EC7212 - Computer Vision and Image Processing

Reg. No: EG/2020/3829

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1 Introduction

This report presents the solutions for the Take Home Assignment 1 of the EC7212 Computer Vision and Image Processing course. The assignment involves implementing four image processing tasks using Python with OpenCV and NumPy libraries. The tasks include reducing intensity levels, performing spatial averaging, rotating images, and reducing spatial resolution.

The code is available in a GitHub repository: https://github.com/pgvda/computer-vision-assignment-1.

The following sections describe each task, the implementation approach, and the results.

2 Task 1: Reducing Intensity Levels

This task involves reducing the number of intensity levels in a grayscale image from 256 to a specified number of levels, which must be a power of 2. The image is first converted to grayscale and then quantized using the formula: (image // factor) * factor, where factor = 256 // levels.

3 Task 2: Spatial Averaging

This task applies spatial averaging to an image using 3x3, 10x10, and 20x20 kernels. The filter replaces each pixel with the average of its neighborhood to smooth the image.

4 Task 3: Image Rotation

This task rotates an image by 45 and 90 degrees. It uses a rotation matrix and resizes the output image to avoid cropping.

5 Task 4: Reducing Spatial Resolution

This task simulates reducing the spatial resolution by replacing each non-overlapping block (3x3, 5x5, 7x7) with its average pixel value.

6 Source Code And Output

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
from google.colab import files
import io
from PIL import Image

uploaded = files.upload()
```

```
import cv2
import matplotlib.pyplot as plt

image_path = 'input.jpg'

image = cv2.imread(image_path)
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

plt.imshow(image)
plt.axis('off')
plt.show()
```

```
def convert_img_to_grayscale(image):
    if len(image.shape) == 3:
        img = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
        return img
    else:
        return image
```

```
def reduce_spatial_resolution(image, block_size):
    if block_size < 1:
        raise ValueError("Block size must be a positive.")
    image =convert_img_to_grayscale(image)</pre>
```

```
# Get image dimensions
h, w = image.shape
# Initialize output image
output = np.zeros((h, w), dtype=np.uint8)

# Process each block
for i in range(0, h - block_size + 1, block_size):
    for j in range(0, w - block_size + 1, block_size):
        # Extract the block
        block = image[i:i+block_size, j:j+block_size]
        # Compute the average
        block_mean = np.mean(block).astype(np.uint8)
        # Assign the average to all pixels in the block
        output[i:i+block_size, j:j+block_size] = block_mean
```

```
def reduce_intensity_level(image, levels):
    if levels < 2 or levels > 256 or (levels & (levels - 1)) != 0:
        raise ValueError("Number of levels must be a power of 2
between 2 and 256.")

image = convert_img_to_grayscale(image)
    factor = 256 // levels
    reduced_img = (image // factor) * factor
    return reduced_img
```

```
def simple_spatial_average(image, kernal_size):
   image = convert_img_to_grayscale(image)
   averaged = cv2.blur(image, (kernal_size, kernal_size))
   return averaged.astype(np.uint8)
```

```
def rotate_img(image, angle):
   (h, w) = image.shape[:2]
   center = (w // 2, h // 2)

M = cv2.getRotationMatrix2D(center, angle, 1.0)

cos = np.abs(M[0, 0])
```

```
sin = np.abs(M[0, 1])
new_w = int((h * sin) + (w * cos))
new_h = int((h * cos) + (w * sin))

M[0, 2] += (new_w / 2) - center[0]
M[1, 2] += (new_h / 2) - center[1]

rotated_image = cv2.warpAffine(image, M, (new_w, new_h))
return rotated_image
```

```
def show result(original img, results dict, title = "Output
Images"):
   cols = 4
    rows = (n results + cols - 1) // cols
   plt.figure(figsize=(16, 4 * rows))
   plt.subplot(rows, cols, 1)
   plt.imshow(original img, cmap='gray')
   plt.title('Original Image')
   plt.axis('off')
    for idx, (name, image) in enumerate(results dict.items(), 2):
       plt.subplot(rows, cols, idx)
       plt.imshow(image, cmap='gray')
       plt.title(name)
       plt.axis('off')
    plt.suptitle(title, fontsize=20, y=1.02)
    plt.tight layout()
   plt.show()
```

```
def main():
    input_img = cv2.imread('input.jpg')

if input_img is None:
    raise FileNotFoundError("Input image not found. Please run
relavant code section")

original_gray = convert_img_to_grayscale(input_img)
```

```
#task : 1
  task 1 = {}
  intensity levels = [2, 4, 8]
  for levels in intensity levels:
   reduced = reduce intensity level(input img, levels)
   task 1[f'Intensity {levels} Levels'] = reduced
    cv2.imwrite(f'reduced intensity {levels} levels.jpg', reduced)
 show_result(original gray, task 1, title="Task 1: Intensity Level
  #task : 2
 task 2 = \{\}
 kernel sizes = [3, 10, 20]
  for k in kernel sizes:
     averaged = simple spatial average(input img, k)
      task 2[f'Spatial Average {k}x{k}'] = averaged
      cv2.imwrite(f'spatial average {k}x{k}.jpg', averaged)
 show result(original gray, task 2, title="Task 2: Spatial
Averaging")
 task 3 = \{\}
 angles = [45, 90]
 for angle in angles:
      rotated = rotate img(input img, angle)
      rotated gray = cv2.cvtColor(rotated, cv2.COLOR BGR2GRAY) if
len(rotated.shape) == 3 else rotated
      task 3[f'Rotated {angle} Degrees'] = rotated gray
      cv2.imwrite(f'rotated {angle} degrees.jpg', rotated)
 show result(original gray, task 3, title="Task 3: Image Rotation")
  task 4 = \{\}
 block sizes = [3, 5, 7]
 for b in block sizes:
      reduced res = reduce spatial resolution(input img, b)
      task 4[f'Reduced Resolution {b}x{b}'] = reduced res
      cv2.imwrite(f'reduced resolution {b}x{b}.jpg', reduced res)
 show result(original gray, task 4, title="Task 4: Spatial
Resolution Reduction")
```

Task 1: Intensity Level Reduction









Task 2: Spatial Averaging









Task 3: Image Rotation







7 Conclusion

The assignment successfully implemented four key image processing tasks using Python and OpenCV. The functions are modular, handle edge cases, and produce the desired visual outcomes. All code and output results are available in the following GitHub repository:

https://github.com/pgvda/computer-vision-assignment-1