

# IMAGE PROCESSING (EDGE DETECTION)

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# Introduction

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# Introduction



- Edges occur at boundaries between regions of different color, intensity, or texture.
- An edge is a location of rapid intensity variation.
- Detecting edges can be helpful for segmenting an image into coherent regions.
- Reduce the amount of data to be processed, disposing less relevant information.
- Important to feature detection and feature extraction.



# Edge Detection and Segmentation using graph searches

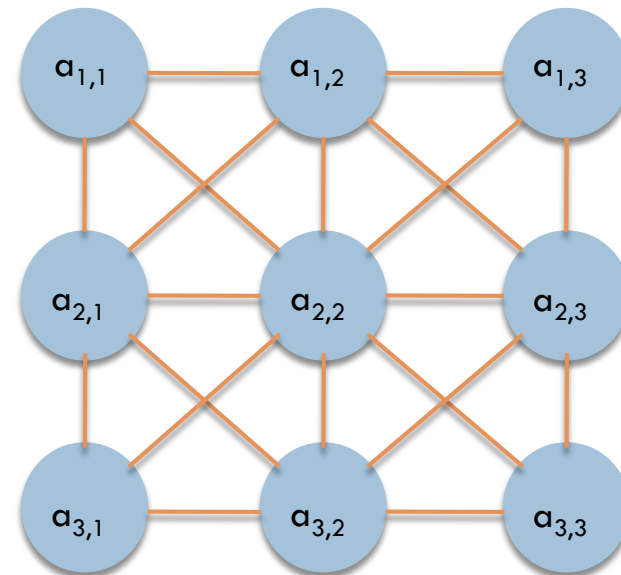
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# Visualization

- If we see an image as a graph, with the pixels as vertices, we can do a search in the image in the same way that we do in a graph.

$a_{1,1}$	$a_{1,2}$	$a_{1,3}$
$a_{2,1}$	$a_{2,2}$	$a_{2,3}$
$a_{3,1}$	$a_{3,2}$	$a_{3,3}$



# Edge Detection



- Visit each vertex and edge of the graph asking if there is a significant change in the intensity in at least one of its neighbors.
- If a certain pixel has at least one neighbor with a significant change in the intensity level, then consider that pixel as an edge.
- Noise can be interpreted as an edge with this approach.

# Edge Detection



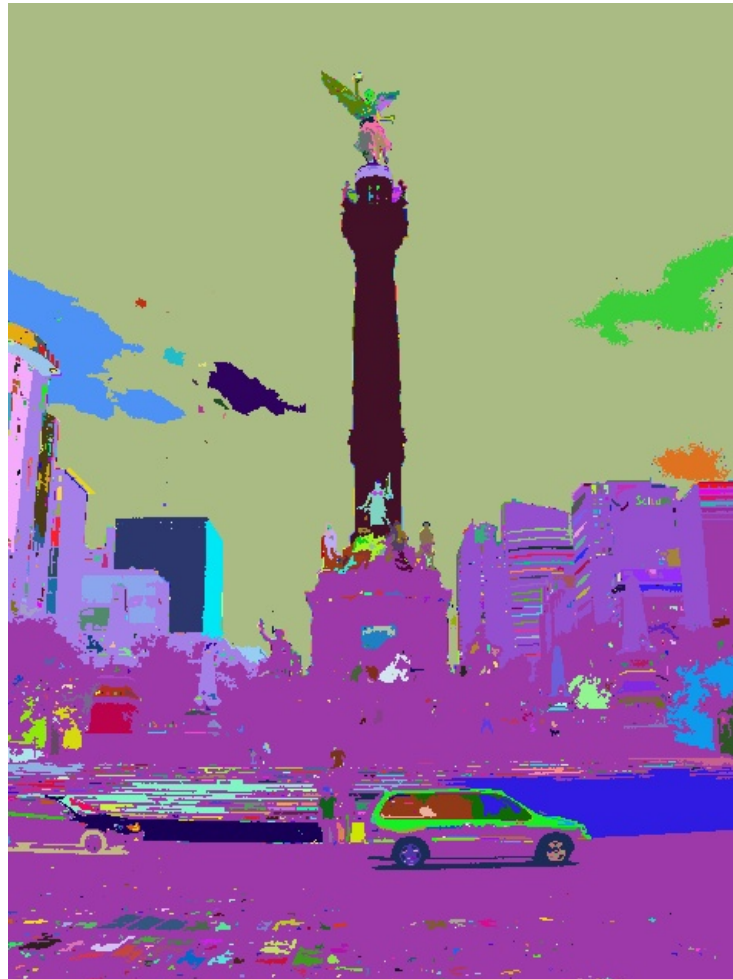
# Segmentation



- Define an initial node and mark it with a color.
- Start a Depth First Search (DFS) starting in that node.
- If the intensity of current node does not present a significant change with the intensity of the initial node, then mark it with the same color.
- The algorithm depends in the selection of the initial vertex.



# Segmentation





# Sobel Operator

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# Sobel Operator

- Commonly used in edge detection algorithms.
- Computes an approximation of the gradient of the image intensity.
- At each point of the image, the magnitude of the gradient is calculated by:

$$|G| = \sqrt{G_x^2 + G_y^2}$$

- Where  $G_x$  and  $G_y$  are the first derivatives in x-direction and y-direction respectively.

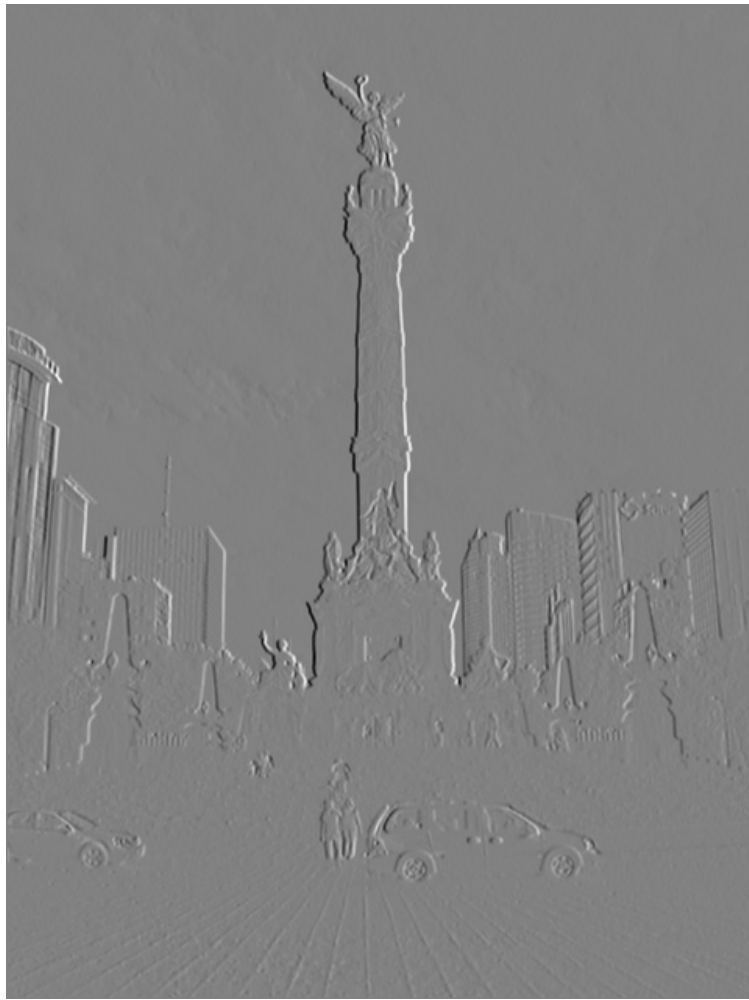
# Sobel Operator

- To obtain the first derivatives in both directions, the Sobel operator use the following kernels

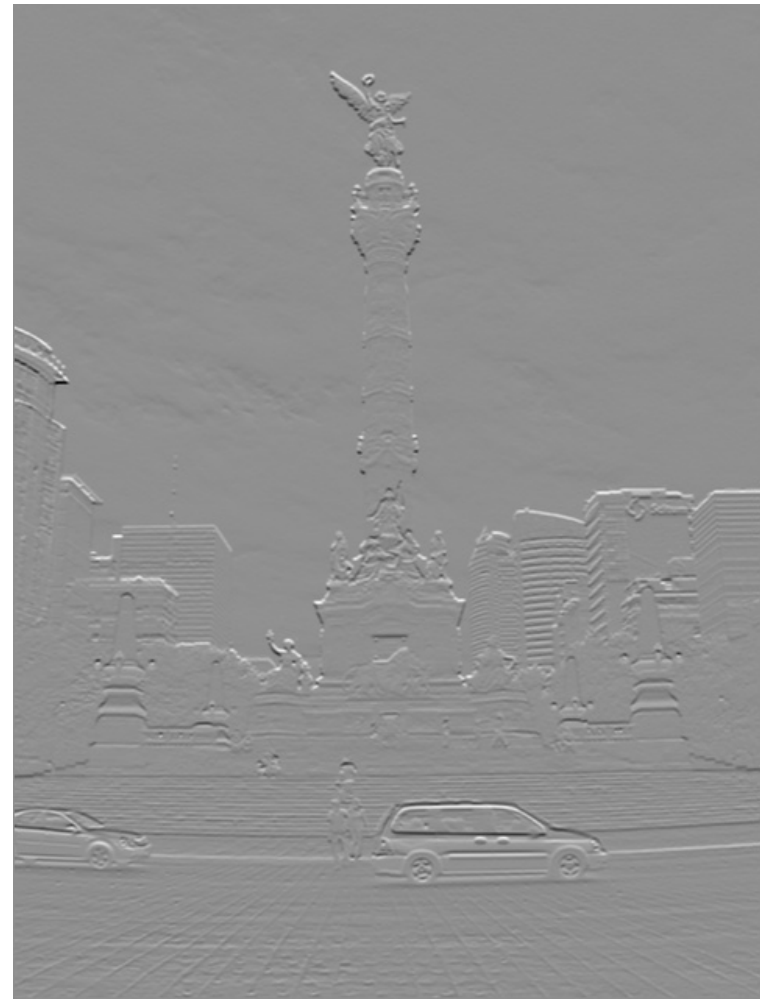
$$K_{Gx} = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad K_{Gy} = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

- The sobel operator convolve an image with the kernels to find the points where the intensity change abruptly.

# Sobel Operator



X Derivate



Y Derivate

# Sobel Operator (Final Result)





# Canny Edge Detector

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# Canny Edge Detector



- Developed by John F. Canny (JFC) in 1986.
- One of the most used algorithms in image processing for detecting edges.
- The goal of JFC was to implement the optimal edge detection algorithm, based on the following points:



# Canny Edge Detector



- **Good Detection:** The algorithm must be able to detect as many real edges as possible in the image.
- **Good Localization:** The detected edges should be marked as close as possible to the real edges.
- **Minimal Response:** A real edge must not result in more than one detected edge, and image noise must not affect in the detection of new edges.

# Canny Edge Detector



- Canny Edge Detector functionality is based in these five steps:
  1. Smoothing
  2. Gradient Detection
  3. Non-maximum suppression
  4. Double Thresholding
  5. Edge Tracking

# Test Image



# Smoothing

- Gaussian Filter is used to reduce the amount of noise in the image.
  - ▣ Noise can be detected as edges by the algorithm.
- The image is then convolved with the Gaussian Filter.
- One commonly used kernel with standard deviation of  $\sigma = 1.4$  is the following:

$$\frac{1}{159} \begin{bmatrix} 2 & 4 & 5 & 4 & 2 \\ 4 & 9 & 12 & 9 & 4 \\ 5 & 12 & 15 & 12 & 5 \\ 4 & 9 & 12 & 9 & 4 \\ 2 & 4 & 5 & 4 & 2 \end{bmatrix}$$

# Smoothing



Original Image



Smoothed Image

# Gradient Detection

- Sobel operator is applied to the image.
  - ▣ Calculates the gradient of the image intensity.
  - ▣ The gradient points in the direction of the largest possible intensity increase.
  - ▣ Areas where intensity changes are notorious, like edges, are detected.
- Each point in the image is replaced by the magnitude of the gradient in the given point.
- The gradient is composed by the first derivatives in the x-direction ( $G_x$ ) and y-direction ( $G_y$ ).

# Gradient Detection

- The Sobel Filter uses these two kernels to convolve the image and calculate the first derivatives of the image:

$$K_{G_x} = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad K_{G_y} = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

- The magnitude of the gradient (G) is given by

$$|G| = \sqrt{G_x^2 + G_y^2}$$

# Gradient Detection



Before Sobel Operator



After Sobel Operator



# Non-maximum Supression

- Each point is compared with it's neighbors, depending on the direction of the gradient in the given point.
- For each point in the image, the direction of the gradient is calculated using the following expression:

$$\theta = \tan^{-1} \left( \frac{|G_x|}{|G_y|} \right)$$

# Non-maximum Supression

- Round the gradient direction  $\theta$  to the nearest  $45^\circ$ .
- Compare its value with the neighbor located in that direction and also in the opposite direction .
- If the intensity of the image in the given point is not greater than the intensity of its two neighbors, the value is suppressed.

# Non-maximum Supression

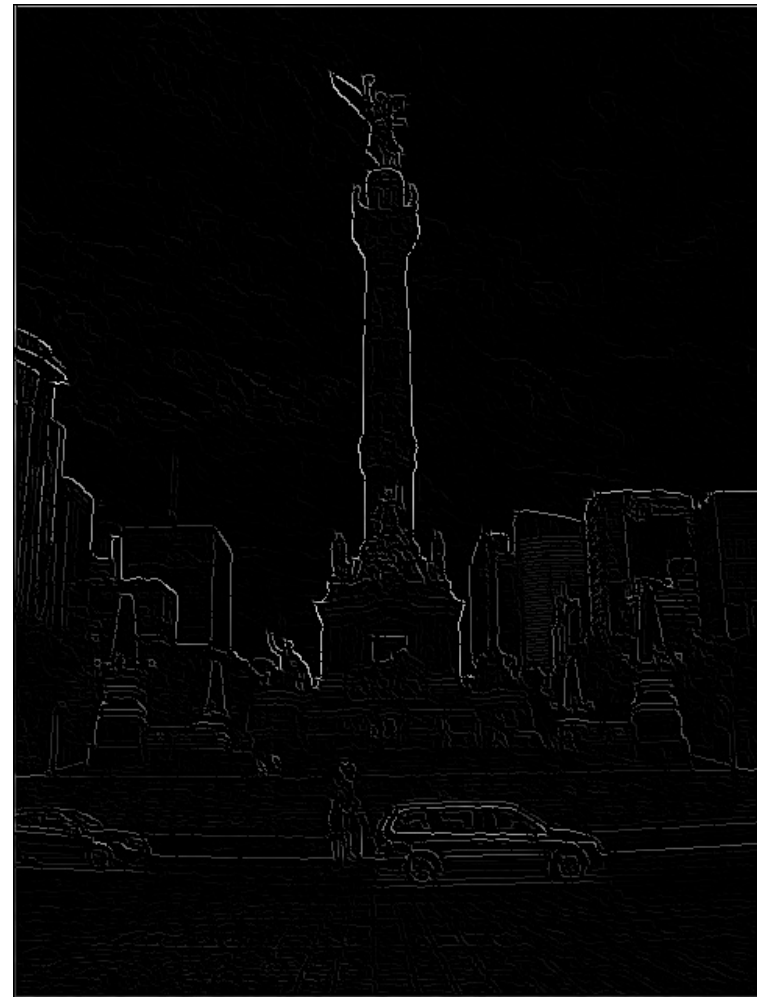


- The objective of doing this is to convert “blurred” edges into “sharp” edges.
- Edge pixels are preserved where the gradient has local maxima.

# Non-maximum Suppression



Before Non-maximum  
Supression



After Non-maximum  
Supression

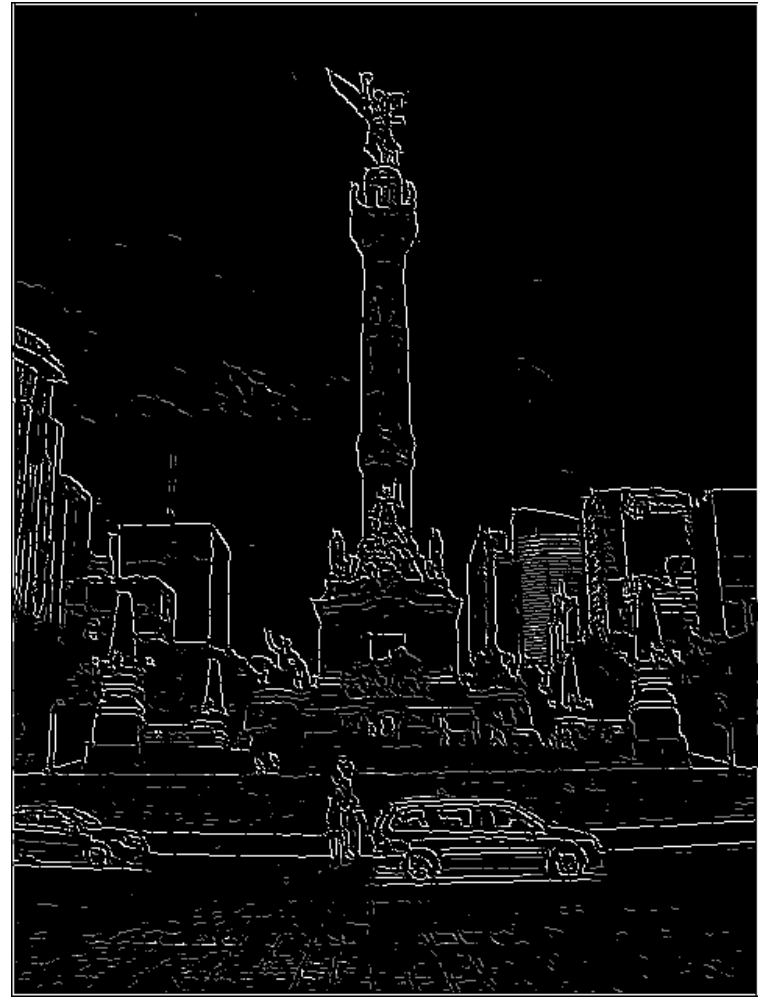
# Double Thresholding

- A high threshold value  $H_{th}$  is defined.
- A low threshold value  $L_{th}$  ( $L_{th} < H_{th}$ ) is defined.
- Values bigger or equal to  $H_{th}$  are painted in white (strong edge).
- Values between  $L_{th}$  and  $H_{th}$  are painted in gray (weak edge).
- Values smaller or equal to  $L_{th}$  are suppressed.

