

# COMPUTER VISION (CAP5415)-PA-1 REPORT

## CANNY EDGE DETECTION

**Step 1:** Firstly, we need to read a gray scale image from the given dataset 'Berkeley Segmentation Dataset' and that image is stored in a matrix named 'I'. In addition, I created a definition to take the input URL automatically without any manual intervention.

**Step2:** I have used the kernel length as 7 i.e., from (-3 to +3) and it is iterated with every position w.r.t gaussian filter

**1D gaussian distribution is stated as follows:**

$$G(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-x^2}{2\sigma^2}}$$

**Step 3:** Here, I performed convolution of 2Dimensional input image along with 1Dimensional Gaussian kernel in both the x and y directions. Named them as Ix and Iy (Iy is transpose of Ix).

**Step 4:** I've used the gaussian derivative formulae as listed below to get the gaussian derivative G(x) which is in X axis direction and applied the transpose of G(x) to get G(y) (where G(y) is gaussian derivative in Y axis direction).

$$G'(x) = \frac{-x}{\sqrt{2\pi}\sigma^3} e^{\frac{-x^2}{2\sigma^2}}$$

**Step 5:** Convolved I(X) and G(X) and named it as (I\_X). Similarly convolved I(y) and G(y) and named it as (I\_Y) {I used the transpose of X to get the I\_Y}.

**Step 6:** I calculated the magnitude of the gradient of each pixel in X and Y direction to detect the edges in an image.

$$M(x,y) = \sqrt{G'(x)^2 + G'(y)^2}$$

**Step 7:** Magnitude, direction and angle is calculated according to each pixel in order to create a Non-Maximum suppression function to identify local maxima containing highest gradient magnitude along the edges by classifying true/ false edges.

**Suppressing the noise direction is calculated using below formulae:**

$$\theta_{x,y} = \arctan(G'(x), G'(y))$$

**Step 8:** Here we used thresholding to identify strong edges and marked them as 1 and weak edges as 0. Where highest value is given as (20% of maximum pixel value) and lowest value (7% of maximum pixel value). Stored the final output image as hysteresis threshold which indeed handled noise and improvising the detected edges quality.

### **CONCLUSION:**

To encapsulate, I took 3 standard deviations/ sigma (0.5,2,4) for my training input images out of which I found that, the lowest sigma value (0.5) handled noise and improvised the detected edges quality. So, I convey that lowest sigma value gave me the best result.

## Output Images:

### Image 1

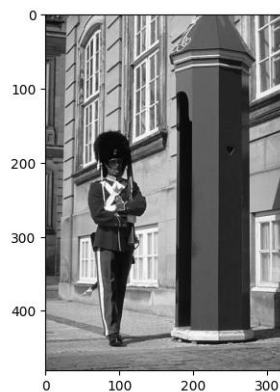
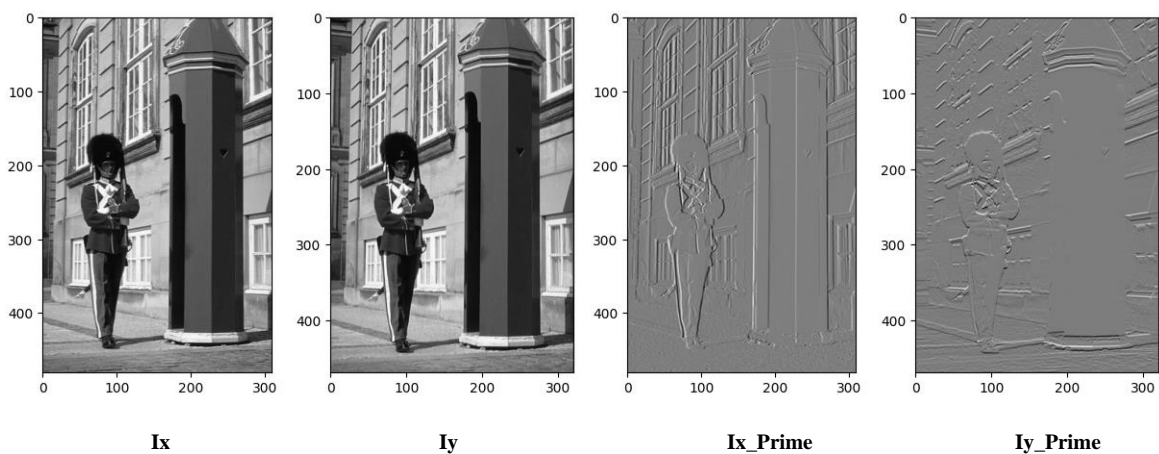


Fig.1 Original image

**Sigma = 0.5**

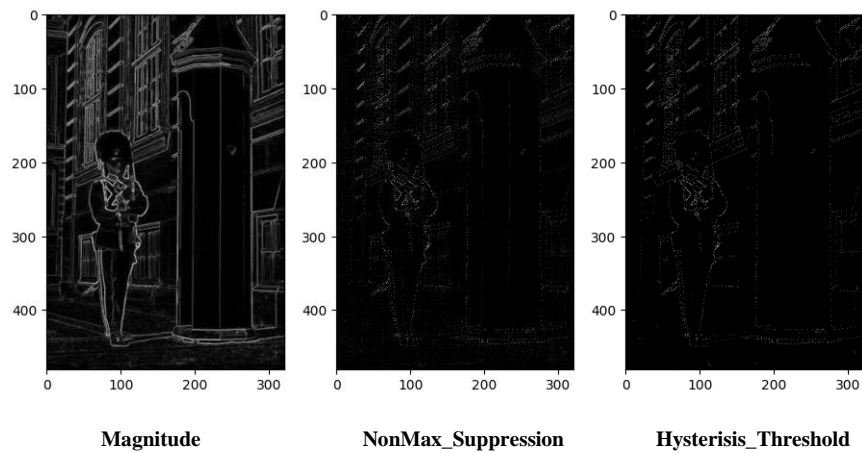


**I<sub>x</sub>**

**I<sub>y</sub>**

**I<sub>x\_Prime</sub>**

**I<sub>y\_Prime</sub>**

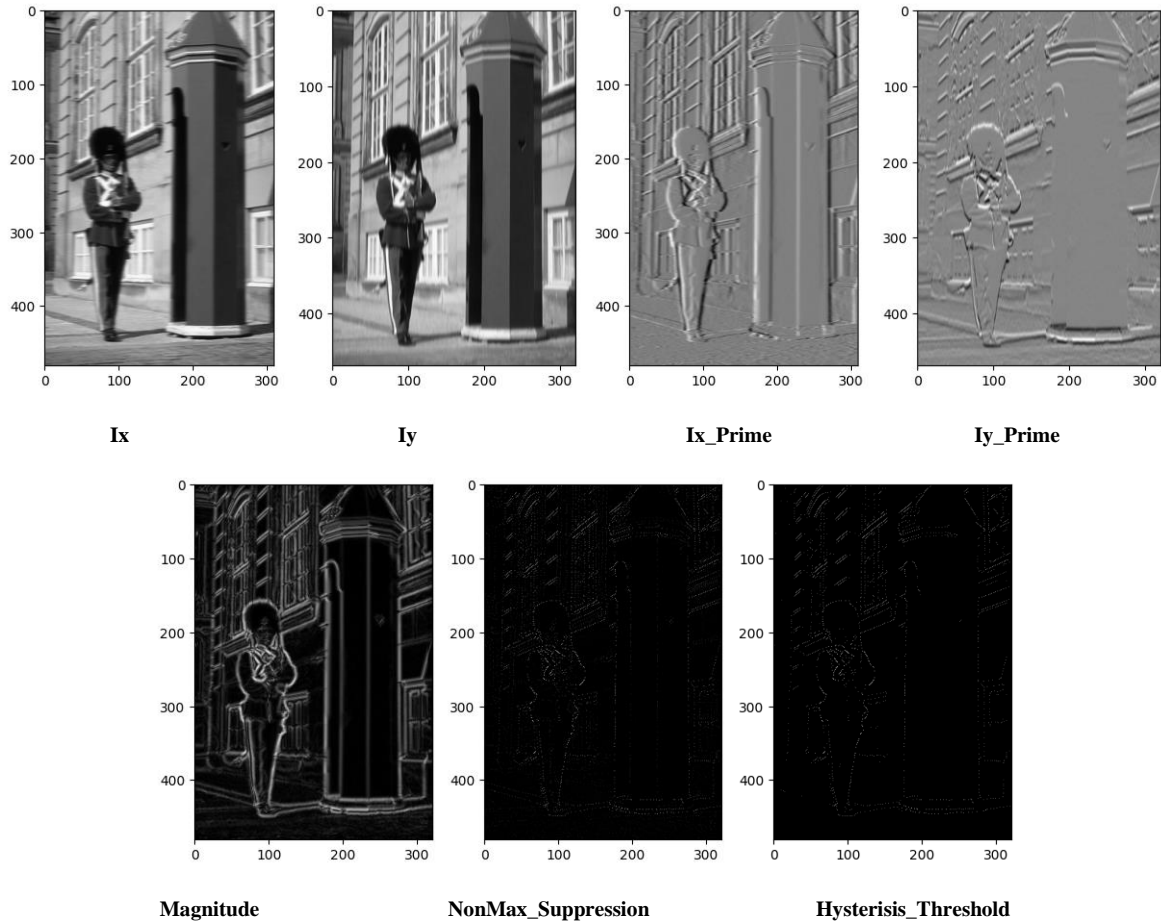


**Magnitude**

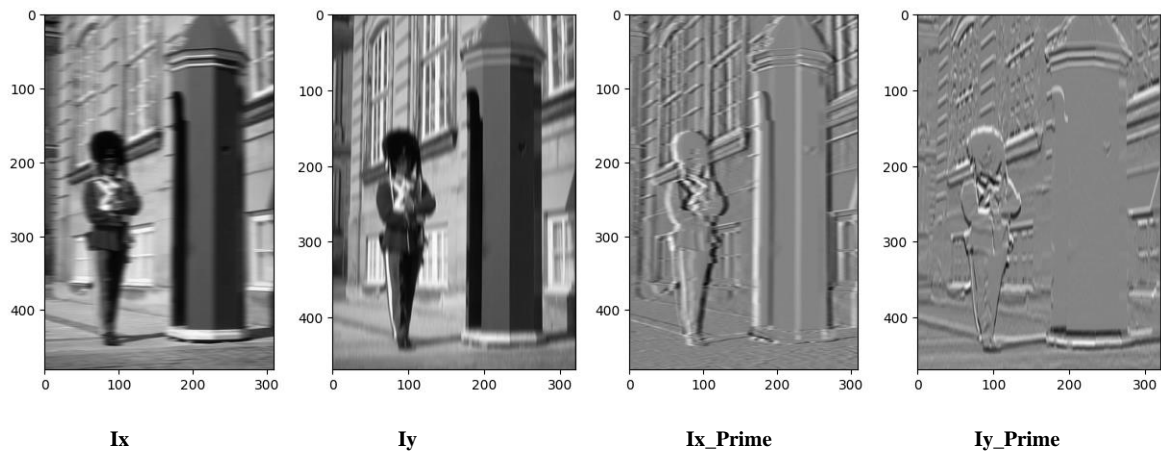
**NonMax\_Suppression**

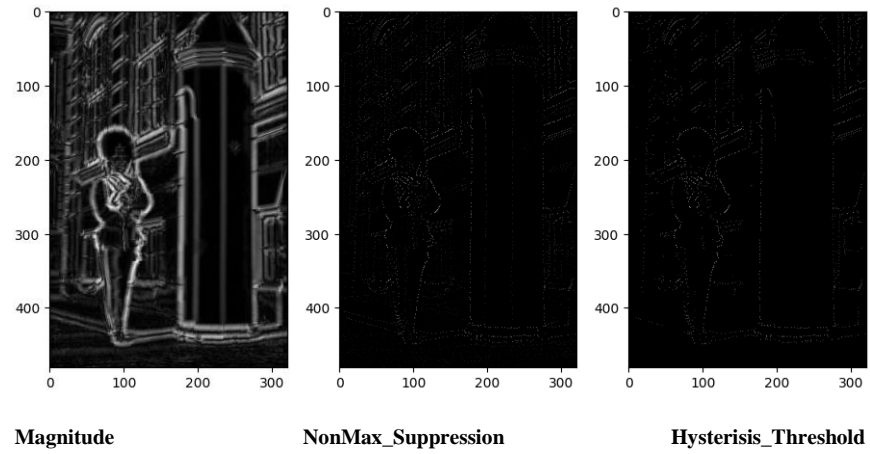
**Hysteresis\_Threshold**

**Sigma = 2**



**Sigma = 4**





## Image 2

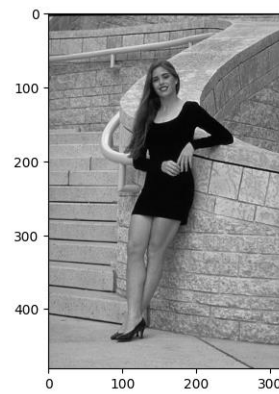
