



# Reading Forsyth & Ponce, chapter 7

What is an image?

# Images as functions

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:
  - $f: [a,b] \times [c,d] \to [0,1]$

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

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

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

$$b(x, y)$$

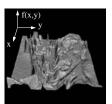
5

# Images as functions









6

# 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  $\Delta$  apart, we can write this as:

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

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

7

# 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?

9

# Image processing

Still other operations operate on both the domain and the range of f.

10

### **Noise**

Image processing is useful for noise reduction...



Salt and





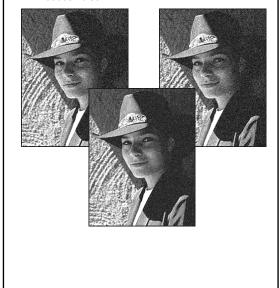
Common types of noise:

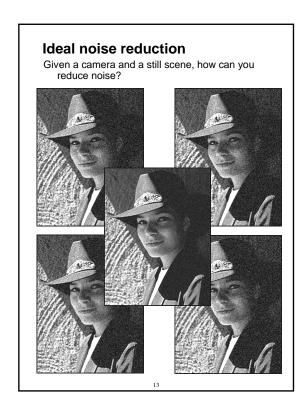
- Salt and pepper noise: contains random occurrences of black and white pixels
- Impulse noise: contains random occurrences of white pixels
- Gaussian noise: variations in intensity drawn from a Gaussian normal distribution

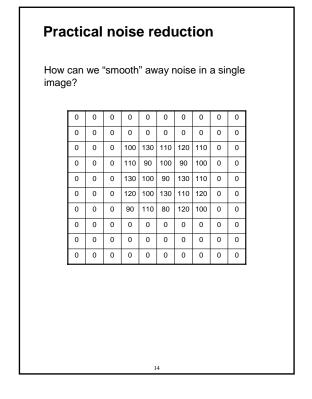
11

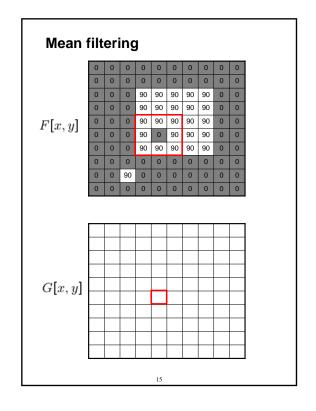
### Ideal noise reduction

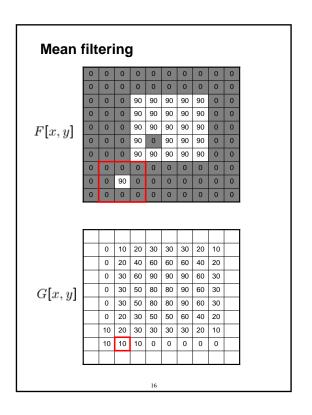
Given a camera and a still scene, how can you reduce noise?



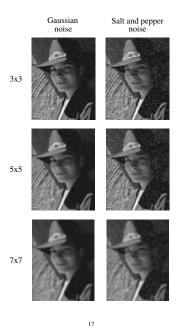








### Effect of mean filters



# **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]

18

### Mean kernel

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

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



# **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

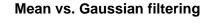


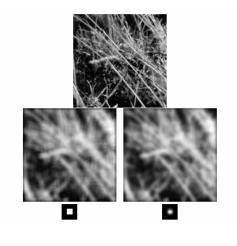
This kernel is an approximation of a Gaussian function:

$$h(u, v) = \frac{1}{2\pi\sigma^2} e^{-\frac{u^2 + v^2}{\sigma^2}}$$



What happens if you increase  $\sigma$ ?





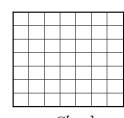
21

# Filtering an impulse





0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0



G[x, y]

22

### 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?

Suppose F is an impulse function (previous slide) What will G look like?

23

### **Continuous Filters**

We can also apply filters to continuous images.

In the case of cross correlation:  $g = h \otimes f$ 

$$g(x,y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u,v) f(x+u,y+v) du dv$$

In the case of convolution:  $g = h \star f$ 

$$g(x,y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u,v) f(x-u,y-v) du dv$$

Note that the image and filter are infinite.

# **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?

