

# An Efficient Diamond Mask Based Sobel Edge Detection Filter

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**Abstract:** This paper describes a new improved sobel edge detection method which can process images with improved performance. We have applied Diamond masking method on sobel kernel and gradients are processed horizontal and vertical wise where negative values are made absolute. Performance wise it shows better with respect to Sobel conventional and other industrial application. In our experiment we use 5×5 mask which represents 3 to 5 fold accuracy compared to other techniques also the processing time could be within the acceptance level.

**Keywords:** Edge detection, Sobel operator, gradient, diamond mask, absolute value.

## 1. Introduction

In imaging science, image processing is any form of signal processing for which the input is an image, such as a photograph or video frame; the output of image processing may be either an image or a set of characteristics or parameters related to the image. Most image-processing techniques involve treating the image as a two-dimensional signal and applying standard signal-processing techniques to it. Modern digital technology has made it possible to manipulate multi-dimensional signals with systems that range from simple digital circuits to advanced parallel computers. The goal of this manipulation can be divided into three categories:

- Image Processing (image in -> image out)
- Image Analysis (image in -> measurements out)
- Image Understanding (image in -> high-level description out)

Edge detection is the name for a set of mathematical methods which aim at identifying points in a digital image at which the image brightness changes sharply or, more formally, has discontinuities. The points at which image brightness changes sharply are typically organized into a set of curved line segments termed *edges*. The same problem of finding discontinuities in 1D signals is known as step detection and the problem of finding signal discontinuities over time is known as change detection. Edge detection is a fundamental tool in image processing, machine vision and computer vision, particularly in the areas of feature detection examples of operators such as Canny, Sobel, Kayyali, etc. and feature extraction.

## 2. Edge Detection Based on Sobel Operator

Standard Sobel operators, for a 3×3 neighborhood, each simple central gradient estimate is vector sum of a pair of orthogonal vectors. Each orthogonal vector is a directional derivative estimate multiplied by a unit vector specifying the derivative's direction. The vector sum of these simple gradient estimates amounts to a vector sum of the 8 directional derivative vectors. Thus a point on Cartesian grid and its eight neighboring density values as shown:

In the directional derivative estimate vector  $G$  was defined such as density difference /distance to neighbor. This vector is determined such that the direction of  $G$  will be given by the unit vector to the approximate neighbor. Note that, the neighbors group into antipodal pairs: (a,i), (b,h), (c,g), (f,d).

Here, this vector is multiplied by 2 because of replacing the divide by 2. The resultant formula of the operator consists of a pair of 3×3 convolution masks as shown in Figure 1. One mask is simply the other rotated by 90°. These masks are designed to respond maximally to edges running vertically and horizontally relative to the pixel grid, one mask for each of the two perpendicular orientations.

In theory at least, the operator consists of a pair of 3×3 convolution kernels as shown in Figure 1. One kernel is simply the other rotated by 90°. This is very similar to the Roberts Cross operator.

-1	0	+1	+1	+2	+1
-2	0	+2	0	0	0
-1	0	+1	-1	-2	-1
$G_x$			$G_y$		

Figure 1: Sobel convolution kernels of order 3×3.

These kernels are designed to respond maximally to edges running vertically and horizontally relative to the pixel grid, one kernel for each of the two perpendicular orientations. The kernels can be applied separately to the input image, to produce separate measurements of the gradient component in each orientation (call these  $G_x$  and  $G_y$ ). These can then be combined together to find the absolute magnitude of the gradient at each point and the orientation of that gradient. The gradient magnitude is given by:

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

Typically, an approximate magnitude is computed using:

$$|G| = |G_x| + |G_y|$$

which is much faster to compute.

### 3. Pseudo-Codes for Sobel Edge Detection

Input: A Sample Image.

Output: Detected Edges.

Step 1: Accept the input image.

Step 2: Apply mask Gx,Gy to the input image.

Step 3: Apply Sobel edge detection algorithm and the gradient.

Step 4: Masks manipulation of Gx, Gy separately on the input image.

Step 5: Results combined to find the absolute magnitude of the gradient.

Step 6: The absolute magnitude is the output edges.

### 4. Implementation of 5×5 Sobel Mask

To implement the enhanced version we remove the unwanted pixels from the given data before finding the edges of the Image. Noise can be removed by using Gaussian filter. In Gaussian Filter a suitable mask is chosen and convolved with image. Mask must be smaller than image size because larger the size of the mask can reduce the noise sensitivity of the detector. Localization error can also increase by increasing mask size.

Standard Sobel operators, for a 5×5 neighborhood, each simple central gradient estimate is vector sum of a pair of orthogonal vectors. Each orthogonal vector is a directional derivative estimate multiplied by a unit vector specifying the derivative's direction. The vector sum of these simple gradient estimates amounts to a vector sum of the 25 directional derivative vectors. Thus a point on Cartesian grid and its 25 neighboring density values.

Here proposed algorithm first calculates Horizontal Gradient and Vertical Gradient by using of proposed Horizontal and Vertical mask. Input image is applied and image is convert into number of 5X5 matrix, then it will convolved with Horizontal mask and generates the vector directional derivates ,in same way the vertical mask is applied to generate Vertical gradient. The main assumption of masking is made by considering the concept of inter pixel correlation. The pixel values in an image are very close to each other and the variation is almost equal to one. Instead of processing the entire pixel in 5×5 kernels, a suitable mask is applied as a filter which passes horizontal and vertical pixels as shown in below figures.

