

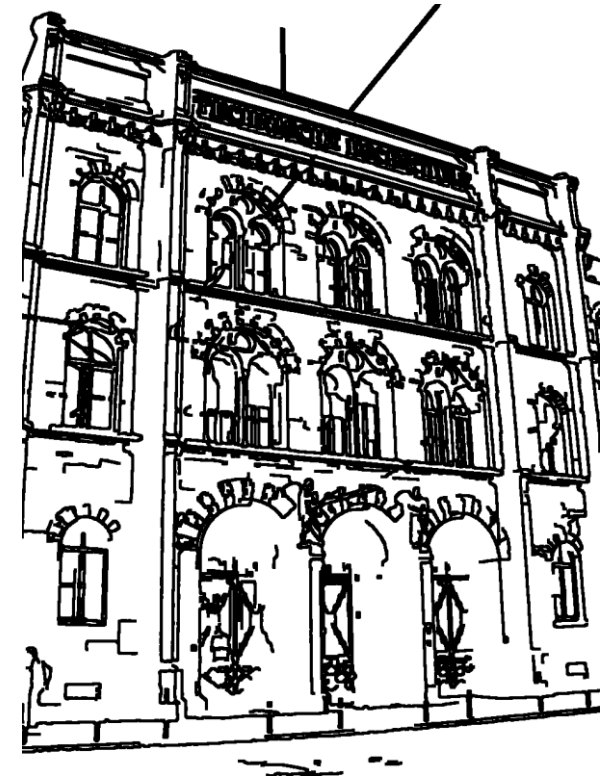
# Machine Vision

## Chapter 3: Edge and Corner Detection

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# Edge Detection



- 亮区和暗区之间变化剧烈的区域 -
- 通常出现在物体边界 -
- 出现在阴影和纹理处 -
- 边缘与图像亮度无关 -

人类视觉皮层的许多部分都在处理灰度边缘问题

## grey level edges:

- areas of hard changes between bright and dark areas
- typically occur at object boundaries
- occur at shadows and texture
- edges independent of image brightness
- many parts of human visual cortex are dealing with grey level edges

# Finding Edges

- edges are areas of rapidly changing grey value

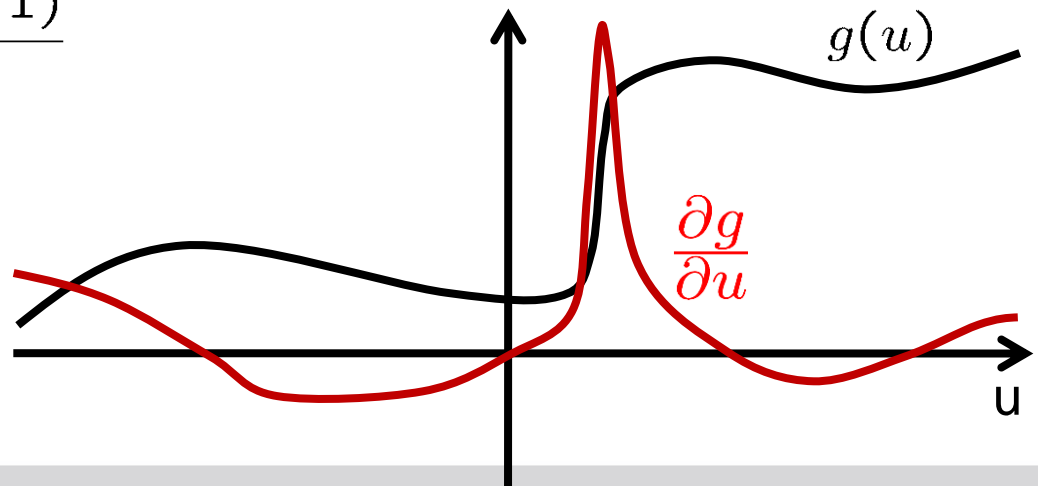
$$|g(u + \epsilon) - g(u - \epsilon)| \text{ large for small } \epsilon$$

- search areas with large derivative of  $g$

$$\frac{\partial g}{\partial u} = \lim_{\epsilon \rightarrow 0} \frac{g(u + \epsilon) - g(u)}{\epsilon} = \lim_{\epsilon \rightarrow 0} \frac{g(u + \epsilon) - g(u - \epsilon)}{2\epsilon}$$

- approximating derivative by difference:

$$\frac{\partial g}{\partial u} \approx \frac{g(u + 1) - g(u - 1)}{2}$$



## Finding Edges cont.

- approximating the derivative can be implemented as convolution with filter mask:

$$\frac{1}{2} \cdot \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}$$

$$\frac{\partial g}{\partial u} \approx \frac{g(u+1) - g(u-1)}{2}$$

- the 2d case:

$$\frac{1}{2} \cdot \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}$$

$$\frac{\partial g}{\partial u} \approx \frac{g(u+1, v) - g(u-1, v)}{2}$$

- analogously:

注意f\*g为卷积运算，需要反向计算

$$\frac{1}{2} \cdot \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$\frac{\partial g}{\partial v} \approx \frac{g(u, v+1) - g(u, v-1)}{2}$$

- noise reduction: additional averaging

# Finding Edges cont.

- Prewitt-operator:

$$\frac{1}{6} \cdot \begin{array}{|c|c|c|} \hline 1 & 0 & -1 \\ \hline 1 & 0 & -1 \\ \hline 1 & 0 & -1 \\ \hline \end{array}$$

$$\frac{1}{6} \cdot \begin{array}{|c|c|c|} \hline 1 & 1 & 1 \\ \hline 0 & 0 & 0 \\ \hline -1 & -1 & -1 \\ \hline \end{array}$$

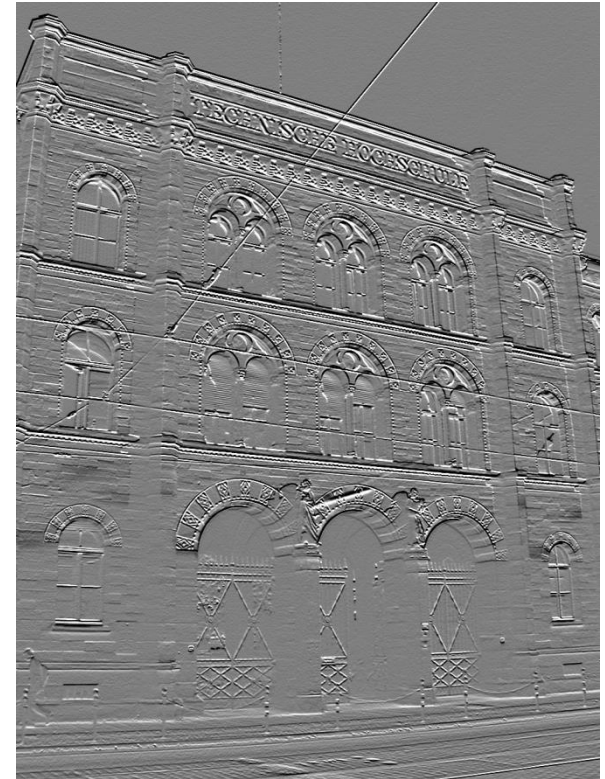
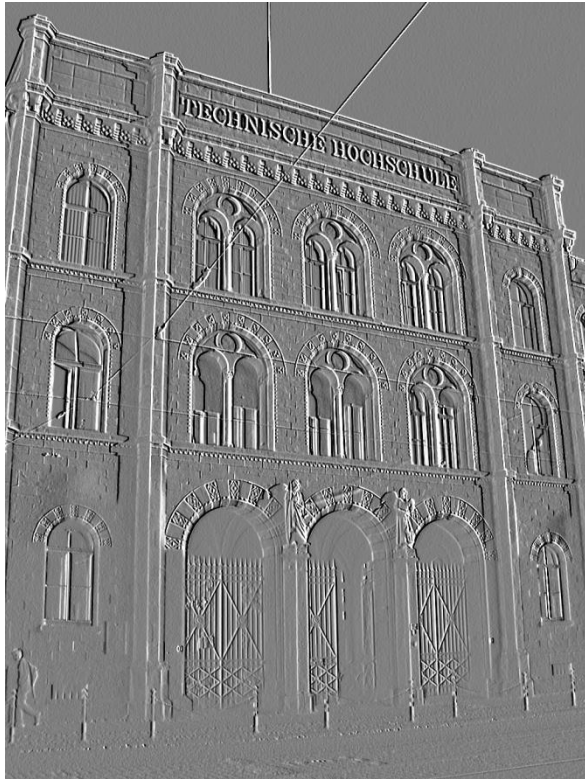
- Sobel-operator:

$$\frac{1}{8} \cdot \begin{array}{|c|c|c|} \hline 1 & 0 & -1 \\ \hline 2 & 0 & -2 \\ \hline 1 & 0 & -1 \\ \hline \end{array}$$

$$\frac{1}{8} \cdot \begin{array}{|c|c|c|} \hline 1 & 2 & 1 \\ \hline 0 & 0 & 0 \\ \hline -1 & -2 & -1 \\ \hline \end{array}$$



# Finding Edges cont.



encoding derivative by gray scale  
for visualization:

- gray: derivative is zero
- bright: derivative is positive
- dark: derivative is negative

$$\frac{\partial g}{\partial u}$$

$$\frac{\partial g}{\partial v}$$

# Finding Edges cont.

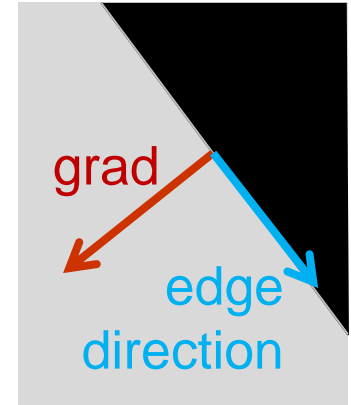
- edge orientation:

$$\text{grad } g = \left( \frac{\partial g}{\partial u}, \frac{\partial g}{\partial v} \right)$$

- grey level gradient points to direction of maximal grey level ascend
- orthogonal directions exhibit no change of grey level

$$\text{grad } g \perp \left( -\frac{\partial g}{\partial v}, \frac{\partial g}{\partial u} \right)$$

- length of gradient is proportional to grey level change rate



# Finding Edges cont.



## gradient plot:

- saturation: gradient length
- color: gradient direction (angle)

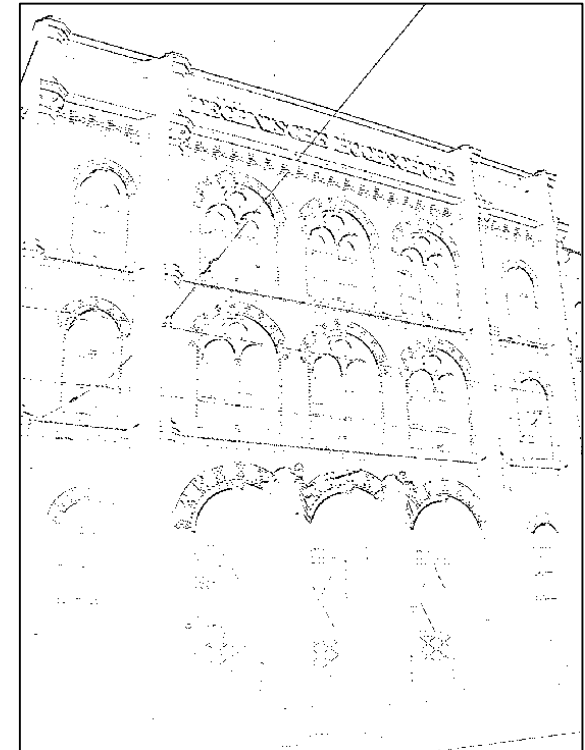
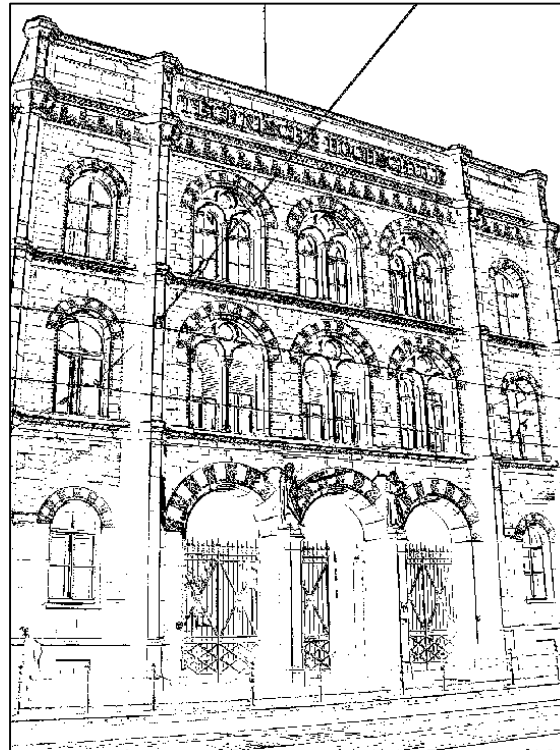




# Finding Edges cont.

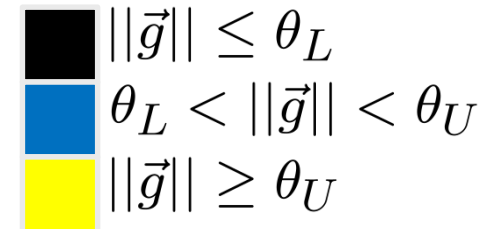
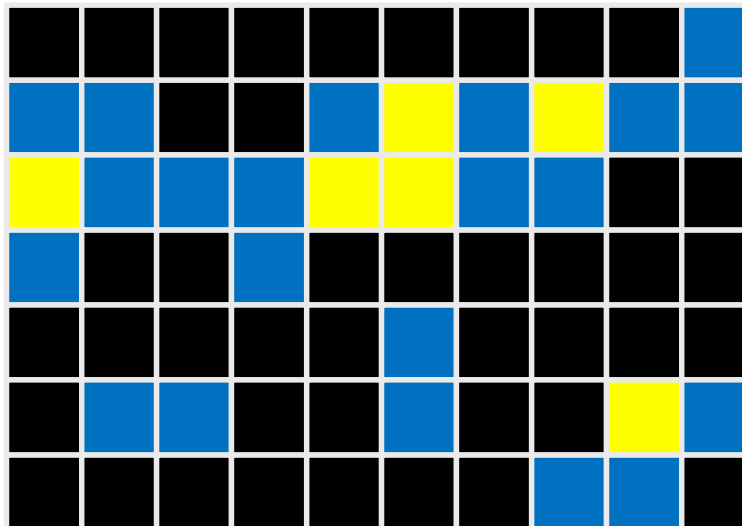
- from which gradient length on are edges relevant?
  - small threshold: too much noise remains
  - large threshold: contours not connected
  - idea: double thresholding

Threshold的大小的影响



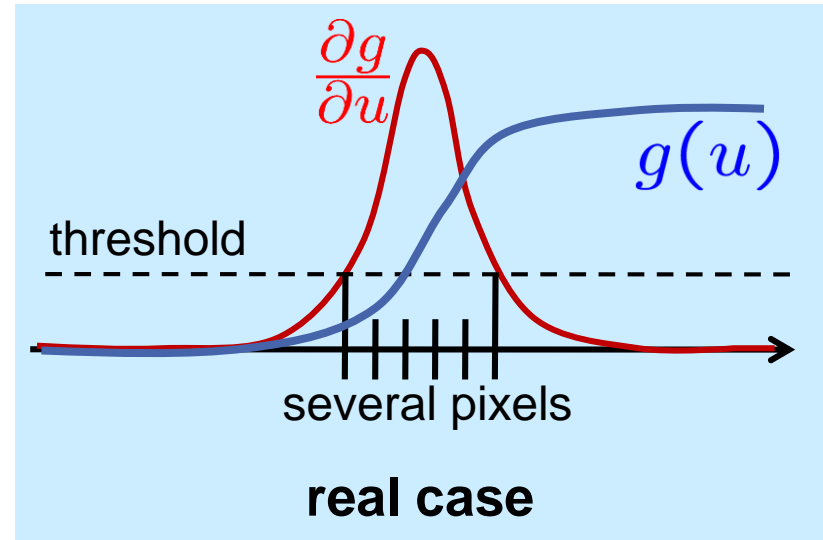
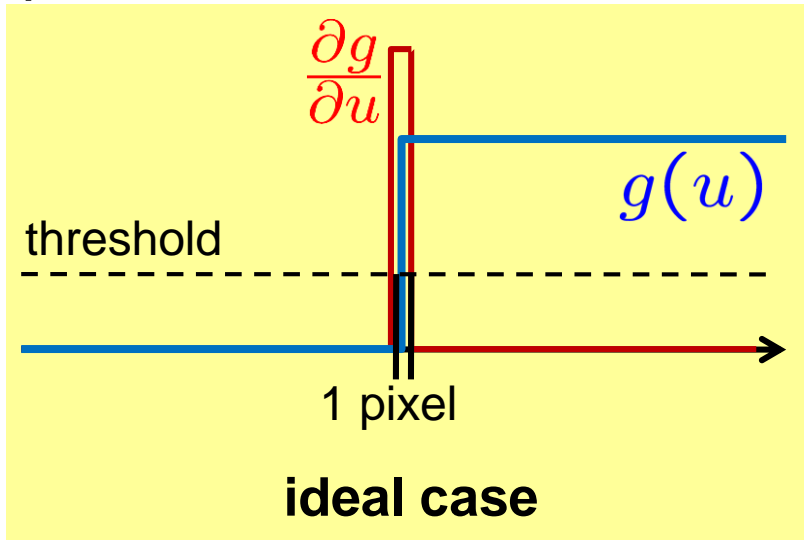
# Finding Edges cont.

- double thresholding:
  - two thresholds:  $\theta_L$ ,  $\theta_U$ ,  $\theta_L < \theta_U$
  - pixels are classified according to gradient length  $\|\vec{g}\|$  :
    - $\|\vec{g}\| \leq \theta_L$  pixel is not edge element
    - $\|\vec{g}\| \geq \theta_U$  pixel is edge element
    - $\theta_L < \|\vec{g}\| < \theta_U$  pixel is edge element if a neighboring pixel is edge element



# Finding Edges cont.

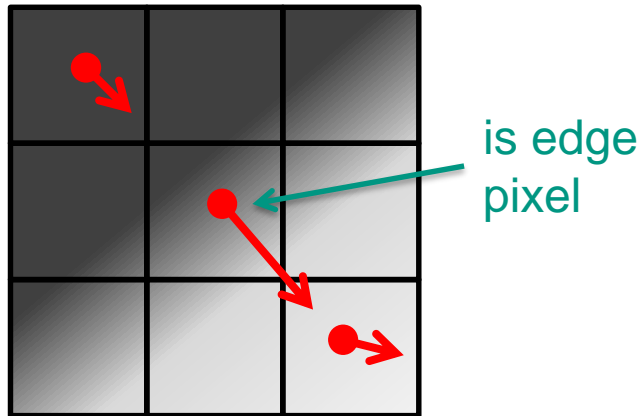
- problem: thick lines



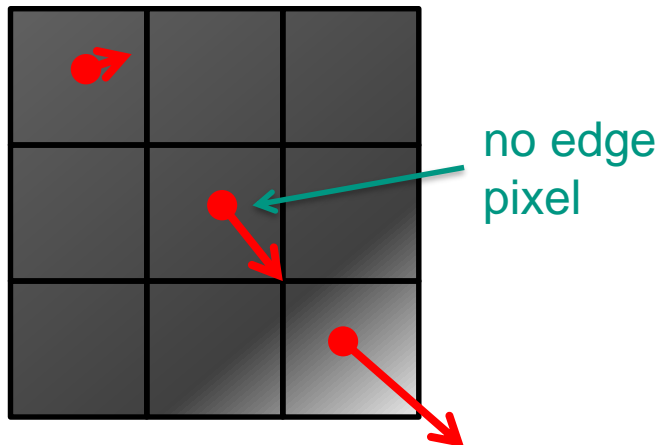
- non-maxima suppression
  - idea: among neighboring pixels, consider only the one with maximal gradient length
  - 2D case: take into account edge direction

# Finding Edges cont.

- Non-maxima suppression



- check gradient direction
- select neighboring pixel in gradient direction and opposite gradient direction
- pixel is edge pixel if gradient length is larger than in those two neighboring pixels

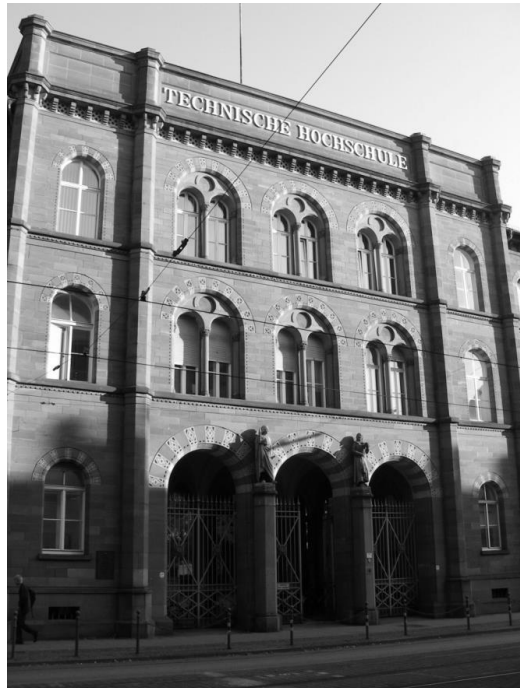




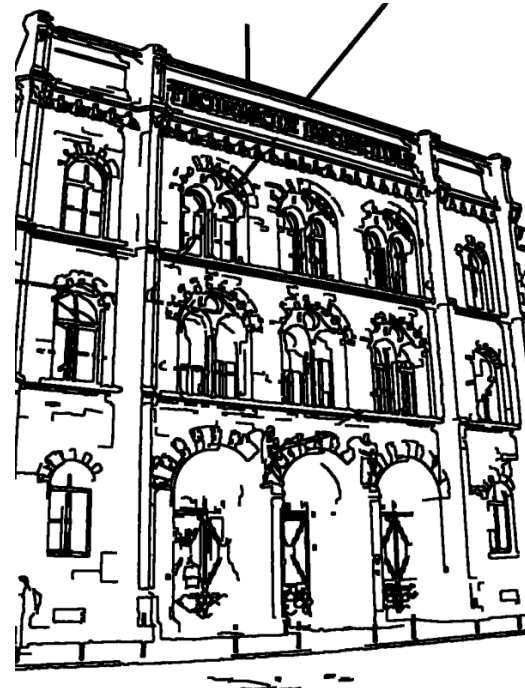
# Canny Edge Detector

Canny edge detector combines the following techniques:

1. smooth image with Gaussian filter
2. compute grey level gradient with Sobel/Prewitt masks
3. apply non-maxima suppression
4. apply double thresholding



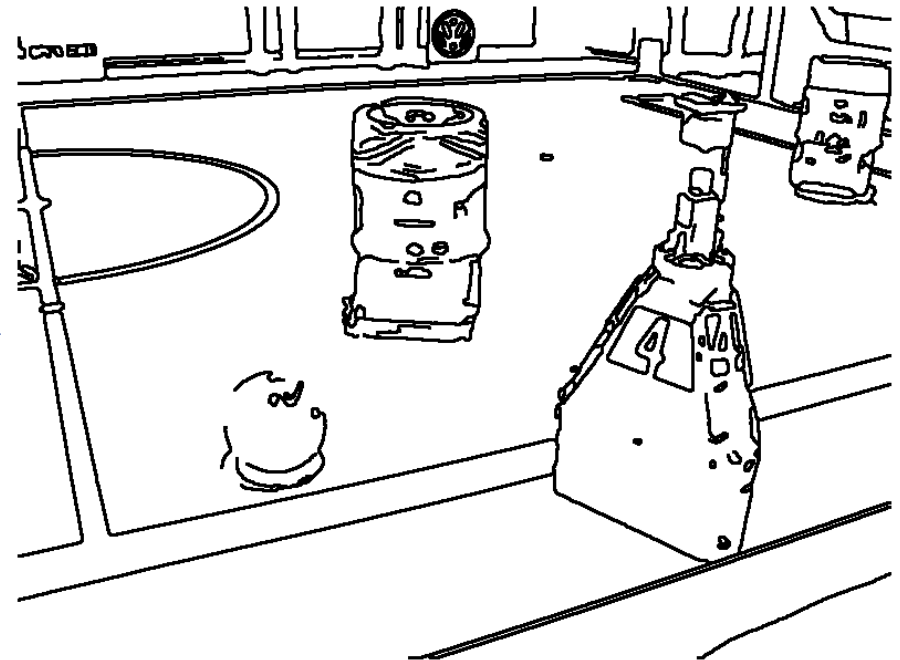
Canny



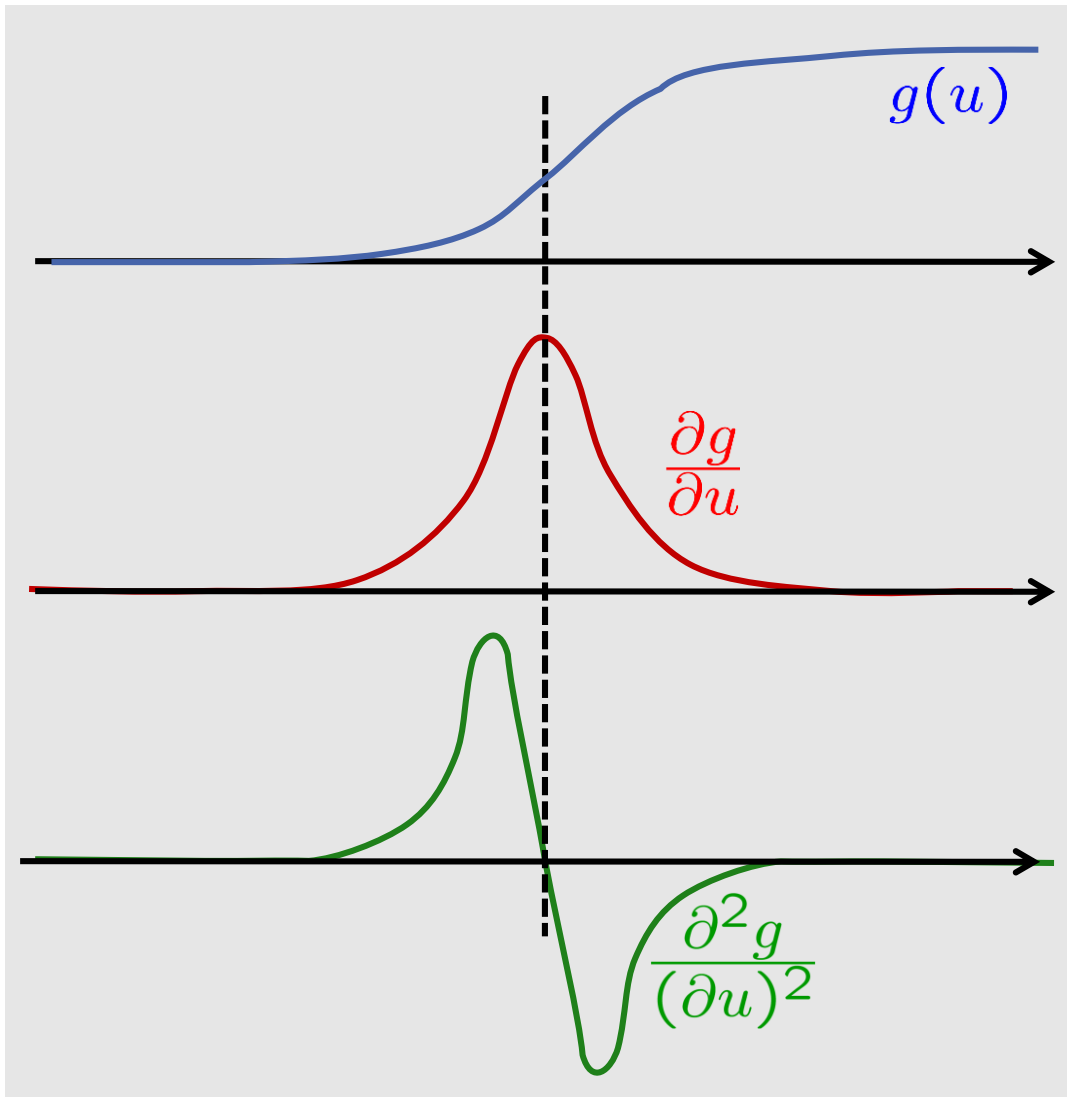
# Canny Edge Detector



Canny



## Finding Edges cont.



grey level edge:

