

# Edge detection

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[Winter in Kraków photographed by Marcin Ryczek](#)

# Edge detection

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- **Goal:** Identify sudden changes (discontinuities) in an image
  - Intuitively, most semantic and shape information from the image can be encoded in the edges
  - More compact than pixels
- **Ideal:** artist's line drawing (but artist is also using object-level knowledge)

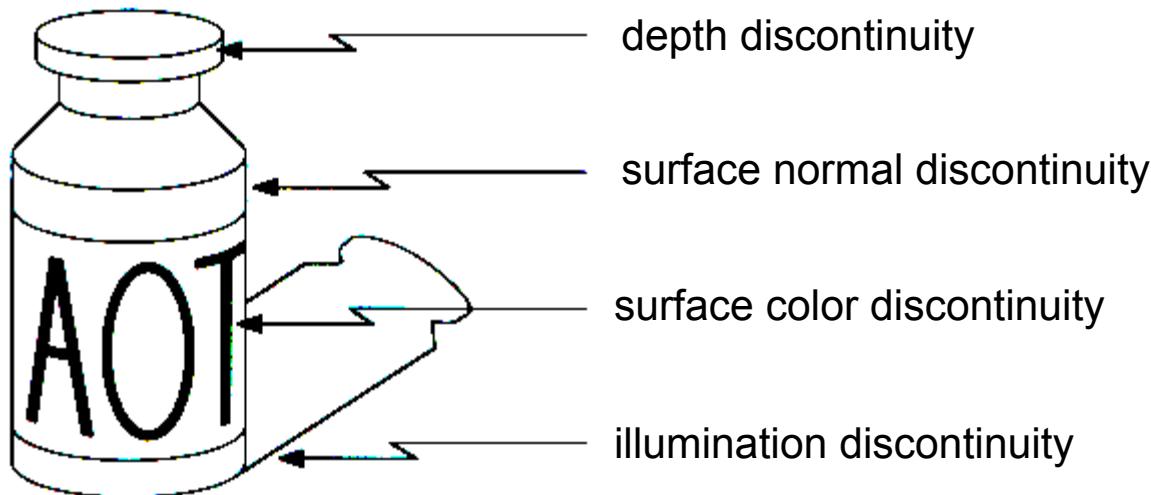


Source: D. Lowe

# Origin of edges

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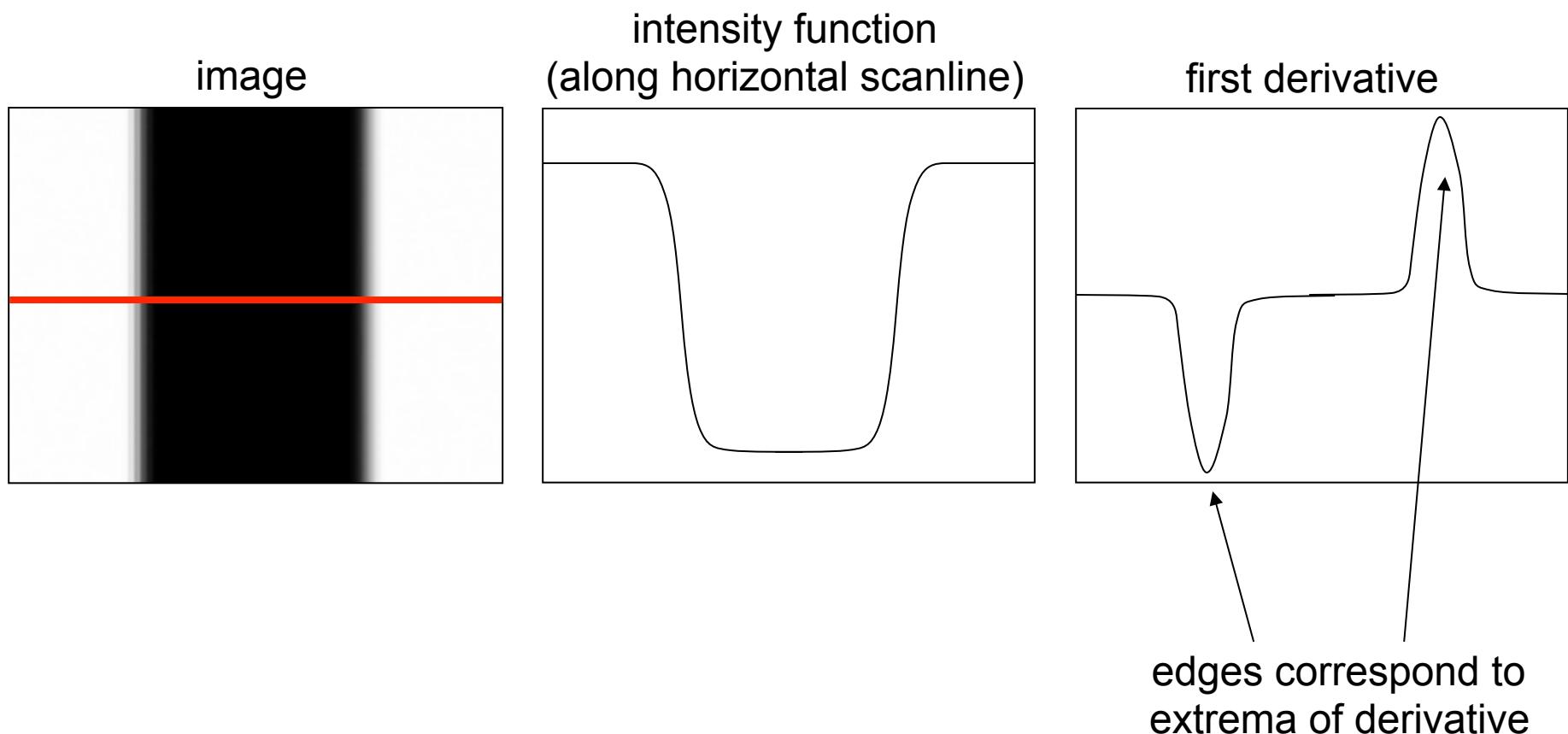
Edges are caused by a variety of factors:



