):
            boop = my_classifier.predict(b[i])
            x_i = np.append(x_i, boop, 0)
        predicted = int(round(np.mean(x_i)))
        return predicted
    return predicted
## perform classification
score = 0
for i, xtest in enumerate(X_test):
  predicted = drew_predict(xtest)
  if (test_labels[i] == 1.0) & (predicted == 1.0):
     score = score + 2
  if (test_labels[i] == 1.0)&(predicted == 0.0):
     score = score - 0.5
  if (test_labels[i] == 0.0)&(predicted == 1.0):
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score = score - 0.5

if (test_labels[i] == 0.0)&(predicted == 0.0):
    score = score + 0.25

print test_labels[i], predicted, score

print 'You final Score is: %.2f' % score
```