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# Bai tap 4

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# 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: <a href="https://www.learnopencv.com/histogram-of-oriented-gradients/">https://www.learnopencv.com/histogram-of-oriented-gradients/</a> (<a href="https://www.learnopencv.com/histogram-of-oriented-gradients/">https://www.learnopencv.com/histogram-of-oriented-gradients/</a>)

#### **Demo Test**

```
In [12]: #import libs
    from time import time
    import numpy as np
    import matplotlib.pyplot as plt
    import pandas as pd
    import matplotlib.image as mpimg
```

```
In [13]: #import scikit-learn
         from sklearn import metrics
         from sklearn.cluster import KMeans, spectral clustering, DBSCAN, AgglomerativeClu
         from sklearn.datasets import load digits
         from sklearn.neighbors import DistanceMetric
         from sklearn.metrics.pairwise import cosine similarity
         from sklearn.datasets import load iris
         from sklearn.datasets import fetch lfw people
         from sklearn.preprocessing import StandardScaler
         from skimage.feature import ORB, hog
         import cv2
         from skimage import data, color, exposure
In [1]: # C:\Users\trang\Desktop\Thay Duy\GTSRB_Final_Test_Images\GTSRB\Final_Test\Images
In [15]: # using traffic test dataset from: http://benchmark.ini.rub.de/?section=qtsrb&sub!
         import glob
         file_list = glob.glob('C:/Users/tranq/Desktop/Thay Duy/GTSRB_Final_Test_Images/GT
In [16]: | print(len(file_list))
         12630
In [17]: #load image and convert into gray image
         gray sample = cv2.imread(file list[0], 0)
In [18]: print(gray_sample.shape)
         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]]
In [19]: fd, hog image = hog(gray sample, orientations=8, pixels per cell=(16, 16),
                             cells_per_block=(1, 1), visualise=True)
         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)

```
In [20]: print(hog image.shape)
         (54, 53)
In [21]:
         print(hog image)
                                     0.]
                0.
                    0. ...,
         [[ 0.
                             0.
                                 0.
                                     0.]
          [ 0.
                0.
                    0. ...,
                             0.
                                 0.
          [ 0.
                0.
                    0. ...,
                             0.
                                 0.
                                     0.]
          [ 0.
                0.
                    0. ...,
                             0.
                                     0.1
                                     0.]
          [ 0.
                0.
                    0. ...,
                             0.
                                 0.
                    0. ...,
                             0.
                                     0.]]
          [ 0.
                0.
                                 0.
In [22]: | print(fd)
         [ 0.07542413  0.44819084
                                   0.27551272
                                               0.03819493
                                                          0.06649688 0.03082425
           0.04924103 0.01611286
                                   0.00806231
                                               0.00436093
                                                           0.31948913
                                                                      0.35425726
           0.2932615
                       0.00553158
                                   0.01211024
                                               0.00292633
                                                           0.01193268
                                                                      0.00847274
           0.01228219 0.06401994
                                   0.27005266
                                               0.46631976
                                                          0.16514645
                                                                      0.0017728
           0.41759222 0.27204966
                                               0.00794486 0.00739411
                                   0.01246415
                                                                      0.00348958
           0.02366572 0.25539889
                                   0.08271567
                                               0.11182527
                                                          0.09531267
                                                                      0.18312603
           0.38925493 0.04878819
                                                           0.23753768 0.02987339
                                   0.04722842
                                               0.0417482
           0.01314065 0.02004655
                                   0.21169811
                                               0.03251646
                                                          0.22866156 0.22652508
           0.03441434 0.00783541
                                   0.00625825
                                               0.00656823
                                                          0.01382613 0.14487574
           0.53099973 0.25522103
                                   0.0053551
                                               0.01479662 0.0354136
                                                                       0.20714963
           0.35470584 0.30839348
                                   0.07305928
                                                           0.09747449
                                                                      0.33413911
                                               0.00112584
           0.37982689 0.10942635
                                   0.05235092
                                               0.01127933 0.01217518
                                                                      0.00332716]
In [23]: print(fd.shape)
         (72,)
In [24]:
         gray sample = cv2.imread(file list[1], 0)
         fd, hog image = hog(gray sample, orientations=8, pixels per cell=(16, 16),
                             cells_per_block=(1, 1), visualise=True)
         print(fd)
         [ 0.19251596  0.3057704
                                   0.21901283
                                               0.03726348
                                                          0.01166032
                                                                      0.01499944
           0.01677023 0.20200677
                                   0.03730958
                                               0.01144912 0.02510078 0.15047019
           0.25345538 0.33599621
                                   0.11425506
                                               0.07196326
                                                          0.27128523
                                                                      0.10601961
           0.05940887 0.07249194
                                               0.08408472
                                   0.08673583
                                                           0.07435032
                                                                      0.24562313
           0.1802568
                       0.12948743
                                   0.06567273
                                               0.04574295
                                                           0.03951589
                                                                      0.08881841
           0.1267734
                       0.32373223]
         c:\users\trang\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)

```
In [25]:
         print(fd.shape)
         print(fd)
         (32,)
         [ 0.19251596  0.3057704
                                  0.21901283
                                              0.03726348 0.01166032 0.01499944
           0.01677023 0.20200677
                                  0.03730958
                                              0.01144912
                                                          0.02510078 0.15047019
           0.25345538 0.33599621
                                  0.11425506
                                              0.07196326 0.27128523
                                                                     0.10601961
                                              0.08408472 0.07435032 0.24562313
           0.05940887 0.07249194
                                  0.08673583
           0.1802568
                      0.12948743
                                  0.06567273
                                              0.04574295
                                                          0.03951589
                                                                     0.08881841
           0.1267734
                      0.32373223]
In [26]:
         gray_sample = cv2.imread(file_list[2], 0)
         fd, hog image = hog(gray sample, orientations=8, pixels per cell=(16, 16),
                             cells per block=(1, 1), visualise=True)
         print(fd)
         print(fd.shape)
         [ 0.01668818  0.06060829
                                              0.21150763 0.46377098 0.07275872
                                  0.13054904
           0.02880035 0.01531532
                                  0.00391811
                                              0.01160133 0.10774882 0.23004025
           0.51640807 0.11568835
                                  0.0138387
                                              0.00075542 0.01146333 0.002967
           0.01736676 0.08170671
                                  0.19700363
                                              0.47111679 0.19943599 0.01893866
           0.12400235 0.07941266
                                  0.03533826
                                              0.06807628 0.1124205
                                                                      0.23207429
           0.19938632 0.14928879
                                  0.06997961
                                              0.01520588 0.03011032 0.02136434
           0.04822098 0.06475232
                                  0.74242242
                                              0.00794362 0.25081311 0.01661556
           0.02625111 0.03687919
                                  0.0467513
                                              0.09225051
                                                         0.18832367
                                                                     0.34211485
           0.05102435 0.02835034
                                  0.07654314
                                              0.05856553 0.28543881 0.28472358
           0.17504065 0.04031307
                                  0.05341589
                                              0.0584742
                                                          0.03810391 0.19479186
           0.39542891 0.15310522
                                  0.07664903
                                              0.0300305
                                                          0.25216953
                                                                      0.13031281
           0.11327148 0.24852658
                                  0.14953921
                                              0.032318
                                                          0.03719213
                                                                     0.03666936]
         (72,)
         c:\users\trang\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)
```

```
In [28]:
         gray sample = cv2.imread(file list[3], 0)
         fd, hog_image = hog(gray_sample, orientations=8, pixels_per_cell=(16, 16),
                              cells_per_block=(1, 1), visualise=True)
         print(fd)
         print(cv2.imread(file_list[3], 0).shape)
         print(cv2.imread(file_list[4], 0).shape)
         print(cv2.imread(file_list[5], 0).shape)
         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\trang\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)
```

