#### Lab 6: SVMs on an Extended MNIST

In addition to the concepts in the MNIST demo (mnist svm.ipynb), you will learn:

- Use the skimage module for some basic pre-processing of images in machine learning
- · Run and test an SVM classifier on a dataset you have created
- · Perform error handling in python

In the MNIST demo (mnist\_svm.ipynb), we saw how SVMs can be used for the classic MNIST problem of digit recognition. In this lab, we are going to extend the MNIST dataset by adding a number of non-digit letters and see if the classifier can distinguish the digits from the non-digits. All non-digits will be lumped as a single 11-th class. In image processing, this is called a 'detection' as opposed to 'classification' problem. Detection is vital in OCR and related problems since the non useful characters must be rejected. For this lab we will create a very simple version of this problem.

# Loading the MNIST data

We first import the standard modules

```
In [1]: import numpy as np
   import matplotlib.pyplot as plt
   %matplotlib inline
   from sklearn import linear_model, preprocessing,datasets
   from sklearn.datasets import fetch_mldata
```

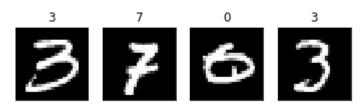
Next, fetch the digits with fetch\_mldata command as shown in the demo. Save the digits data matrix and labels to variables Xdig and ydig. Also, recall that the pixel values in Xdig are between 0 and 255. Create a scaled version of Xdig called Xdigs where the components are between -1 and 1.

```
In [3]: # TODO
# Xdig = ... Load MNIST data
Xdig = mnist.data
# ydig = ...
ydig = mnist.target
# Xdigs = ... Rescale MNIST data
print(Xdig.shape)
Xdigs = Xdig/255.0*2-1
#for i in range(Xdigs.shape[0]):
# Xdigs[i] = Xdigs[i].astype(float)/255*2-1
print(np.max(Xdigs))
(70000, 784)
```

(70000, 784) 1.0

Create a function plt\_digit that plots the digits. You can use the code from the demo. Test the function by plotting four random digits. Use the plt.title command to print the numeric label in ydig above each digit.

```
In [4]: # TODO: Select four random digits and plot them using the subplot command
        def plt_digit(x):
            nrow = 28
            ncol = 28
            xsq = x.reshape((nrow,ncol))
            plt.imshow(xsq, cmap='Greys_r')
            plt.xticks([])
            plt.yticks([])
        # Convert data to a matrix
        X = mnist.data
        y = mnist.target
        # Select random digits
        nplt = 4
        nsamp = X.shape[0]
        Iperm = np.random.permutation(nsamp)
        # Plot the images using the subplot command
        for i in range(nplt):
            ind = Iperm[i]
            plt.subplot(1,nplt,i+1)
            plt digit(X[ind,:])
            plt.title(ydig[ind].astype(int))
```



#### **Exception Handling**

In the routines we will develop below, we will need to handle error conditions, called exceptions. A very nice description of how to perform exception handling in python is given in

https://docs.python.org/3/tutorial/errors.html (https://docs.python.org/3/tutorial/errors.html)

As described there, errors are described by a class that derives from a base class Exception. When the error occurs, the program raises the exception with the raise command. The calling function can catch the exception with the try ... except control flow. We will define our exception as follows which has an optional string argument.

```
In [5]: class ImgException(Exception):
    def __init__(self, msg='No msg'):
        self.msg = msg
```

Exceptions are used as follows: First, when there is an error in some function, you raise the exception as follows:

```
foo():
    ...
    if (error):
        raise ImgException("File not found")

# Code that will not execute if the error condition occured
```

The function that calls foo() can catch the error using the following syntax:

```
try:
    foo()

# Continue processing in case when there was no exception
....

except ImgException as e:
    print("foo() didn't work")
    print("Error msg = %s" % e.msg)
```

## **Get Non-Digit Characters**

We will now build a set of non-digit characters. As a simple source, we will get hand-written lowercase letters 'a' to 'z' and process them with the skimage package. The skimage module is a very powerful package that has a similar interface as OpenCV. We first import the relevant modules.

```
In [6]: import matplotlib.image as mpimg
   import skimage.io
   from skimage.filters import threshold_otsu
   from skimage.segmentation import clear_border
   from skimage.measure import label, regionprops
   from skimage.morphology import closing, square
   from skimage.color import label2rgb
   from skimage.transform import resize
   import matplotlib.patches as mpatches
   from skimage import data
   import skimage
```

We can get a set of character images from a very nice website

http://www.ee.surrey.ac.uk/CVSSP/demos/chars74k/ (http://www.ee.surrey.ac.uk/CVSSP/demos/chars74k/)

Go to this website, and download the file EnglishHnd.tgz. After you untar this file, there are a large number of .png files in the directory:

EnglishHnd\English\Hnd\Img

Each directory has about 55 samples of hand-written letters and numbers. After you have downloaded this file, complete the function load\_img to load an image from a character and sample index.

Alternatively, the files are available on Google Drive:

https://drive.google.com/file/d/0BxOz-SM9a1h4UksxSXBjQ0dabUk/view?usp=sharing (https://drive.google.com/file/d/0BxOz-SM9a1h4UksxSXBjQ0dabUk/view?usp=sharing)

You can download and unzip the file.

The code at the end will test the function to see if it working correctly. For one sample, it should print the image and a second it should say the file was not found.

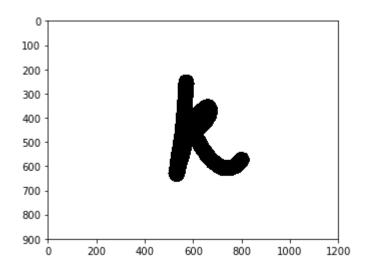
```
In [7]: import os.path
        def load_img(char_ind, samp_ind):
            Returns the image from the dataset given a character and sample index.
            If the file doesn't exist, it raises an Exception with the filename.
            # TODO: Set the file name based on char ind and samp ind
            # fname = ...
            fname = '../English/Hnd/Img/Sample0'+str(int(char_ind/10))+str(int(char_ind%1
             print(fname)
            # TODO: Use the os.path.isfile command to check if the file exists.
            # If not raise an ImgException with the message "[fname] not found"
            try:
                 if(os.path.isfile(fname)!=True):
                     raise ImgException
            except ImgException:
                 print(fname+' not found')
                raise
            # TODO: Use the skimage.io.imread() command to read the png file and return
            \# img = ...
            else:
                 img = skimage.io.imread(fname)
            return img
```

Test the load\_img function. This should:

- Plot the image in Sample047\img047-006.png
- Say that the Sample047\img047-070.png is not found.

```
In [8]: char_ind = 47
    samp_inds = [6,70]
    for samp_ind in samp_inds:
        try:
        img = load_img(char_ind=char_ind, samp_ind=samp_ind)
        print("Char = %d samp=%d" % (char_ind, samp_ind))
        plt.imshow(img)
    except ImgException as e:
        print(e.msg)
```

Char = 47 samp=6
../English/Hnd/Img/Sample047/img047-070.png not found
No msg



The images in the sample directory have very high resolution. Complete the following method to find the image and place it in a 28 x 28 box. You can look at this very nice demo of the skimage methods here:

http://scikit-image.org/docs/dev/auto\_examples/segmentation/plot\_label.html (http://scikit-image.org/docs/dev/auto\_examples/segmentation/plot\_label.html)

