

04_Exercise1_K-means

May 20, 2018

1 Task01: Image compression with K-means

Implement K-Means algorithm and apply it to compress an image "NAORelease.jpg" for various K (see slides for details). As a feature vector use RGB-representation of each pixel from the image. Analyse running time, what could you suggest to improve it? Compare your implementation with the existing k-mean algorithm given in python.

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In [1]: from IPython.display import Image
import numpy as np
import scipy as sp
import cv2
from sklearn.cluster import KMeans
from skimage import io
import matplotlib.pyplot as plt
import time
```

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In [2]: Image(filename='NAORelease.jpg')
```

Out[2]:



```
In [3]: class compare_KMeans():
def __init__(self,filename):
    image = cv2.cvtColor(cv2.imread(filename), cv2.COLOR_BGR2RGB)
    image = np.array(image, dtype=np.float64) / 255
    self.width = image.shape[0]
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self.height = image.shape[1]
self.image= image.reshape((self.width * self.height, image.shape[2]))

def skLearn_Kmeans(self,k):
    t0 = time.time()
    kmeans = KMeans(n_clusters = k).fit(self.image)
    labels = kmeans.predict(self.image)
    cluster_centers = kmeans.cluster_centers_
    compressed_image = np.reshape(cluster_centers[labels],\
                                   (self.width, self.height,\
                                    cluster_centers.shape[1]))\
                                   )

    t1 = time.time()-t0
    return compressed_image, t1

def _Kmeans(self, k, MAX_ITERATIONS=200):
    t0 = time.time()
    np.random.seed(0)
    center_idx = np.random.randint(self.image.shape[0], size=k)
    means=self.image[center_idx,:]
    clustered_data=self.image
    clustered_data=np.insert(clustered_data, 3, -1, axis= 1)

    iteration=0
    while(True):
        iteration+=1
        distances = sp.spatial.distance.cdist(self.image,means)
        clustered_data[:,3]=np.argmin(distances, axis=1)
        old_means=means.copy()
        for i,cluster in enumerate(np.unique(clustered_data[:,3])):
            data_in_cluster=clustered_data[np.where(clustered_data[:,3]==cluster)]
            means[i]=np.mean(data_in_cluster[:,3],axis=0)

        if iteration>=MAX_ITERATIONS or np.allclose(old_means,means):
            break

    distances = sp.spatial.distance.cdist(self.image,means)
    clustered_data[:,3]=np.argmin(distances, axis=1)
    compressed_image = self.image.copy()
    for i,cluster in enumerate(np.unique(clustered_data[:,3])):
        idx=np.where(clustered_data[:,3]==cluster)
        compressed_image[idx] = means[i]
    compressed_image = np.reshape(compressed_image, \
                                   (self.width, self.height,\
                                    means.shape[1]))

    #print "Total iterations: ",iteration
    t1 = time.time()-t0
    return compressed_image, t1

```

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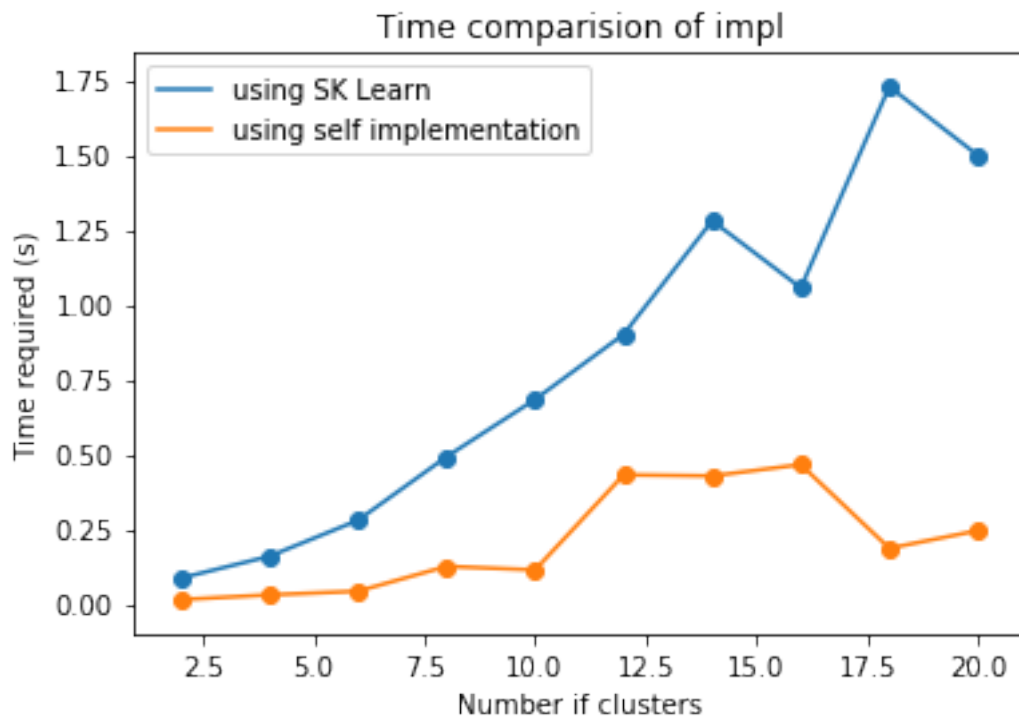
In [4]: Kmean_compare = compare_KMeans('NAORelease.jpg')

In [5]: K = np.arange(0, 22, 2)[1:]
sk_compressed = list()
self_compressed = list()

for k in K:
    sk_img,sk_time=Kmean_compare.skLearn_Kmeans(k)
    self_img,self_time=Kmean_compare._Kmeans(k)
    sk_compressed.append(np.array([sk_img,sk_time]))
    self_compressed.append(np.array([self_img,self_time]))

In [6]: plt.plot(K,np.array(sk_compressed[:])[:,1],label = "using SK Learn")
plt.scatter(K,np.array(sk_compressed[:])[:,1])
plt.plot(K,np.array(self_compressed[:])[:,1],label = "using self implementation")
plt.scatter(K,np.array(self_compressed[:])[:,1])
plt.title("Time comparision of impl")
plt.xlabel("Number if clusters")
plt.ylabel("Time required (s)")
plt.legend()
plt.show()

```



```

In [7]: images_sk=np.array(sk_compressed[:])[:,0]
images_self=np.array(self_compressed[:])[:,0]

```

```

figure = plt.figure(figsize=(10,40))

img_number = 1
for i in np.arange(0,len(K),1):
    plt.subplot(10,2,2*i+1)
    plt.title('K = %i with SK Learn'%K[i])
    plt.imshow(images_sk[i])
    plt.subplot(10,2,2*i+2)
    plt.title('K = %i with Self Implementation'%K[i])
    plt.imshow(images_self[i])
plt.tight_layout()
plt.show()

```