### =>> We need to resize images into the samesize

# Let's get start

```
In [2]: #import libs
        from time import time
        import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        import matplotlib.image as mpimg
        #import scikit-learn
        from sklearn import metrics
        from sklearn.cluster import KMeans, spectral_clustering, DBSCAN, AgglomerativeClu
        from sklearn.datasets import load_digits
        from sklearn.neighbors import DistanceMetric
        from sklearn.metrics.pairwise import cosine similarity
        from sklearn.datasets import load iris
        from sklearn.datasets import fetch lfw people
        from sklearn.preprocessing import StandardScaler
        from skimage.feature import ORB, hog
        from skimage import data, color, exposure
        import cv2
```

# Get all paths of images

t0 = time()

graph = cosine\_similarity(data)

labels spectral = spectral clustering(graph, n clusters=7)

t\_spectral = time()- t0

```
In [3]: # using traffic test dataset from: http://benchmark.ini.rub.de/?section=gtsrb&sub.
        import glob
        file_list = glob.glob('C:/Users/tranq/Desktop/Thay Duy/GTSRB_Final_Test_Images/GT
        print('Total images: ', len(file list))
        Total images: 12630
In [4]: | #load image, resize them into the same size, and convert into greyscale
        def load_image_and_pre_processing(image_path):
            gray img = cv2.imread(image path, 0)
            gray_img = cv2.resize(gray_img,(20,20))
            return gray_img
In [5]: | # input: list of file paths
        # output: data - list of hog_vectors
        def HOG data measurement(file list ):
            data = []
            for path in file_list:
                grey_img = load_image_and_pre_processing(path)
                hog_data,hog_image = hog(grey_img, orientations=8, pixels_per_cell=(8, 8)
                             cells_per_block=(1, 1), visualise=True)
                 data.append(hog data)
            return data
In [6]: | data = HOG_data_measurement(file_list)
        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)
        Clustering
In [7]: #Kmeans
        nClusters = 7
        t0 = time()
        kmeans_model = KMeans(nClusters)
        t_kmeans = time() - t0
        labels_kmeans = kmeans_model.fit_predict(data)
In [8]: #Spectral_clustering
```

# **Visulization**

```
In [11]: from sklearn.decomposition import PCA
         %matplotlib inline
         nComponents = 2
         vPCA = PCA(nComponents)
         digitData to 2D = vPCA.fit transform(data)
         fig = plt.figure(figsize=(15,15))
         fig.suptitle('Comparition results of methods', fontsize=20)
         ax = fig.add_subplot(3,2,1)
         plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_spectral, s=20
         ax.set_title('Spectral Clustering')
         ax = fig.add subplot(3,2,2)
         plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_dbscan, s=20)
         ax.set_title('DBSCAN Clustering')
         ax = fig.add_subplot(3,2,3)
         plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_kmeans, s=20)
         ax.set title('K-means')
         ax = fig.add_subplot(3,2,4)
         plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= labels_AgglomerativeC
         ax.set_title('Agglomerative clustering')
         \# ax = fig.add subplot(3,2,5)
         # plt.scatter(digitData_to_2D[:,0], digitData_to_2D[:,1], c= lfw_people.target,
         # ax.set_title('Target Result')
```

Out[11]: <matplotlib.text.Text at 0x1b42a911ac8>

# Comparition results of methods