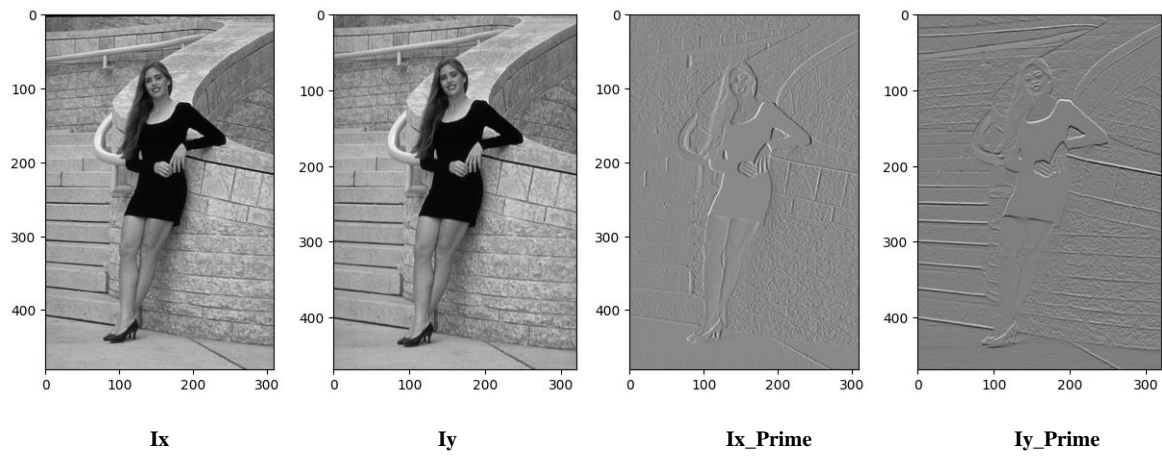
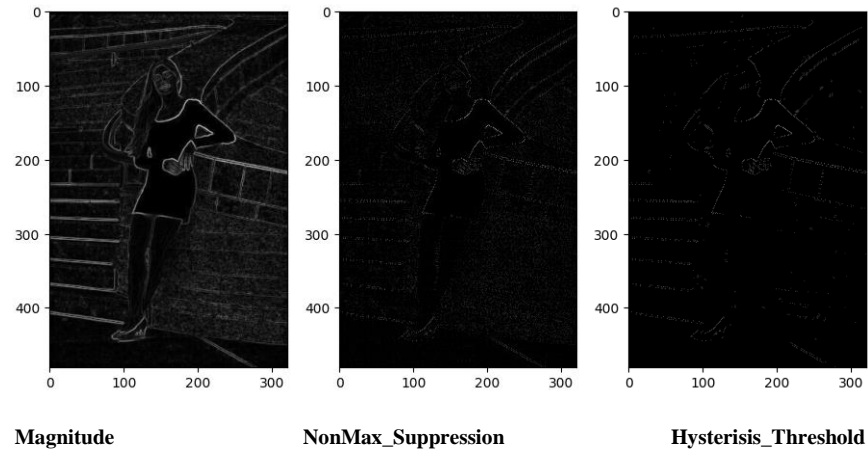
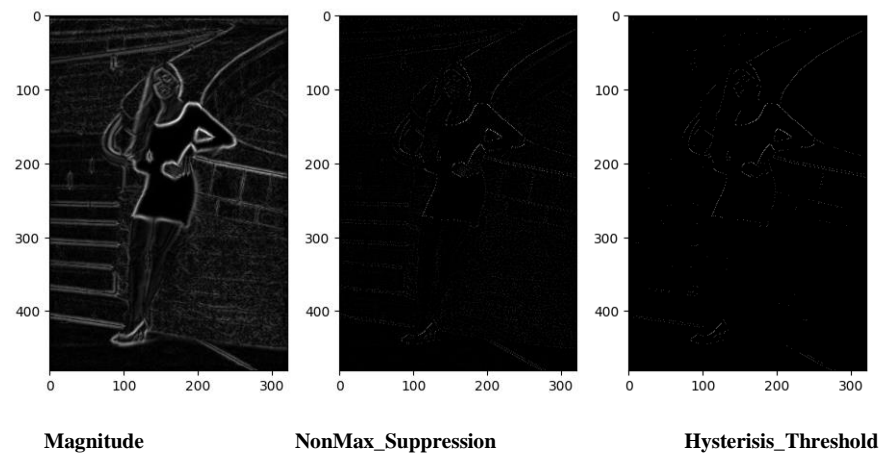
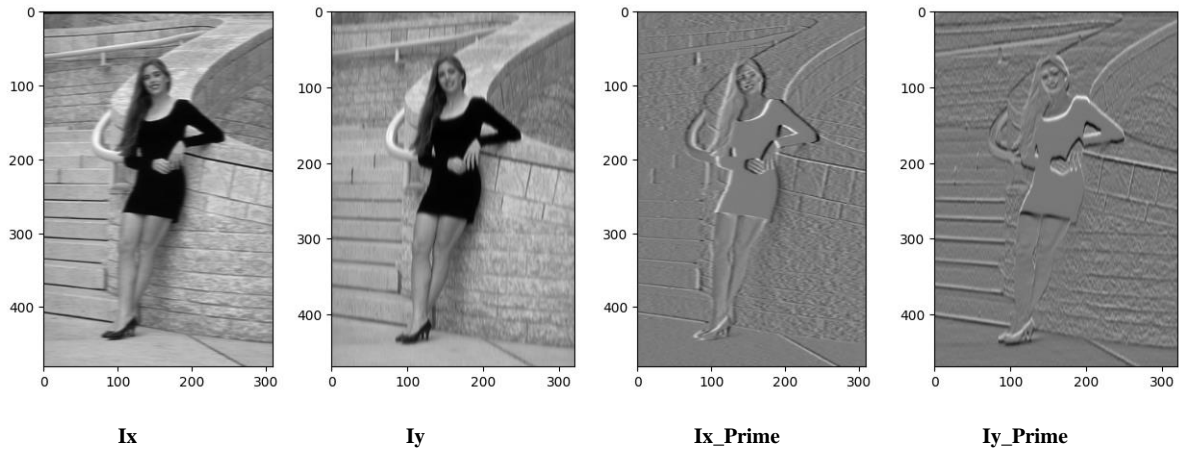


Fig.2 original



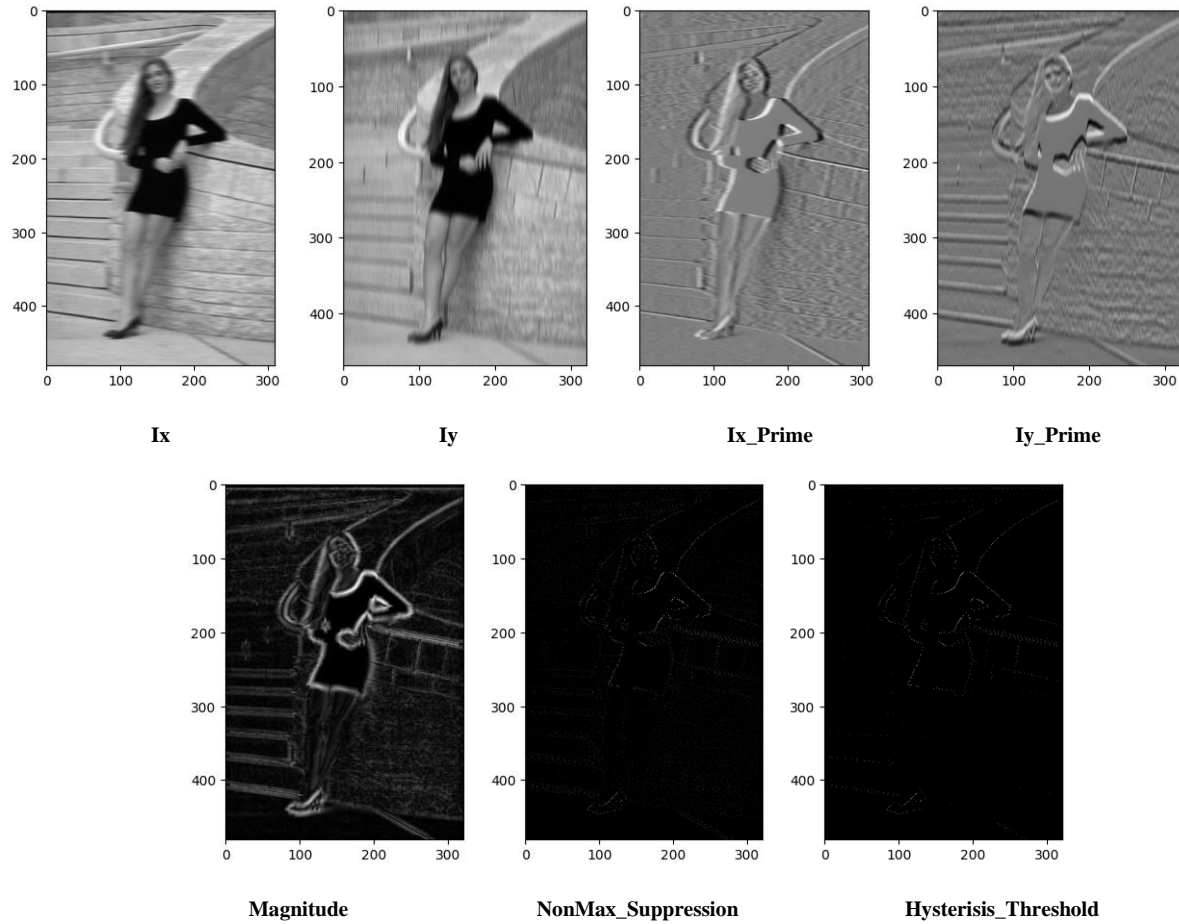


**Sigma = 2**





**Sigma = 4**



**Image 3**



**Fig.3 Original**

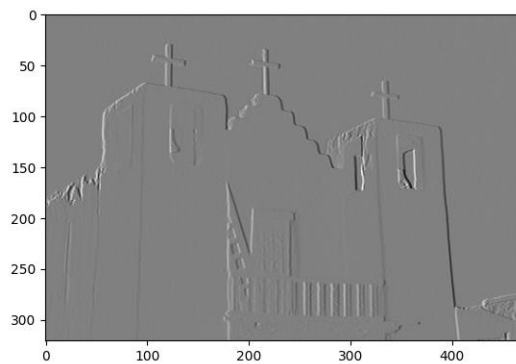
**Sigma =0.5**



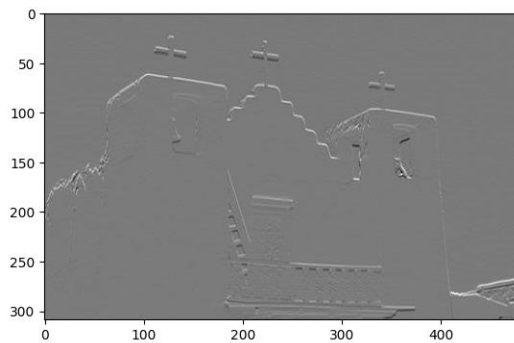
**Ix**



**Iy**



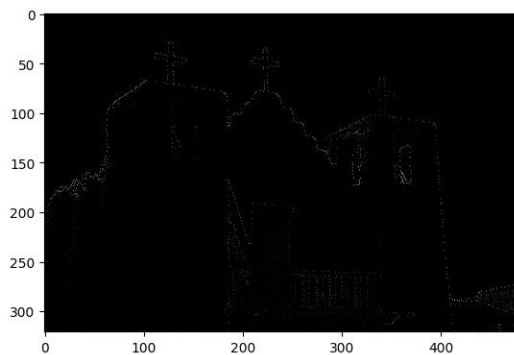
**Ix\_Prime**



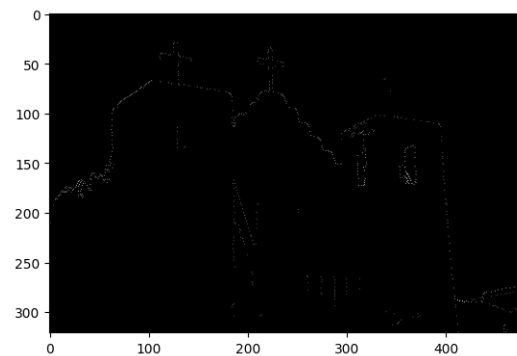
**Iy\_Prime**



**Magnitude**



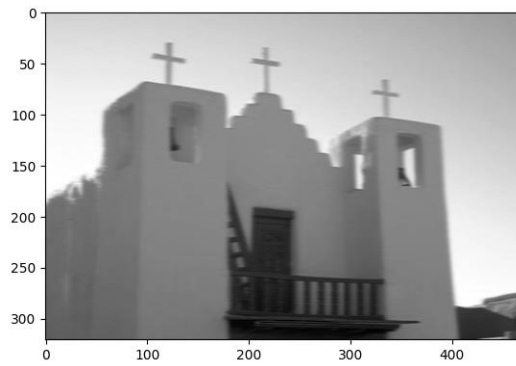
**NonMax\_Suppression**



**Hysteresis\_Threshold**



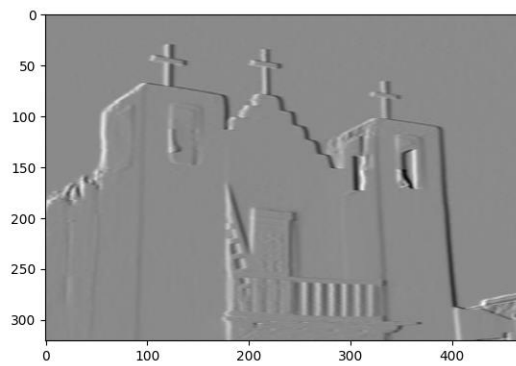
**Sigma = 2**



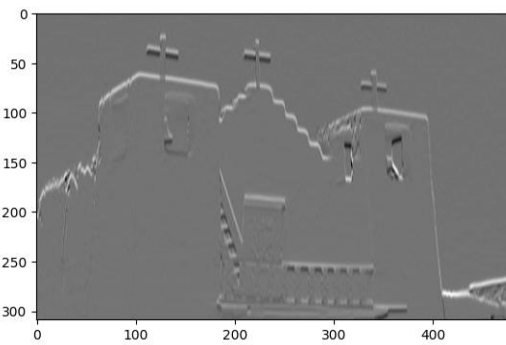
**I<sub>x</sub>**



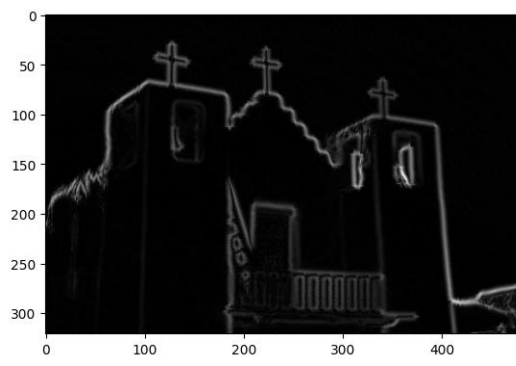
**I<sub>y</sub>**



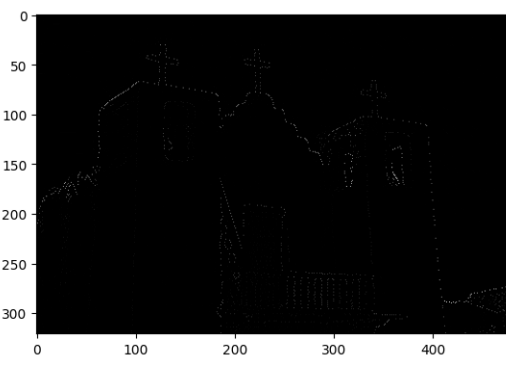
**I<sub>x</sub>\_Prime**



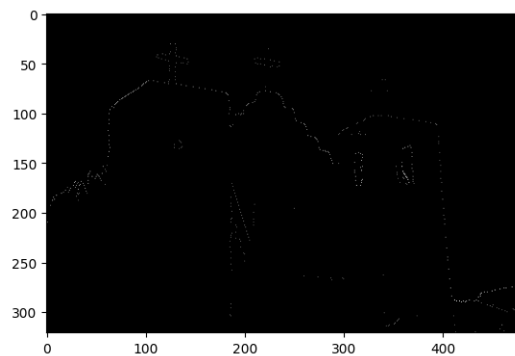
**I<sub>y</sub>\_Prime**



**Magnitude**



**NonMax\_Suppression**

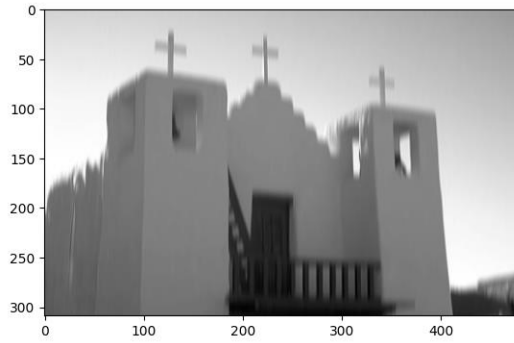


