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In [ ]: import cv2
from google.colab.patches import cv2_imshow
import numpy as np
import numpy.linalg as la
from matplotlib.colors import LogNorm
from scipy import signal
from skimage.transform import rescale
import matplotlib.pyplot as plt
import utils
import scipy
import skimage
from sklearn.neighbors import KDTree
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In [ ]: # To connect to drive for input images. We just used Local images

# from google.colab import drive
# drive.mount('/content/drive')
```

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In [ ]: #Declare global variables (Hyperparameters)
L = 1
total_levels = 4
kappa = 1

# kernel3 denotes gaussian weighting for a 3 x 3
kernel3 = np.array([[1,2,1],
                    [2,4,2],
                    [1,2,1]])
kernel3 = kernel3 * 1/16

# kernel5 denotes gaussian weighting for a 5 x 5
kernel5 = np.array([[1,4,7,4,1],
                    [4,16,26,16,4],
                    [7,26,41,26,7],
                    [4,16,26,16,4],
                    [1,4,7,4,1]])
kernel5 = kernel5 * 1/273
```

In []: *#Brute Force Solution for coarsest level and corner and edge pixels of each level*

```
def bruteforce(total_feature_vectors, A1_pyramid, Alf_pyramid, B1_pyramid, Blf_pyramid):
    feature_B, weights_B = GetFeatureVectors(q[0], q[1], l, B1_pyramid, Blf_pyramid)
    feature_B = np.array(feature_B)
    idx = np.where(np.array(weights_B) != 0)
    feature_B = feature_B[idx]
    mindist = -1
    for y in range(A1_pyramid[1].shape[0]):
        for x in range(A1_pyramid[1].shape[1]):
            if y > 1 and y < A1_pyramid[1].shape[0] - 2 and x > 1 and x < A1_pyramid[1].shape[1] - 2:
                feature_A = total_feature_vectors[1][(y-2)*(A1_pyramid[1].shape[1]-4) + x-2]
            else:
                feature_A, weights_A = GetFeatureVectors(y, x, l, A1_pyramid, Alf_pyramid)
            feature_A = np.array(feature_A)[idx]
            diff = feature_B - feature_A
            dist = np.dot(diff, diff)
            if mindist == -1 or dist < mindist:
                mindist = dist
                pix = (y, x)
    return pix
```



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In [ ]: #Main function to return filtered B
def CreateImageAnalogy(A, A_filtered, B, B_filtered):
    #Convert each image to YIQ color space
    A_yiq = YIQConvert(A)
    Af_yiq = YIQConvert(A_filtered)
    B_yiq = YIQConvert(B)
    Bf_yiq = YIQConvert(B_filtered)

    #Create YIQ pyramids
    A_pyramids = create_pyramid(A_yiq, total_levels)
    Af_pyramids = create_pyramid(Af_yiq, total_levels)
    B_pyramids = create_pyramid(B_yiq, total_levels)
    Bf_pyramids = create_pyramid(Bf_yiq, total_levels)

    Al_pyramid, Alf_pyramid, Bl_pyramid, Blf_pyramid = [],[],[],[]

    #Get the Luminance pyramids
    for i in range(total_levels):
        Al_pyramid.append(A_pyramids[i][:,:,0])
        Alf_pyramid.append(Af_pyramids[i][:,:,0])
        Bl_pyramid.append(B_pyramids[i][:,:,0])
        Blf_pyramid.append(Bf_pyramids[i][:,:,0])

    #Initialize the search structures, which can also be used as a cache
    total_feature_vectors = []
    total_weight_vectors = []
    for level in range(total_levels):
        features, weights = InitializeTree(Al_pyramid, Alf_pyramid, level)
        total_feature_vectors.append(features)
        total_weight_vectors.append(weights)

