# Edge: Canny

Dr. Tushar Sandhan

Single point thick edges

input



Single point thick edges

input



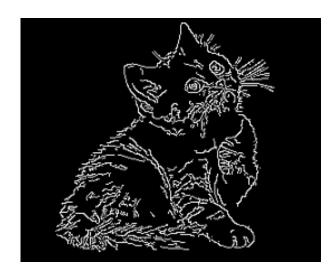


Single point thick edges

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Canny edges



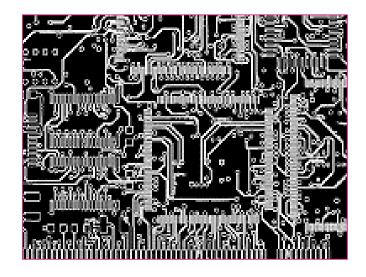
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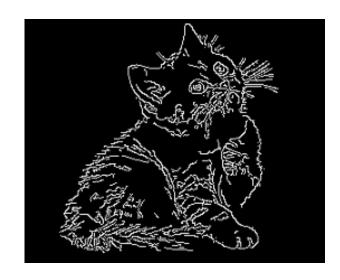


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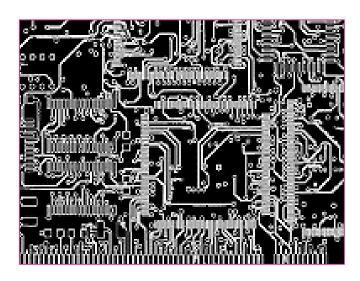
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Canny edges



Canny PCB edges

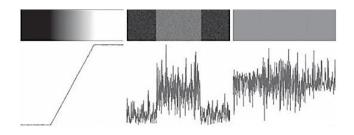


What would be important steps in edge det.

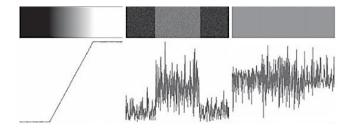
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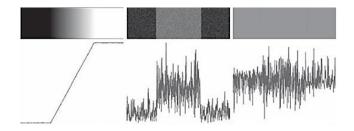
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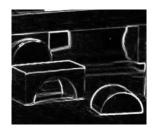


- What would be important steps in edge det.
  - Smooth derivatives
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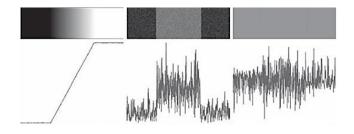


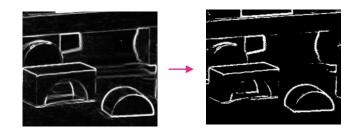
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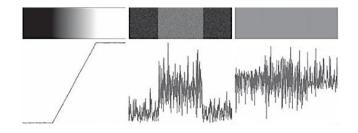


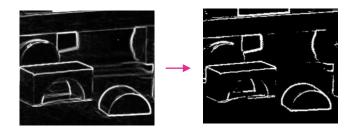
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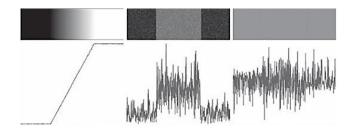


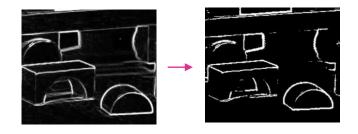
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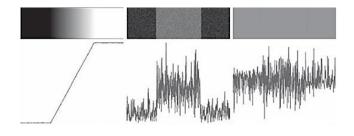


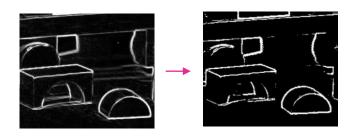
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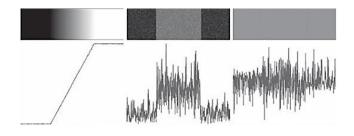
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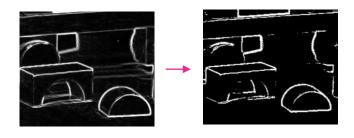