The code is somewhat complex, so I have provided some of the steps, esp. for the thresholding.

```
In [9]: def mnist resize(img):
            Extracts a character from the image, and places in a 28x28 image to match the
            Returns:
            img1: MNIST formatted 28 x 28 size image with the character from img
                   A bounding box indicating the locations where the character was found
            # Image sizes (fixed for now). To match the MNIST data, the image
            # will be first resized to 20 x 20. Then, the image will be placed in center
            # offet by 4 on each side.
            nx img = 20
            ny img = 20
            nx box = 28
            ny box = 28
            offx = 4
            offy = 4
            # TODO: Convert the image to gray scale using the skimage.color.rgb2gray metl
            \# bw = ...
            bw = skimage.color.rgb2gray(img)
            # Threshold the image using OTSU threshold
            thresh = threshold otsu(bw)
            bw = closing(bw < thresh, square(3)).astype(int)</pre>
            # Get the regions in the image.
            # This creates a list of regions in the image where the digit possibly is.
            regions = regionprops(bw)
            # TODO: Find region with the largest area. You can get the region area from
            \# region_max = ...
            area max = 0
            region_max = regions[0]
            for region in regions:
                 if(area max < region.area):</pre>
                     area max = region.area
                     region max = region
            # Raise an ImgException if no region with area >= 100 was found
            if (area max < 100):
                 raise ImgException("No image found")
            # Get the bounding box of the character from region_max.bbox
            minr, minc, maxr, maxc = region_max.bbox
            box = [minr,minc,maxr,maxc]
            # TODO: Crop the image in bw to the bounding box
            # bw crop = bw[...]
            bw crop = bw[minr:maxr,minc:maxr]
            # TODO: Resize the cropped image to a 20x20 using the resize command.
            # You will need to use the mode = 'constant' option
```

```
# bw_resize = ...
bw_resize = skimage.transform.resize(bw*1.0,(nx_img,ny_img), mode = 'constant

# TODO: Threshold back to a 0-1 image by comparing the pixels to their mean
bw_resize_mean = np.mean(bw_resize)
bw_resize[bw_resize >= bw_resize_mean] = 1
bw_resize[bw_resize < bw_resize_mean] = 0

# TODO: Place extracted 20 x 20 image in larger image 28 x 28
# img1 = ...
img1 = resize(bw_resize, (28, 28), mode='constant')
img1 = np.zeros((28, 28), dtype=np.uint8)
img1[offx:offx+nx_img,offy:offy+ny_img] = bw_resize

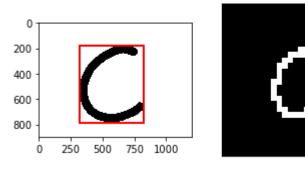
return img1, box</pre>
```

Now test the mnist\_resize program by completing the following code. Create two subplots:

- subplot(1,2,1): The original image with the bounding box for the character that was found in the image.
- subplot(1,2,2): The MNIST resized image.

```
In [10]: # Load an image
         img = load img(13,9)
         print(img.shape)
         try:
             # Resize the image
             # img1, box = mnist_resize(img)
             img1,box = mnist_resize(img)
             # TODO: Plot the original image, ima, along with a red box around the capture
             # Use the mpatches.Rectangle and ax.add_patch methods to construct the rectangle
             plt.imshow(img)
             ax = plt.subplot(121)
             plt.imshow(img)
         #
              plt.grid()
              plt.xticks([])
         #
              plt.yticks([])
              box = [minr,minc,maxr,maxc]
             rect = mpatches.Rectangle((box[1], box[0]), box[3] - box[1], box[2] - box[0],
                                            fill=False, edgecolor='red', linewidth=2)
             ax.add patch(rect)
             # TODO: Plot the resized 28 x 28 image, img1. You can use the plt_digit(img.
             plt.subplot(122)
             plt_digit(img1)
         except ImgException as e:
             print(e.msg)
```

(900, 1200, 3)



Now, run the command nlet=1000 times to get 1000 letter images. In each iteration, select a random image from a lowercase letter and add it to a matrix Xlet.

```
In [11]: # Dimensions
         nlet = 1000
         nrow = 28
         ncol = 28
         npix = nrow*ncol
         Xlet = np.zeros((nlet, npix))
         i = 0
         while i < nlet:</pre>
             # TODO: Generate a random character and sample
             # char ind = random number corresponding to a Lowercase Letter except 'O' and
             # samp ind = random number from 0 to 49
             char ind = np.random.randint(26)
             while char ind == 9 or char ind == 15:
                  char ind = np.random.randint(26)
             char ind += 10
             samp ind = np.random.randint(50)
              print(char_ind)
             try:
                  # TODO: Load the image with load_img function
                  \# img = ...
                  img = load_img(char_ind,samp_ind)
                  # TODO: Reize the image with mnist resize function
                  # imq1, box = \dots
                  img1,box = mnist_resize(img)
                 # TODO: Store the image in a row of XLet[i,:] and increment i
                 Xlet[i,:] = np.resize(img1,(1,28*28))
                  i += 1
                 # Print progress
                  if (i % 50 == 0):
                      print ('images captured = {0:d}'.format(i))
             except ImgException:
                  # Skip if image loading or resizing failed
                  pass
```

```
../English/Hnd/Img/Sample015/img015-000.png not found
../English/Hnd/Img/Sample028/img028-000.png not found
images captured = 50
../English/Hnd/Img/Sample016/img016-000.png not found
images captured = 100
images captured = 150
images captured = 200
../English/Hnd/Img/Sample021/img021-000.png not found
../English/Hnd/Img/Sample033/img033-000.png not found
../English/Hnd/Img/Sample028/img028-000.png not found
images captured = 250
images captured = 300
images captured = 350
../English/Hnd/Img/Sample026/img026-000.png not found
../English/Hnd/Img/Sample021/img021-000.png not found
```

```
images captured = 400
images captured = 450
../English/Hnd/Img/Sample035/img035-000.png not found
images captured = 500
images captured = 550
../English/Hnd/Img/Sample030/img030-000.png not found
../English/Hnd/Img/Sample014/img014-000.png not found
images captured = 600
../English/Hnd/Img/Sample010/img010-000.png not found
images captured = 650
images captured = 700
../English/Hnd/Img/Sample027/img027-000.png not found
images captured = 750
../English/Hnd/Img/Sample024/img024-000.png not found
../English/Hnd/Img/Sample011/img011-000.png not found
../English/Hnd/Img/Sample022/img022-000.png not found
images captured = 800
../English/Hnd/Img/Sample033/img033-000.png not found
images captured = 850
../English/Hnd/Img/Sample026/img026-000.png not found
../English/Hnd/Img/Sample018/img018-000.png not found
images captured = 900
images captured = 950
../English/Hnd/Img/Sample024/img024-000.png not found
../English/Hnd/Img/Sample033/img033-000.png not found
images captured = 1000
```

Since this takes a long time to generate, save the matrix X1et to a file X1et.p using the pickle.dump command.

```
In [12]: import pickle

# TODO
pickle.dump( Xlet, open( "Xlet.p", "wb" ) )
```

Reload the data Xlet from the file Xlet.p

```
In [13]: # TODO
Xlet = pickle.load( open( "Xlet.p", "rb" ))
#plt.imshow(Xlet[0])
```

## **Create Extended Training Data**

Now, create an extended data set by combining ndig=5000 randomly selected digit samples and nlet=1000 letters.

- Select ndig=5000 random samples from Xdigs and their labels in ydig.
- Rescale the letters Xlet to a new matrix Xlets = 2\*Xlet-1 to make the pixel values go from -1 to 1.

- Use the np.vstack command to create a 6000 element alpha-numeric data set X
- Create a corresponding label vector y where all the non-digit characters are labeled with a non-digit label, letter lbl=10.

```
In [14]:
          # TODO
          nsamp = Xdigs.shape[0]
          Iperm = np.random.permutation(nsamp)
          # X = \dots Array with 6000 characters (5000 digits + 1000 letters)
          ntr = 5000
          Xtr = Xdigs[Iperm[:ntr],:]
          Xlet = 2*Xlet-1
          #plt digit(Xdigs[0].reshape(28,28))
          Xex = np.vstack((Xtr,Xlet))
          print(Xex.shape)
          \#print(Xex[0].reshape(28,28))
          # y = ... Array with 6000 labels (0-9 for the digits, 10 = non-digit)
          ytr = ydig[Iperm[:ntr]]
          \#yex = np.vstack([y,np.zeros(1000)])
          z = np.zeros(1000)+10
          yex = np.concatenate((ytr,z))
          print(yex[4950:5050])
          (6000, 784)
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                        6.
                                                               2.
             4.
                   6.
                        4.
                              4.
                                    2.
                                         2.
                                               6.
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                                                          0.
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                                                                                      1.
                                                                                           2.
             5.
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                                                               9.
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```

# Run the SVM classifier

First create the SVM classifer. Use an "rbf" classifier with C=2.8 and gamma=.0073. Not sure if these are the best parameters, you could try to search for better ones.

```
In [15]: from sklearn import svm

# TODO: Create a classifier: a support vector classifier
# svc = ...
svc = svm.SVC(probability=False, kernel="rbf", C=2.8, gamma=.0073, verbose=10)
```

Get 5000 training samples Xtr, ytr and 1000 test samples Xts, yts. Remember to randomly select them.

```
In [17]: print(np.min(Xtr[0]))
```

-1.0

Use the svc.fit command to fit on the training data. This may take a few minutes

```
In [18]: # TODO
svc.fit(Xtr,ytr)

[LibSVM]
```

Measure the accuracy on the test samples. You should get about 96% accuracy. You can get better by using more training samples, but it will just take longer to run.

```
In [19]: # TODO
    yhat_ts = svc.predict(Xts)
    acc = np.mean(yhat_ts == yts)
    print('Accuaracy = {0:f}'.format(acc))
```

Accuaracy = 0.963000

Print the normalized confusion matrix

```
In [20]:
           # TODO
           from sklearn.metrics import confusion matrix
           C = np.zeros((11,11))
           C += confusion matrix(yts, yhat ts)
           print(C)
                82.
           [[
                         0.
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                                               1.
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                                        0.
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                                                      1.
                                                              0.
                                                                     0.
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                                                                                    0.
                                                                                         174.]]
```

# Plotting some error samples

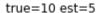
We now plot some errors. Plot up to four images where yhat == 10 but yts != 10. That is, the true image was a digit, but the classifier classified it as a non-digit. Note there may be less than four such errors (when I ran it I got only three such errors). In that case, just plot only the errors you got. If there are no errors, print "No such error found"

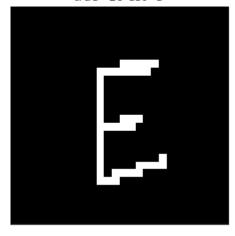
```
In [21]:
         # TODO
          Ierr1 = np.where((yhat_ts != yts))[0]
          Ierr = np.where(yhat_ts[Ierr1] == 10)
          Ierr = Ierr1[Ierr]
          \#Ierr = np.array([])
          print(Ierr.shape)
          if Ierr.shape[0]<4:</pre>
              nplt = Ierr.shape[0]
          else:
              nplt = 4
          if Ierr.shape[0]==0:
              print("No such error found")
          else:
              plt.figure(figsize=(10, 4))
              for i in range(nplt):
                  plt.subplot(1,nplt,i+1)
                  ind = Ierr[i]
                  plt digit(Xts[ind,:])
                  title = 'true={0:d} est={1:d}'.format(yts[ind].astype(int), yhat_ts[ind].
                  plt.title(title)
```

(0,) No such error found

Now plot up to four images where yhat != 10, but yts == 10. That is, the image was a non-digit, but the classifier thought it was an image. I happened to get no such images. If you find no such examples, print "No such error found".

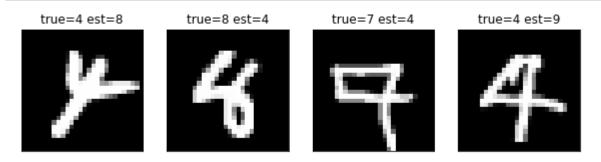
```
In [22]: # TODO
         Ierr1 = np.where((yhat_ts != yts))[0]
         Ierr = np.where(yts[Ierr1] == 10)
         Ierr = Ierr1[Ierr]
         if Ierr.shape[0]<4:</pre>
              nplt = Ierr.shape[0]
         else:
              nplt = 4
         if Ierr.shape[0]==0:
              print("No such error found")
         else:
              plt.figure(figsize=(10, 4))
              for i in range(nplt):
                  plt.subplot(1,nplt,i+1)
                  ind = Ierr[i]
                  plt_digit(Xts[ind,:])
                  title = 'true={0:d} est={1:d}'.format(yts[ind].astype(int), yhat_ts[ind].
                  plt.title(title)
```





Finally, plot up to four images where yts != yhat and both yts < 10 and yhat < 10.

```
In [23]: # TODO
         Ierr1 = np.where((yhat_ts != yts))[0]
         Ierr = np.where(yts[Ierr1] != 10)
         Ierr = Ierr1[Ierr]
         Ierr = np.where(yhat_ts[Ierr1] != 10)
         Ierr = Ierr1[Ierr]
         if Ierr.shape[0]<4:</pre>
              nplt = Ierr.shape[0]
         else:
              nplt = 4
         if Ierr.shape[0]==0:
              print("No such error found")
         else:
              plt.figure(figsize=(10, 4))
              for i in range(nplt):
                  plt.subplot(1,nplt,i+1)
                  ind = Ierr[i]
                  plt_digit(Xts[ind,:])
                  title = 'true={0:d} est={1:d}'.format(yts[ind].astype(int), yhat_ts[ind].
                  plt.title(title)
```



In [ ]: