# Blur Filter

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# 1 Blurring Filter

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#### 1.1.1 Gaussian Method

The Gaussian method of blurring also called Gaussian smoothing, reduces noise in an image by assigning a weighted average of surrounding pixel values to each pixel in an image. It is often used as a pre-processing step such as prior to detecting edges to reduce the amount of noise/detail in an image.

```
Gaussian blurring works by convolving an image with a Gaussian kernel such as [[1, 4, 6, 4, 1], [2, 8, 12, 8, 2], [6, 24, 36, 24, 6], [2, 8, 12, 8, 2], [1, 4, 6, 4, 1]]
```

and then normalizing the result by the sum of the weights in the filter. The resulting value is thus a weighted average of surrounding pixels values and is assigned to the center pixel to reduce the gradients in the image. The kernel controls the amount of blurring: a larger kernel convolved over the image produces more blurring because it smooths out the pixel values over a larger area. Other kernels, such as a simple box

```
'[[1 1 1] [1 1 1] [1 1 1]]
```

can also be used for blurring.

The basic process is as follows:

- 1. Create a filter with larger filters resulting in more blurring.
- 2. Separate the image into three color channels
- 3. For each channel, iterate through the rows and columns of the image
- 4. At each step, element-wise multiply the filter times the patch
- 5. Take the weighted average of the patch and assign the pixel value to all pixels in the patch
- 6. Normalize the multiplication results by the sum of weights in the kernel and convert to integers between 0 and 255 for image plotting.
- 7. Visualize Images
- 8. Try different filters to adjust the level of blurring

```
from matplotlib.image import imread
# PIL for loading images
from PIL import Image
```

### 1.2 Implementation

```
In [4]: def apply_blur_filter(blur_filter, image_path):
            # Load in the image
            image = Image.open(image_path)
            # Crop to correct size
            image = image.crop(box=(0, 0,
                               int(image.size[0] / blur_filter.shape[0]) * blur_filter.shape[0]
                               int(image.size[1] / blur_filter.shape[1]) * blur_filter.shape[1]
            im_array = np.array(image)
            # Horizontal and vertical moves, using a stride of filter shape
            h_moves = int(im_array.shape[1] / blur_filter.shape[1])
            v_moves = int(im_array.shape[0] / blur_filter.shape[0])
            new_image = np.zeros(shape = im_array.shape)
            k = np.sum(blur_filter)
            # Iterate through 3 color channels
            for i in range(im_array.shape[2]):
                # Extract the layer and create a new layer to fill in
                layer = im_array[:, :, i]
                new_layer = np.zeros(shape = layer.shape, dtype='uint8')
                # Left and right bounds are determined by columns
                l_border = 0
                r_border = blur_filter.shape[1]
                # Iterate through the number of horizontal and vertical moves
                for h in range(h_moves):
                    # Top and bottom bounds are determined by rows
                    b_border = 0
                    t_border = blur_filter.shape[0]
                    for v in range(v_moves):
                        patch = layer[b_border:t_border, l_border:r_border]
                        # Take the element-wise product of the patch and the filter
                        product = np.multiply(patch, blur_filter)
```

```
product = np.sum(product) / k
                        new_layer[b_border:t_border, l_border:r_border] = product
                        b_border = t_border
                        t_border = t_border + blur_filter.shape[0]
                    l_border = r_border
                    r_border = r_border + blur_filter.shape[1]
                new_image[:, :, i] = 255 * ( (new_layer - np.min(new_layer)) /
                                            (np.max(new_layer) - np.min(new_layer)) )
            # Convert to correct type for plotting
           new_image = new_image.astype('uint8')
           plt.imshow(image); plt.title('Original Image'); plt.axis('off')
           plt.show()
           plt.imshow(new_image); plt.title('Blurred Image'); plt.axis('off')
           plt.show()
            return new_image
1.3 Results
In [5]: blur_filter = np.array([[1, 4, 6, 4, 1],
                                [2, 8, 12, 8, 2],
                                [6, 24, 36, 24, 6],
                                [2, 8, 12, 8, 2],
                                [1, 4, 6, 4, 1]])
        blurred_image = apply_blur_filter(blur_filter, 'images/president-barack-obama.jpg')
```

# Find the weighted average of the patch

Original Image



Blurred Image



blurred\_image = apply\_blur\_filter(blur\_filter, 'images/president-barack-obama.jpg')

Original Image



Blurred Image



blurred\_image = apply\_blur\_filter(blur\_filter, 'images/president-barack-obama.jpg')

Original Image



Blurred Image



blurred\_image = apply\_blur\_filter(blur\_filter, 'images/mountains.jpg')

Original Image



Blurred Image



Original Image



Blurred Image



## 1.4 Conclusions

We see that by changing the parameters of the blur filter we can control the amount of smoothing in the image. A blur filter is a useful pre-processing step because it reduces the amount of noise in an image which can be helpful for processes such as edge detection and feature recognition.