Assignment05

April 11, 2019

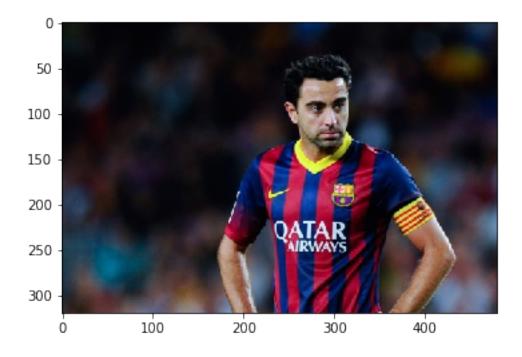
Assignment05: K-means algorithm on color image Software Engineering 20154652 Lee Dong Jae

In [1]: import PIL.Image as pilimg
 import numpy as np
 import matplotlib.pyplot as plt

In [2]: # Read image
 im = pilimg.open('xavi.jpg')

In [4]: plt.imshow(data)

Out[4]: <matplotlib.image.AxesImage at 0x2e4f137d8d0>



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In [5]: hor = list(data.shape)[0]
        ver = list(data.shape)[1]
In [6]: def distance(x, y):
            d = (x - y) ** 2
            \#s = np.sum(d)
            \#r = np.sqrt(s)
            return(d)
In [7]: def initialize_label(k):
            #Assign a dictionary that contain energy
            info = {'energy_list': []}
            #initialize label randomly
            first_label = np.random.randint(k, size = (hor, ver))
            clusters_hor = {idx: [] for idx in range(k)}
            clusters_ver = {idx: [] for idx in range(k)}
            new_clusters = {idx: [] for idx in range(k)}
            #append corresponding [hor, ver] of image to cluster
            for i in range(k):
                clusters_hor[i], clusters_ver[i] = np.where(first_label == i)
                for j in range(len(clusters_hor[i])):
                    new_clusters[i].append([clusters_hor[i][j], clusters_ver[i][j]])
            make_new_centroid(new_clusters, k, info)
In [8]: def energy_function(centroid, clusters, k):
            energy = 0
            #get energy
            for m in range(k):
                for n in clusters[m]:
                    for u in range(3):
                        energy += distance(data[n[0]][n[1]][u], centroid[m][u])
            return energy / (hor*ver)
In [9]: def make_new_centroid(clusters, k, info, centroid = 0):
            new_centroid = np.zeros((k,3))
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#sum previous data contained in same cluster
            for i in range(k):
                for j in clusters[i]:
                    new_centroid[i][0] += data[j[0]][j[1]][0]
                    new_centroid[i][1] += data[j[0]][j[1]][1]
                    new_centroid[i][2] += data[j[0]][j[1]][2]
            for m in range(k):
                new_centroid[m][0] = new_centroid[m][0] / len(clusters[m])
                new_centroid[m][1] = new_centroid[m][1] / len(clusters[m])
                new_centroid[m][2] = new_centroid[m][2] / len(clusters[m])
            #if clustering does not change over, plot images and information
            if np.array_equal(centroid, new_centroid):
                print('end')
                plot_image(centroid, clusters, k)
                plot_charts(info)
            else:
                do_clustering(new_centroid, k, info)
