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Bai tap 4

File's contents:

- 1. Test functions on an image set: test some functions in need
- 2. Let's get start: homework 4

Dataset:

- http://www.cs.columbia.edu/CAVE/software/softlib/coil-20.php (http://www.cs.columbia.edu/CAVE/software/softlib/coil-20.php)
- 2. http://benchmark.ini.rub.de/?section=gtsrb&subsection=news (http://benchmark.ini.rub.de/?section=news)

Using HOG - sklearn

In brief, a HOG descriptor is computed by calculating image gradients that capture contour and silhouette information of grayscale images. Compute a Histogram of Oriented Gradients (HOG) by:

- 1. (optional) global image normalization
- 2. computing the gradient image in x and y
- 3. computing gradient histograms
- 4. normalizing across blocks
- 5. flattening into a feature vector

read more and go to details at: https://www.learnopencv.com/histogram-of-oriented-gradients/)

Test functions on an image set

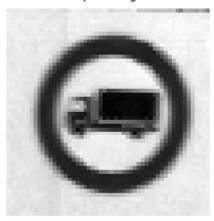
```
In [2]:
          1 #import scikit-learn
             from sklearn import metrics
          3 from sklearn.cluster import KMeans, spectral_clustering, DBSCAN, Agglomerative
          4 from sklearn.datasets import load digits
             from sklearn.neighbors import DistanceMetric
             from sklearn.metrics.pairwise import cosine_similarity
          7
             from sklearn.datasets import load iris
          8
          9
             from sklearn.datasets import fetch lfw people
             from sklearn.preprocessing import StandardScaler
         10
         11
         12 from skimage.feature import ORB, hog
         13 import cv2
             from skimage import data, color, exposure
         14
In [3]:
          1 # C:\Users\trang\Desktop\Thay Duy\GTSRB Final Test Images\GTSRB\Final Test\Im
In [4]:
          1
             # using traffic test dataset from: http://benchmark.ini.rub.de/?section=qtsrb
          3
             import glob
             file list = glob.glob('C:/Users/trang/Desktop/Thay Duy/GTSRB Final Test Image
             print(len(file_list))
In [5]:
        12630
In [6]:
          1 #load image and convert into gray image
             gray_sample = cv2.imread(file_list[0], 0)
In [7]:
          1 print(gray sample.shape)
          2 print(gray_sample)
        (54, 53)
        [[135 135 134 ..., 106 98 79]
         [140 137 135 ..., 139 136 134]
         [139 135 136 ..., 138 136 137]
         [134 133 132 ..., 134 135 135]
         [132 132 130 ..., 134 135 135]
         [132 130 131 ..., 136 136 136]]
```

Hiển thị kết quả ảnh xám và ảnh HOG

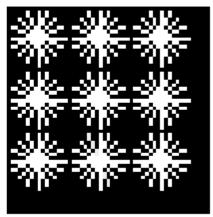
```
In [8]:
            fd, hog image = hog(gray sample, orientations=8, pixels per cell=(16, 16),
                                cells per block=(1, 1), visualise=True)
          3
            fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(8, 4), sharex=True, sharey=True
          4
          5
           ax1.axis('off')
          6
            ax1.imshow(gray_sample, cmap=plt.cm.gray)
          7
           ax1.set title('Input image')
         9
            ax1.set_adjustable('box-forced')
        10
        11 # Rescale histogram for better display
        12
            hog_image_rescaled = exposure.rescale_intensity(hog_image, in_range=(0, 0.02)
        13
        14 ax2.axis('off')
        15 ax2.imshow(hog_image_rescaled, cmap=plt.cm.gray)
        16 ax2.set_title('Histogram of Oriented Gradients')
            ax1.set adjustable('box-forced')
        17
        18
            plt.show()
```

c:\users\tranq\appdata\local\programs\python\python36\lib\site-packages\skimage
\feature_hog.py:119: skimage_deprecation: Default value of `block_norm`==`L1`
is deprecated and will be changed to `L2-Hys` in v0.15
 'be changed to `L2-Hys` in v0.15', skimage_deprecation)

Input image



Histogram of Oriented Gradients



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 'be changed to `L2-Hys` in v0.15', skimage deprecation)

Kiểm tra kích thước các ảnh

```
In [28]:
              gray_sample = cv2.imread(file_list[3], 0)
           2
              fd, hog_image = hog(gray_sample, orientations=8, pixels_per_cell=(16, 16),
           3
                                  cells_per_block=(1, 1), visualise=True)
           4 print(fd)
            print(cv2.imread(file list[3], 0).shape)
           6 print(cv2.imread(file_list[4], 0).shape)
             print(cv2.imread(file_list[5], 0).shape)
           8 print(cv2.imread(file_list[6], 0).shape)
              print(cv2.imread(file_list[7], 0).shape)
             print(cv2.imread(file_list[8], 0).shape)
             print(cv2.imread(file list[9], 0).shape)
         [ 0.12876587  0.17094845  0.18110611  0.14761291  0.10575664  0.07341477
           0.10465705 0.08773796]
         (29, 27)
         (57, 60)
         (56, 52)
         (130, 147)
         (33, 32)
         (50, 45)
         (86, 81)
         c:\users\tranq\appdata\local\programs\python\python36\lib\site-packages\skimage
         \feature\ hog.py:119: skimage deprecation: Default value of `block norm`==`L1`
          is deprecated and will be changed to `L2-Hys` in v0.15
            'be changed to `L2-Hys` in v0.15', skimage_deprecation)
```

=>> Nhận thấy với ảnh có kích thước khác nhau thì fd cũng có kích thước không gian khác nhau

