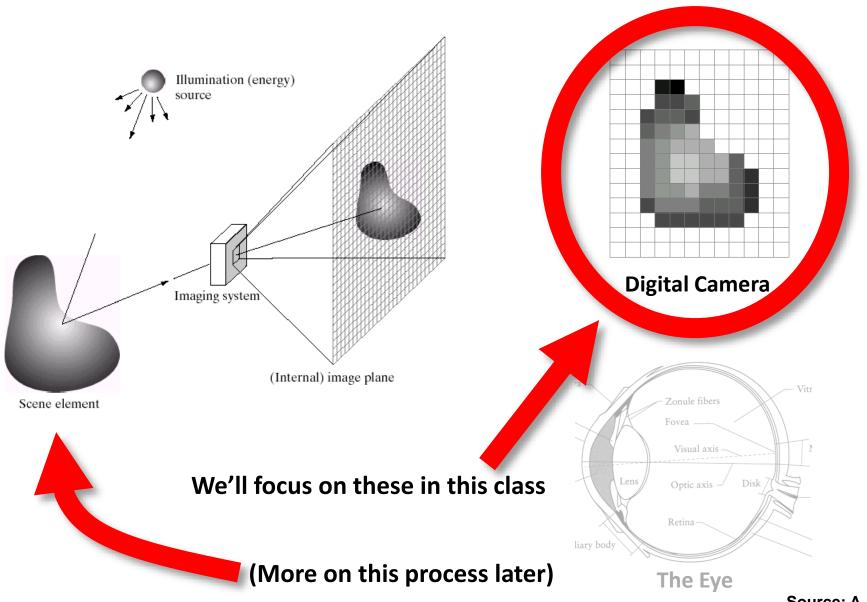
CS 231

Images and image filtering

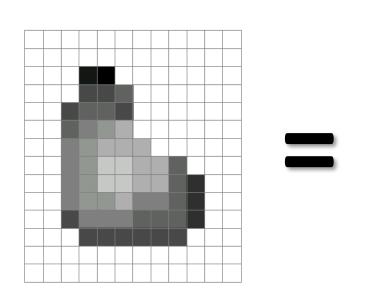
Slide credit: Cornell Univ. - CS4670





Source: A. Efros

A grid (matrix) of intensity values



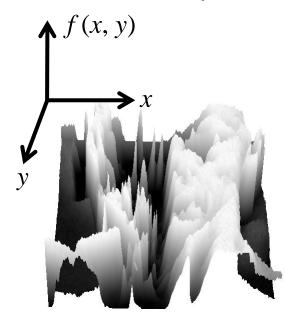
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 20 | 0 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 75 | 75 | 75 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 75 | 95 | 95 | 75 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 96 | 127 | 145 | 175 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 175 | 175 | 175 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 200 | 200 | 175 | 175 | 95 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 200 | 200 | 175 | 175 | 95 | 47 | 255 | 255 |
| | | | | | | | | | | | |
| 255 | 255 | 127 | 145 | 145 | 175 | 127 | 127 | 95 | 47 | 255 | 255 |
| 255 | 255 | 74 | 127 | 127 | 127 | 95 | 95 | 95 | 47 | 255 | 255 |
| 255 | 255 | 255 | 74 | 74 | 74 | 74 | 74 | 74 | 255 | 255 | 255 |
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |

(common to use one byte per value: 0 = black, 255 = white)

- We can think of a (grayscale) image as a function, f, from R² to R:
 - -f(x,y) gives the **intensity** at position (x,y)



snoop

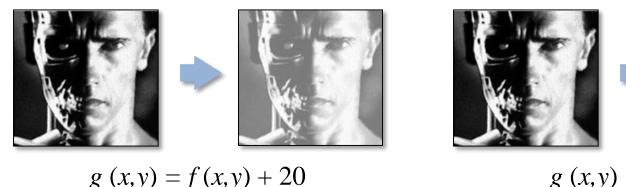


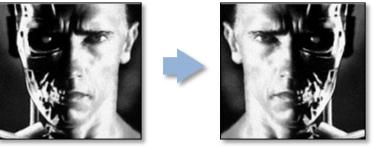
3D view

A digital image is a discrete (sampled, quantized) version of this function

Image transformations

 As with any function, we can apply operators to an image





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

 We'll talk about a special kind of operator, convolution (linear filtering)

Question: Noise reduction

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



Take lots of images and average them!

What's the next best thing?

Image filtering

 Modify the pixels in an image based on some function of a local neighborhood of each pixel

| 10 | 5 | 3 | | |
|----|---|---|--|--|
| 4 | 5 | 1 | | |
| 1 | 1 | 7 | | |



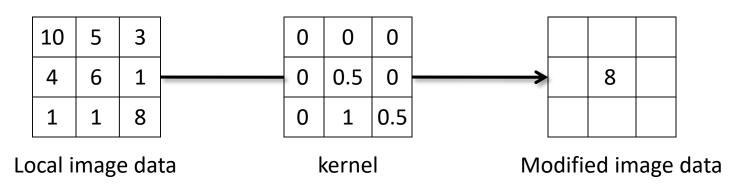




Modified image data

Linear filtering

- One simple version: linear filtering (cross-correlation, convolution)
 - Replace each pixel by a linear combination (a weighted sum) of its neighbors
- The prescription for the linear combination is called the "kernel" (or "mask", "filter")



Cross-correlation

Let F be the image, H be the kernel (of size $2k+1 \times 2k+1$), and G be the output image

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

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

$$G = H \otimes F$$

 Can think of as a "dot product" between local neighborhood and kernel for each pixel

Convolution

 Same as cross-correlation, except that the kernel is "flipped" (horizontally and vertically)

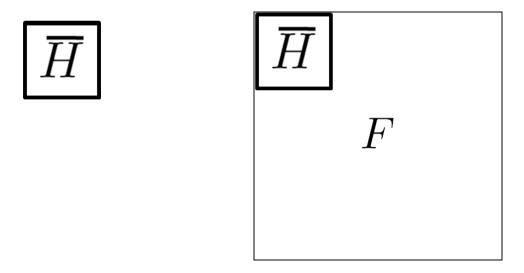
$$G[i,j] = \sum_{u=-k}^{k} \sum_{v=-k}^{k} H[u,v]F[i-u,j-v]$$

This is called a **convolution** operation:

$$G = H * F$$

Convolution is commutative and associative

Convolution

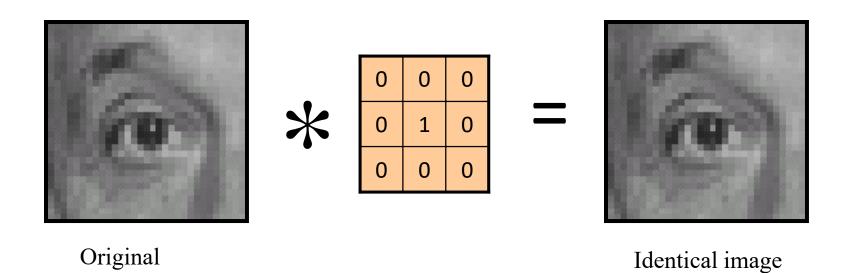


Mean filtering

| | 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 |
| H | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 11 | 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 | |
| | | | | | | | | |

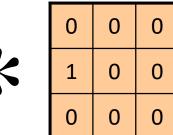
(



Source: D. Lowe



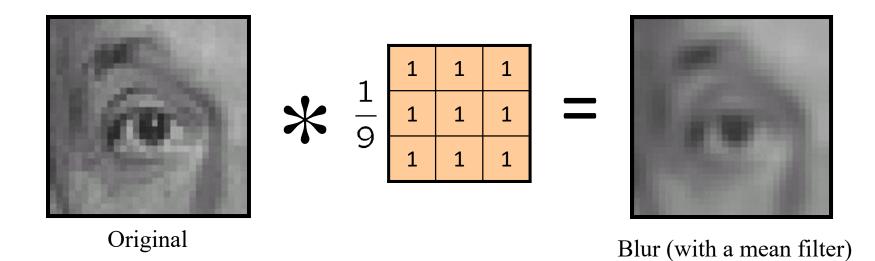




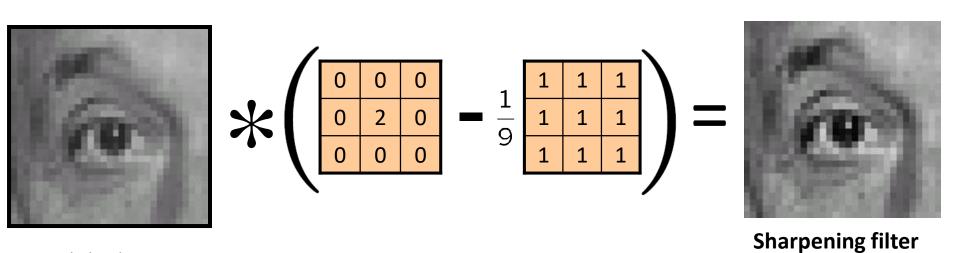
0 0 =



Shifted left By 1 pixel



Source: D. Lowe

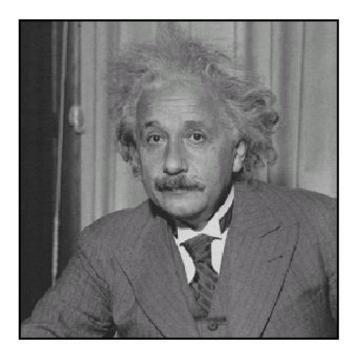


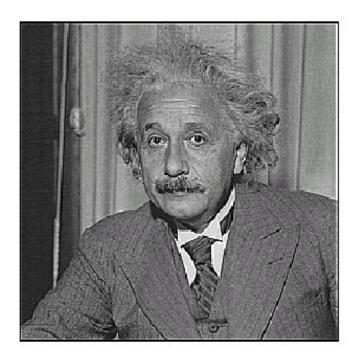
Original

Source: D. Lowe

(accentuates edges)