- $g$  rapidly changing
- maximum of  $\frac{\partial g}{\partial u}$
- zero crossing of  $\frac{\partial^2 g}{(\partial u)^2}$

- 2D analogon to 2nd order derivative is Laplace operator:

$$\begin{aligned}\nabla^2 g &= \frac{\partial^2 g}{(\partial u)^2} + \frac{\partial^2 g}{(\partial v)^2} \\ &= \text{trace}(H) \\ &\quad (H \text{ Hessian})\end{aligned}$$

# Laplace Operator

- Approximation to Laplace operator:

$$\frac{\partial g}{\partial u}(u, v) \approx g(u + 1, v) - g(u, v)$$

$$\begin{aligned}\frac{\partial^2 g}{(\partial u)^2}(u, v) &\approx \frac{\partial g}{\partial u}(u, v) - \frac{\partial g}{\partial u}(u - 1, v) \\ &\approx g(u + 1, v) - 2g(u, v) + g(u - 1, v)\end{aligned}$$

$$\frac{\partial^2 g}{(\partial v)^2}(u, v) \approx g(u, v + 1) - 2g(u, v) + g(u, v - 1)$$

$$\nabla^2 g \approx g(u + 1, v) + g(u - 1, v) + g(u, v + 1) + g(u, v - 1) - 4g(u, v)$$

- implementation as filter mask:

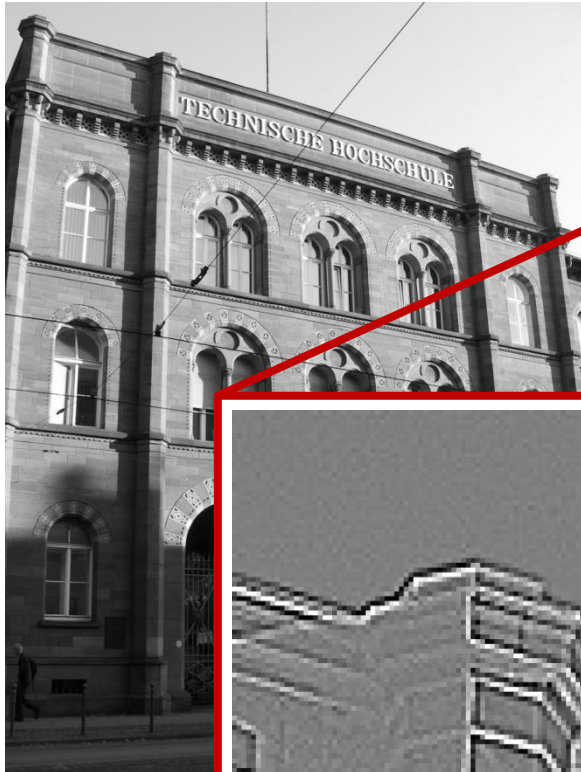
0	1	0
1	-4	1
0	1	0

1	4	1
4	-20	4
1	4	1

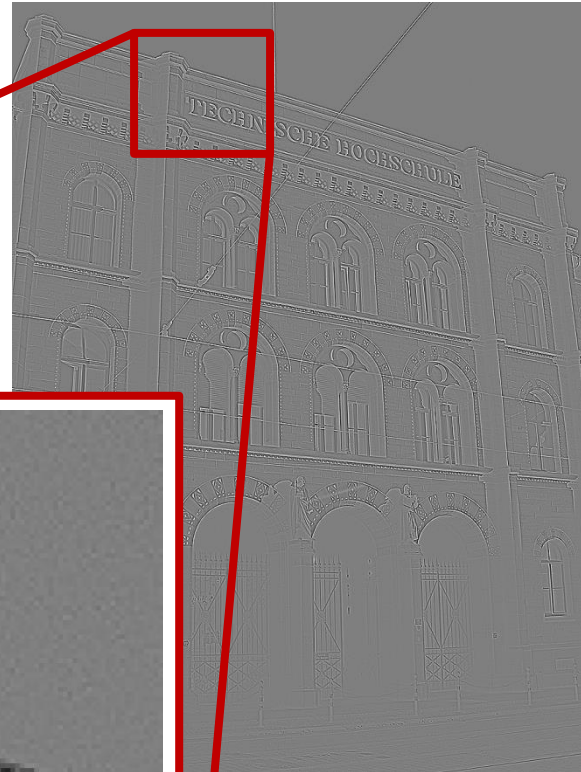
$\frac{1}{6} \cdot$



# Laplace Operator cont.



Laplace



Laplace operator

- grey: zero
- white: positive
- black: negative

# Laplace Operator cont.

- 2nd order derivative is very noisy
- combine Laplacian with Gaussian smoothing

$$\nabla^2(G * g) = (\nabla^2 G) * g$$

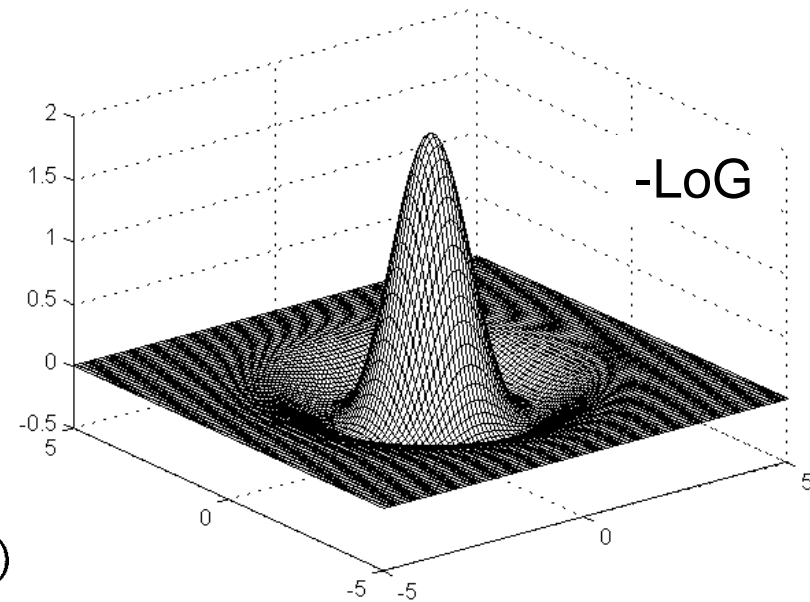
( $G$  Gaussian)

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

$$\begin{aligned} \frac{\partial G}{\partial u} &= \frac{1}{2\pi\sigma^2} \left(-\frac{1}{\sigma^2}\right) 2u e^{-\frac{1}{2\sigma^2}(u^2 + v^2)} \\ &= -\frac{u}{\sigma^2} G(u, v) \end{aligned}$$

$$\begin{aligned} \frac{\partial^2 G}{(\partial u)^2} &= -\frac{1}{\sigma^2} G(u, v) - \frac{u}{\sigma^2} \left(-\frac{u}{\sigma^2} G(u, v)\right) \\ &= \frac{u^2 - \sigma^2}{\sigma^4} G(u, v) \end{aligned}$$

$$\nabla^2 G = \frac{u^2 + v^2 - 2\sigma^2}{\sigma^4} G(u, v)$$



“Laplacian of Gaussian”  
(LoG)

“mexican hat”

## Laplace Operator cont.

- filter masks for LoG:

0	0	1	0	0
0	1	2	1	0
1	2	-16	2	1
0	1	2	1	0
0	0	1	0	0

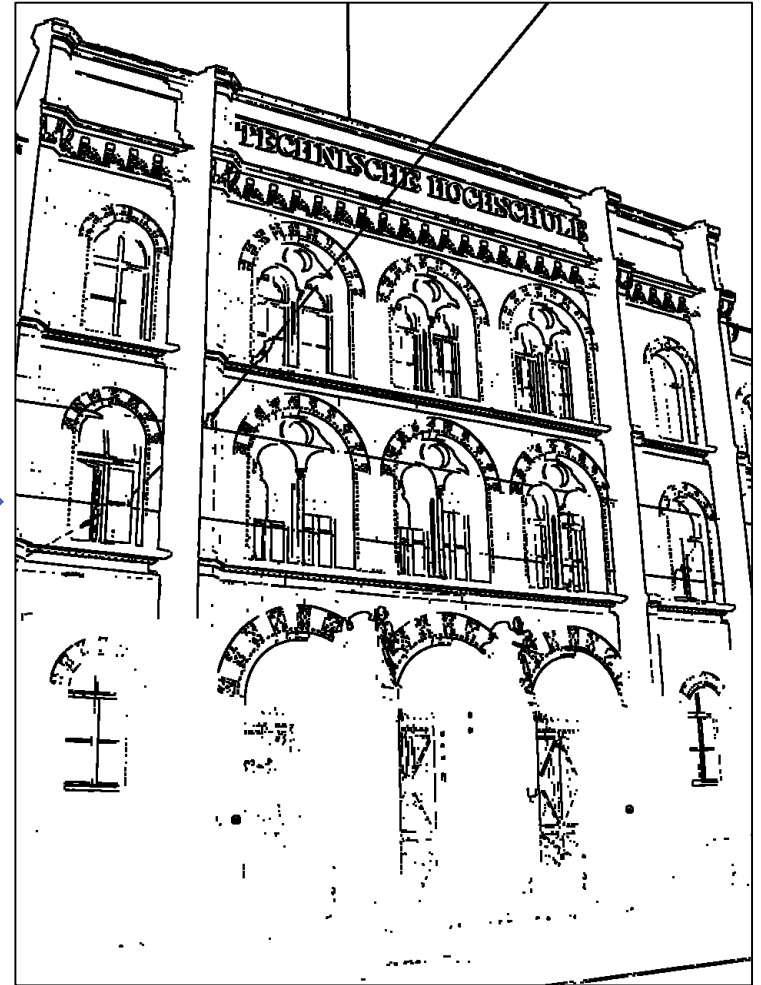
- LoG can be approximated by DoG “Difference of Gaussian”

$$DoG(u, v) = G_{\sigma_1}(u, v) - G_{\sigma_2}(u, v)$$

# Laplace Operator cont.



LoG +  
Threshold



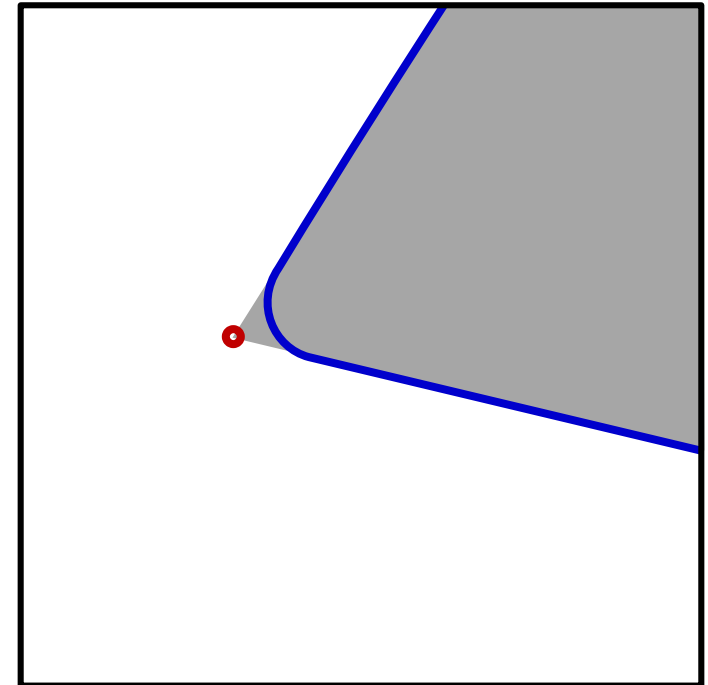
- edge detection approach according to Marr/Hildreth



# CORNER DETECTION

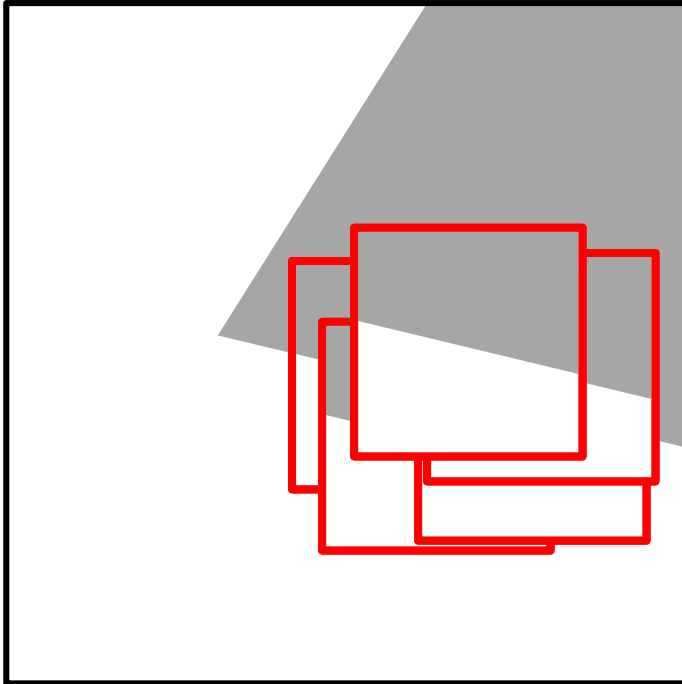
# Corner Detection

- graylevel corners important due to:
    - good features to find again in another image
    - corners as feature points, e.g. for calibration
    - edge detector usually round off corners
- special filter to detect corners

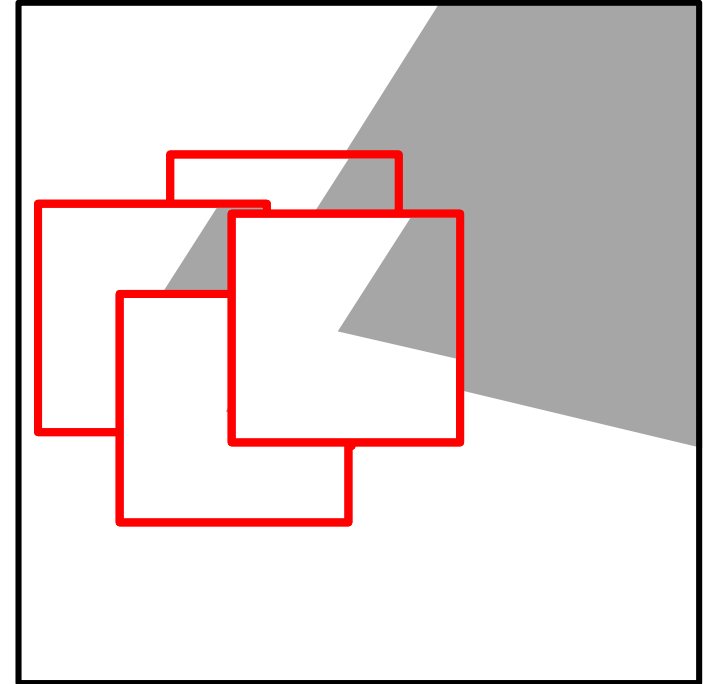


## Corner Detection cont.

- Idea: find patches of maximal dissimilarity for local moves



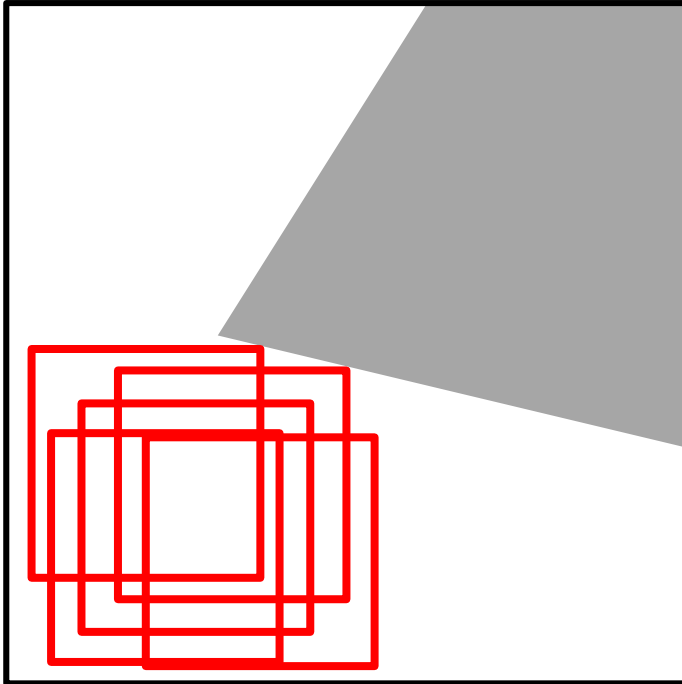
**edge:** similar when moving along edge, dissimilar when moving orthogonal



**corner:** dissimilar in all direction

## Corner Detection cont.

- Idea: find patches of maximal dissimilarity for local moves



**homogeneous areas:**  
similar in all directions

dissimilarity measure:

$$\begin{aligned} & \sum_{(u,v) \in \text{rectangle}} (g(u + \Delta u, v + \Delta v) - g(u, v))^2 \\ & \approx \sum_{(u,v) \in \text{rectangle}} \left( g(u, v) + \Delta u \frac{\partial g}{\partial u} + \Delta v \frac{\partial g}{\partial v} - g(u, v) \right)^2 \\ & = \sum_{(u,v) \in \text{rectangle}} \left( \left( \Delta u \frac{\partial g}{\partial u} \right)^2 + 2 \Delta u \frac{\partial g}{\partial u} \Delta v \frac{\partial g}{\partial v} + \left( \Delta v \frac{\partial g}{\partial v} \right)^2 \right) \\ & = \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix}^T \underbrace{\begin{pmatrix} \sum \left( \frac{\partial g}{\partial u} \right)^2 & \sum \frac{\partial g}{\partial u} \frac{\partial g}{\partial v} \\ \sum \frac{\partial g}{\partial u} \frac{\partial g}{\partial v} & \sum \left( \frac{\partial g}{\partial v} \right)^2 \end{pmatrix}}_{=:S} \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix} \end{aligned}$$



# Corner Detection cont.

- Dissimilarity measure:

$$d := \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix}^T \underbrace{\begin{pmatrix} \sum (\frac{\partial g}{\partial u})^2 & \sum \frac{\partial g}{\partial u} \frac{\partial g}{\partial v} \\ \sum \frac{\partial g}{\partial u} \frac{\partial g}{\partial v} & \sum (\frac{\partial g}{\partial v})^2 \end{pmatrix}}_{=:S} \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix}$$

- dissimilarity should be large for all unit vectors  $(\Delta u, \Delta v)$
- for special choice of coordinate system  $S$  becomes a diagonal matrix (Eigenvector coordinate system)

$$d := \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix}^T \underbrace{\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}}_{=:S} \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix} = \lambda_1 (\Delta u)^2 + \lambda_2 (\Delta v)^2$$

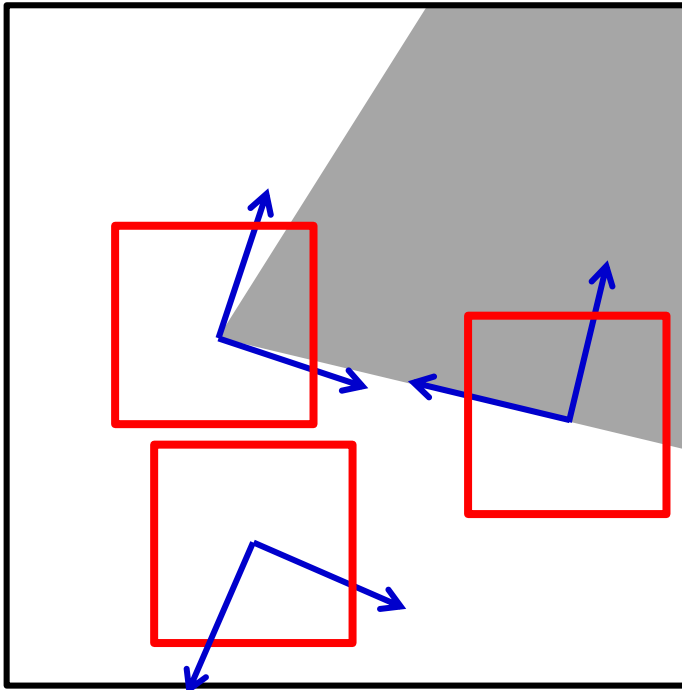
w.l.o.g.  $\lambda_1 \geq \lambda_2 \geq 0$

# Corner Detection cont.

$$d := \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix}^T \underbrace{\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}}_{=S} \begin{pmatrix} \Delta u \\ \Delta v \end{pmatrix} = \lambda_1 (\Delta u)^2 + \lambda_2 (\Delta v)^2$$

w.l.o.g.  $\lambda_1 \geq \lambda_2 \geq 0$

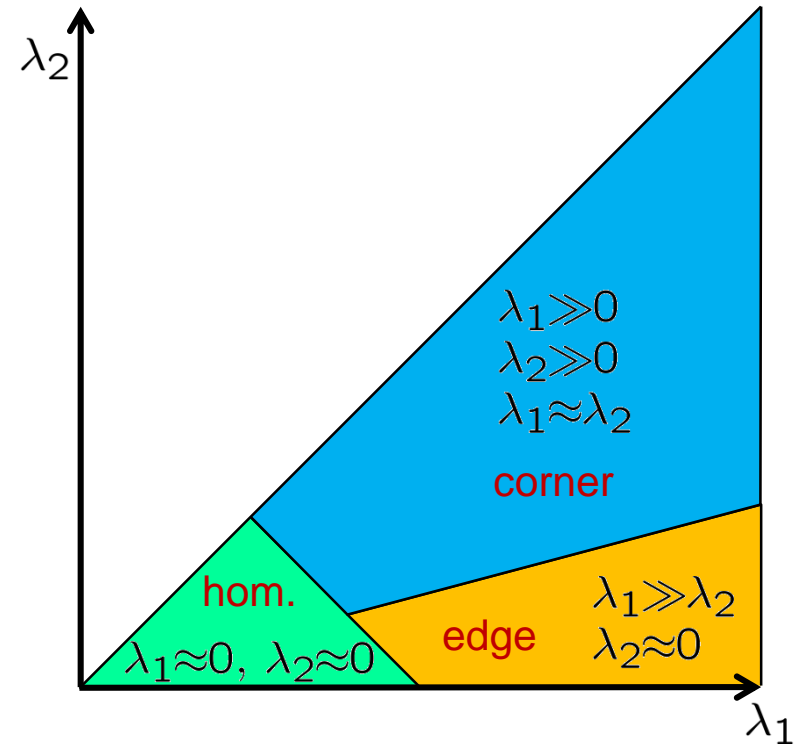
- typical cases:



	$\lambda_1$	$\lambda_2$
edge	large	small
corner	large	large
homogeneous area	small	small

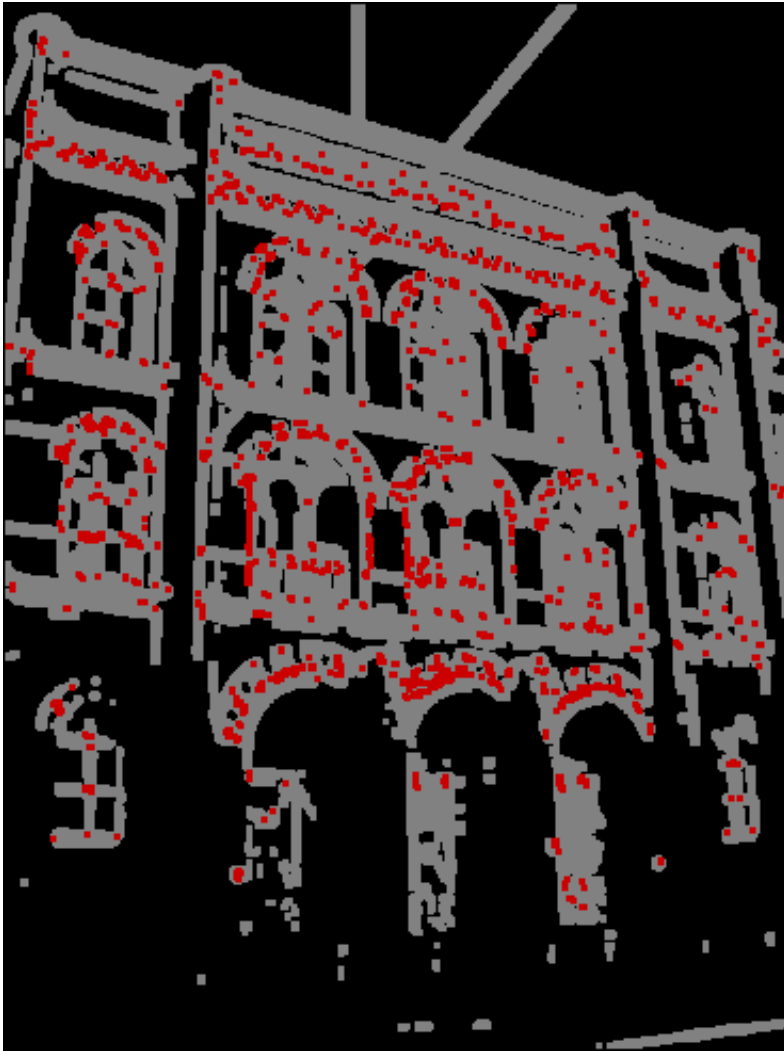
# Corner Detection cont.

- decision rule
  - pixel is in homogeneous area if
$$\text{trace}(S) = \lambda_1 + \lambda_2 < \theta$$
  - otherwise, pixel is corner if
$$\lambda_2 > \alpha \lambda_1$$
$$\Leftrightarrow \det(S) - \frac{\alpha}{(1 + \alpha)^2} (\text{trace}(S))^2 > 0$$
  - otherwise, pixel is edge
- parameters  $\theta$  and  $\alpha$  have to be tuned manually



Harris corner detector

# Corner Detection cont.



# **SUMMARY: EDGE AND CORNER DETECTION**



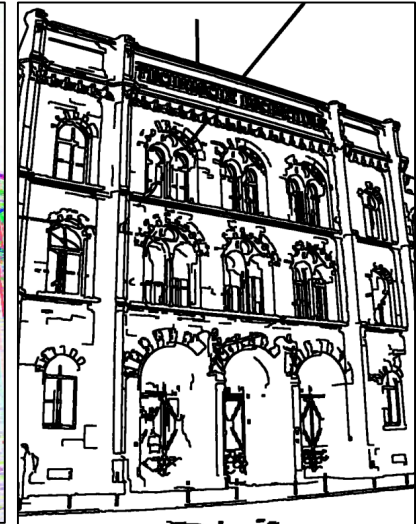
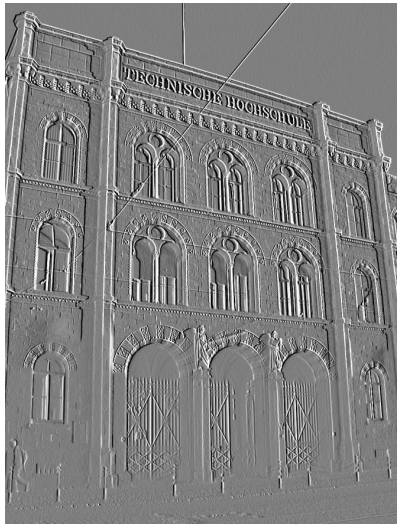
# Summary

## – edge detection

- gradient: Sobel, Prewitt
- thresholding, double thresholding
- non-maxima suppression
- Canny operator
- Laplace operator, Marr/Hildreth approach

1	0	-1
2	0	-2
1	0	-1

1	4	1
4	-20	4
1	4	1



# Summary cont.

- edge detection
- corner detection
  - Harris corner detector

