

K-Means for Image Segmentation/ Image Compression

Import required libraries

In [6]:

```
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import numpy as np

plt.figure(figsize=(10,10))
# Read Images
img = mpimg.imread('kids.jpg')

# Output Images
plt.imshow(img)
```

Out[6]:

<matplotlib.image.AxesImage at 0x291e8fbc6d0>



Preprocess Image

In [9]:

```
data = img.reshape((-1,3))  
data = data/255  
print(data[:5])
```

```
[[0.60392157 0.61568627 0.63529412]  
 [0.60784314 0.61960784 0.63921569]  
 [0.60784314 0.61960784 0.63921569]  
 [0.61176471 0.62352941 0.64313725]  
 [0.61176471 0.62352941 0.64313725]]
```

K Means Algorithm Implementation

In [15]:

```

def segment_image_using_KMeans(no_of_clusters):

    K = no_of_clusters
    N = len(data) # no of points in dataset where each point is 3-dimensional (R,G,B)

    # initialize randomly K centroids  $\mu_1(0), \dots, \mu_K(0)$  as K distinct points of dataset
     $\mu$  = data[tuple([np.random.choice(range(len(data)), size=K, replace=False))]] #  $\mu = [\mu_1$ 
     $\mu = [ [\mu[j]]$  for j in range(K)] # list to store all previous updates  $\mu_1(t), \dots, \mu_K(t)$  ti

    Z = [ np.random.choice(range(1,K+1)) for i in range(N)]

    # E-Step :

    # Assign each point to 1 of K clusters based on euclidean distance from
    # centroid's and assign every point to closest of these K clusters

    t = 0
    epsilon = 10**(-2)
    print("Beginning K-Means algorithm for K=%d"%K)
    while True :

        #E-Step begins
        for i in range(N) :
            point = data[i]
            distances = [ np.linalg.norm( $\mu[j][t]$  - point) for j in range(K) ]
            Z[i] = np.argmin(distances) + 1
        # End of E-Step

        # M- Step : Update Centroids  $\mu_1(t+1), \mu_2(t+1), \dots, \mu_K(t+1)$ 
        for j in range(1,K+1) :
             $\mu_j$ _new = np.array([ data[i] for i in range(N) if Z[i]==j ])
             $\mu_j$ _new = np.sum( $\mu_j$ _new, axis=0) / len( $\mu_j$ _new)
             $\mu[j-1]$ .append( $\mu_j$ _new)
        # End of M-Step

        tmp = np.array([ ( np.linalg.norm( $\mu[j][t+1]$  -  $\mu[j][t]$ ) < epsilon) for j in range(K) ])
        if ( tmp.all() ) :
            print("Finishing K-Means Algorithm. Bye!")
            print("Total iterations : %d"%t)
            img_updated = np.array([ $\mu[Z[i]-1][-1]$  for i in range(N)])
            img_updated = img_updated.reshape(img.shape)
            return img_updated
        t = t + 1

```

In [16]:

```

plt.figure(figsize=(10,5))
result = segment_image_using_KMeans(2)
plt.imshow(result)

```

Beginning K-Means algorithm for K=2
 Finishing K-Means Algorithm. Bye!
 Total iterations : 11

In [33]:

```

fig,axes = plt.subplots(2,3,figsize=(20,7))
i,j=0,0
for K in [2,5,10] :
    result = segment_image_using_KMeans(K)
    axes[i,j].imshow(result)
    axes[i,j].axis('off')
    axes[i,j].set_title('K = %d'%K)
    j += 1
    if j == 3 :
        j = 0
        i += 1

```

Beginning K-Means algorithm for K=2

Finishing K-Means Algorithm. Bye!

Total iterations : 1

Beginning K-Means algorithm for K=5

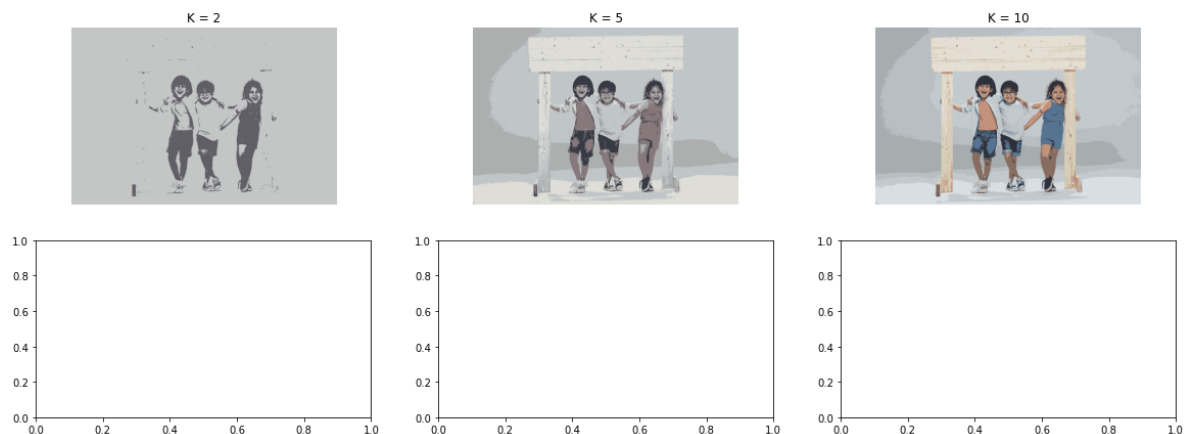
Finishing K-Means Algorithm. Bye!

Total iterations : 6

Beginning K-Means algorithm for K=10

Finishing K-Means Algorithm. Bye!

Total iterations : 18



In []: