Image filtering



Images by Pawan Sinha

What is an image?

We can think of an **image** as a function, f, from R^2 to R:

- *f*(*x*, *y*) gives the **intensity** at position (*x*, *y*)
- Realistically, we expect the image only to be defined over a rectangle, with a finite range:
 - $f: [a,b] \mathbf{x}[c,d] \rightarrow [0,1]$

A color image is just three functions pasted together. We can write this as a "vector-valued" function:

$$f(x, y) = \begin{bmatrix} r(x, y) \\ g(x, y) \\ b(x, y) \end{bmatrix}$$

Images as functions









What is a digital image?

In computer vision we usually operate on **digital** (**discrete**) images:

- **Sample** the 2D space on a regular grid
- Quantize each sample (round to nearest integer)

If our samples are Δ apart, we can write this as:

 $f[i, j] = \text{Quantize} \{ f(i \Delta, j \Delta) \}$

The image can now be represented as a matrix of integer values



Image processing

An **image processing** operation typically defines a new image g in terms of an existing image f.

We can transform either the domain or the range of f.

Range transformation:

$$g(x,y) = t(f(x,y))$$

What's kinds of operations can this perform?

Image processing

Some operations preserve the range but change the domain of f:

$$g(x,y) = f(t_x(x,y), t_y(x,y))$$

What kinds of operations can this perform?

Filtering Operations Use Masks

- Masks operate on a neighborhood of pixels.
- A mask of coefficients is centered on a pixel.
- The mask coefficients are multiplied by the pixel values in its neighborhood and the products are summed.
- The result goes into the corresponding pixel position in the output image.

Input Image

1/9 1/9 1/9 1/9 1/9 1/9 1/9 1/9 1/9

3x3 Mask



Output Image

Noise

Image processing is useful for noise reduction...



Original



Salt and pepper noise



Impulse noise

Gaussian noise

Common types of noise:

- Salt and pepper noise: contains random occurrences of black and white pixels
- Impulse noise: contains random occurrences of white pixels
- Gaussian noise: variations in intensity drawn from a Gaussian normal distribution

Practical noise reduction

How can we "smooth" away noise in a single image?

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	100	130	110	120	110	0	0
0	0	0	110	90	100	90	100	0	0
0	0	0	130	100	90	130	110	0	0
0	0	0	120	100	130	110	120	0	0
0	0	0	90	110	80	120	100	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

Mean filtering

F[x, y]

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0



G[x, y]





Mean filtering

L

F[x,	y]
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0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

0	10	20	30	30	30	20	10	
0	20	40	60	60	60	40	20	
0	30	60	90	90	90	60	30	
0	30	50	80	80	90	60	30	
0	30	50	80	80	90	60	30	
0	20	30	50	50	60	40	20	
10	20	30	30	30	30	20	10	
10	10	10	0	0	0	0	0	

I

G[x, y]

Effect of mean filters



3x3

5x5

7x7

Cross-correlation filtering

Let's write this down as an equation. Assume the averaging window is (2k+1)x(2k+1):

$$G[i,j] = \frac{1}{(2k+1)^2} \sum_{u=-k}^{k} \sum_{v=-k}^{k} F[i+u,j+v]$$

We can generalize this idea by allowing different weights for different neighboring pixels:

$$G[i,j] = \sum_{u=-k}^{k} \sum_{v=-k}^{k} H[u,v]F[i+u,j+v]$$

This is called a **cross-correlation** operation and written:

$$G = H \otimes F$$

H is called the "filter," "kernel," or "mask."

The above allows negative filter indices. When you implement need to use: H[u+k,v+k] instead of H[u,v]

Mean kernel

What's the kernel for a 3x3 mean filter?

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

H[u, v]

F[x, y]

Gaussian Filtering

A Gaussian kernel gives less weight to pixels further from the center of the window

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

1	1	2	1
<u>т</u>	2	4	2
16	1	2	1

H[u, v]

F[x, y]

This kernel is an approximation of a Gaussian function: $1 \quad u^2 + v^2$



Mean vs. Gaussian filtering



Convolution

A **convolution** operation is a cross-correlation where the filter is flipped both horizontally and vertically before being applied to the image:

$$G[i,j] = \sum_{u=-k}^{k} \sum_{v=-k}^{k} H[u,v]F[i-u,j-v]$$

It is written:
$$G = H \star F$$

Suppose H is a Gaussian or mean kernel. How does convolution differ from cross-correlation?

Median filters

A **Median Filter** operates over a window by selecting the median intensity in the window.

What advantage does a median filter have over a mean filter?

Is a median filter a kind of convolution?

Comparison: salt and pepper noise



Comparison: Gaussian noise

Gaussian Median Mean 3x3 5x5 7x7