## University of San Francisco

# Physics 301 – Computational Physics Fall, 2015

#### HW08

Due: Thursday Dec 3, at 8 PM

Write a program and name it

For the images of the 26 letters in the English alphabet (provided), please do the following.

I. <u>Image Preparation</u>

For each image,

- 1. Display the image (pyplot.imshow()) and, by index slicing, crop the image remove as much of the empty area around each letter as possible and remove a narrow band that sometimes appears near the right edge of the image.
- 2. Resample the images onto a coarser grid so that each image now has 16 × 16 pixels. You can use scipy.interpolate.interp2d() to do this. We will cover the usage of this function on Tuesday, Dec 1. If you want to get started sooner than that, feel free to consult

http://docs.scipy.org/doc/scipy/reference/generated/scipy.interpolate.interp2d.html

3. Subtract from each image its mean.

The following function should take an image file, let\_file, and perform the above three tasks:

The keyword arguments y\_lo, y\_hi, x\_lo, x\_hi contain the default values of the indices for cropping the image. The argument edge\_pix is specified to remove the a narrow band that sometimes appears to the right of edge\_pix (mentioned in 1. above) — the value of this argument varies depending on the image. If plot\_let is True, plot the cleaned-up 16 × 16 image.

The function should return  $let_im$ ,  $let_im_flat$ . The first is a cleaned-up  $16 \times 16$  image of a letter and the second is that image reshaped into a 1d-array of 256 elements (a "flattened" image).

## II. Data array construction

Construct a 2d-array that is  $26 \times 256$  and consists of the image data for the 26 letters; call this array X.

#### III. Principal Component Analysis

Perform PCA on X. Call the PCA-projected array Xproj. Decide on how many PCA components are necessary for the reconstructed images for *all* 26 letters to be recognizable by eye.

First write a function to perform PCA on X:

```
alphabet_pca(X, n_comp = 5)
```

n comp specifies the number of PCA components you want. It should return

```
pca, Xproj, pca_comps
```

where

pca: an object of the class sklearn.decomposition.PCA,

Xproj: contains the PCA projections (coefficients) in the form of an array with dimension (26, n\_comps),

pca comps: PCA components (eigenimages).

Then write the following function that can be used to display the eigenimages and the PCA constructed image of a single letter, which is specified by let\_idx (e.g., let\_idx = 0 corresponds to the letter "A"; let\_idx = 1, the letter "B"; etc.):

```
show pca im(Xproj, pca comps, dim = 16, let idx = 0)
```

## IV. Overall Structure

You should specific let\_idx and n\_comp at command line by using argparse in the program, and the program should take the following overall structure:

```
,,,
Docstring
, , ,
# imports
import ...
import ...
import ...
# function definitions in alphabetic order
def alphabet pca(X, n comp = 5):
     . . .
def (...):
def (...):
     . . .
# ******** Main Program ***********
# Input arguments
parser = argparse.ArgumentParser()
# Prepare letter images
A, A_flat = make_let_im('letterA.png', edge_pix = 135)
B, B_flat = make_let_im('letterB.png', edge_pix = 130)
```

```
.
.
.
Z, Z_flat = make_let_im('letterZ.png', edge_pix = 115)

# Construct data array for all 26 letters, X
...

# Perform PCA
alfbet_pca, Xproj, pca_comps = alphabet_pca(X, n_comp = n_comp)

# plot PCA reconstructed letters, one at a time.
# e.g., let_idx = 0 would show the image for letter A,
# and the PCA eigenimages.
let_coef = show_pca_im(Xproj, pca_comps, let_idx = let_idx)

At the command line, with
> python pca_letters.py -let_idx 5 -n_comp 8

You should get
```



















The first image shows the 8 eigenimages, and the second image is the PCA reconstructed letter "F".

Please note that in order to do character recognition well, we really need a large data set that consists of all commonly used fonts of the 26 letters in the alphabet. In this homework, we have only looked at one font. It is a warmup exercise for you to become familiar with PCA.