Sharpening

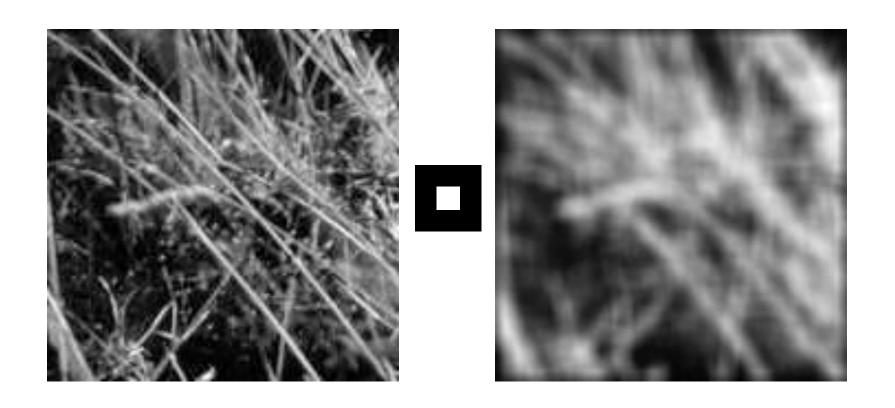




before after

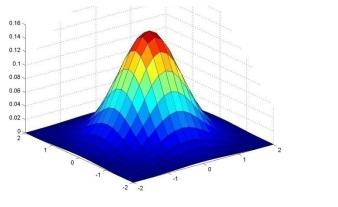
Source: D. Lowe

Smoothing with box filter revisited



Source: D. Forsyth

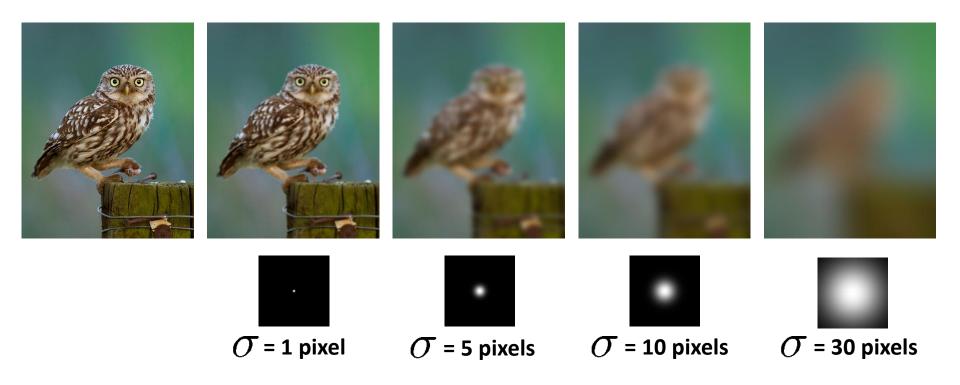
Gaussian Kernel



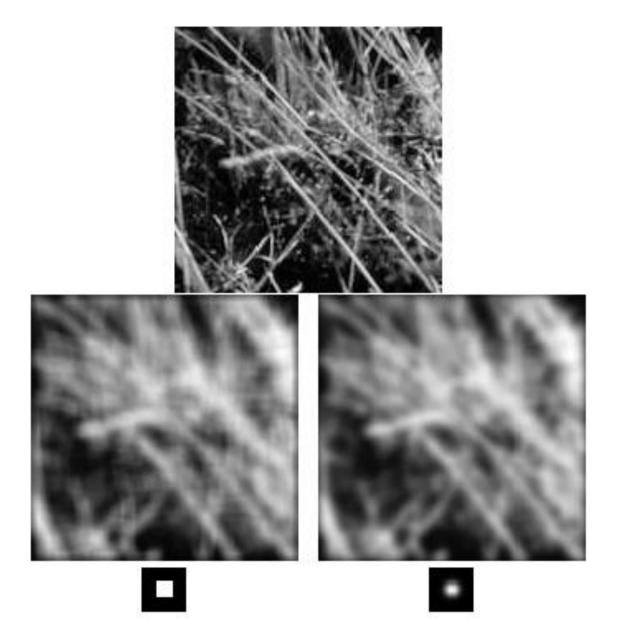
$$(m^2 + a^2)$$

$$G_{\sigma} = \frac{1}{2\pi\sigma^2} e^{-\frac{(x^2+y^2)}{2\sigma^2}}$$

Gaussian filters

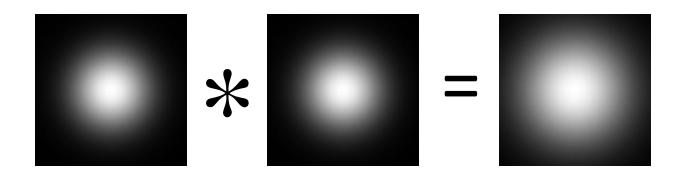


Mean vs. Gaussian filtering



Gaussian filter

- Removes "high-frequency" components from the image (low-pass filter)
- Convolution with self is another Gaussian



– Convolving twice with Gaussian kernel of width σ = convolving once with kernel of width $\sigma\sqrt{2}$

Sharpening revisited

What does blurring take away?







Let's add it back:

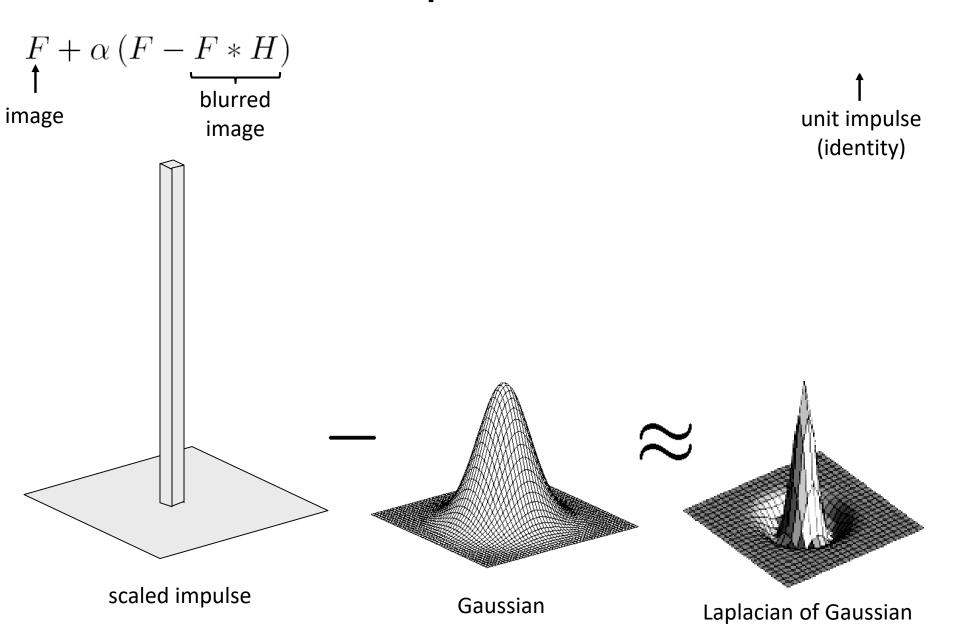




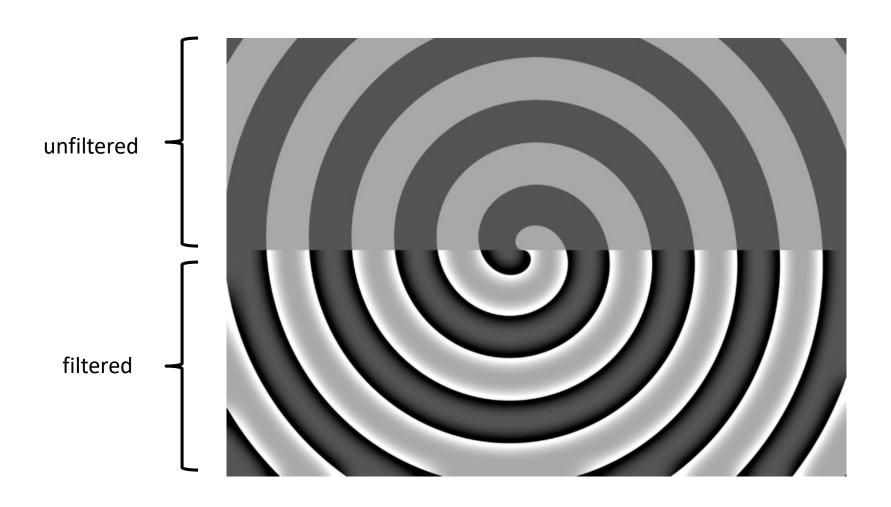


Source: S. Lazebnik

Sharpen filter



Sharpen filter



"Optical" Convolution

Camera shake



Source: Fergus, et al. "Removing Camera Shake from a Single Photograph", SIGGRAPH 2006

Bokeh: Blur in out-of-focus regions of an image.



Source: http://lullaby.homepage.dk/diy-camera/bokeh.html