# Double Thresholding



Before Double Thresholding



After Double Thresholding

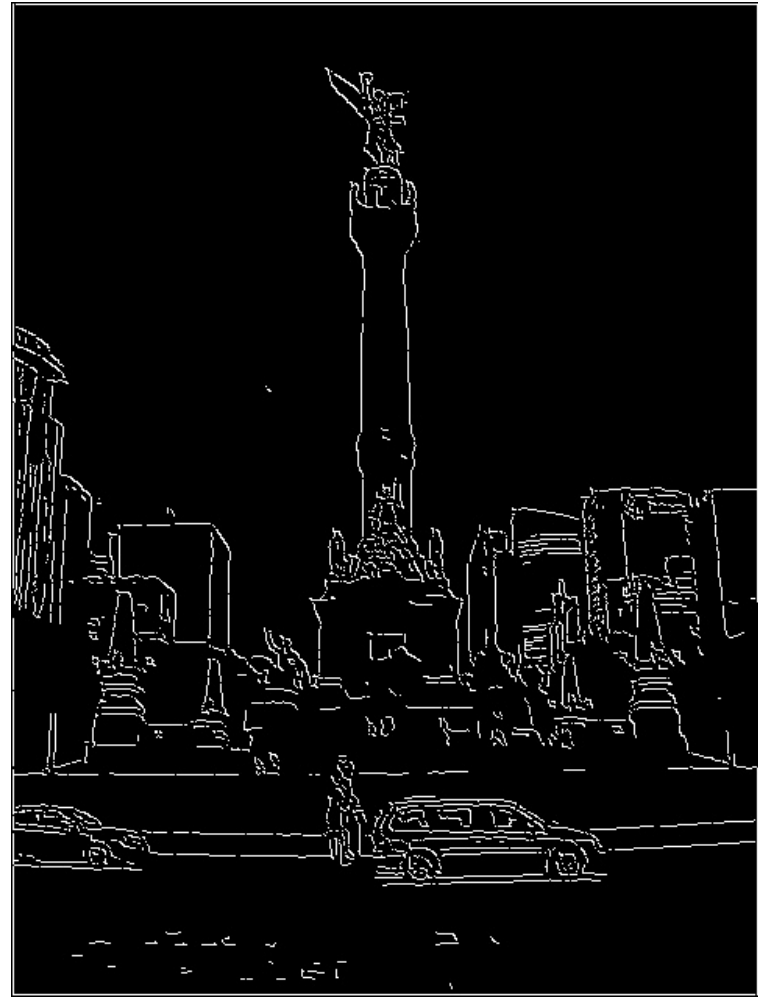
# Edge Tracking

- If a certain pixel of weak edge is not connected directly or indirectly to a strong edge, is suppressed. Otherwise is marked as a strong edge.
- The image can be seen as a graph, with each pixel representing a vertex connected to its neighbors.
- A depth first search (DFS) can be used to track the edges

# Edge Tracking



Before Edge Tracking



After Edge Tracking



# Result



Original Image

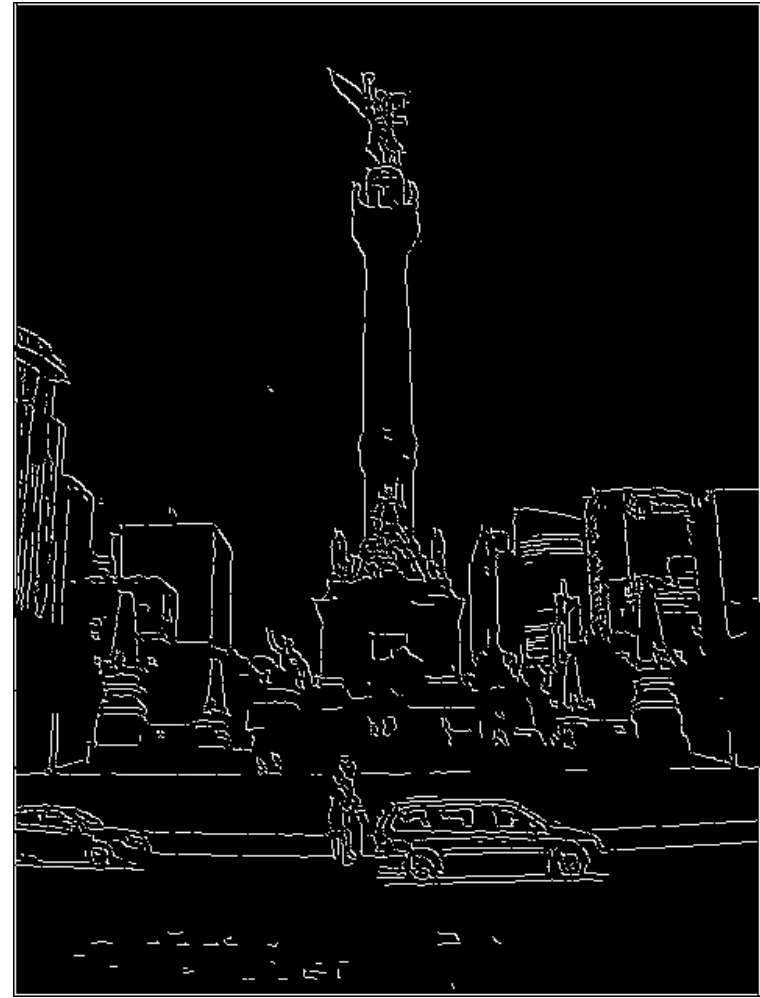


Image after applying Canny  
Edge Detector



# Future Work

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# Future Work



- Study and implement the most common methods for image processing
  - ▣ Edge Detection
  - ▣ Corner Detection
  - ▣ Segmentation
  - ▣ Filters
  - ▣ Etc.
- Continue studying about the Dirichlet Process and machine learning techniques to apply them in image segmentation algorithms.