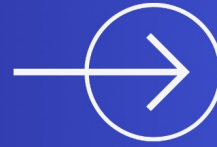
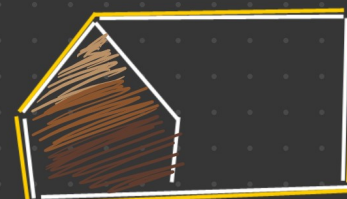
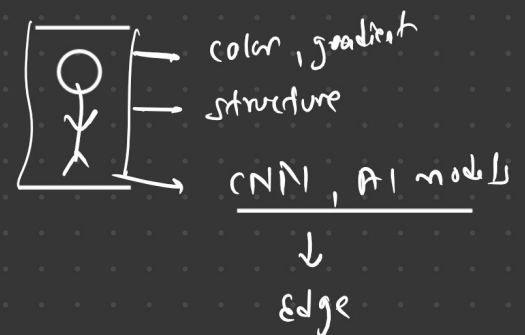
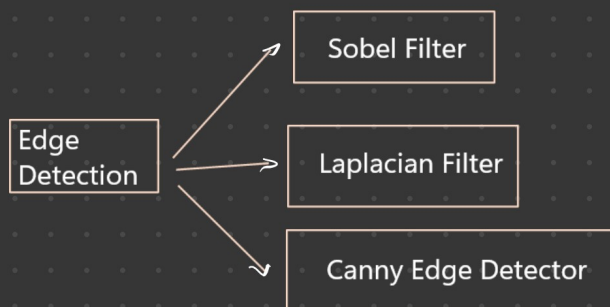
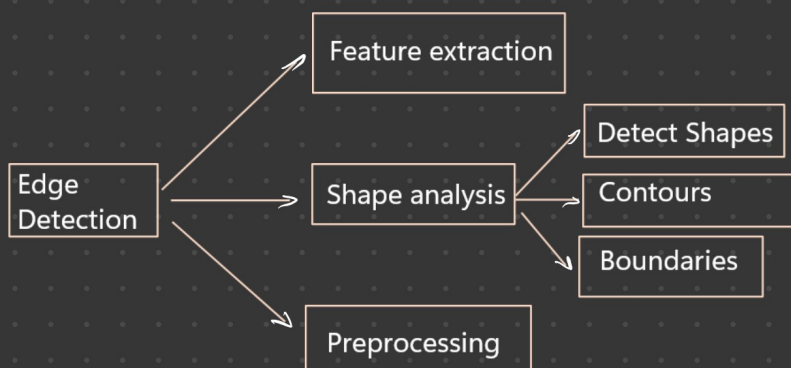


# Edge Detection Using Sobel, Canny & Laplacian



[www.krishnaik.in](http://www.krishnaik.in)

Edges represent sharp changes in intensity (or color) in an image. These changes often correspond to object boundaries, textures, or other key features.



# Sobel Filter

The Sobel filter detects edges in horizontal and vertical directions by calculating gradients. Below filters mathematically mirrors the derivative operation.

Horizontal kernel  $(G_x)$   $\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$       Vertical kernel  $(G_y)$   $\begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$

Convolution

$I_{G_x}, I_{G_y}$

Combining gradients

$$G = \sqrt{G_x^2 + G_y^2} \rightarrow \text{Calculate magnitude of gradient of each pixel}$$

$$G = \sqrt{I_{G_x}[0][0]^2 + I_{G_y}[0][0]^2} = \text{value}$$

Magnitude

$G =$

10	15	3	8
6	10	10	2
10	18	5	12
13	17	10	18

→ Avg → Ex. **10** ← Threshold


0	1	0	0
0	0	0	0
0	1	0	1
0	1	0	1

value if > threshold else 0


## Laplacian Filter

Laplacian detects edges by finding regions of rapid intensity change. Unlike Sobel, it combines x- and y-direction derivatives in a single operation.

The Laplacian kernel below is a discrete approximation of this continuous mathematical derivative. Below filters mathematically mirrors the derivate operation.


$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

Kernel 1


$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & -8 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Kernel 2

Both kernels are performing 2nd order Derivative

Laplacian detects sudden change in intensity, highlighting the edge.

## Canny Edge

The Canny algorithm detects edges using multi-step processing, combining Sobel and thresholding techniques.

Step 1 : Convert to Grayscale because edges are present in color change information not in colors.

Step 2 : Apply Gaussian Blur which adds smoothness to image making it less sensitive to small color change variations.

Step 3.1 : Compute Intensity Gradients using the Sobel operator to approximate the first derivatives in x and y-directions

Step 3.2 : Step 3.1 calculate magnitude but we also need angle of the edge and we can get using :

$$\theta = \arctan\left(\frac{y}{x}\right)$$

### Step 4 : Non-Maximum Suppression.

- For each pixel, compare the gradient magnitude to its two neighbors along the gradient direction
- If the pixel's gradient is not the largest, suppress (set to 0).



Example:

Gradient Direction:  $45^\circ$

compare  $G(x, y) \rightarrow G(x+1, y+1)$   
 $\searrow$   
 $G(x-1, y-1)$

This keeps only the strongest edge points.

## Step 5 : Double Thresholding

- High Threshold: Pixels above this are strong edges.
- Low Threshold: Pixels below this are suppressed entirely.
- Pixels between the two thresholds are classified as weak edges.

- High threshold = 100

- Low threshold = 50

- $G > 100$  : Strong edge

- $50 < G \leq 100$  : Weak edge

- G <= 50 Suppressed (Not and edge)

## Step 6 : Edge Tracking by Hysteresis

- In this final step, weak edges are either kept or discarded based on their connectivity to strong edges.
- If a weak edge is connected to a strong edge, it is Retained else Discarded.

Canny combines several techniques (Gaussian blur, Sobel, thresholds) to ensure edges are clean & noise-free.