In [10]: def do_clustering(centroid, k, info):
             #make a place for put indexes of data to each clusters
             clusters = {idx: [] for idx in range(0, k)}
             #a temporary array for keeping the distance
             temp_distance = np.zeros(k)
             for i in range(hor):
                 for j in range(ver):
                     for t in range(k):
                         for u in range(3):
                             temp_distance[t] += distance(data[i][j][u], centroid[t][u])
                     #find the argmin of distance. And append a idx of data to the cluster[arg.
                     clusters[np.argmin(temp_distance)].append([i, j])
                     temp_distance = np.zeros(k)
             #calcuate energy, training accuracy, test accuracy at each time
             energy2 = energy_function(centroid, clusters, k)
             print(energy2)
             #append calculated information to each list
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info['energy_list'].append(energy2)
             make_new_centroid(clusters, k, info, centroid)
In [11]: def plot_charts(info):
             plt.figure(figsize=(10, 8))
             plt.title("Energy")
             plt.plot(range(len(info['energy_list'])), info['energy_list'])
             plt.show()
In [12]: def plot_image(centroid, clusters, k):
             plt.figure(1)
             new_image = np.zeros((hor, ver, 3), dtype = np.uint8)
             for i in range(k):
                 for j in clusters[i]:
                     for u in range(3):
                         new_image[j[0]][j[1]][u] = centroid[i][u]
             plt.title("new image")
             plt.imshow(new_image)
             frame = plt.gca()
             frame.axes.get_xaxis().set_visible(False)
             frame.axes.get_yaxis().set_visible(False)
             plt.show()
In [14]: initialize_label(4)
6869.347145829909
2632.3436863733496
2156.7014151616113
1925.5632320079064
1839.7810285356727
1796.7839507007075
1759.2190443437503
1719.933738742148
1679.1925231701755
1639.4560519961944
1602.5394862145065
1569.5916946911132
1540.7853650136726
1517.319957454133
1496.9830975237378
1482.2068080044485
1473.7848105219116
1469.6899814944143
1467.6212540682548
1466.5799718565652
1466.1350378451189
1465.9126944474576
```

1465.8100874181919

1465.7520904499872

1465.725012997234

1465.7129018934477

1465.7069835486

1465.7033975805637

1465.700760211445

1465.6987400011617

1465.6964498050959

1465.6947154969462

1465.6935003966457

1465.6928674434214

1465.6923717896977

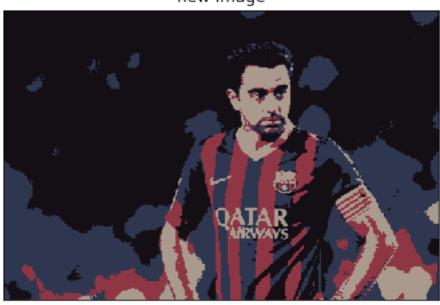
1465.6921019325973

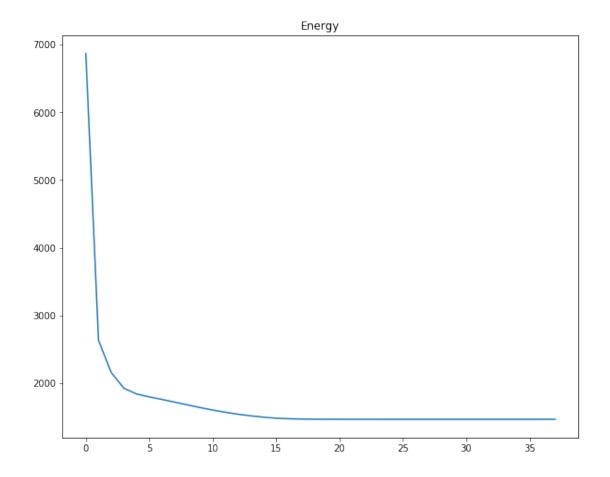
1465.692032569517

1465.6920228778195

end

new image





In [13]: initialize_label(8)