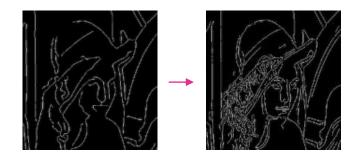




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Objectives

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  - 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  $g_x(x, y)$ ,  $g_y(x, y)$

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- Thinning
  - o  $M_s(x,y)$  wide ridges around local maxima
  - o ridges thinning is needed
  - non-max suppression
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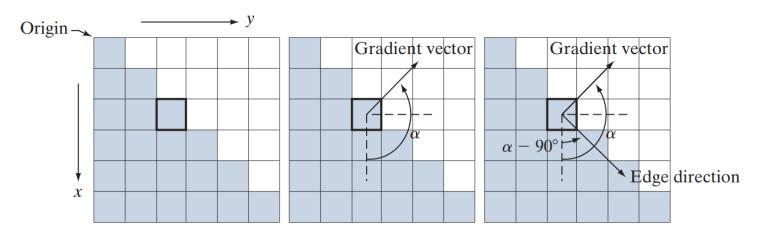
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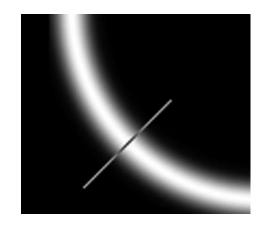


#### Thinning

- non-max suppression:
   checks whether pixel is local maxima in grad direction
- linear interpolation for missing locations e.g. r, p

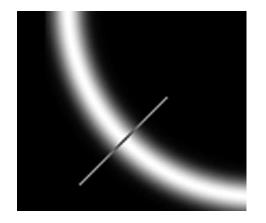
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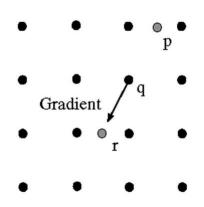
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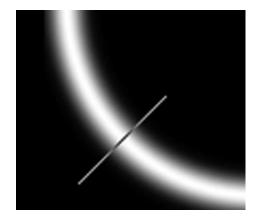
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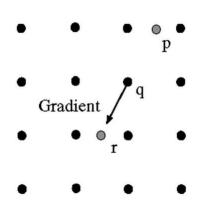


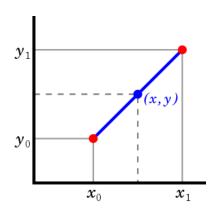


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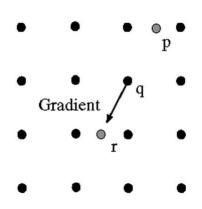


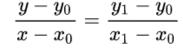


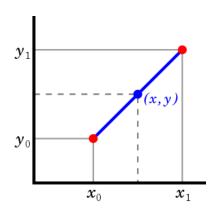


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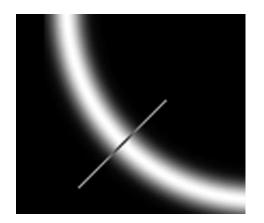


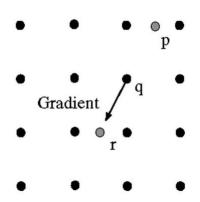




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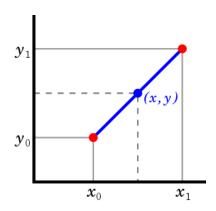
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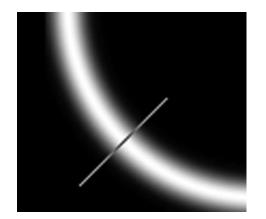
$$\frac{y-y_0}{x-x_0} = \frac{y_1-y_0}{x_1-x_0}$$

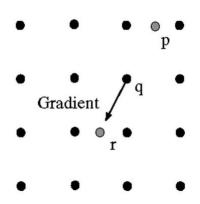
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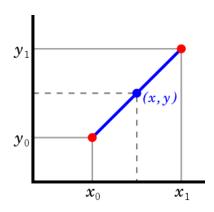


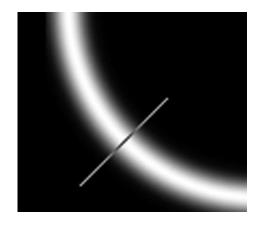


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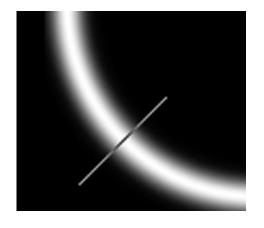
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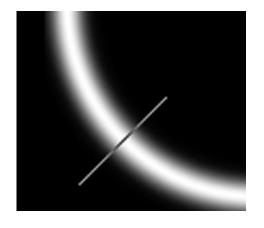




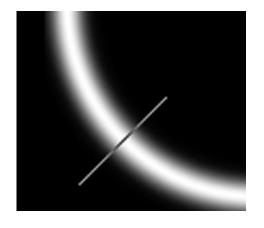
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  - Canny edge detector
    - It starts with one thing: gradients
    - In the end, it doesn't even matter: which operators have been used
  - o at last we have to link the non-suppressed ones!



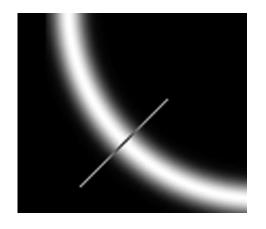
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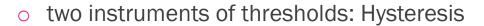


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  - two instruments of thresholds: Hysteresis

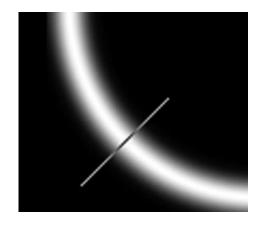


#### LINKINg Points

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



Entire algorithm composition:

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  - 1. Filter image with derivatives of Gaussian
  - 2. Get  $M, \alpha$
  - 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  $\alpha$  (angles)
  - 4 directions

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Horizontal

+45 degrees

Vertical

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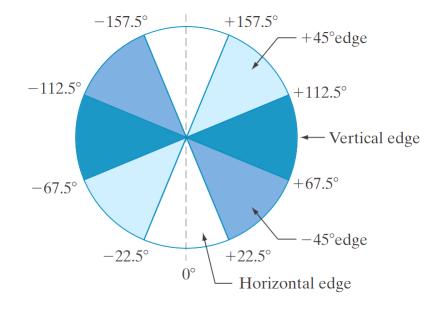
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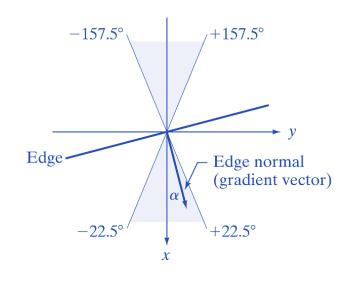
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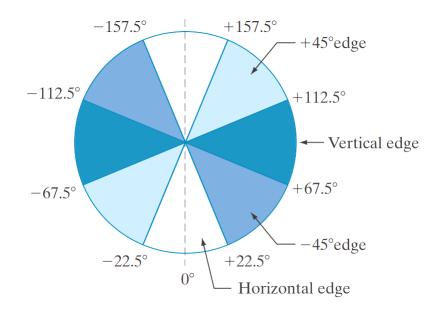
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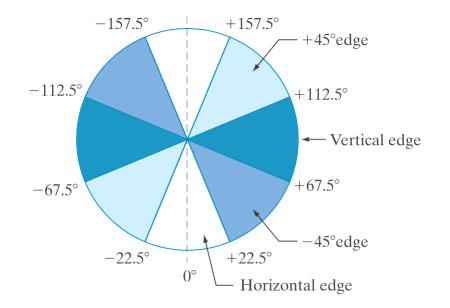
Vertical

