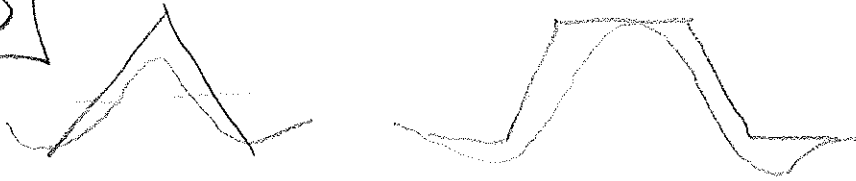
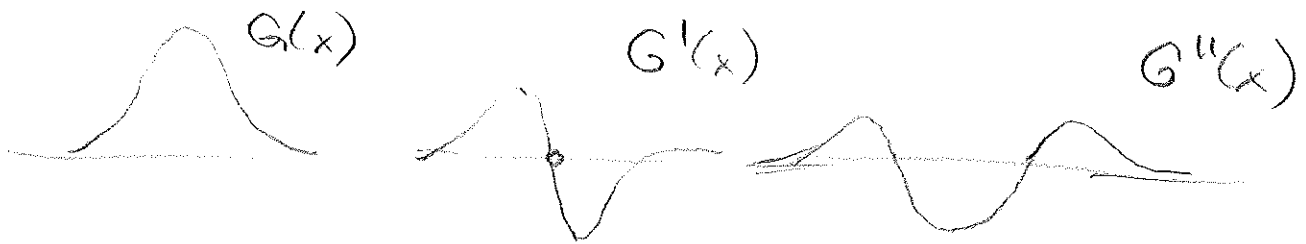


Canny Operator for ridge/line profiles

1-D



⇒ resembles 2nd derivative of Gaussian



Filtering:  $\left(\frac{d^2}{dx^2} G(x, \sigma)\right) \otimes I(x)$

linear operation:

$$\frac{d^2}{dx^2} \underbrace{(G(x, \sigma) \otimes I(x))}_{\text{Gaussian smoothed image}}$$

discrete  $\begin{bmatrix} 1 & -2 & 1 \end{bmatrix}^* \otimes (G(x, \sigma) \otimes I(x))$

convolute:  $\frac{d}{dx} \cdot \frac{d}{dx} : \begin{bmatrix} 1 & 1 \end{bmatrix} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} -1 & 2 & -1 \end{bmatrix}$

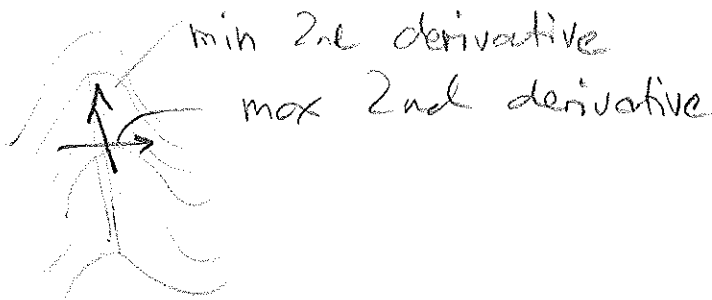
11/3/2010

②

Canny for ridges / lines etc.

12-D

2nd derivative in direction  
orthogonal to ridge = direction  
of maximum 2nd derivative



how to calculate?

$$\begin{bmatrix} \frac{\partial^2}{\partial x^2} & \frac{\partial^2}{\partial x \partial y} \\ \frac{\partial^2}{\partial x \partial y} & \frac{\partial^2}{\partial y^2} \end{bmatrix} \cdot I(x) = \begin{bmatrix} I_{xx} & I_{xy} \\ I_{yx} & I_{yy} \end{bmatrix}$$

"Hessian"  $H$

Directions of min and max 2nd derivatives?

Diagonalization:  $|H - \lambda I| = 0$

Characteristic system

principal 2nd derivatives: eigenvalues

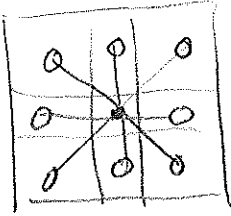
$$[\lambda_1, \lambda_2]$$

principal directions: eigenvectors  $[\bar{e}v_1, \bar{e}v_2]$

11/3/2010

## Simplification for discrete implementation

(3)



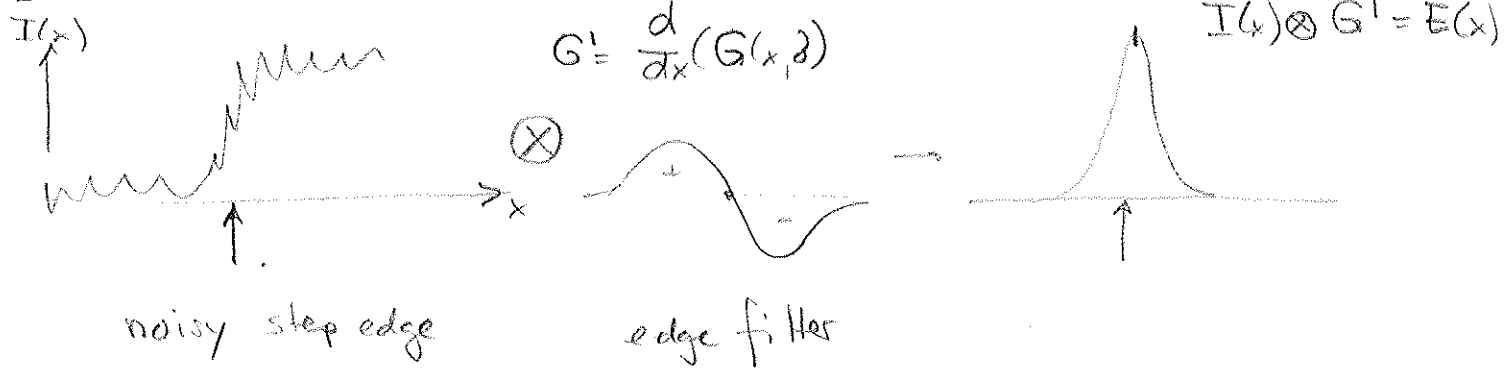
- calculate 2nd derivatives in 4 raster directions
- choose direction where 2nd derivative is extremal (max or min)
- discrete marks for  $x, y$ :  $\begin{bmatrix} 1 & -2 & 1 \end{bmatrix}$
- discrete marks for diagonals:  $\frac{1}{2} \begin{bmatrix} 1 & -2 & 1 \end{bmatrix}$   
(factor due to  $\sqrt{2}$  spacing of pixels)

## Summary 2nd derivative operator

- Gaussian blurring of image (scale):  $G(x, \sigma) * I(x)$
- build 2nd derivatives (discrete or via Hessian)
- choose extremal 2nd derivative perpendicular to ridge  $\rightarrow$  pixel output
- please note that you get positive and negative output for dark and bright lines

# Non maximum suppression

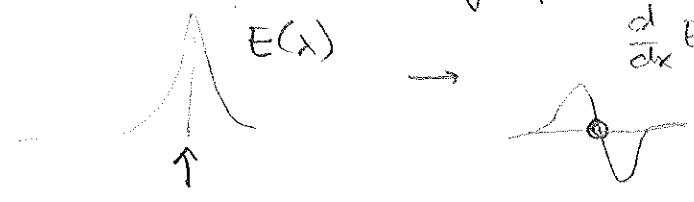
1-D: (i.e., detection of maxima)



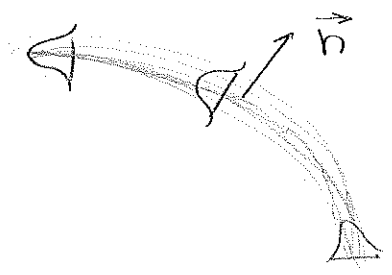
detection of peak: - neighbors smaller than center



- analytical: zero crossing of derivative



2-D: Peak detection along normal direction = gradient direction



notation:  $\nabla = \begin{pmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \end{pmatrix}$

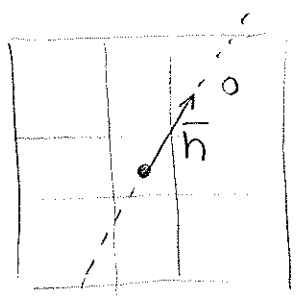
$\vec{n} = \frac{\nabla(G \otimes I)}{|\nabla(G \otimes I)|}$

← gradient vector

← normalization to length 1

11/3/2010

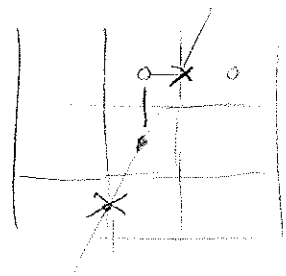
5



- since  $\bar{n}$  is continuous, exact values have to be interpolated:

$$G_A = \frac{u_x}{u_y} G(x+1, y+1) + \frac{u_y - u_x}{u_y} G(x, y+1)$$

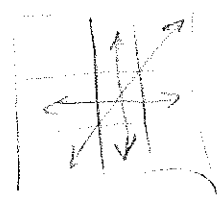
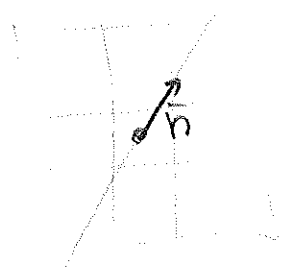
$$G_B = \dots$$



Rule: keep center pixel if neighbors have smaller values

• simplification:

- take  $\bar{n}$
- find neighbors closest to direction of  $\bar{n}$ 
  - calculate angle  $\alpha$ , choose discrete direction of neighbors
- take neighbor pixels and determine peak



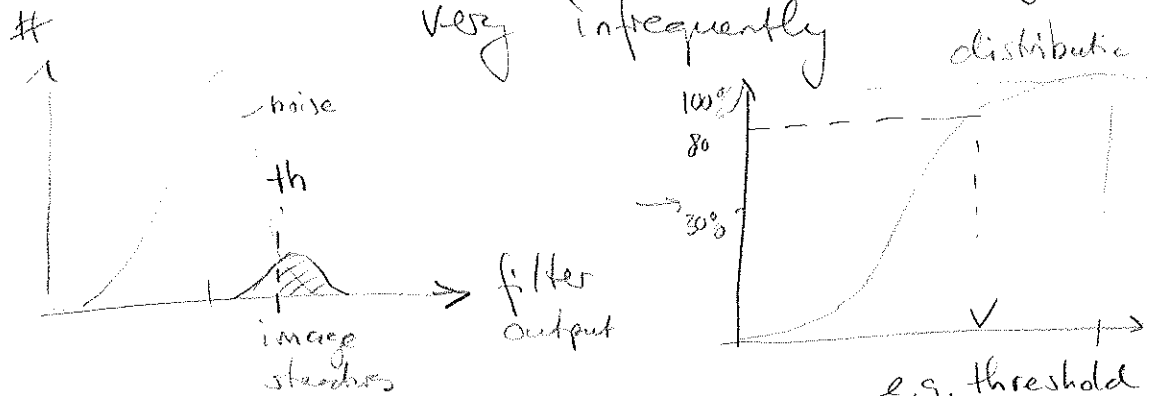
Slide

# Canny: Hysteresis thresholding

goal: separation between signal and noise:

- noise: Gaussian distributed, amplitude in general lower than image structures

- edges/lines: large values occurring very infrequently



① Automatic choice of threshold: e.g. threshold cut 80% quantile

→ create histogram of filter output (after non-maximum suppression),

- calculate distribution:

$$d(x) = \int_0^x h(x) dx$$

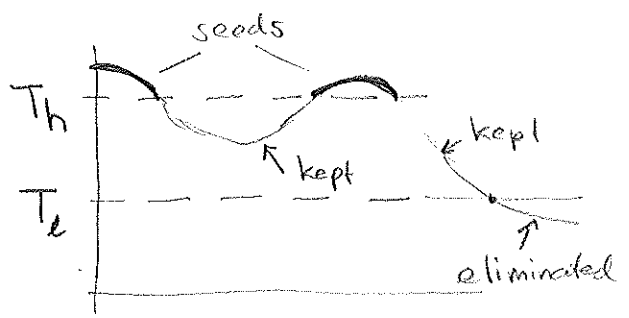
- choose 80% or other high percentile for thresholding the filter output image

# Hysteresis thresholding, ctd

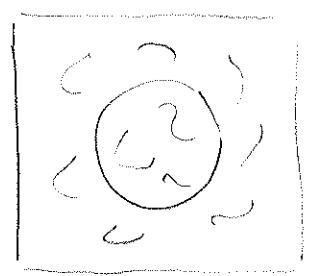
## ② Hysteresis thresholding

- observation: after single threshold, edge contours still broken up since values fluctuate above and below the length of a contour

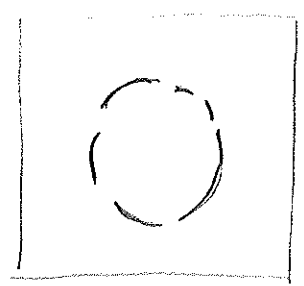
- idea: - low and high thresholds  $T_e, T_h$



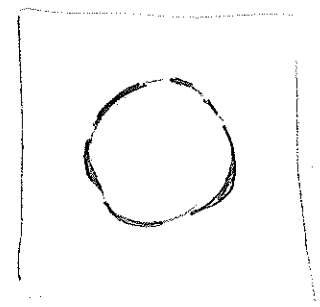
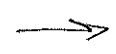
- contour parts above  $T_h$ : taken as output
- segments connected to those "seed" contours and above lower threshold  $T_e$ : taken as output
- connectivity for segments:  $d_m = 4/8$
- segments with values above  $T_e$  and below  $T_h$ , but not connected to "seed  $> T_h$ " are eliminated



threshold  $T_e$



threshold  $T_h$   
"seed" contours



hysteresis