=>> Cần resize các ảnh về cùng kích thước

Let's get start

```
In [1]:
          1 #import libs
            from time import time
          3 import numpy as np
          4 import matplotlib.pyplot as plt
            import pandas as pd
           import matplotlib.image as mpimg
          8 #import scikit-learn
         9 from sklearn import metrics
         10 from sklearn.cluster import KMeans, spectral_clustering, DBSCAN, Agglomerativ
         11 from sklearn.datasets import load digits
         12 from sklearn.neighbors import DistanceMetric
        13 from sklearn.metrics.pairwise import cosine_similarity
            from sklearn.datasets import load iris
            from sklearn.datasets import fetch lfw people
         16
            from sklearn.preprocessing import StandardScaler
        17
        18 from skimage.feature import ORB, hog
         19
            from skimage import data, color, exposure
         20
            import cv2
```

Get all paths of images

Total images: 1440

```
In [3]: 1 #load image, resize them into the same size, and convert into greyscale
2 def load_image_and_pre_processing(image_path):
3     gray_img = cv2.imread(image_path, 0)
4     #gray_img = cv2.resize(gray_img,(20,20))
5     return gray_img
```

```
1 # input: list of file paths
In [4]:
          2 # output: data - list of hog_vectors
             def HOG_data_measurement(file_list_):
          3
                 data = []
          4
          5
                 for path in file list:
                     grey_img = load_image_and_pre_processing(path)
          6
          7
                     hog_data,hog_image = hog(grey_img, orientations=8, pixels_per_cell=(8
                                 cells_per_block=(1, 1), visualise=True)
          8
          9
                     data.append(hog data)
         10
                 return data
```

```
Clustering
In [6]:
          1 #Kmeans
          2 nClusters = 20
          3 \mid t0 = time()
         4 kmeans_model = KMeans(nClusters)
          5 t_kmeans = time() - t0
          6 labels_kmeans = kmeans_model.fit_predict(data)
In [7]:
         1 #Spectral_clustering
          2 t0 = time()
          3 graph = cosine_similarity(data)
          4 t spectral = time()- t0
          5 labels_spectral = spectral_clustering(graph, n_clusters=20)
In [8]:
          1 #DBSCAN
          2 t0 = time()
          3 data = StandardScaler().fit_transform(data)
          4 labels_dbscan = DBSCAN(eps=0.3, min_samples=1,algorithm='kd_tree').fit_predid
          5 t_dbscan = time() - t0
In [9]:
          1 #Agglomerative Clustering
          2 t0 = time()
          3 Agglomerative_model = AgglomerativeClustering(n_clusters = nClusters)
          4 labels_AgglomerativeClustering = Agglomerative_model.fit_predict(data)
          5 t_agg = time() - t0
```

c:\users\tranq\appdata\local\programs\python\python36\lib\site-packages\skimage
\feature_hog.py:119: skimage_deprecation: Default value of `block_norm`==`L1`

data = HOG_data_measurement(file_list)

is deprecated and will be changed to `L2-Hys` in v0.15 'be changed to `L2-Hys` in v0.15', skimage_deprecation)

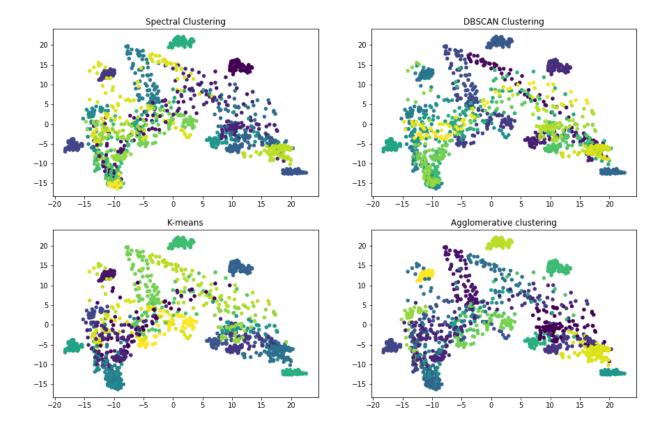
Visulization

In [5]:

```
In [10]:
           1 from sklearn.decomposition import PCA
             %matplotlib inline
           2
           3 \mid nComponents = 2
           4 vPCA = PCA(nComponents)
           5 digitData to 2D = vPCA.fit transform(data)
           7 fig = plt.figure(figsize=(15,15))
          8 fig.suptitle('Comparition results of methods', fontsize=20)
          9
         10 ax = fig.add_subplot(3,2,1)
         11
             plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_spectral,
         12
             ax.set_title('Spectral Clustering')
         13
         14 ax = fig.add subplot(3,2,2)
         15
             plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_dbscan, s=
         16 ax.set_title('DBSCAN Clustering')
         17
         18 ax = fig.add_subplot(3,2,3)
             plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_kmeans, s=
         19
         20 ax.set title('K-means')
         21
         22 ax = fig.add_subplot(3,2,4)
         23 plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_Agglomerat
         24 | ax.set_title('Agglomerative clustering')
         25
         26 \# ax = fig.add subplot(3,2,5)
         27 # plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= lfw_people.targ
         28 # ax.set_title('Target Result')
         29
```

Out[10]: <matplotlib.text.Text at 0x26e6ee78208>

Comparition results of methods



References

- 1. http://scikit-image.org/docs/dev/auto_examples/features_detection/plot_hog.html (http://scikit-image.org/docs/dev/auto_examples/features_detection/plot_hog.html)
- 2. https://www.learnopencv.com/histogram-of-oriented-gradients/ (https://www.learnopencv.com/histogram-of-oriented-gradients/)

Dataset:

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- 2. http://benchmark.ini.rub.de/?section=gtsrb&subsection=news (http://benchmark.ini.rub.de/?section=news)