

Exercise 2: Convolution & Edge Detection

Due date: 18.4.2022

The purpose of this exercise is to help you understand the concept of the convolution and edge detection by performing simple manipulations on images.

This exercise covers:

- Implementing convolution on 1D and 2D arrays
- Performing image derivative and image blurring
- Edge detection
- Hough Circles
- Bilateral filter

1 Convolution 10 pt

Write two functions that implement convolution of 1 1D discrete signal and 2D discrete signal. The two functions should have the following interfaces:

```
def conv1D(in_signal: np.ndarray, k_size: np.ndarray) -> np.ndarray:
    """
    Convolve a 1-D array with a given kernel
    :param in_signal: 1-D array
    :param k_size: 1-D array as a kernel
    :return: The convolved array
    """

def conv2D(in_image: np.ndarray, kernel: np.ndarray) -> np.ndarray:
```

```

"""
Convolve a 2-D array with a given kernel
:param in_image: 2D image
:param kernel: A kernel
:return: The convolved image
"""

```

The result of conv1D should match `np.convolve(signal, kernel, 'full')` ([link](#)) and conv2D should match `cv2.filter2D` ([link](#)) with option 'borderType'=cv2.BORDER_REPLICATE.

2 Image derivatives & blurring 20 pt

2.1 Derivatives 10 pt

Write a function that computes the magnitude and the direction of an image gradient. You should derive the image in each direction separately (rows and column) using simple convolution with $[1, 0, -1]^T$ and $[1, 0, -1]$ to get the two image derivatives. Next, use these derivative images to compute the magnitude and direction matrix and also the x and y derivatives.

```

def convDerivative(in_image: np.ndarray) -> (np.ndarray, np.ndarray):
    """
    Calculate gradient of an image
    :param in_image: Grayscale iamge
    :return: (directions, magnitude)
    """

```

Reminder:

$$Mag_G = ||G|| = \sqrt{I_x^2 + I_y^2} \quad (1)$$

$$Direction_G = \tan^{-1} \left(\frac{I_y}{I_x} \right) \quad (2)$$

2.2 Blurring: Bonus 10 pt

You should write two functions that performs image blurring using convolution between the image f and a Gaussian kernel g . The functions should have the following interface:

```

def blurImage1(in_image: np.ndarray, k_size: int) -> np.ndarray:

```

```

"""
Blur an image using a Gaussian kernel
:param in_image: Input image
:param k_size: Kernel size
:return: The Blurred image
"""

def blurImage2(in_image: np.ndarray, k_size: int) -> np.ndarray:
    """
    Blur an image using a Gaussian kernel using OpenCV built-in functions
    :param in_image: Input image
    :param k_size: Kernel size
    :return: The Blurred image
    """

```

blurImage1 should be fully implemented by you, using your own implementation of convolution (conv2D) and Gaussian kernel. blurImage2 should be implemented by using python's internal functions: filter2D and [getGaussianKernel](#).

Comments:

- In your implementation, the Gaussian kernel g should contain approximation of the Gaussian distribution using the binomial coefficients. A consequent 1D convolutions of $[1 \ 1]$ with itself is an elegant way for obtaining a row of the binomial coefficients. Explore how you can get a 2D Gaussian approximation using the 1D binomial coefficients. **Remember:**

$$\sum_{i,j} kernel_{i,j} = 1$$

- The border of the images should be padded same as in the 'Convolution' section (cv2.BORDER_REPLICATE).
- The size of the Gaussian' $kernelSize$, should always be an odd number.

3 Edge detection 10 pt

You should implement *edgeDetectionZeroCrossingLOG* OR *edgeDetectionZeroCrossingSimple*

- ```
def edgeDetectionZeroCrossingSimple(img: np.ndarray) -> np.ndarray:
 """
 Detecting edges usint "ZeroCrossing" method
 :param img: Input image
 :return: opencv solution, my implementation
 """
```
- ```
def edgeDetectionZeroCrossingLOG(img: np.ndarray) -> np.ndarray:
    """
    Detecting edges usint "ZeroCrossingLOG" method
    :param img: Input image
    :return: opencv solution, my implementation
    """
```

I is the intensity image and edgeImage is binary image (zero/one) with ones in the places the function identifies edges. Each function implements edge detections according to a different method. edgeImage1 is the edge image of your own implementation and edgeImage2 is the edge image returned by python's edge function with appropriate parameters. You can find the description of each of the methods at <https://docs.opencv.org/4.0.0/>.

For simple zero-crossing use a simple image like the 'codeMonkey', and for LoG zero-crossing try something more challenging like 'boxMan', adjust the image/Gaussian kernel size to get good results.

4 Hough Circles 30 pt

You should implement the Hough circles transform.

```
def houghCircle(img: np.ndarray, min_radius: int, max_radius: int) -> list:
    """
    Find Circles in an image using a Hough Transform algorithm extension
    To find Edges you can Use OpenCV function: cv2.Canny
    :param img: Input image
    :param min_radius: Minimum circle radius
    :param max_radius: Maximum circle radius
    :return: A list containing the detected circles,
             [(x,y,radius),(x,y,radius),...]
```

```
"""
```

I is the intensity image, min_radius, max_radius should be positive numbers and $min_radius < max_radius$. Use the Canny Edge detector as the edge detector, you can use the function: `cv.Canny()` ([link](#)). The functions should return a list of all the circles found, each circle will be represented by: (x,y,radius). Circle center x, Circle center y, Circle radius.

* This function is costly in run time, be sure to keep the min/maxRadius values close and the images small.

5 Bilateral filter 30 pt

You should implement the Bilateral filter, compare your implementation with OpenCV implementation `cv.bilateralFilter()` ([link](#)).

```
def bilateral_filter_implement(in_image: np.ndarray, k_size: int, sigma_color: float, sigma_sp
> (np.ndarray, np.ndarray):
    """
    :param in_image: input image
    :param k_size: Kernel size
    :param sigma_color: represents the filter sigma in the color space.
    :param sigma_space: represents the filter sigma in the coordinate.
    :return: OpenCV implementation, my implementation
    """
```

6 Important Comments

- Self-submission
- The input of all the above functions will be grayscale images.
- Your edges should be reasonable, but don't worry if they are not as good as OpenCV's.
- Don't waste your time on input validation.
- Do not have any plots in `ex2_utils.py`!

7 Submission

You should submit a zip file ,under the name "Ex2", containing:

- `ex2_utils.py` - This file will have all the functions above.
- `ex2_main.py` - This file will be the main file that executes all the functions, including your thresholds which gave you the best results. The program should print your ID at the beginning.
- All the images you used in `ex2_main.py`

8 Good Luck