# Characterizing edges

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- An edge is a place of rapid change in the image intensity function



# Derivatives with convolution

For 2D function  $f(x,y)$ , the partial derivative is:

$$\frac{\partial f(x, y)}{\partial x} = \lim_{\varepsilon \rightarrow 0} \frac{f(x + \varepsilon, y) - f(x, y)}{\varepsilon}$$

For discrete data, we can approximate using finite differences:

$$\frac{\partial f(x, y)}{\partial x} \approx \frac{f(x + 1, y) - f(x, y)}{1}$$

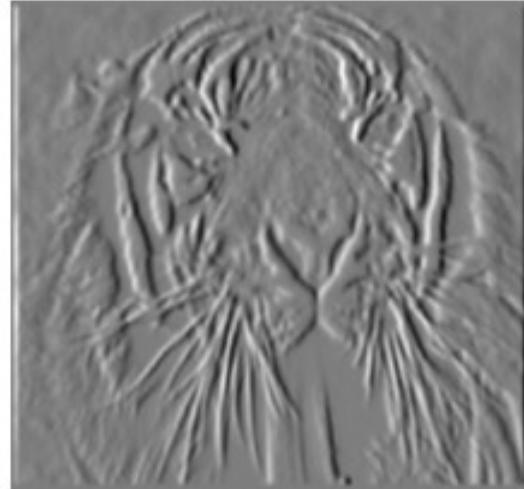
To implement above as convolution, what would be the associated filter?

# Partial derivatives of an image



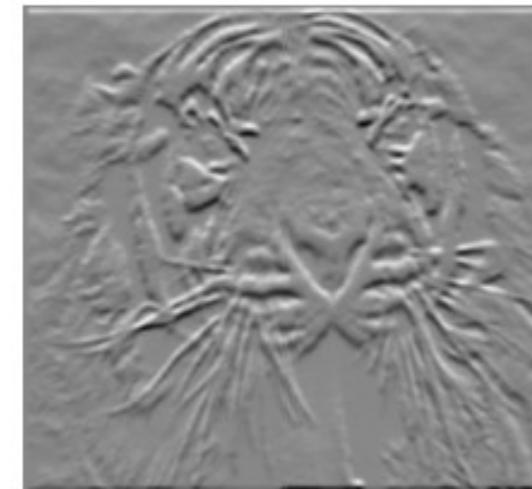
$$\frac{\partial f(x, y)}{\partial x}$$

-1	1
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$$\frac{\partial f(x, y)}{\partial y}$$

-1	1
1	-1



Which shows changes with respect to x?

# Finite difference filters

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Other approximations of derivative filters exist:

**Prewitt:**  $M_x = \begin{bmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{bmatrix}$  ;  $M_y = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ -1 & -1 & -1 \end{bmatrix}$

**Sobel:**  $M_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$  ;  $M_y = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$

**Roberts:**  $M_x = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$  ;  $M_y = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$

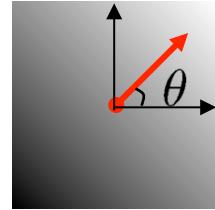
# Image gradient

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The gradient of an image:  $\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$


$$\nabla f = \left[ \frac{\partial f}{\partial x}, 0 \right]$$


$$\nabla f = \left[ 0, \frac{\partial f}{\partial y} \right]$$


$$\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

The gradient points in the direction of most rapid increase in intensity

- How does this direction relate to the direction of the edge?

The gradient direction is given by  $\theta = \tan^{-1} \left( \frac{\partial f}{\partial y} / \frac{\partial f}{\partial x} \right)$

The *edge strength* is given by the gradient magnitude

$$\|\nabla f\| = \sqrt{\left( \frac{\partial f}{\partial x} \right)^2 + \left( \frac{\partial f}{\partial y} \right)^2}$$

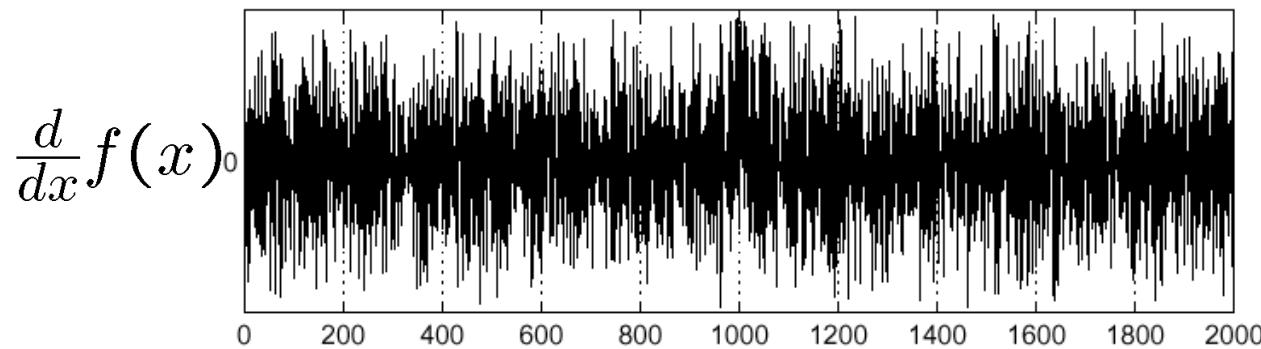
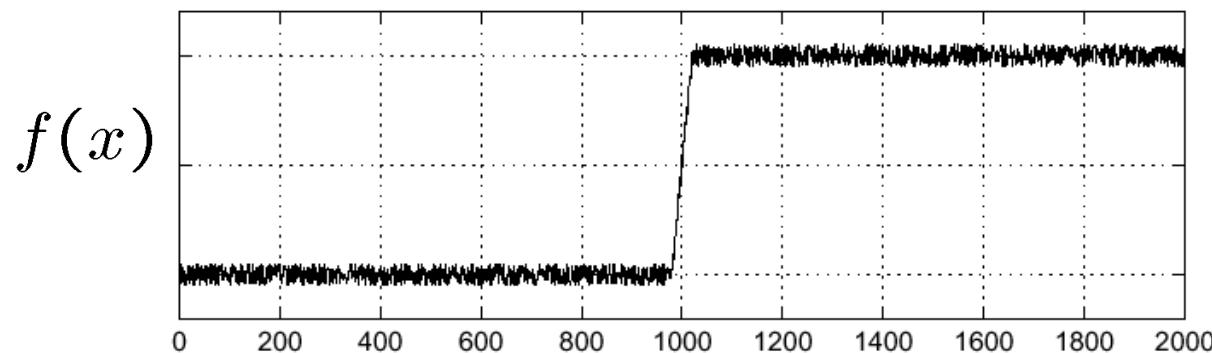
Source: Steve Seitz

