

Computer Vision

Assignment N^o3

Theoretical Questions
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1 Convolution with Padding

1. Zero Padding:

As can be seen, with zero padding we add one row of zeros to each side. Because filter size is 3×3 and Image size is 4×4 .

For convolution, First we rotate the kernel for 180° .

$$\begin{array}{|c|c|c|c|c|c|} \hline 0 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 1 & 2 & 1 & 6 & 0 \\ \hline 0 & 7 & 1 & 1 & 1 & 0 \\ \hline 0 & 3 & 1 & 2 & 0 & 0 \\ \hline 0 & 1 & 4 & 0 & 2 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 \\ \hline \end{array} * \begin{array}{|c|c|c|} \hline 1 & 2 & 1 \\ \hline 1 & 0 & 1 \\ \hline 1 & 3 & 1 \\ \hline \end{array} = \begin{array}{|c|c|c|c|} \hline 24 & 13 & 13 & 5 \\ \hline 15 & 22 & 19 & 16 \\ \hline 23 & 28 & 11 & 11 \\ \hline 11 & 8 & 11 & 2 \\ \hline \end{array}$$

2. Border-Reflect Padding:

With Border-Reflect Padding the result for non-border pixels are the same as Zero Padding. But for borders the values are significantly greater because of the repetition of non-zero / large pixel values in the border.

$$\begin{array}{|c|c|c|c|c|c|} \hline 1 & 1 & 2 & 1 & 6 & 6 \\ \hline 1 & 1 & 2 & 1 & 6 & 6 \\ \hline 7 & 7 & 1 & 1 & 1 & 1 \\ \hline 3 & 3 & 1 & 2 & 0 & 0 \\ \hline 1 & 1 & 4 & 0 & 2 & 2 \\ \hline 1 & 1 & 4 & 0 & 2 & 2 \\ \hline \end{array} * \begin{array}{|c|c|c|} \hline 1 & 2 & 1 \\ \hline 1 & 0 & 1 \\ \hline 1 & 3 & 1 \\ \hline \end{array} = \begin{array}{|c|c|c|c|} \hline 37 & 19 & 23 & 31 \\ \hline 26 & 22 & 19 & 23 \\ \hline 34 & 28 & 11 & 14 \\ \hline 23 & 21 & 17 & 12 \\ \hline \end{array}$$

2 Kernels

1. Vertical Edge Detector:

This filter can detect vertical edges of the image. Because of semi-symmetric shape this filter will have a value close to zero regarding pixels with the same range of values, But in case of an edge, the pixel value difference is magnificent, so the filter will produce a high (higher than the normal range) value and makes the vertical edge become a lot more sharper.

The transpose of this filter can be used as a Horizontal Edge Detector.

2. Averaging Smoother Filter:

This filter is used to make the image more smooth using averaging mechanism. It performs the averaging (smoothing) horizontally, because of vertical shape ($3 * 1$).

The filter has a concentration on the central value (central value is 2) so that when convolving on the border pixels, it can preserve the value of these (border) pixels.

3 CLAHE

in my approach, I use the nearest transition function computed for the border pixels. I don't use any kind of padding.

For performing CLAHE on this image with a $3 * 3$ filter size, the image is partitioned into two parts (showed in two colors, amber and light cyan):

46	51	57	59
46	52	58	60
46	52	58	60

1. Clip Limit : 1

Computing the equalization for the left and right parts of the image using Clip Limit of 1:.

Left Part (Light cyan):

k	46	51	52	57	58
n_k	3	1	2	1	2
$n_k(Clipped)$	1.015	1.015	1.015	1.015	1.015
$\sum_{j=k}^k n_j$	1.71	2.79	3.81	4.89	5.9
$\sum_{j=k}^k \frac{n_j}{n}$	0.19	0.31	0.423	0.543	0.656
$(L - 1) \sum_{j=k}^k \frac{n_j}{n}$	48.69	79.24	108.02	138.56	167.34
\approx	49	79	108	139	167

Right Part (Amber):

k	51	52	57	58	59	60
n_k	1	2	1	2	1	2
$n_k(Clipped)$	1.011	1.011	1.011	1.011	1.011	1.011
$\sum_{j=k}^k n_j$	1.597	2.609	3.667	4.679	5.691	6.703
$\sum_{j=k}^k \frac{n_j}{n}$	0.177	0.289	0.407	0.519	0.632	0.744
$(L-1) \sum_{j=k}^k \frac{n_j}{n}$	45.266	73.932	103.925	132.591	161.256	189.921
\approx	45	74	104	133	161	190

The final image after applying transitions is:

49	79	104	161
49	108	133	190
49	108	133	190

2. Clip Limit : 2

Computing the equalization for the left and right parts of the image using Clip Limit of 2:.

Left Part (Light cyan):

k	46	51	52	57	58
n_k	3	1	2	1	2
$n_k(Clipped)$	2.003	1.003	2.003	1.003	2.003
$\sum_{j=k}^k n_j$	2.179	3.199	5.203	6.222	8.226
$\sum_{j=k}^k \frac{n_j}{n}$	0.242	0.355	0.578	0.691	0.914
$(L-1) \sum_{j=k}^k \frac{n_j}{n}$	61.757	90.644	147.421	176.308	233.085
\approx	62	91	147	176	233

Right Part (Amber):

k	51	52	57	58	59	60
n_k	1	2	1	2	1	2
$n_k(Clipped)$	1	2	1	2	1	2
$\sum_{j=k}^k n_j$	1	3	4	6	7	9
$\sum_{j=k}^k \frac{n_j}{n}$	0.111	0.333	0.444	0.666	0.777	1
$(L-1) \sum_{j=k}^k \frac{n_j}{n}$	28.33	85	113.33	170	198.33	255
\approx	28	85	113	170	198	255

The final image after applying transitions is:

62	91	113	198
62	147	170	255
62	147	170	255

3. OpenCV CLAHE Result

When using OpenCV CLAHE object, the result is different from the one calculated above.

It is identical for Clip Limit 1 and 2, as shown below:

128	191	160	191
128	255	192	255
128	223	176	223

The reason for this difference is that OpenCV uses BORDER_REFLECT_101¹ method for padding before applying CLAHE in contrast with above method that uses no padding.

In addition, for removing artifacts in tile borders, OpenCV uses Bilinear Interpolation².

The implementation and outputs of this section can be found on "CLAHE" section of HW3_PA_Report.pdf file.

¹Source: [OpenCV Source code](#)

²Source: [CLAHE Tutorial](#)