VIT-Vellore SCOPE

CSE6037 - Deep Learning and its Applications

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Assessment 1

Problem 3

Image Classifier using simple SVM.

```
In [46]:
```

```
from pathlib import Path
import matplotlib.pyplot as plt
import numpy as np
from sklearn import svm, metrics, datasets
from sklearn.utils import Bunch
from sklearn.model_selection import GridSearchCV, train_test_split
from skimage.io import imread
from skimage.transform import resize
import cv2
%matplotlib notebook
```

Loading the images form local file system

 Loading the dataset of images from a local system where they are stored in seperate folders for each class and also train and test images altogeather.

```
In [47]:
```

```
def load image files (container path, dimension=(64, 64)):
   image dir = Path(container path)
   folders = [directory for directory in image dir.iterdir() if directory.is dir()]
   categories = [fo.name for fo in folders]
   descr = "A image classification dataset"
   images = []
   flat data = []
   target = []
   for i, direc in enumerate(folders):
       for file in direc.iterdir():
            img = imread(file)
            img resized = resize(img, dimension, anti aliasing=True, mode='reflect')
            flat data.append(img resized.flatten())
           images.append(img resized)
           target.append(i)
   flat data = np.array(flat data)
   target = np.array(target)
   images = np.array(images)
   return Bunch (data=flat data,
                 target=target,
                 target names=categories,
                 images=images,
                 DESCR=descr)
```

In [48]:

iman datast = 100d iman filos/# /imant/.....ii/ima datast#N

```
_ tmage_aataset = toaa_tmage_tites("../tmput/xyzyutyt/tmg_aataset")
```

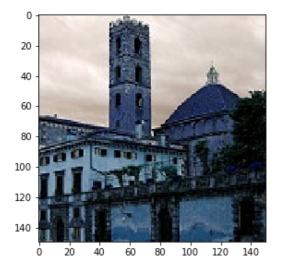
In [49]:

X_train, X_test, y_train, y_test = train_test_split(image_dataset.data, image_dataset.ta
rget, test_size=0.3,random_state=109)

Feature Extraction from images

In [50]:

```
from skimage.io import imread, imshow
from skimage import filters
image1 = cv2.imread('../input/xyzyuiyi/img_dataset/buildings/20061.jpg')
imshow(image1)
```

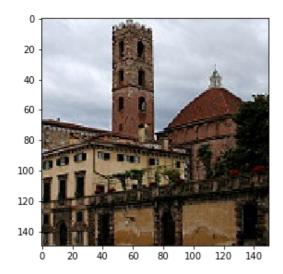


Out[50]:

<matplotlib.image.AxesImage at 0x7f033fb0bf90>

In [51]:

```
from skimage.io import imread, imshow
image2 = imread('../input/xyzyuiyi/img_dataset/buildings/20061.jpg', 0)
imshow(image2)
```



Out[51]:

<matplotlib.image.AxesImage at 0x7f033fb2fd10>

Analyzing both the images

Shape and Size of the image(i.e. the number of pixels in the img)

```
In [52]:
```

```
print("Color image will have 3 layers namely R, G, B so the shape is",image1.shape)
print("Gray scale image only has a single layer so its shape is",image2.shape)
print("Size of the RGB image is", image1.size)
print("Size of the gray scale is", image2.size)
```

```
Color image will have 3 layers namely R, G, B so the shape is (150, 150, 3) Gray scale image only has a single layer so its shape is (150, 150, 3) Size of the RGB image is 67500 Size of the gray scale is 67500
```

Pixel Features

The number of pixels in an image is the same as the size of the image for grayscale images we can find the pixel features by reshaping the shape of the image and returning the array form of the image.

```
In [57]:
```

```
print("Color image:",np.reshape(image2, (225 * 100 * 3)))
Color image: [227 233 247 ... 49 45 34]
```

• Region-Based Segmentation

We use this segment object from a background. We will use our image and try region-based segmentation on it.

In [58]:

```
from skimage.exposure import histogram
hist, hist_centers = histogram(image2)
#Plotting the Image and the Histogram of gray values
fig, axes = plt.subplots(1, 2, figsize=(8, 3))
axes[0].imshow(image2, cmap=plt.cm.gray)
axes[0].axis('off')
axes[1].plot(hist_centers, hist, lw=2)
axes[1].set_title('histogram of gray values')
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:2: UserWarning: This might be a color image. The histogram will be computed on the flattened image. You can instead a pply this function to each color channel.



histogram of gray values 2500 -2000 -1500 -1000 -500 -0 50 100 150 200 250

```
Out[58]:
```

```
Text(0.5, 1.0, 'histogram of gray values')
```

In [59]:

```
param_grid = [
    {'C': [1, 10, 100, 1000], 'kernel': ['linear']},
    {'C': [1, 10, 100, 1000], 'gamma': [0.001, 0.0001], 'kernel': ['rbf']},
]
svc = svm.SVC()
clf = GridSearchCV(svc, param_grid)
clf.fit(X_train, y_train)
```

```
Out[59]:
GridSearchCV(estimator=SVC(),
           param grid=[{'C': [1, 10, 100, 1000], 'kernel': ['linear']},
                      {'C': [1, 10, 100, 1000], 'gamma': [0.001, 0.0001],
                       'kernel': ['rbf']}])
In [60]:
y pred = clf.predict(X test)
In [61]:
t(y test, y pred)))
Classification report for -
GridSearchCV(estimator=SVC(),
           param_grid=[{'C': [1, 10, 100, 1000], 'kernel': ['linear']},
                      {'C': [1, 10, 100, 1000], 'gamma': [0.001, 0.0001],
                      'kernel': ['rbf']}]):
                      recall f1-score
            precision
                                       support
         0
                0.41
                        0.54
                                  0.47
                                            13
         1
                         0.33
                                  0.39
                0.46
                                            18
         2
                0.46
                         0.55
                                  0.50
                                            22
         3
                0.08
                         0.07
                                 0.07
                                            15
         4
                0.73
                         0.73
                                 0.73
                                            15
         5
                0.50
                         0.50
                                  0.50
                                            16
                                  0.45
                                            99
   accuracy
                                 0.44
               0.44
                         0.45
                                            99
  macro avg
               0.45
                         0.45
                                 0.45
                                            99
weighted avg
```