

Edge detection

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Edges

- Edges are significant local changes of intensity in a digital image.
- An edge can be defined as a set of connected pixels that forms a boundary between two disjoint regions.
- There are three types of edges:
 - Horizontal edges
 - Vertical edges
 - Diagonal edges





Edge Detection

- Edge Detection is a method of segmenting an image into regions of discontinuity. It is a widely used technique in digital image processing like
 - pattern recognition
 - image morphology
 - feature extraction
- Edge detection allows users to observe the features of an image for a significant change in the gray level. This texture indicating the end of one region in the image and the beginning of another.
- It reduces the amount of data in an image and preserves the structural properties of an image.





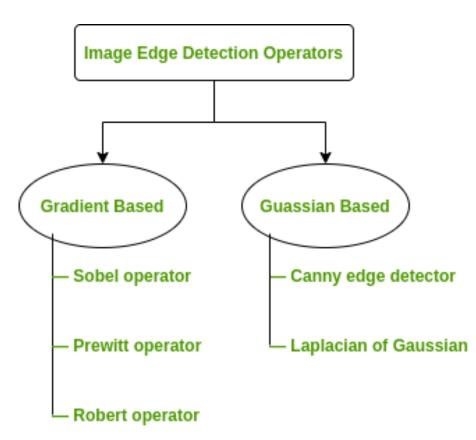
Edge Detection

- Edge Detection Operators are of two types:
 - Gradient
 - based operator which computes first-order derivations in a digital image like, Sobel operator, Prewitt operator, Robert operator
 - Gaussian
 - based operator which computes second-order derivations in a digital image like, Canny edge detector, Laplacian of Gaussian



Edge Detection









Edge Detection: How?

- Image smoothing: The image is first smoothed to reduce noise and enhance edges. This is often done using filters like Gaussian blur.
- Gradient calculation: The gradient of the image is calculated, which measures the change in intensity between neighboring pixels. Edges are typically located where the gradient magnitude is high.
- Edge thresholding: A threshold is applied to the gradient magnitude to identify pixels that are likely to be part of an edge.
- Edge linking: The identified edge pixels are connected to form continuous edge lines.



Edge Detection?







Sobel Edge Detection



- It is a discrete differentiation operator. It computes the gradient approximation of image intensity function for image edge detection.
- At the pixels of an image, the Sobel operator produces either the normal to a vector or the corresponding gradient vector.
- It uses two 3 x 3 kernels or masks which are convolved with the input image to calculate the vertical and horizontal derivative approximations respectively



Sobel Edge Detection



 It uses two 3 x 3 kernels or masks which are convolved with the input image to calculate the vertical and horizontal derivative approximations respectively –

$$M_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad M_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$





• Convolution with kernels:

- The Sobel algorithm uses two 3x3 kernels, one for horizontal edges and one for vertical edges. Each kernel element has a specific weight assigned to it.
- These kernels are convolved with the grayscale image, meaning they are slid across the image, and the weighted sum of the corresponding pixel values under the kernel mask is calculated at each pixel location.
- The horizontal kernel emphasizes intensity changes in the horizontal direction, while the vertical kernel emphasizes changes in the vertical direction.





- Gradient calculation:
 - The convolution with each kernel results in two gradient images, one representing the horizontal gradient and another representing the vertical gradient.
 - Each pixel in these gradient images holds the magnitude of the intensity change in the corresponding direction.





- Combining gradients:
 - The final edge strength at each pixel is calculated by combining the horizontal and vertical gradients. This can be done using different approaches:
 - Magnitude: Square root of the sum of squared horizontal and vertical gradients.
 - Gradient direction: arctangent of the vertical gradient divided by the horizontal gradient.





- A threshold is applied to the combined gradient image to classify pixels as edge pixels or nonedge pixels.
- Pixels with gradient values exceeding the threshold are considered edges.
- The threshold value can be chosen manually or determined automatically using techniques like Otsu's method.





- Post-processing (optional):
 - Additional steps like non-maximum suppression can be applied to refine the detected edges by removing single-pixel edge segments and ensuring only the strongest edges remain.



Sobel Edge Detection:



Practical





- Scharr edge detection is an image processing technique used to identify and highlight edges or boundaries of objects within an image.
- It is a type of gradient-based edge detection, which means it calculates the magnitude and direction of the gradient at each pixel in the image.
- Edges are then identified as pixels with high gradient magnitudes.





- The Scharr operator is a 3x3 filter that is applied to the image to calculate the gradient.
- The filter kernel for the Scharr operator in the xdirection is:

```
[-3 0 3]
[-10 0 10]
[-3 0 3]
```





• The filter kernel for the Scharr operator in the ydirection is:

[-3 -10 -3] [0 0 0] [3 10 3]





- The Scharr operator has several advantages over other gradient-based edge detectors, such as the Sobel operator.
 - First, it is more rotationally symmetric, which means that it is less sensitive to the orientation of edges in the image.
 - Second, it has a slightly higher response to edges, which can improve the accuracy of edge detection.







- Convolution with Scharr filters:
 - Applies 3x3 Scharr filters to the image to compute gradients in two directions:
 - Horizontal filter (dx = 1, dy = 0) emphasizes vertical edges.
 - Vertical filter (dx = 0, dy = 1) emphasizes horizontal edges.
- Gradient magnitude calculation:
 - Combines the horizontal and vertical gradients using the following formula: gradient_magnitude = sqrt(dx^2 + dy^2)





Practical





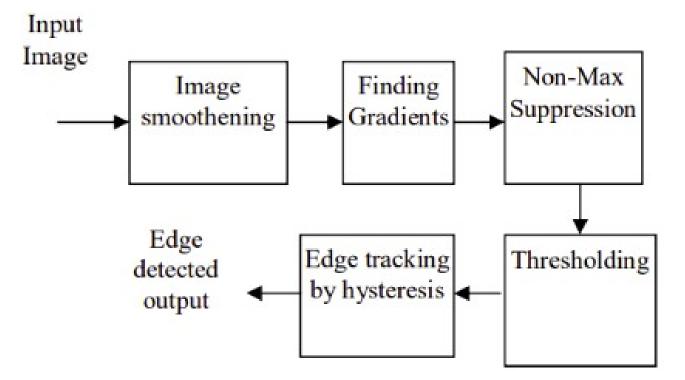
Canny Edge Detection

- A multi-stage edge detection algorithm developed by John F. Canny in 1986.
- Aims to find wide range of edges while considering factors like:
 - Good detection: Finding real edges accurately.
 - Localization: Placing detected edges precisely.
 - Minimizing single responses: Not detecting the same edge multiple times.
 - Minimizing noise: Not detecting noise as edges.



Canny Edge Detection











skillologies

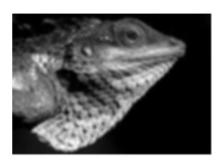




• Image Smoothening

- In this step, we convert the image to grayscale as edge detection does not dependent on colors.
- Then we remove the noise in the image with a Gaussian filter as edge detection is prone to noise.









- Finding Intensity Gradients of the Image
 - We then apply the Sobel kernel in horizontal and vertical directions to get the first derivative in the horizontal direction (Gx) and vertical direction (Gy) on the smoothened image. We then calculate the edge gradient(G) and Angle(θ) as given below,

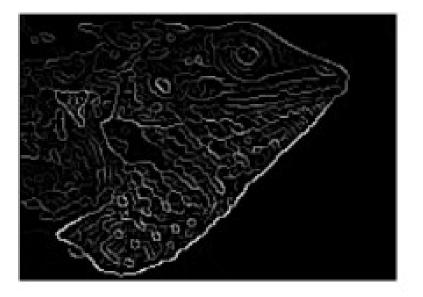
Edge_Gradient(G) = $\sqrt{(Gx^2+Gy^2)}$

Angle(θ)=tan⁻¹(Gy/Gx)

 We know that the gradient direction is perpendicular to the edge. We round the angle to one of four angles representing vertical, horizontal, and two diagonal directions.





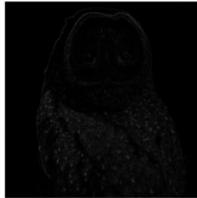


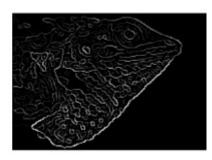
skillologies





- Non-Max Suppression
 - Now we remove all the unwanted pixels which may not constitute the edge.
 - For this, every pixel is checked in the direction of the gradient if it is a local maximum in its neighbourhood.
 - If it is a local maximum, it is considered for the next stage, otherwise, it is darkened with 0. This will give a thin line in the output image.





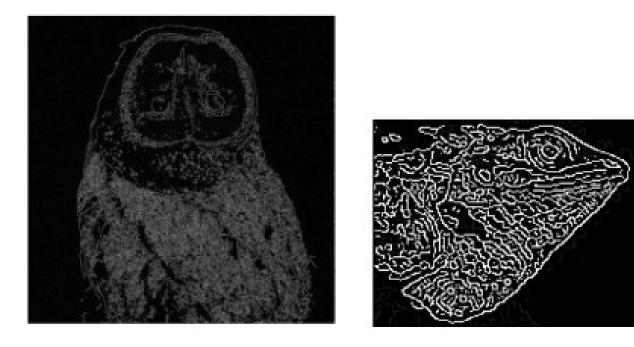




- Double Threshold
 - Pixels due to noise and color variation would persist in the image. So, to remove this, we get two thresholds from the user, lowerVal and upperVal.
 - We filter out edge pixels with a weak gradient(lowerVal) value and preserve edge pixels with a high gradient value(upperVal).
 - Edges with an intensity gradient more than upperVal are sure to edge, and those below lowerVal are sure to be non-edges, so discarded.
 - The pixels that have pixel value lesser than the upperVal and greater than the lowerVal are considered part of the edge if it is connected to a "sure-edge". Otherwise, they are also discarded.



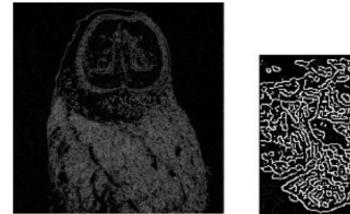
skillologies







- A pixel is made as a strong pixel if either of the 8 pixels around it is strong(pixel value=255) else it is made as 0.
- That's pretty much about Canny Edge Detection. As you can see here, the edges are detected from an image.

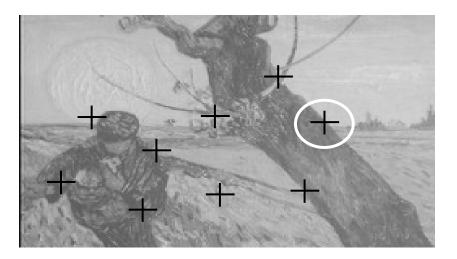






Interest Points

- A point in an image which has a well-defined position and can be robustly detected.
- Typically associated with a significant change of one or more image properties simultaneously (e.g., intensity, color, texture).

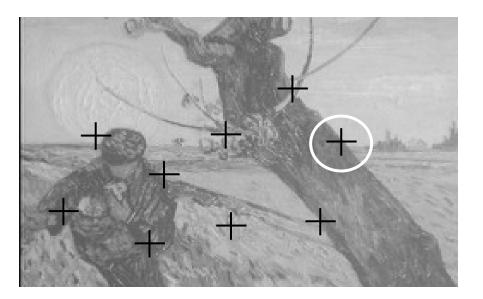




Interest Points



- Corners is a special case of interest points.
- However, interest points could be more generic than corners.





Interest Points: Why useful?

 Could be used to find corresponding points between images which is very useful for numerous applications!

stereo matching

left camera



right camera





oloaies







Corner Detection

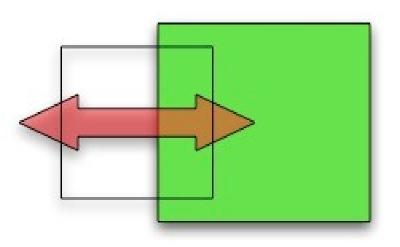
- Corner detection is very useful for shape description and matching.
- Corner detection works on the principle that if you place a small window over an image, if that window is placed on a corner then if it is moved in any direction there will be a large change in intensity.
- If the window is over a flat area of the image then there will be obviously be no intensity change when the window moves.
- If the window is over an edge there will only be an intesity change if the window moves in one direction.







 If the window is over a corner then there will be a change in all directions, and therefore we know there must be a corner.

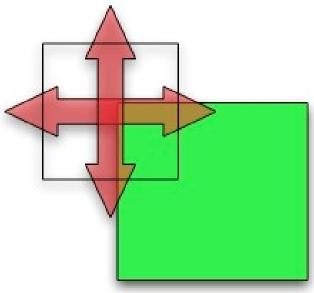






Corner Detection

- The Harris corner detector, measures the strength of detected corners, and only marks those above a given strength as actual corners.
- The number detected can be altered by varying the value of k.

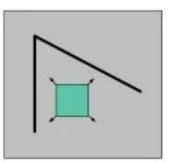




Corner Detection



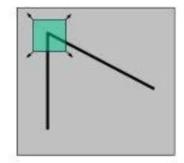
 Corners are the important features in the image, and they are generally termed as interest points which are invariant to translation, rotation, and illumination.



"flat" region: no change in all directions



"edge": no change along the edge direction



"corner": significant change in all directions



Harris Corner Detection



- 1. Take the grayscale of the original image
- 2. Apply a Gaussian filter to smooth out any noise
- 3. Apply Sobel operator to find the x and y gradient values for every pixel in the grayscale image
- 4. For each pixel p in the grayscale image, consider a 3×3 window around it and compute the corner strength function. Call this its Harris value.
- 5. Find all pixels that exceed a certain threshold and are the local maxima within a certain window (to prevent redundant dupes of features)
- 6. For each pixel that meets the criteria in 5, compute a feature descriptor.



Harris Corner Detection



Practical



Thank you

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