# Effects of noise

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Consider a single row or column of the image

- Plotting intensity as a function of position gives a signal

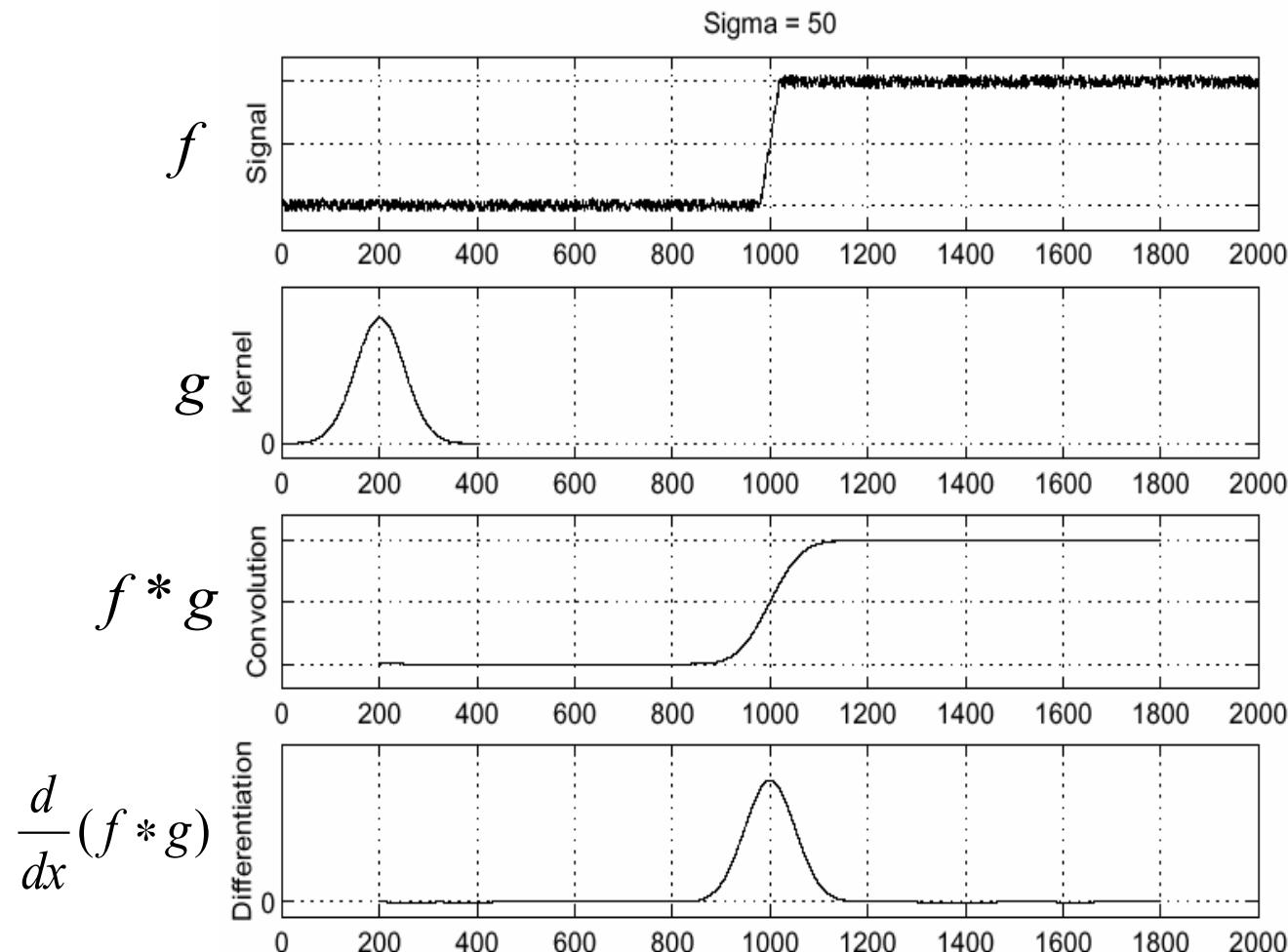


Where is the edge?

Source: S. Seitz

# Solution: smooth first

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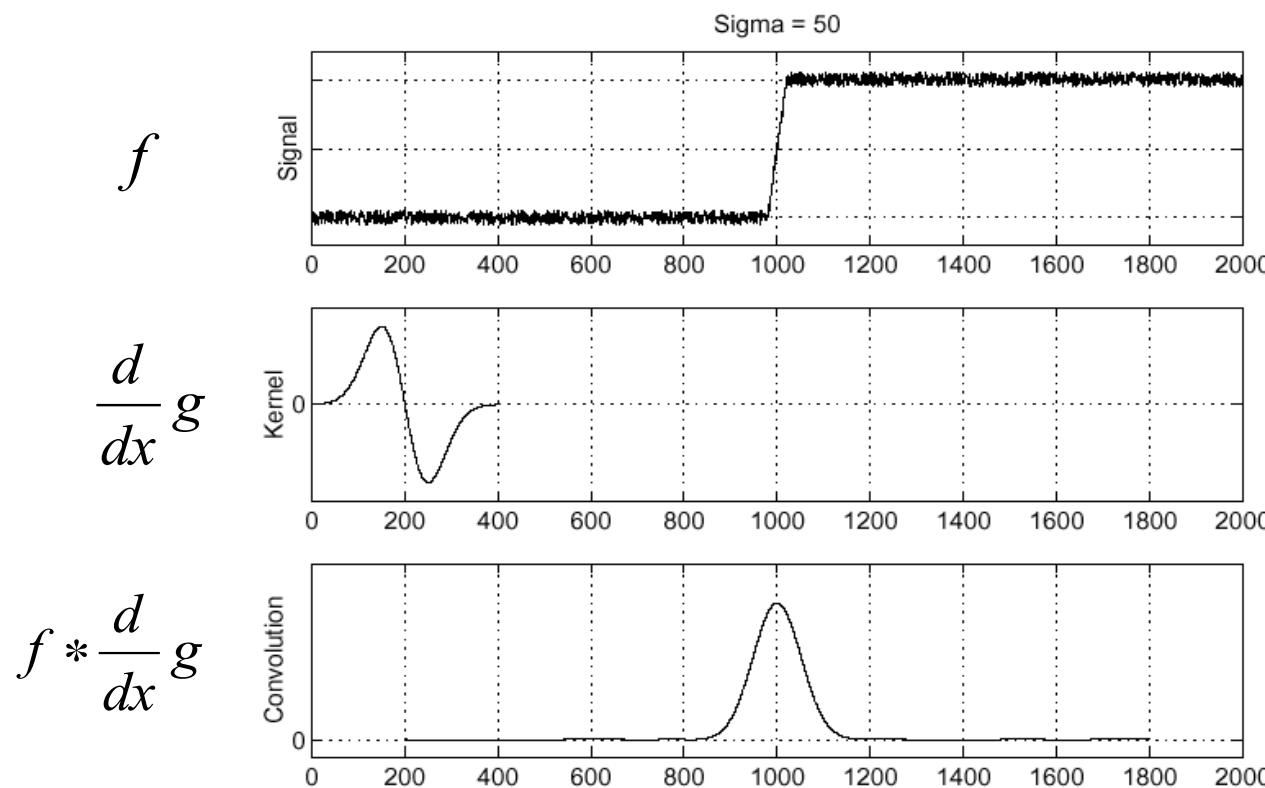
- To find edges, look for peaks in  $\frac{d}{dx}(f * g)$

Source: S. Seitz

# Derivative theorem of convolution

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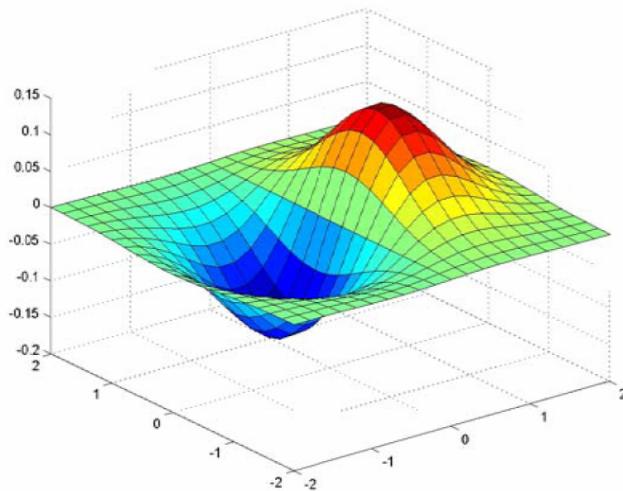
- Differentiation is convolution, and convolution is associative:  $\frac{d}{dx}(f * g) = f * \frac{d}{dx}g$
- This saves us one operation:



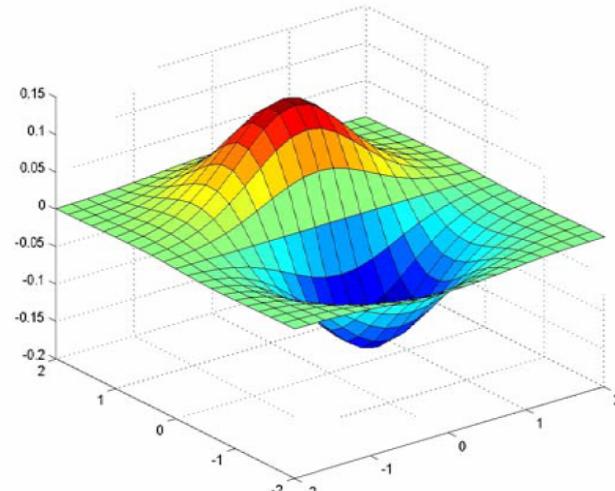
Source: S. Seitz

# Derivative of Gaussian filter

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x-direction

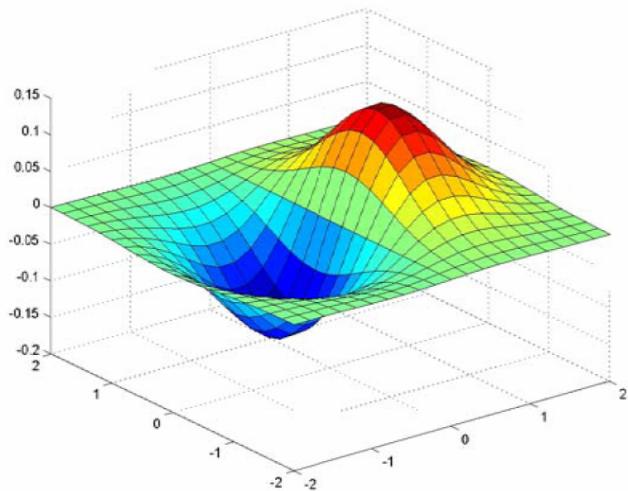


y-direction

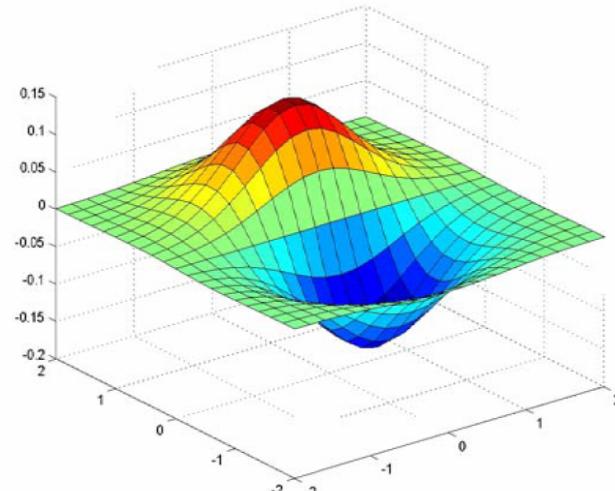
Are these filters separable?

# Derivative of Gaussian filter

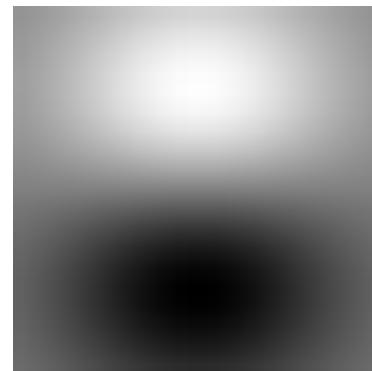
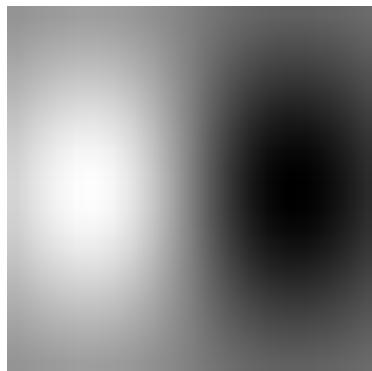
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x-direction



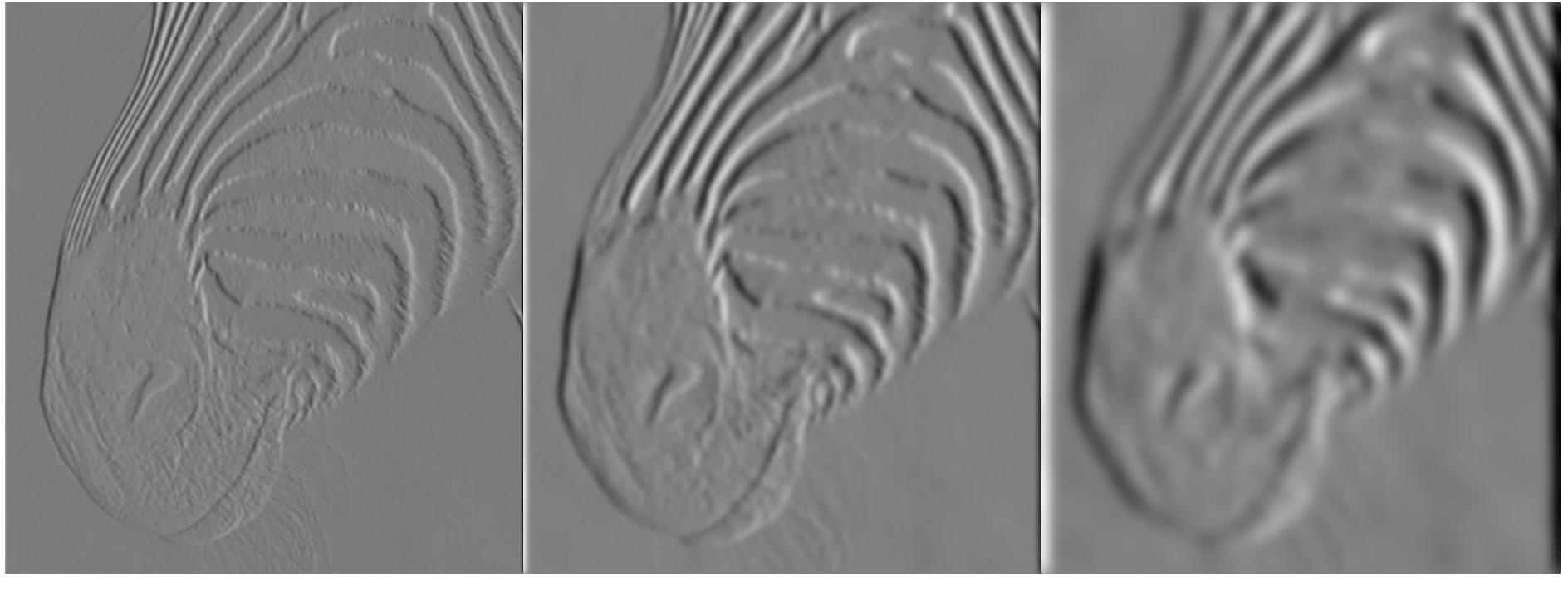
y-direction



Which one finds horizontal/vertical edges?

# Scale of Gaussian derivative filter

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1 pixel

3 pixels

7 pixels

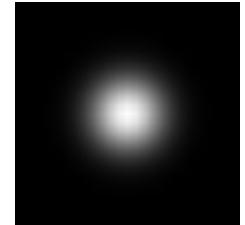
Smoothed derivative removes noise, but blurs edge. Also finds edges at different “scales”

Source: D. Forsyth

# Review: Smoothing vs. derivative filters

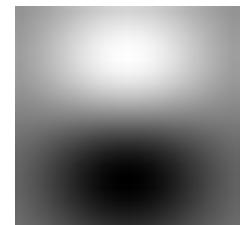
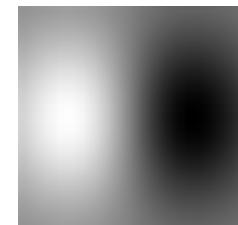
## Smoothing filters

- Gaussian: remove “high-frequency” components; “low-pass” filter
- Can the values of a smoothing filter be negative?
- What should the values sum to?
  - **One:** constant regions are not affected by the filter



## Derivative filters

- Derivatives of Gaussian
- Can the values of a derivative filter be negative?
- What should the values sum to?
  - **Zero:** no response in constant regions
- High absolute value at points of high contrast



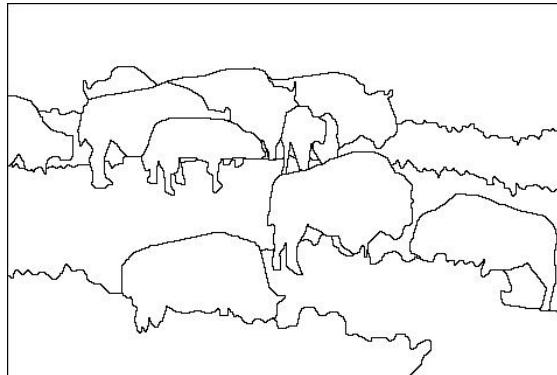
# Edge detection is just the beginning...

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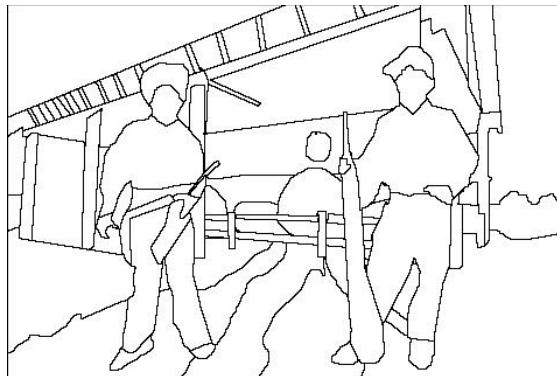
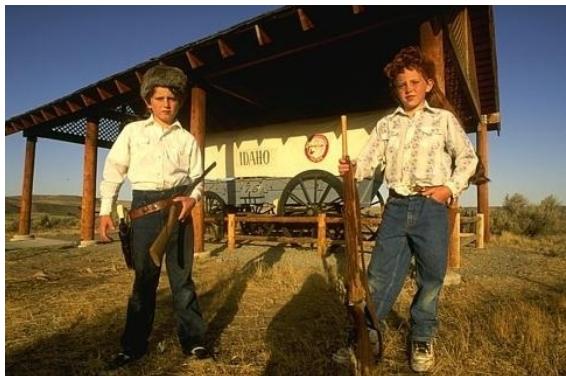
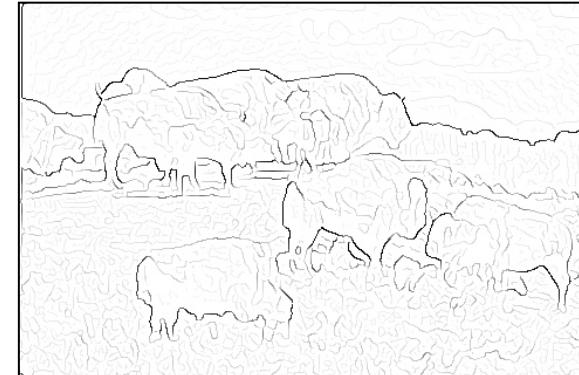
image



human segmentation



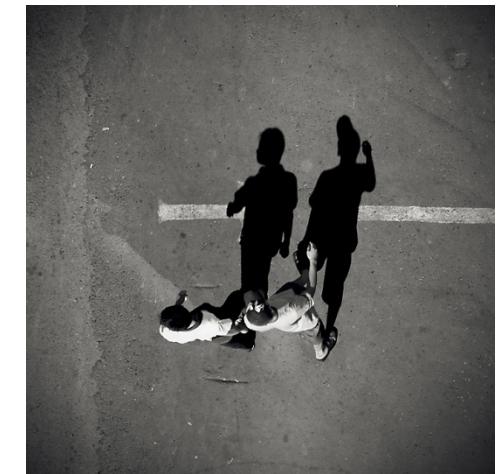
gradient magnitude



Berkeley segmentation database:

<http://www.eecs.berkeley.edu/Research/Projects/CS/vision/grouping/segbench/>

# Low-level edges vs. perceived contours



**Background**

**Texture**

**Shadows**