-45 degrees





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



• Varying  $\sigma$ 

input



• Varying  $\sigma$ 

input





• Varying  $\sigma$ 

input







• Varying  $\sigma$ 

input  $\sigma$  small







• Varying  $\sigma$ 

input  $\sigma$  small  $\sigma$  large







Comparing other edge detectors

input



Comparing other edge detectors

input

Sobel with TH





Comparing other edge detectors

Sobel with TH LoG zero crossings



input





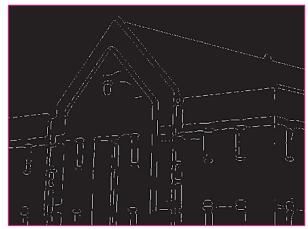
Comparing other edge detectors

input Sobel with TH LoG zero crossings Canny



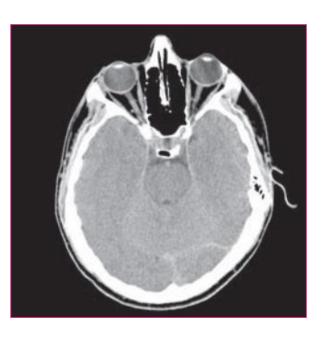






Comparing other edge detectors

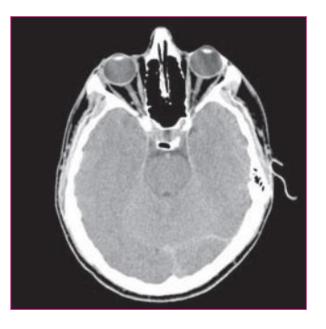
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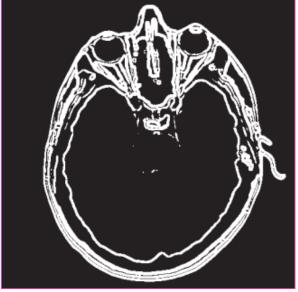


Comparing other edge detectors

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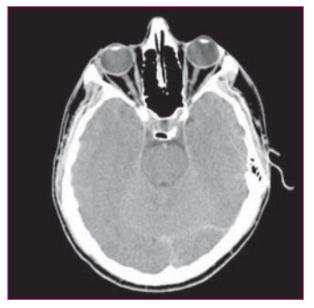


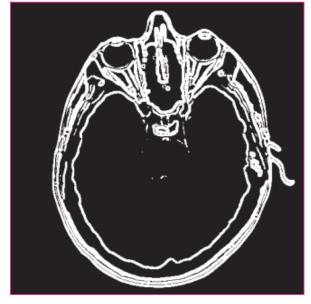
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Comparing other edge detectors



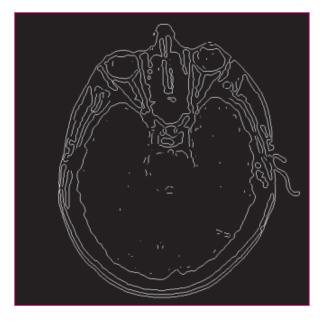
Sobel with TH

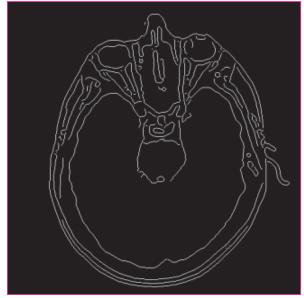
LoG zero crossings

Canny









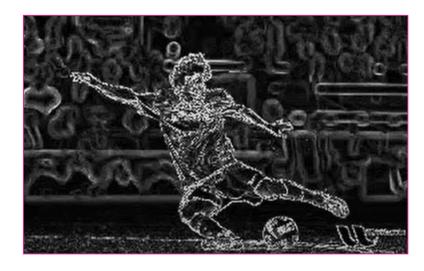




Messi

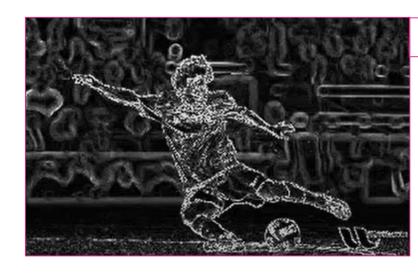


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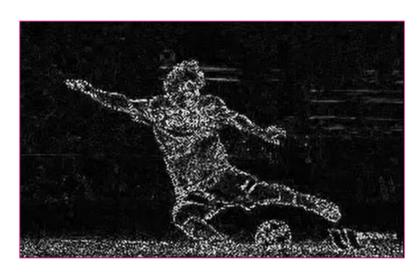
Sobel



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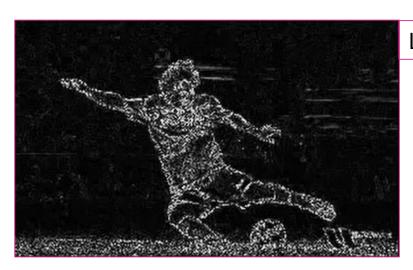


Sobel

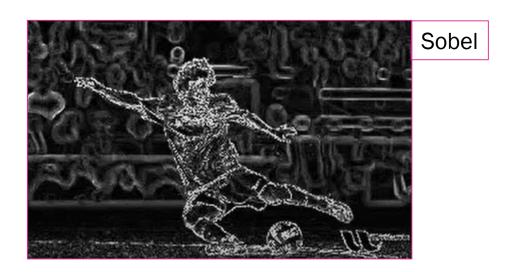




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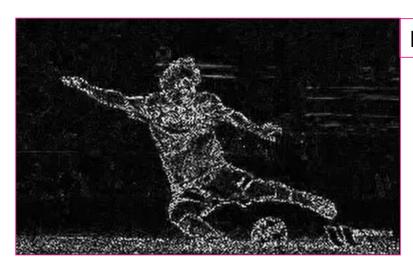


Laplacian

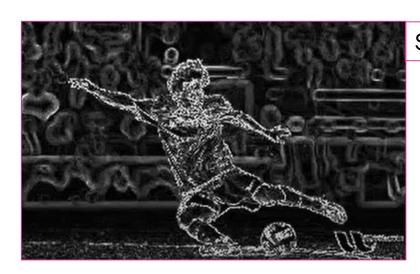


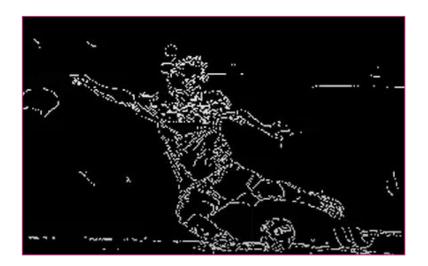


Messi



Laplacian



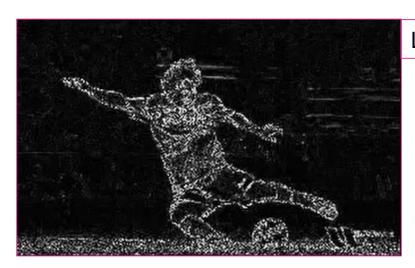


EE604: IMAGE PROCESSING sandhan@iitk.ac.in

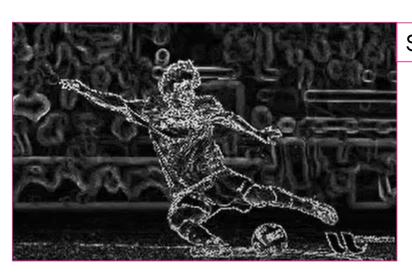
Sobel



Messi



Laplacian



Sobel



Canny

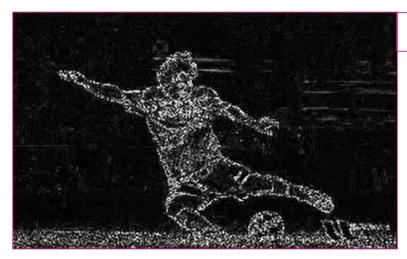


Messi

who will you go with?



Sobel



Laplacian



Canny

- Canny edge detector

Canny edge detector

- ☐ 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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