    #Initialize Blf_pyramid:
    # Blf_pyramid = []
    # for i in range(total_levels):
    #     Blf_pyramid.append(B_pyramid[i].shape))

    s = {} # collection of points indexed by q that map to p, s(q) = p
    for level in range(total_levels-1,-1,-1): # for each level •, from coarser to finer
        print (level)
        if level < total_levels-1:
            KDtree = KDTree(total_feature_vectors[level]) #scipy.spatial.KDTree
        for row in range(Blf_pyramid[level].shape[0]): # for each pixel q ∈ Q
            for col in range(Blf_pyramid[level].shape[1]):
                #Find the best matching pixel
                if row > 1 and row < Blf_pyramid[level].shape[0] - 2 and col > 1 and col < Blf_pyramid[level].shape[1] - 2:
                    pixel = BestMatch(KDtree, total_feature_vectors, total_weight_vectors, row, col)
                else:
                    pixel = bruteforce(total_feature_vectors, Al_pyramid, Alf_pyramid, B_pyramids, Bf_pyramids, level, row, col)
                Blf_pyramid[level][row][col] = Alf_pyramid[level][int(pixel[0])][int(pixel[1])]
                s[(row,col)] = pixel

    #Take Luminance channel from finest level of pyramid copy I and Q channels
    image = np.zeros((B.shape[0],B.shape[1],3))
    image[:, :, 0] = Blf_pyramid[0]
    image[:, :, 1:] = B_yiq[:, :, 1:]

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    return Blf_pyramid, image

#Finds the Best Match for a pixel in a Level of B filtered
def BestMatch(KDtree, total_feature_vectors, total_weight_vectors, Al_pyramid,
    #B_filtered is partially synthesized
    #s = source information level, collection of points indexed by q that map to
    #l = level
    #q = the pixel being synthesized in B_filtered, tuple of row and col

    pixel_approximate = BestApproximateMatch(KDtree, total_feature_vectors, Al_pyramid, q)
    pixel_coherence = BestCoherenceMatch(total_feature_vectors, total_weight_vectors, Al_pyramid, q)

    if (pixel_coherence) == (None, None):
        return pixel_approximate

    feature_B, weights_B = GetFeatureVectors(q[0],q[1],l, Bl_pyramid, Blf_pyramid)

    # Get features and weights from cache
    features = total_feature_vectors[l]
    weights = total_weight_vectors[l]
    one_d_approx = int(pixel_approximate[0]-2)*(Al_pyramid[l].shape[1]-4) + int(pixel_approximate[1]-2)*(Al_pyramid[l].shape[1]-4)
    one_d_coherence = int(pixel_coherence[0]-2)*(Al_pyramid[l].shape[1]-4) + int(pixel_coherence[1]-2)*(Al_pyramid[l].shape[1]-4)
    feature_approx = features[int(one_d_approx)]
    weights_approx = weights[int(one_d_approx)]
    feature_coherence = features[int(one_d_coherence)]
    weights_coherence = weights[int(one_d_coherence)]

    d_coherence = la.norm(Distance(feature_coherence, feature_B, weights_coherence, weights_B))
    d_approx = la.norm(Distance(feature_approx, feature_B, weights_approx, weights_B))
    kappanew = -1

    if (d_coherence <= d_approx*(1+(2**(1 - total_levels))*kappanew)):
        return pixel_coherence
    return pixel_approximate

#Finds Best Match for a pixel based on Coherence
def BestCoherenceMatch(total_feature_vectors, total_weight_vectors, Al_pyramid,
    #s = source information level
    #l = level
    #q = the pixel being synthesized in Bf_pyramid[l], tuple of row and col
    #total_feature_vectors = cache of feature vectors of A from tree initialization
    #total_weight_vectors = cache of weights of A from tree initialization, l X

    r_star = (None,None) #r*, tuple of best coherence match

    initial_row = q[0]
    initial_col = q[1]
    min = np.inf

    #Loop through 5x5 neighborhood around q, Looking for best match
    flag = False
    for row in range(-2,1):
        if (flag):
            break
        for col in range(-2,3):
            if (row == 0) and (col == 1):

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        flag = True
        break
    r = (initial_row + row, initial_col + col)
    #check if pixel is in bounds:
    if (r[0] >= 0) and (r[0] < Blf_pyramid[1].shape[0]) and (r[1] >= 0) and (r[1] < Blf_pyramid[1].shape[1]):
        if (r in s): #check if pixel has already been synthesized
            potential_point = (int(s[r][0] + (q[0] - r[0])), int(s[r][1] + (q[1] - r[1])))
            if (potential_point[0] >= 0) and (potential_point[0] < Blf_pyramid[1].shape[0]) and (potential_point[1] >= 0) and (potential_point[1] < Blf_pyramid[1].shape[1]):
                # Get feature_A and weights_A from the caches:
                total_feature = total_feature_vectors[1]
                total_weight = total_weight_vectors[1]
                one_d_coord = (potential_point[0]-2)*(Alf_pyramid[1].shape[1]-4)
                if int(one_d_coord) < len(total_feature):
                    feature_A = total_feature[int(one_d_coord)]
                    weights_A = total_weight[int(one_d_coord)]
                    feature_B, weights_B = GetFeatureVectors(q[0], q[1], 1, Bl_pyramid, Blf_pyramid)
                    # compute weighted norm of feature_A and feature_B:
                    cost = la.norm(Distance(feature_A, feature_B, weights_A, weights_B))
                    if (cost < min):
                        min = cost
                        r_star = r

if (r_star) == (None, None):
    return r_star
best_match = (int(s[r_star][0] + (q[0]-r_star[0])), int(s[r_star][1] + (q[1]-r_star[1])))
return best_match

#Finds Best Match of a pixel based on nearest neighbour
def BestApproximateMatch(KDtree, total_feature_vectors, Al_pyramid, Alf_pyramid, Bl_pyramid, Blf_pyramid):
    feature_B, weights_B = GetFeatureVectors(q[0], q[1], 1, Bl_pyramid, Blf_pyramid)
    result = KDtree.query([feature_B], k=1)
    index = result[1]
    row = (index / (Al_pyramid[1].shape[1]-4)) + 2
    col = (index % (Al_pyramid[1].shape[1]-4)) + 2

    return (row, col)

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In [ ]: #Helper Function: Gets the features from a particular level and filtered or unfiltered
def GetFeatures(row,col, l, neighborhood_size, luminance_pyramid):
    features = []
    weights = []
    kernel = 0
    # set kernel to corresponding gaussian kernel either 5 x 5 or 3 x 3
    if (neighborhood_size == 2):
        kernel = kernel5
    else:
        kernel = kernel3
    for y_off in range(-neighborhood_size, neighborhood_size + 1):
        for x_off in range(-neighborhood_size, neighborhood_size + 1):
            x_pos = col + x_off
            y_pos = row + y_off
            if (x_pos >= 0) and (x_pos < luminance_pyramid[l].shape[1]) and (y_pos >= 0) and (y_pos < luminance_pyramid[l].shape[0]):
                features.append(luminance_pyramid[l][int(y_pos)][int(x_pos)])
                weights.append(kernel[y_off + neighborhood_size][x_off + neighborhood_size])
            else:
                features.append(0)
                weights.append(0)
    return features, weights

#Helper Function: Same as GetFeatures just for special case of getting from current level
def GetFeaturesPrime(row,col, l, neighborhood_size, luminance_pyramid):
    features = []
    weights = []
    kernel = 0
    # set kernel to corresponding gaussian kernel either 5 x 5 or 3 x 3
    if (neighborhood_size == 2):
        kernel = kernel5
    else:
        kernel = kernel3
    flag = False
    for y_off in range(-neighborhood_size, 1):
        if (flag):
            break
        for x_off in range(-neighborhood_size, neighborhood_size + 1):
            if (y_off == 0) and (x_off == 1):
                flag = True
                break
            x_pos = col + x_off
            y_pos = row + y_off
            if (x_pos >= 0) and (x_pos < luminance_pyramid[l].shape[1]) and (y_pos >= 0) and (y_pos < luminance_pyramid[l].shape[0]):
                features.append(luminance_pyramid[l][int(y_pos)][int(x_pos)])
                weights.append(kernel[y_off + neighborhood_size][x_off + neighborhood_size])
            else:
                features.append(0)
                weights.append(0)
    return features, weights

#Computes the complete feature of a given pixel in either A or B
def GetFeatureVectors(row, col, level, luminance_pyramid, luminance_pyramid_prime):
    features = []
    weights = []
    #Compute features at fine level
    fl_features, fl_weights = GetFeatures(row,col,level, 2, luminance_pyramid)
    fl_features_prime, fl_weights_prime = GetFeaturesPrime(row,col,level, 2, luminance_pyramid_prime)

```



```
In [ ]: #Low Pass and High Pass Filters  
#Only used to create filtered image A' if testing on blur filter  
  
def lowPass(im1, sigma_low):  
    low_pass_gaussian = utils.gaussian_kernel(sigma_low, 3*sigma_low)  
    low_passed_image = cv2.filter2D(im1, -1, low_pass_gaussian)  
    return low_passed_image  
  
def highPass(im2, sigma_high):  
    high_pass_gaussian = utils.gaussian_kernel(sigma_high, 3*sigma_high)  
    high_passed_image = im2 - cv2.filter2D(im2, -1, high_pass_gaussian)  
    return high_passed_image
```

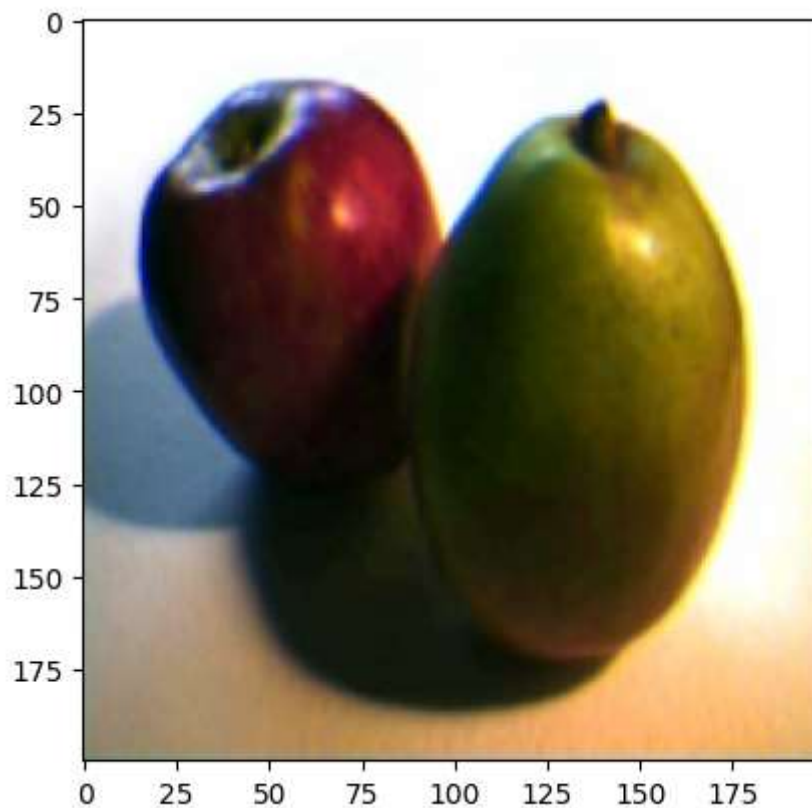
```
In [ ]: #Setup input images
n = 200
A = cv2.imread('fruit.jpg')[:, :, (2,1,0)]
A = cv2.resize(A, (n,n))
# Af = LowPass(A, 1)
Af = cv2.imread('fruit_filtered.jpg')[:, :, (2,1,0)]
Af = cv2.resize(Af, (n,n))
B = cv2.imread('landscape.jpg')[:, :, (2,1,0)]
B = cv2.resize(B, (n,n))
Bf = np.zeros(B.shape)

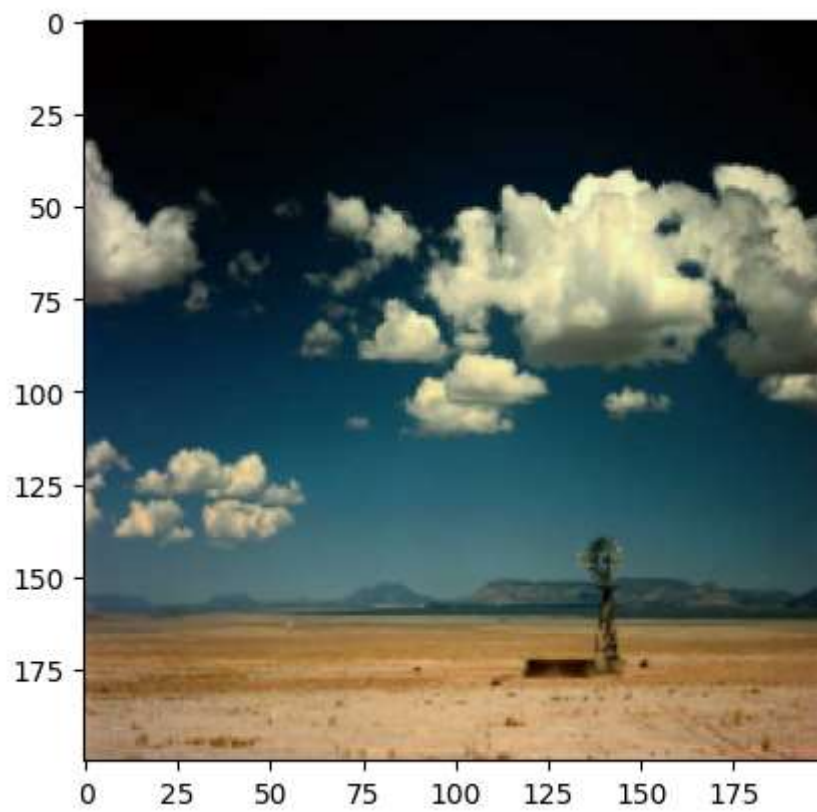
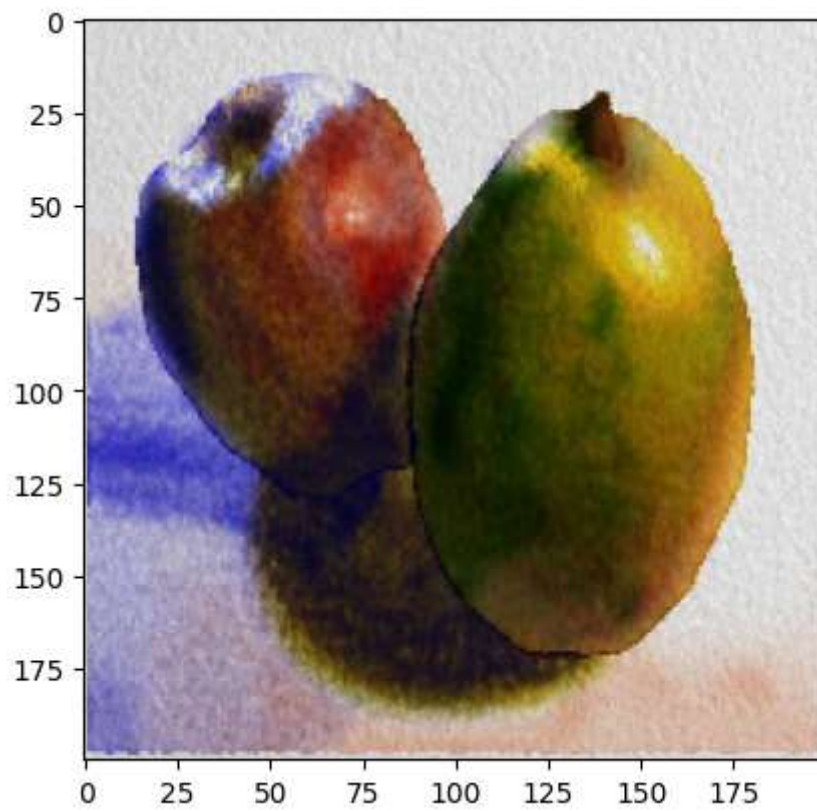
plt.figure()
plt.imshow(A)

plt.figure()
plt.imshow(Af)

plt.figure()
plt.imshow(B)
```

Out[14]: <matplotlib.image.AxesImage at 0x7f4280ee5e40>





```
In [ ]: # Call create Image Analogy (Bf is just an image of zeros)  
B_filtered,image = CreateImageAnalogy(A,Af,B,Bf)
```

3
2
1
0

```

In [ ]: plt.figure()
plt.imshow(B)

plt.figure()

# Just for TESTING
# Viewing the entire resulting pyramid
# for i in range(total_levels):
#     plt.figure()
#     plt.imshow(B_filtered[i])
#     plt.imshow(Bf_pyramids[i], cmap = 'gray')

Bf_rgb = skimage.color.yiq2rgb(image)

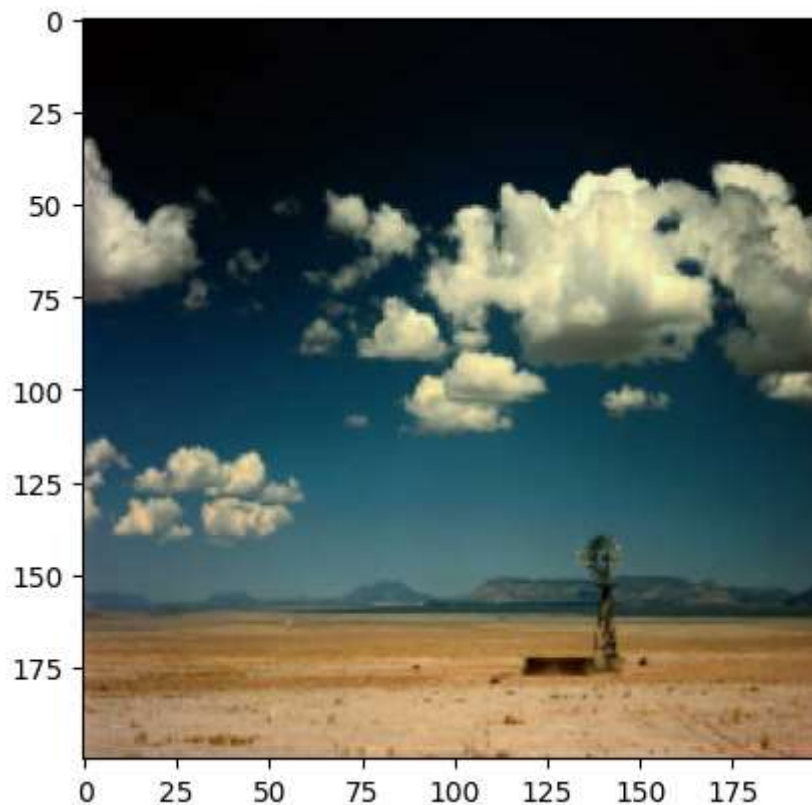
plt.figure()
plt.imshow(Bf_rgb)

print("done")

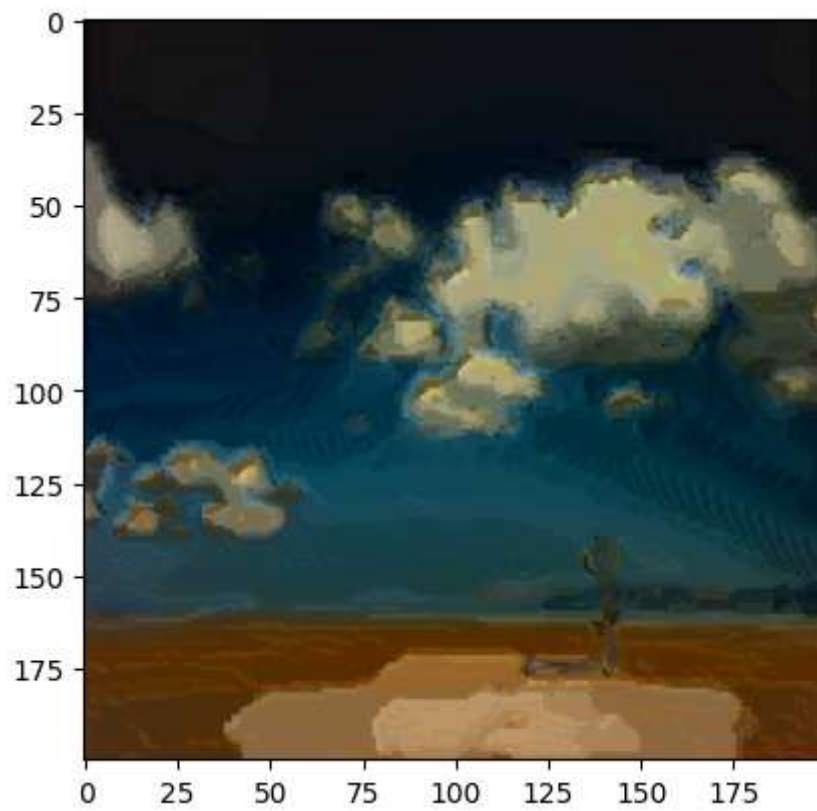
```

WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

done



<Figure size 640x480 with 0 Axes>



In []: