Edge: Canny

Dr. Tushar Sandhan

Directions with operators

Horizontal

-1	-1	-1
2	2	2
-1	-1	-1

Directions with operators

Horizontal

-1	-1	-1
2	2	2
-1	-1	-1

+45 degrees

2	-1	-1
-1	2	-1
-1	-1	2

EE604: IMAGE PROCESSING

Directions with operators

Horizontal

-1	-1	-1
2	2	2
-1	-1	-1

+45 degrees

2	-1	-1
-1	2	-1
-1	-1	2

Vertical

-1	2	-1
-1	2	-1
-1	2	-1

Directions with operators

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+45 degrees

2	-1	-1
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Vertical

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-1	2	-1
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-1	-1	2
-1	2	-1
2	-1	-1

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Horizontal

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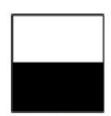
+45 degrees

2	-1	-1
-1	2	-1
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Vertical

-1	2	-1
-1	2	-1
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-1	-1	2
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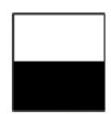
+45 degrees

2	-1	-1
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Vertical

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Directions with operators

Horizontal

-1	-1	-1
2	2	2
-1	-1	-1

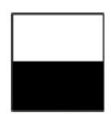
+45 degrees

2	-1	-1
-1	2	-1
-1	-1	2

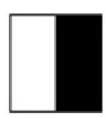
Vertical

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2	-1	-1









- Directions with Kirsch
 - direction is defined by the mask that produces max edge magnitude

$$h_{n,m} = \max_{z=1,\ldots,8} \sum_{i=-1}^1 \sum_{j=-1}^1 g_{ij}^{(z)} \cdot f_{n+i,m+j}$$

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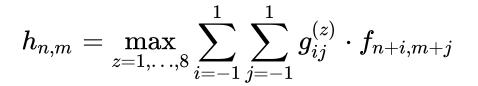




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 $g^{(2)}$

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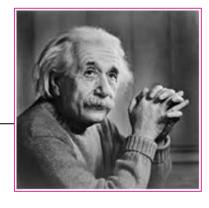
 $g^{(2)}$

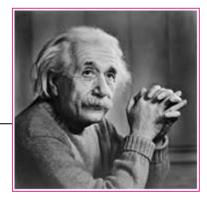
 $g^{(4)}$



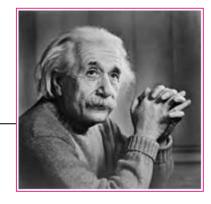






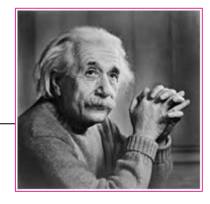






North

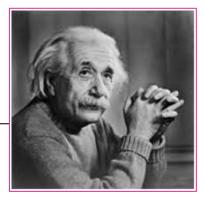




North





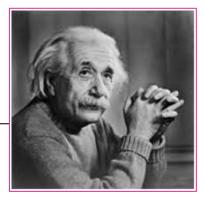


North



North West





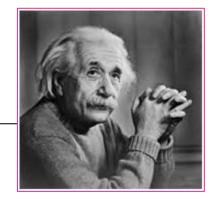
North



North West







North

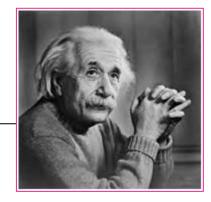


North West



West





North



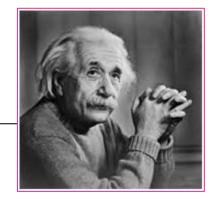
North West



West







North



North West

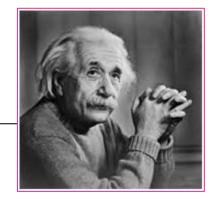


West



South West





North



North West



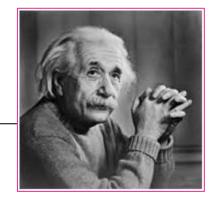
West



South West







North



North West



West

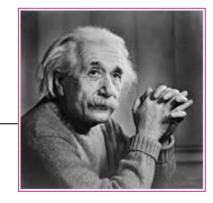


South West



South





North



North West



West



South West

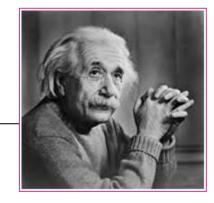


South



South East





North



North West



West



South West



South

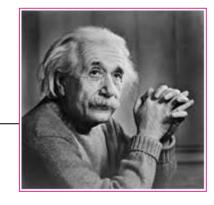


South East



East





North



North West



West



South West



South



South East



East



North East



Single point thick edges

input



Single point thick edges

input



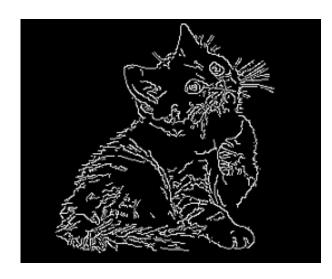


Single point thick edges

input



Canny edges



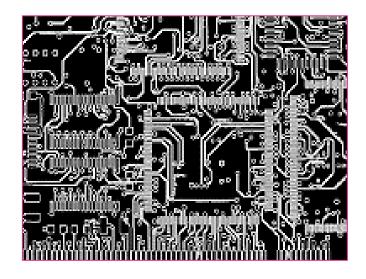
Single point thick edges

input



Canny edges





Introduction

Single point thick edges

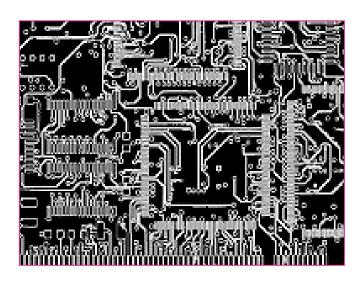
input



Canny edges



Canny PCB edges

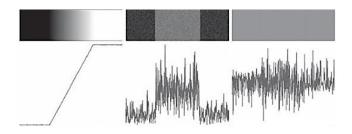


What would be important steps in edge det.

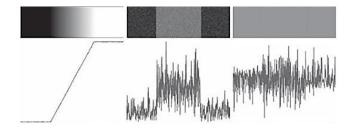
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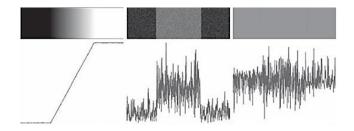
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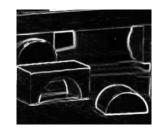


- What would be important steps in edge det.
 - Smooth derivatives
 - Thresholding

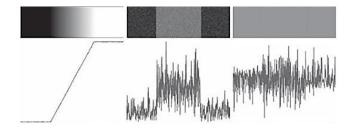


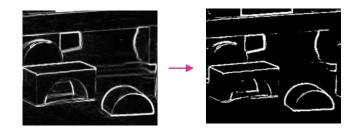
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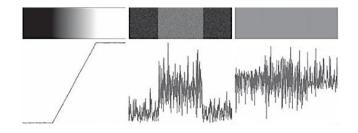


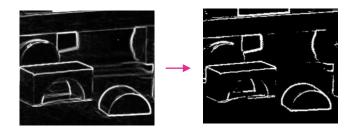
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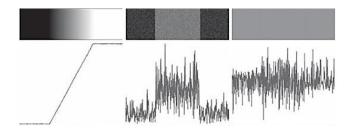


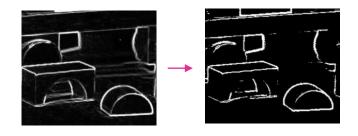
- What would be important steps in edge det.
 - Smooth derivatives
 - Thresholding
 - Thinning



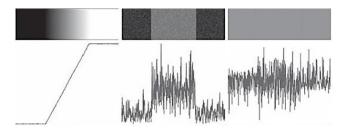


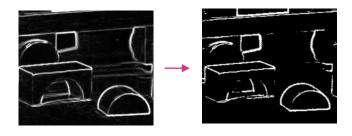
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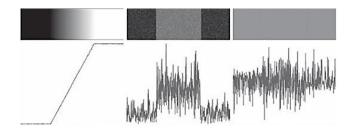
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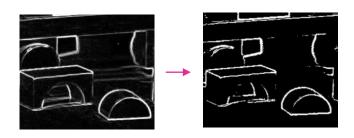


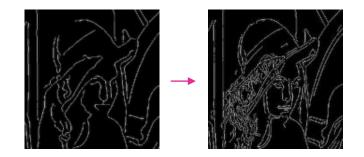




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Objectives

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 - obtained edge point at i^{th} pixel: e_i
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 - good localization of edges
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 - obtained edge point at i^{th} pixel: e_i
 - minimize the distance $||c_i e_i||_2$
 - single point edge response
 - 1 point for each true edge point

- Image derivatives
 - o input image f(x, y)
 - smoothed $f_S(x,y)$
 - o any operator can be used to get $g_x(x, y)$, $g_y(x, y)$

$$G(x,y) = e^{-\frac{x^2 + y^2}{2\sigma^2}}$$

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 $g_y(x,y) = \partial f_s(x,y)/\partial y$

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$$g_{v}(x,y) = \partial f_{s}(x,y)/\partial y$$

$$M_s(x,y) = \|\nabla f_s(x,y)\| = \sqrt{g_x^2(x,y) + g_y^2(x,y)}$$

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$$\alpha(x,y) = \tan^{-1} \left[\frac{g_y(x,y)}{g_x(x,y)} \right]$$

- Thinning
 - o $M_s(x,y)$ wide ridges around local maxima
 - o ridges thinning is needed
 - non-max suppression
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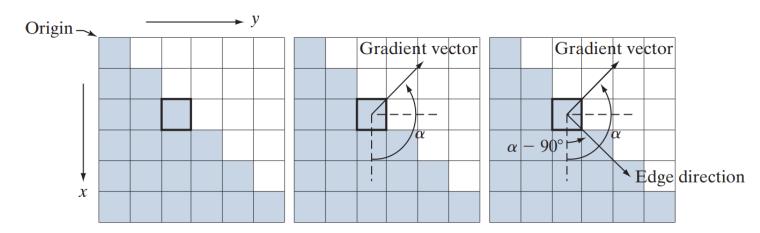
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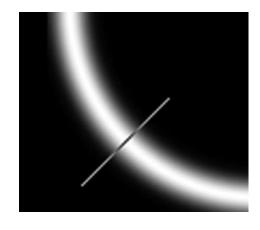
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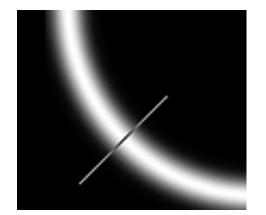
- Thinning
 - non-max suppression: checks whether pixel is local maxima in grad direction

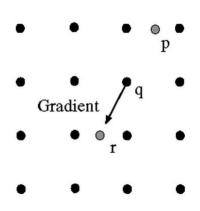
Thinning

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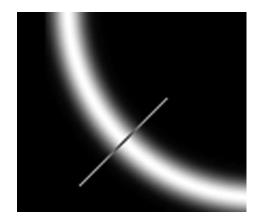
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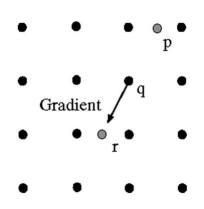




Thinning

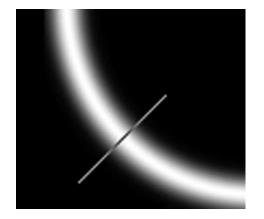
- non-max suppression: checks whether pixel is local maxima in grad direction
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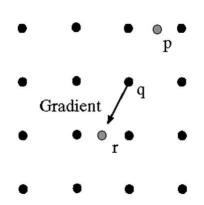


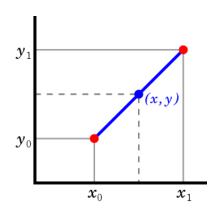


Thinning

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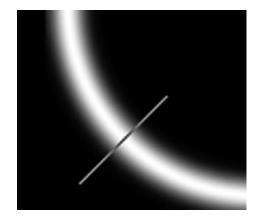


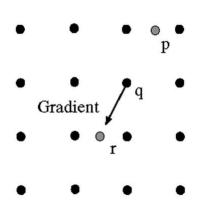


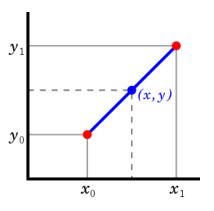
Thinning

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$$\frac{y-y_0}{x-x_0} = \frac{y_1-y_0}{x_1-x_0}$$





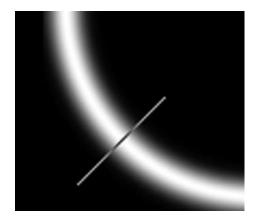


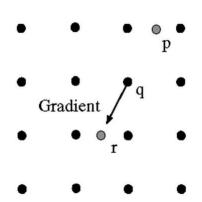
Thinning

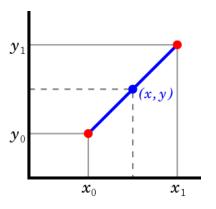
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$$rac{y-y_0}{x-x_0} = rac{y_1-y_0}{x_1-x_0}$$

$$y=y_0+(x-x_0)rac{y_1-y_0}{x_1-x_0}$$







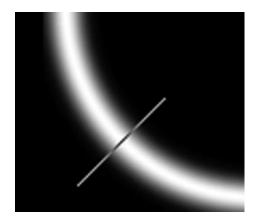
Thinning

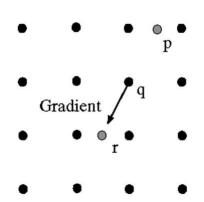
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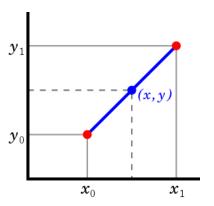
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$$y_0 = y_0 \left(1 - rac{x - x_0}{x_1 - x_0}
ight) + y_1 \left(rac{x - x_0}{x_1 - x_0}
ight) + y_1 \left(rac{x - x_0}{x_1 - x_0}
ight)$$

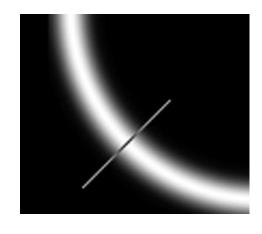






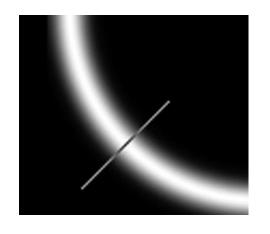
Thinning

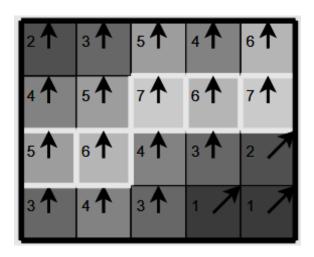
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Thinning

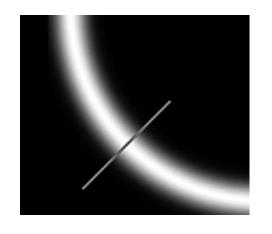
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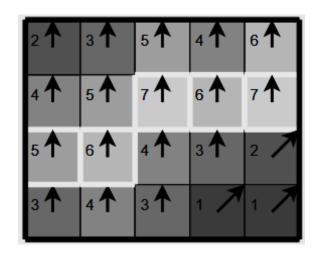




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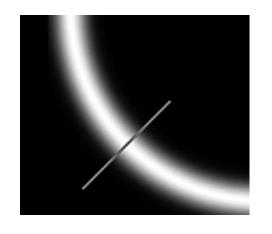


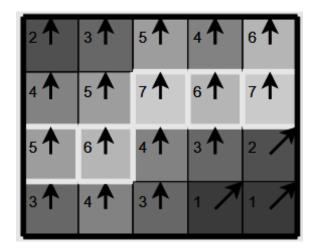




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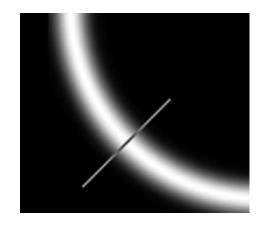


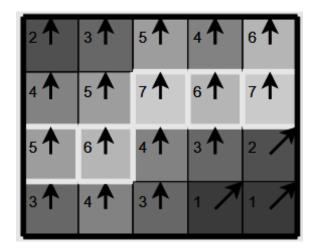




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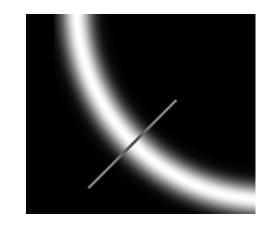






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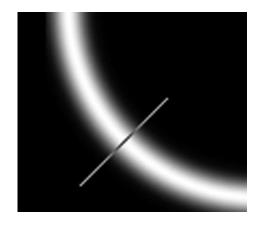




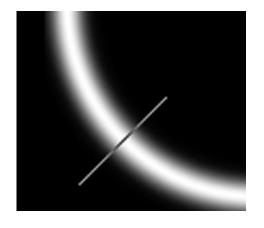


after non-max supp

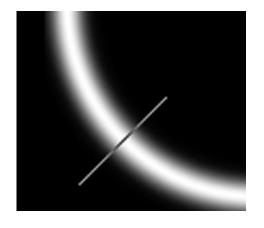




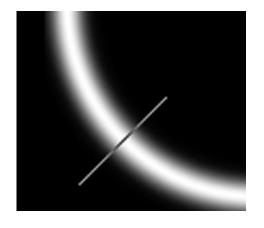
- LINKINg Points
 - Canny edge detector
 - It starts with one thing: gradients
 - In the end, it doesn't even matter: which operators have been used
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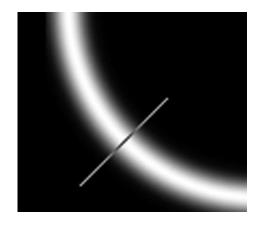
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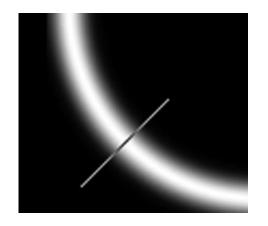


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- a. find all edge points using TH^{high}
- **b.** from each strong point follow the both side direction $oldsymbol{\perp}$ to the edge normal
- c. in that directions, construct the contours of connected edge points
- d. mark all points greater than TH_{low}



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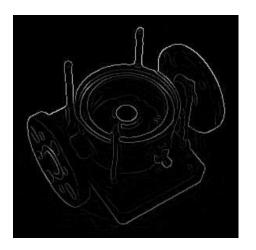
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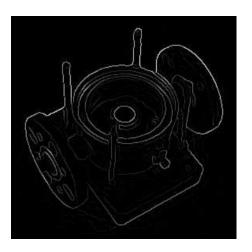
after non-max supp



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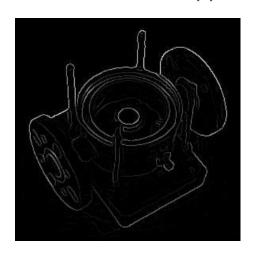
double thresh



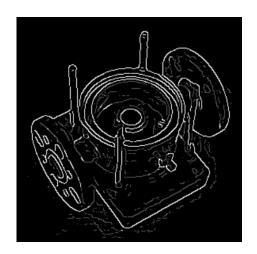
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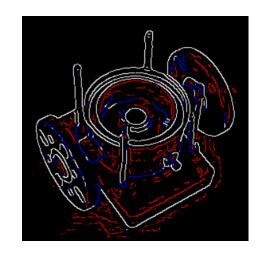
after non-max supp



double thresh



hysteresis



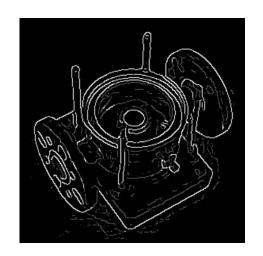
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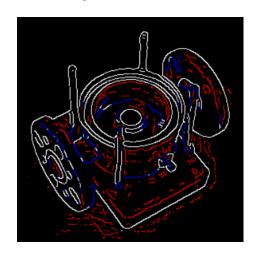
after non-max supp



double thresh



hysteresis

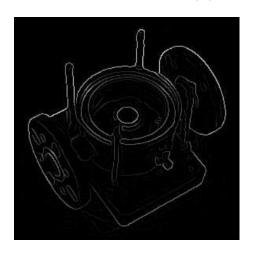


White – strong edges
Blue – weak connect
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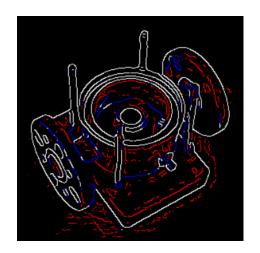
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Ref: P. Kalra

Entire algorithm composition:

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 - 1. Filter image with derivatives of Gaussian
 - 2. Get M, α
 - 3. Non-max suppression
 - thin multi-pixel wide edges to a single pixel widths
 - 4. Linking: the hysteresis
 - \circ 2 thresholds: TH_{low} , TH^{high}
 - o TH^{high} : to start an edge
 - \circ TH_{low} : continue started edge

- Speeding up the beats of operations
 - \circ binning the α (angles)
 - 4 directions

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Horizontal

+45 degrees

Vertical

-45 degrees

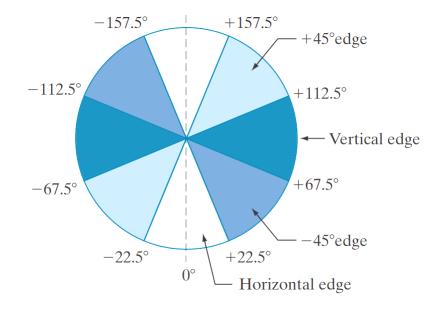
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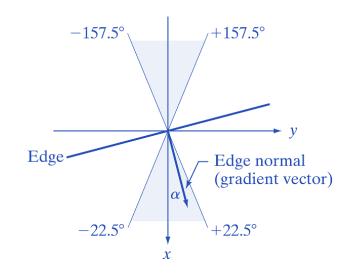
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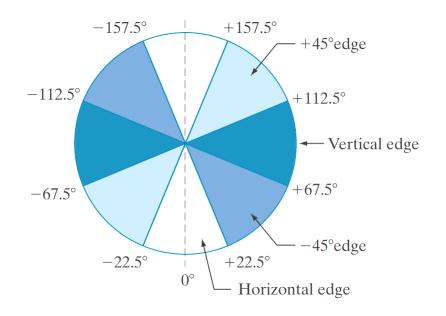
Horizontal

+45 degrees

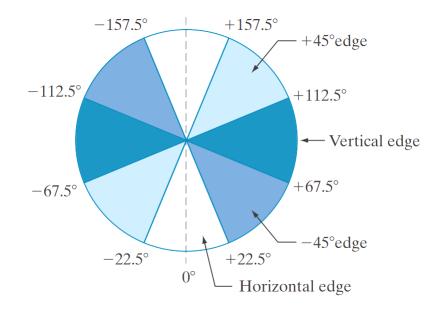
Vertical

-45 degrees





- Speeding up the beats of operations
 - \circ binning the α (angles)
 - get the directional bin Bin() closest to α
 - from previous operations edge: M(x, y)
 - suppression
 - If M(x', y') > M(x, y) then $M(x, y) \to 0$
 - where neighbors $x', y' \leftarrow Bin(x, y)$



- 1. Smoothing: Blurring of the image to remove noise.
- 2. **Finding gradients:** The edges should be marked where the gradients of the image has large magnitudes.
- 3. Non-maximum suppression: Only local maxima should be marked as edges.
- 4. **Double thresholding:** Potential edges are determined by thresholding.
- 5. Edge tracking by hysteresis: Final edges are determined by suppressing all edges that are not connected to a very certain (strong) edge.

Ref: P. Kalra

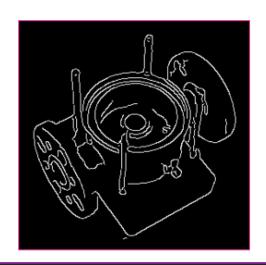
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Ref: P. Kalra

• Varying σ

input



• Varying σ

input





• Varying σ

input







• Varying σ

input σ small







• Varying σ

input σ small σ large







Comparing other edge detectors

input

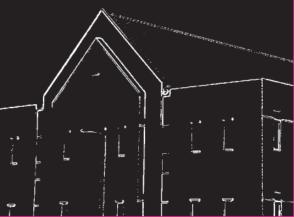


Comparing other edge detectors

input

Sobel with TH





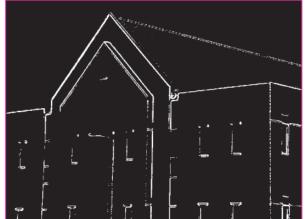
Comparing other edge detectors

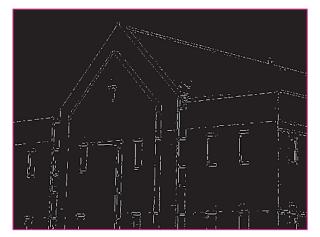
input

Sobel with TH

LoG zero crossings





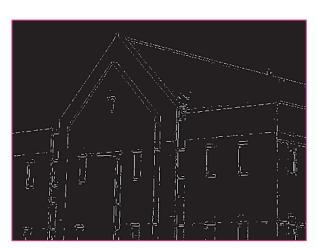


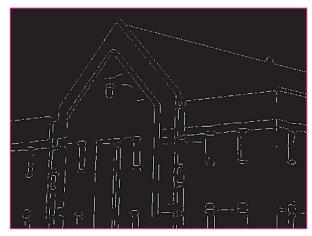
Comparing other edge detectors

input Sobel with TH LoG zero crossings Canny



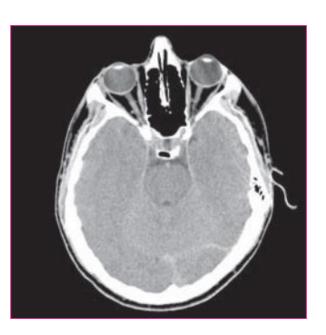






Comparing other edge detectors

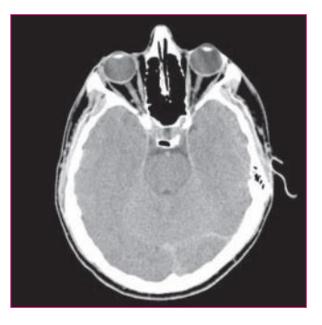
input

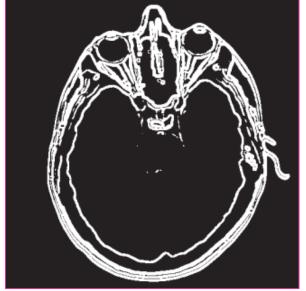


Comparing other edge detectors

input

Sobel with TH





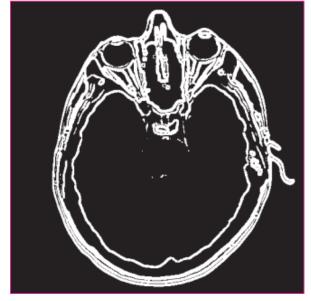
Comparing other edge detectors

input

Sobel with TH

LoG zero crossings







Comparing other edge detectors



Sobel with TH

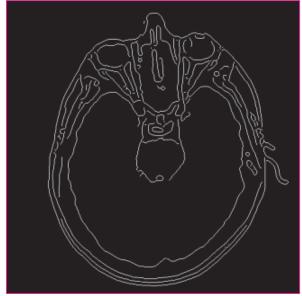
LoG zero crossings

Canny









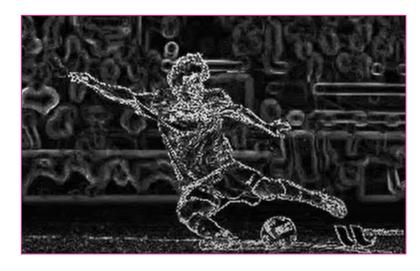




Messi

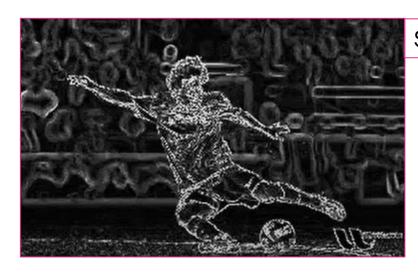


Messi





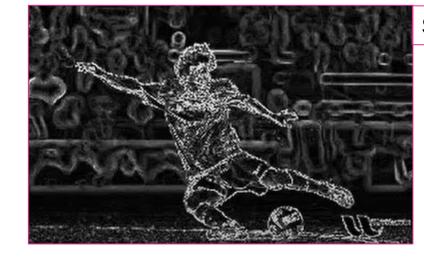
Messi



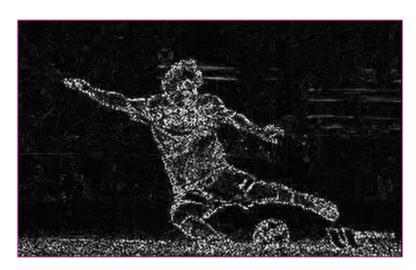
Sobel



Messi

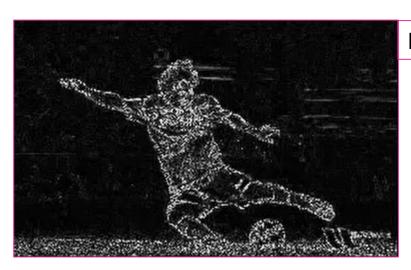


Sobel

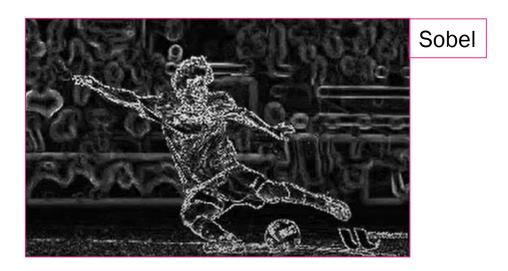




Messi

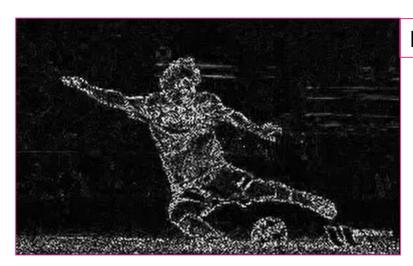


Laplacian

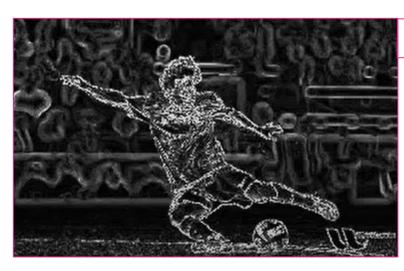




Messi



Laplacian

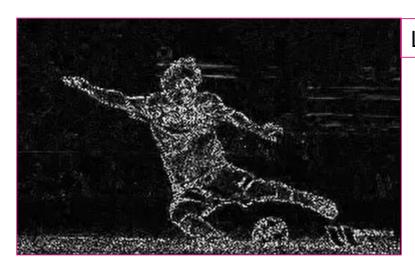


Sobel

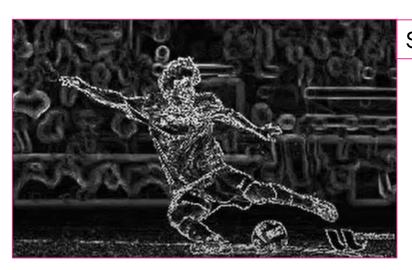




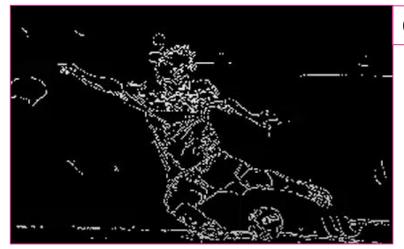
Messi



Laplacian



Sobel

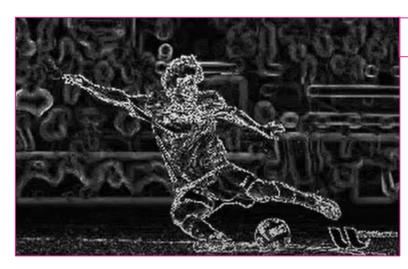


Canny

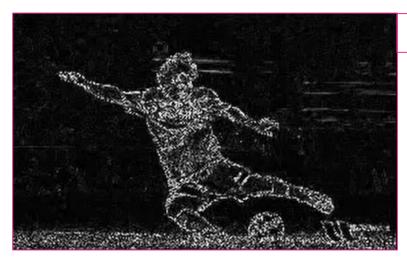


Messi

who will you go with?



Sobel



Laplacian



Canny

EE604: IMAGE PROCESSING

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- ☐ Single point thick edges
- Canny operations
 - Thinning: non-max suppression
 - Linking: double TH hysteresis
 - High accuracy is paid via computational expenses

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