Programming Assignment 10 Report

November 25, 2020

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Importing prerequisites

```
[1]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import os
import time
```

```
[2]: def plotter(img_list, r, w, gray, wr, hr, fig_name = None):
         Plots images' list with its' caption and saves result image if you want.
         Parameters:
             img_list (list): The list of tuples of image and its' caption.
             r (int): The number of row(s).
             w (int): The number of column(s).
             gray (bool): The flag for plotting images in grayscale mode.
             wr (int): The width of one figure.
             hr (int): The height of one figure.
             fig_name (str): The name of the image of the plot. if not set this \sqcup
      \rightarrow parameter the plot doesn't save.
         111
         plt.rcParams['figure.figsize'] = (wr, hr)
         for i in range(len(img_list)):
             plt.subplot(r, w, i + 1)
             if img_list[i][2] == 'img':
                 if gray:
                     plt.imshow(img_list[i][0], cmap = 'gray')
                 else:
                     plt.imshow(img_list[i][0])
                 plt.xticks([])
                 plt.yticks([])
             elif img_list[i][2] == 'hist':
                 plt.bar(np.arange(len(img_list[i][0])), img_list[i][0], color = 'c')
             else:
```

```
raise Exception("Only image or histogram. Use third parameter of 

→tuples in img_list and set it to img or hist.")

plt.title(img_list[i][1])

if fig_name is not None:

plt.savefig(fig_name + '.png')

plt.show()
```

1 Shape and Color Detection

We will be performing shape and color detection using contours and LAB color space.

detect_shape(contour): Gets a contour as input, predicts the shape of the contour by calculating primeter. Returns the label of shape predicted.

detect_color(image, contour): Gets image and a contour as input, predicts the color of the region determined by contour using distanmee comparison between mean of color intensities and LAB color space. Returns the label of color predicted.

```
[3]: from scipy.spatial import distance as dist
     def detect_shape(contour):
         shape = "unidentified"
         perimeter = cv2.arcLength(contour, True)
         approx = cv2.approxPolyDP(contour, 0.05 * perimeter, True)
         if len(approx) == 3:
             shape = 'triangle'
         elif len(approx) == 4:
             (x, y, w, h) = cv2.boundingRect(approx)
             ratio = w / float(h)
             if ratio >= 0.95 and ratio <= 1.05:</pre>
                 shape = 'square'
             else:
                 shape = 'rectangle'
         elif len(approx) == 5:
             shape = 'pentagon'
         else:
             shape = 'circle'
         return shape
     def detect_color(image, contour):
```

```
colors = np.array([
    [[255, 0, 0]],
    [[0, 255, 0]],
    [[0, 0, 255]],
], dtype="uint8")
color_names = ['red', 'green', 'blue']
lab = cv2.cvtColor(colors, cv2.COLOR_RGB2LAB)
mask = mask = np.zeros(image.shape[:2], np.uint8)
cv2.drawContours(mask, [contour], -1, 255, -1)
mask = cv2.erode(mask, None, iterations=1)
mean = cv2.mean(image, mask=mask)[:3]
min_dist = (np.inf, None)
for (i, row) in enumerate(lab):
    d = dist.euclidean(row[0], mean)
    if d < min dist[0]:</pre>
        min_dist = (d, i)
return color_names[min_dist[1]]
```

detect_shape_color(image): Gets image as input. Detects the shapes and colors of the shapes in the image, draws contour borders and (color + shape) labels on the shapes in the image.

```
contours, _ = cv2.findContours(thresh.copy(), cv2.RETR_EXTERNAL, cv2.
GCHAIN_APPROX_SIMPLE)

for contour in contours:
   if cv2.contourArea(contour) > 0:

   M = cv2.moments(contour)

   cX = int(M["m10"] / M["m00"])
   cY = int(M["m01"] / M["m00"])

   cv2.drawContours(image, [contour], -1, (0, 255, 0), 2)

   shape = detect_shape(contour)

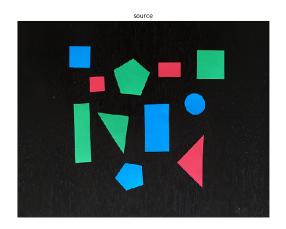
   color = detect_color(image, contour)

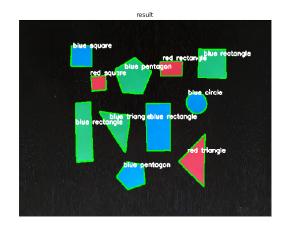
   text = color + " " + shape
   cv2.putText(image, text, (cX - 20, cY - 20), cv2.

GONT_HERSHEY_SIMPLEX, 0.5, (255, 255, 255), 2)

return image
```

Test of implementation.





2 Erosion and Dilation

make_slice(image, offset, i, j): Slices the image using the coordinates and offset provided. get_indices(shape, offset): Returns the valid indices for a particular kernel offset to be sliding over the image.

```
[6]: def make_slice(image, offset, i, j):
    return image[i - offset : i + offset + 1, j - offset : j + offset + 1]

def get_indices(shape, offset):
    indices = []
    for val in range(shape):
        if val - offset >= 0 and val + offset + 1 < shape:
            indices.append(val)
    return indices</pre>
```

```
[7]: structuring_element = np.ones((25, 25))
```

```
[8]: |def your_dilate(image, structuring_element):
         Applies your dilation.
         Parameters:
             image (numpy.ndarray): The input image.
             structuring_element (numpy.ndarray): The structuring element must be ...
      \hookrightarrow square.
         Returns:
             numpy.ndarray: The result image.
         offset = structuring_element.shape[0] // 2
         rows = get indices(image.shape[0], offset)
         cols = get_indices(image.shape[1], offset)
         dilated_image = np.zeros_like(image)
         for i in rows:
             for j in cols:
                  image_slice = make_slice(image, offset, i, j)
                  if np.logical_and(structuring_element, image_slice).any():
                      dilated_image[i, j] = 1
         return dilated_image
```

```
[9]: def your_erode(image, structuring_element):
          Applies your erosion.
          Parameters:
              image (numpy.ndarray): The input image.
              structuring\_element (numpy.ndarray): The structuring element must be \sqcup
       \hookrightarrow square.
          Returns:
              numpy.ndarray: The result image.
          offset = structuring_element.shape[0] // 2
          rows = get_indices(image.shape[0], offset)
          cols = get_indices(image.shape[1], offset)
          eroded_image = np.zeros_like(image)
          for i in rows:
              for j in cols:
                   image_slice = make_slice(image, offset, i, j)
                   logical_and = np.logical_and(structuring_element, image_slice)
                   if np.array_equal(logical_and, structuring_element):
                       eroded_image[i, j] = 1
          return eroded_image
[10]: def cv dilate(image, structuring element):
          Applies OpenCV dilation.
          Parameters:
              image (numpy.ndarray): The input image.
              structuring_element (numpy.ndarray): The structuring element must be ⊔
       \hookrightarrow square.
          Returns:
              numpy.ndarray: The result image.
          return cv2.dilate(image, structuring_element, iterations=1)
[11]: def cv_erode(image, structuring_element):
```

Applys OpenCV erosion.

```
Parameters:
    image (numpy.ndarray): The input image.
    structuring_element (numpy.ndarray): The structuring element must be

⇒square.

Returns:
    numpy.ndarray: The result image.

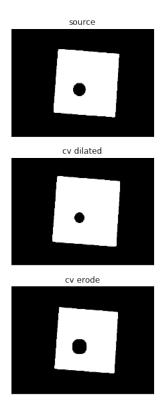
'''

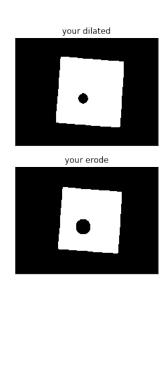
return cv2.erode(image, structuring_element, iterations=1)
```

Test of implementation.

```
[12]: image list = []
      image = cv2.imread(os.path.join('images', 'q4.png'), cv2.IMREAD_GRAYSCALE)
      image_list.append([image, 'source', 'img'])
      t1 = time.time()
      y dilate = your dilate(image, structuring element)
      t2 = time.time()
      print('time for your dilate: %f s' % (t2 - t1))
      t1 = time.time()
      c_dilate = cv_dilate(image, structuring_element)
      t2 = time.time()
      print('time for cv_dilate: %f s' % (t2 - t1))
      t1 = time.time()
      y_erode = your_erode(image, structuring_element)
      t2 = time.time()
      print('time for your_erode: %f s' % (t2 - t1))
      t1 = time.time()
      c_erode = cv_erode(image, structuring_element)
      t2 = time.time()
      print('time for cv_erode: %f s' % (t2 - t1))
      image_list.append([y_dilate, 'your dilated', 'img'])
      image_list.append([c_dilate, 'cv dilated', 'img'])
      image_list.append([y_erode, 'your erode', 'img'])
      image_list.append([c_erode, 'cv erode', 'img'])
      plotter(image_list, 3, 2, True, 20, 10, 'Q4')
```

time for your_dilate: 12.550562 s time for cv_dilate: 0.001987 s time for your_erode: 21.180147 s time for cv_erode: 0.002227 s





in terms of speed, the OpenCV implementation is much much faster than the regular naive manual approach. The naive approach has two nested for loops which increases the order of iterations. in terms of result both implementations use 1 iteration, and we can see the results are pretty similar.