6818.508878821344

2171.027567719056

1454.2400775765177

1130.550739323829

972.3903000420127

938.45151379406

932.3785488116956

928.8938866107779

925.7166721369014

922.4704766763533

918.5866222927281

913.2517584393966

904.9931546206287

891.4203780201897

874.7268051882824

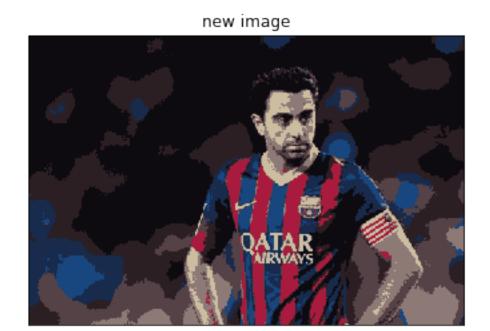
858.9224305464621

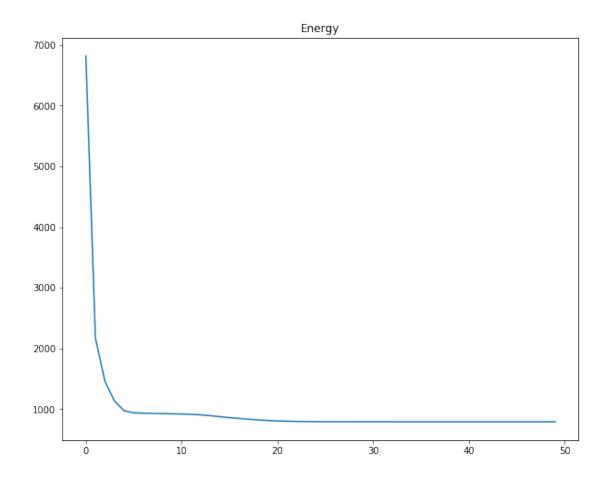
845.8096972477477

832.6247578496118

- 820.8891827066001
- 811.2240896245225
- 804.3900664849914
- 799.871961306284
- 796.7536431685215
- 794.6823822846449
- 793.2482915504431
- 792.2810672625548
- 791.6823492873867
- 791.3208123458546
- 791.0694567178322
- 790.874811658933
- 790.7312856303241
- 790.6417142808629
- 790.5782885424502
- 790.5354933256189
- 790.5028868603781
- 790.4795823557183
- 790.4656119384291
- 790.4513172593504
- 790.4438202903559
- 790.4390301198158
- 790.4354176139917
- 790.4324901327739
- 790.4310848206418
- 790.4298358856993
- 790.4285969657419
- 790.427908896846
- 790.4276833213734
- 790.4272819431948
- 790.4270748425315
- 790.4270726179872

end





In [15]: initialize_label(6)

6857.188169556307

2241.5153072699695

1466.2006817547642

1171.128211294202

1080.8937352663406

1051.9189897727645

1042.648311613631

1038.6413229186264

1036.0300174900296

1033.7600275764635

1031.668468741558

1029.7496385258737

1028.0382873636095

1026.4445388931485

1024.7702863309457

1022.8565383912006

1020.7348261064482

1018.3726820355429

4045 000000000000

1015.8728990733243

1013.5275124236996

1010.8927955253913

1008.1934194449321

1005.8823311896048

1004.0076919655661

1002.6109542282444

1001.5463874233889

1000.7588502420014

1000.199191736293

999.8190065548052

999.5292557455305

999.289455619304

999.1383262001704

999.0038728430345

998.9231097003111

998.878319162669

998.850165533962

998.8354117048776

998.8253861233113

998.8191733525517

998.8156058644868

998.8127265083461

998.8108415198955

998.8099656368821

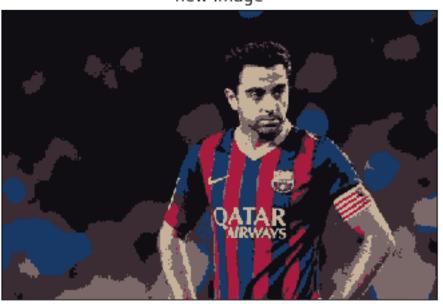
998.8080439148315 998.8069651178537 998.806309412233 998.8054797458691 998.8051576221903 998.8050158380197

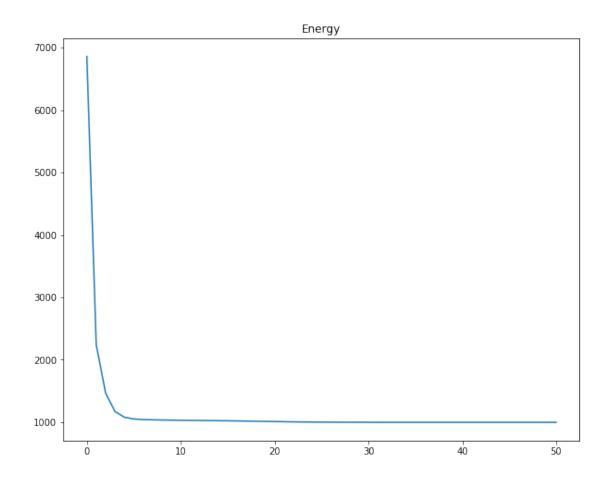
998.8048434926222

998.8048428401086

end

new image





In [13]: initialize_label(2)

6916.552024507432

3350.1244937035394

3105.9615917426827

3015.78655090859

2978.326949881929

2962.636000170432

2955.9351202025746

2953.392506901132

2952.505476838586

2952.2497508718393

2952.152741001759

2952.116256445785

2952.099447548904

2952.092607404813

2952.0911018739553

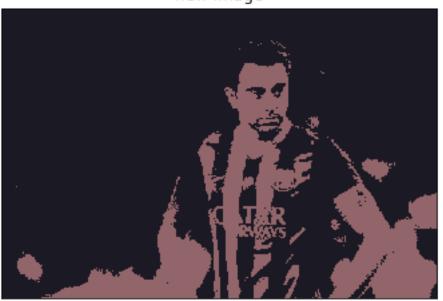
2952.0908195228526

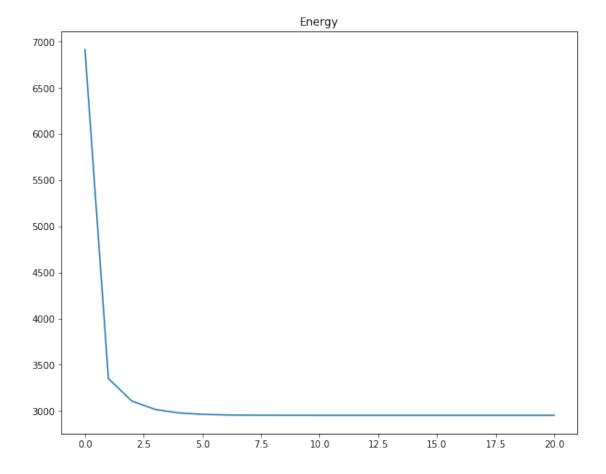
2952.090721736558

2952.0906863018017

2952.090656784123 2952.090652824963 2952.090651103949 end

new image





In []: