### **Date Information**

• Due: 2022.11.08

Last Modified: 2022.11.08

# **Environment Requirement**

- python 3.0 or newer for f-strings f"Something {variable}".
- Another requirements are written in requirements.txt, just type pip install -r requirements.txt in the terminal.
  - o matplotlib
  - o numpy
  - opency-python == 4.5.5.62 (for auto-complete working on pycharm)
    - <a href="https://stackoverflow.com/questions/73174194/opencv-autocomplete-not-working-on-pycharm">https://stackoverflow.com/questions/73174194/opencv-autocomplete-not-working-on-pycharm</a>

### **Execution**

The main python code is main.py, type the following command and then you can run the program.

python main.py

There are some parameters in main.py.

Note that the program will always save all images in Img.

#### Warning: The program will always delete existing folder Img and recreate it.

- is\_show: Tell the program if you want to show the result on the screen or not.
  - True: Show the result on the screen.
  - False: Don't show. Save it only.
- save\_eps: Tell the program if you want to save the histograms by vector (.eps) or bitmap (.png) images.
  - True: Vector images (.eps)
  - False: Bitmap images (.png), default resolution is 600 dpi.
- dpi: The resolution for plotting histogram, default is 600 dpi.

All the methods is implemented in spatial\_image\_enhancement.py and import in main as sie.

# **Technical Description**

### Histogram

We set the bins to 256 because the grayscale intensity is [0, 255].

#### **Gamma Correction**

The power-law transformations, or Gamma correction, have the basic form:

 $s = cr^{\gamma}$ , where c and  $\gamma$  are positive constants.

Note that s and r are normalized to [0, 1].

Therefore, we first divide all value by 255, then do the gamma correction, and then transfer back to [0, 255] by multiplying all value by 255.

### **Histogram Equalization**

- 1. Count the intensity value for every pixel by using <code>np.bincount</code>.
- 2. Normalize the intensity value to get the probability mass function.
- 3. Calculate the cumulative distribution function by using np.cumsum.
- 4. Create the mapping table by the CDF transfer back to [0, 255].
- 5. Map all pixels to create the new image.

### **2D Convolution**

First of all, we have to apply the **cross correlation** to the kernel, which is **flipping the kernel matrix horizontally and vertically**.

Then, we need to know the final size of image.

$$n_{out} = \left\lfloor rac{n_{in} + 2p - k}{s} 
ight
floor + 1$$

, where

- $n_{in}$  is the number of input feature.
- $n_{out}$  is the number of output feature.
- *k* is the kernel size.
- *p* is the padding size. We perform zero-padding here.
- s is the stride step.

Next, we replace the inner portion of the padded image with the actual image

```
imagePadded[int(padding):int(-1 * padding), int(padding):int(-1 * padding)] = image
```

If padding = 1, then the effect will be like this.

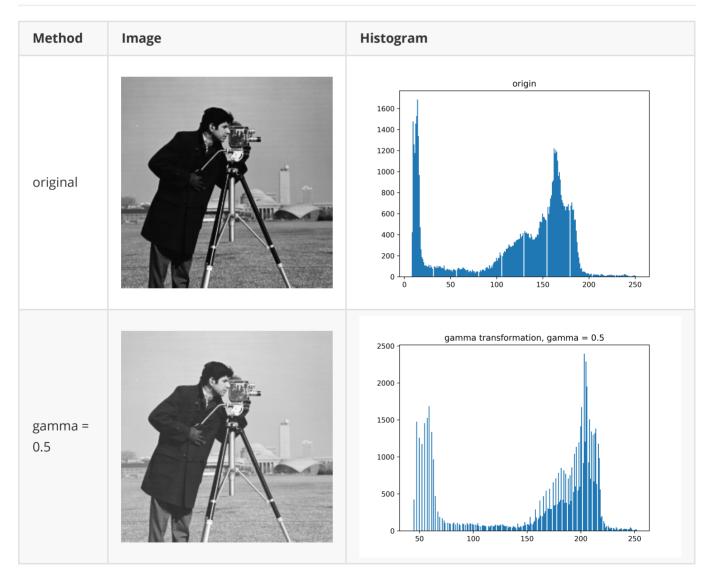
Finally, we iterate through the entire image, Let A be (x, y) and the filter size is 3 \* 3, then

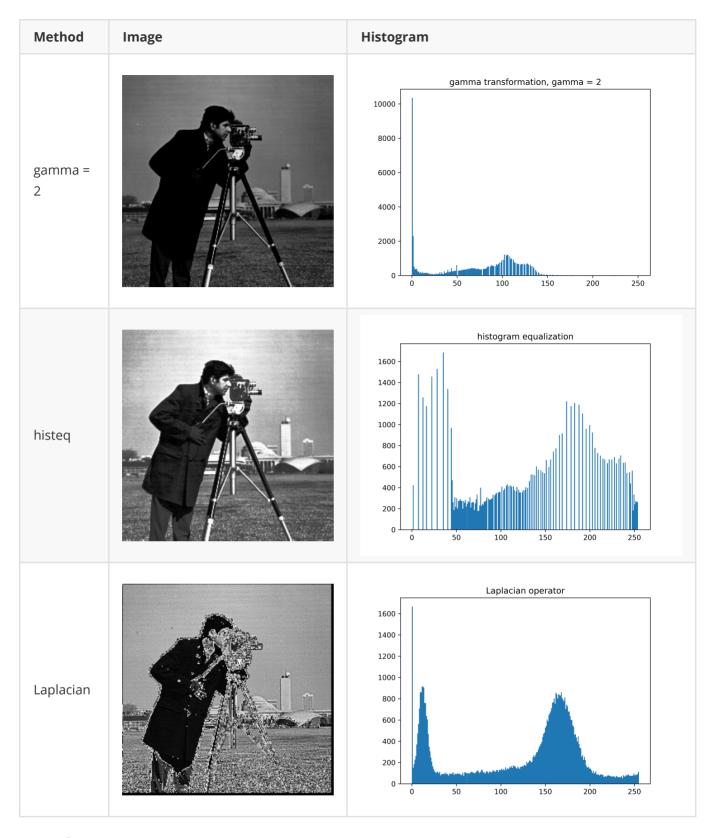
We only do convolution for specified strides

```
if x % strides == 0 and y % strides == 0
```

# **Experimental results**

#### **Cameraman**

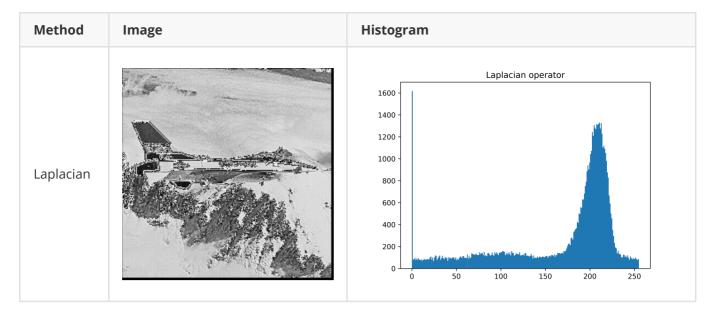




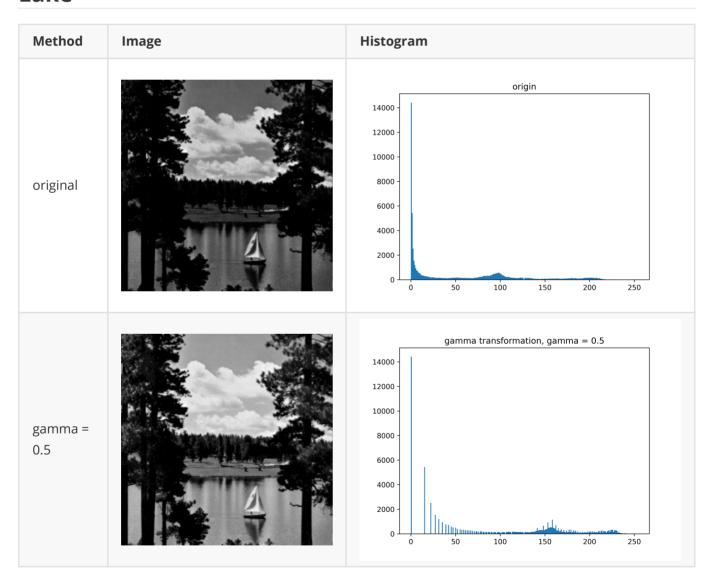
# Jetplane

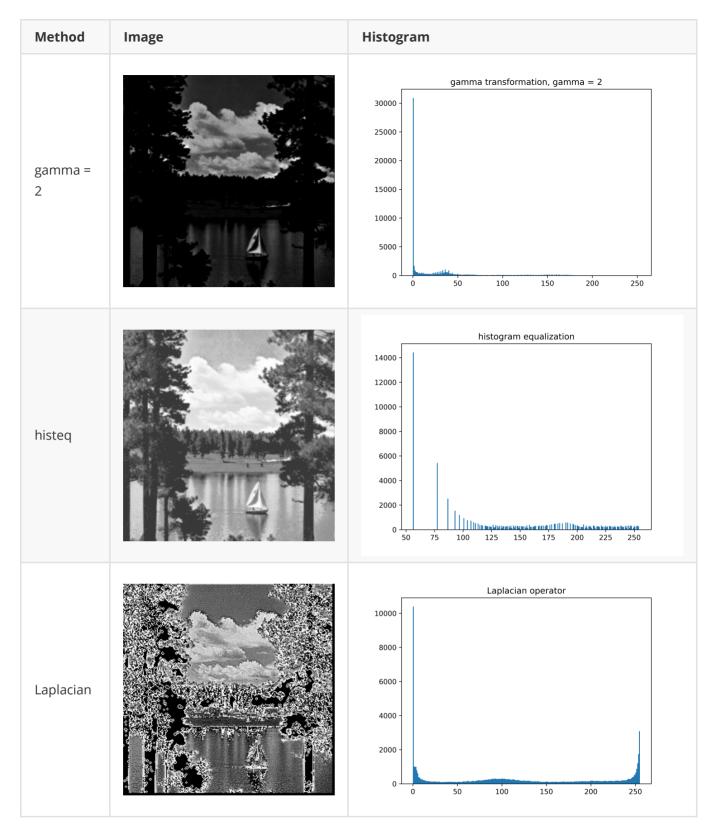
| Method | Image | Histogram |
|--------|-------|-----------|
|--------|-------|-----------|

| Method         | Image           | Histogram  |
|----------------|-----------------|--|
| original       | U.S.A.R. FORCE  | origin  2000 - 1750 - 1500 - 1000 - 750 - 500 - 250 - 0 - 50 100 150 200   |
| gamma =<br>0.5 | OSS US MA FORCE | gamma transformation, gamma = 0.5  4000 -  |
| gamma =<br>2   | USARFORCE       | gamma transformation, gamma = 2  2000 - 1750 - 1500 - 1250 - 1000 - 750 - 1000 - 750 - 1000 - 1250 - 1000 - |
| histeq         | U.S.AM FORCE    | histogram equalization  2000 -   |



## Lake

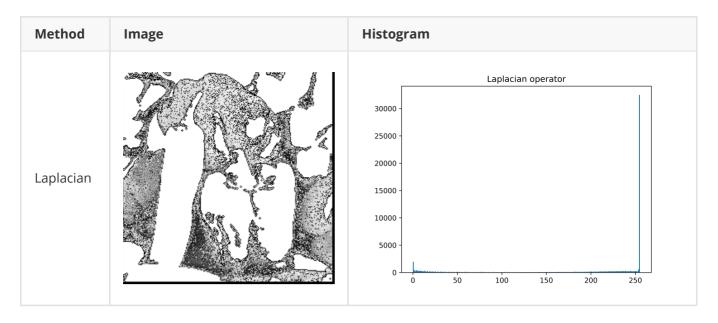




# **Peppers**

| Method | Image | Histogram |
|--------|-------|-----------|
|--------|-------|-----------|

| Method         | Image | Histogram  |
|----------------|-------|--|
| original       |       | origin  35000 - 30000 - 25000 - 20000 - 15000 - 10000 - 5000 - 0   |
| gamma =<br>0.5 |       | gamma transformation, gamma = 0.5  35000 - 25000 - 20000 - 15000 - 10000 - 5000 - 5000 - 0 - 500 - 100 150 200 250 |
| gamma =<br>2   |       | gamma transformation, gamma = 2  35000 - 30000 - 25000 - 15000 - 10000 - 5000 - 0                                  |
| histeq         |       | histogram equalization  35000 - 30000 - 25000 - 15000 - 10000 - 5000 - 0   |



## **Discussions**

Take Peppers for example, the picture have many white pixels and thus the histogram equalization performs worse than other pictures.

The Laplacian operator tends to detect too many edge points and thus the performance is worse.

### Reference

- <a href="https://geek-docs.com/opencv/opencv-examples/gamma-correction.html">https://geek-docs.com/opencv/opencv-examples/gamma-correction.html</a>
- <a href="https://levelup.gitconnected.com/introduction-to-histogram-equalization-for-digital-image-enhancement-4">https://levelup.gitconnected.com/introduction-to-histogram-equalization-for-digital-image-enhancement-4</a>
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- <a href="https://medium.com/analytics-vidhya/2d-convolution-using-python-numpy-43442ff5f381">https://medium.com/analytics-vidhya/2d-convolution-using-python-numpy-43442ff5f381</a>