**Hysteresis\_Threshold**

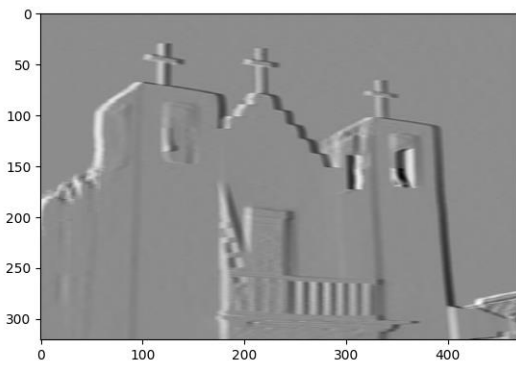
**Sigma = 4**



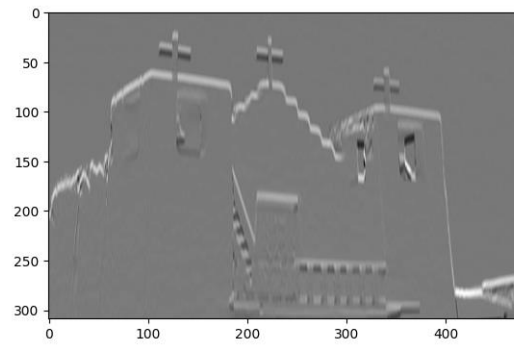
**Ix**



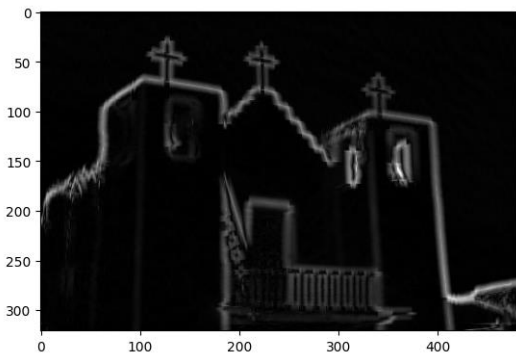
**Iy**



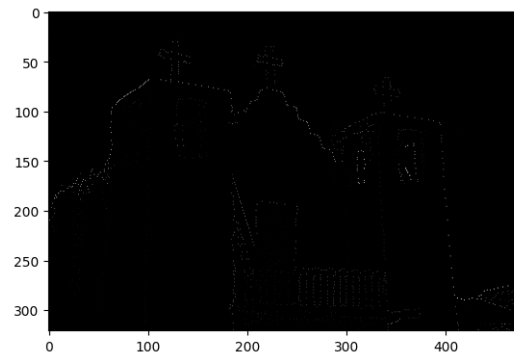
**Ix\_Prime**



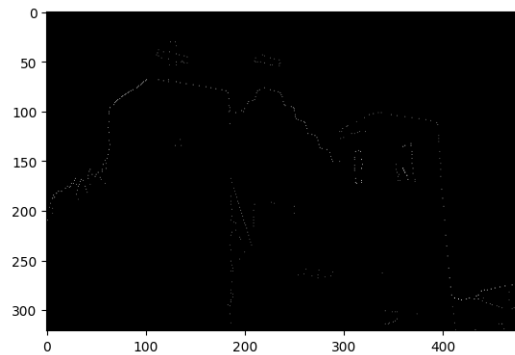
**Iy\_Prime**



**Magnitude**



**NonMax\_Suppression**



**Hysteresis\_Threshold**