#### Proposed Horizontal Mask

$$G_x = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \end{bmatrix}$$

Figure 2: Horizontal mask of order 5×5

#### Proposed Vertical Mask

$$G_y = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Figure 3: Vertical mask of order 5×5

Overall 5×5 mask is following Diamond Structure

$$G_x + G_y = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \end{bmatrix}$$

Figure 4: Filter mask kernel for Sobel operator

**Gx1=Absolute (Gx)**

**Gy1=Absolute (Gy)**

**Sobel Edge Pixel = Gx1+Gy1**

Generated filter mask consists of few values which are to be processed which results in fast computation and low area and power consumption at architectural level. The new filter mask consists of negative and positive values. By applying absolute on the result values and summing up them generates the same conventional function with reduces complexity.

During first stage the proposed method is implemented on MATLAB to thoroughly investigate the required time to detect edges with in an object and compare output image with various parameters.

### 5. Parameter Comparison

Various parameters are used to evaluate the proposed algorithm. The various parameters are

1. PSNR (Peak Signal-To-Noise Ratio)
2. SSIM (Structural-Similarity-Based Image Quality Assessment)
3. FSIM (Feature Similarity Index for Image Quality Assessment)

DIGITAL images are subject to a wide variety of distortions during acquisition, processing, compression, storage, transmission and reproduction, any of which may result in a degradation of visual quality. For applications in which images are ultimately to be viewed by human beings, the only "correct" method of quantifying visual image quality is through subjective evaluation. In practice, however, subjective evaluation is usually too inconvenient, time-consuming and expensive.

#### 1. PSNR (Peak signal-to-noise ratio)

Peak signal-to-noise ratio, often abbreviated PSNR, is an engineering term for the ratio between the maximum possible power of a signal and the power of corrupting noise that affects the fidelity of its representation. Because many signals have a very wide dynamic range, PSNR is usually

expressed in terms of the logarithmic decibel scale. PSNR is most commonly used to measure the quality of reconstruction of lossy compression codecs (e.g., for image compression). The signal in this case is the original data, and the noise is the error introduced by compression. When comparing compression codecs, PSNR is an *approximation* to human perception of reconstruction quality. Although a higher PSNR generally indicates that the reconstruction is of higher quality, in some cases it may not. One has to be extremely careful with the range of validity of this metric; it is only conclusively valid when it is used to compare results from the same codec (or codec type) and same content. PSNR is most easily defined via the mean squared error (MSE). Given a noise-free  $m \times n$  monochrome image  $I$  and its noisy approximation  $K$ , MSE is defined as:

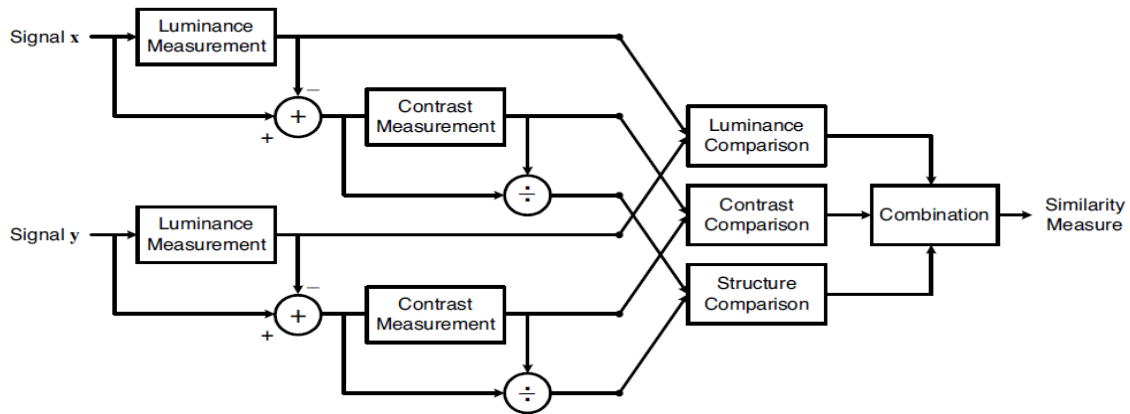
$$MSE = \frac{1}{m \cdot n} \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} [I(i, j) - K(i, j)]^2$$

The PSNR is defined as

$$\begin{aligned} PSNR &= 10 \cdot \log_{10} \left( \frac{MAX_I^2}{MSE} \right) \\ &= 20 \cdot \log_{10} \left( \frac{MAX_I}{\sqrt{MSE}} \right) \\ &= 20 \cdot \log_{10}(MAX_I) - 10 \cdot \log_{10}(MSE) \end{aligned}$$

## 2. SSIM (Structural-Similarity-Based Image Quality Assessment)

Natural image signals are highly structured: their pixels exhibit strong dependencies, especially when they are spatially proximate, and these dependencies carry important information about the structure of the objects in the visual scene. The Minkowski error metric is based on point wise signal differences, which are independent of the underlying signal structure. Although most quality measures based on error sensitivity decompose image signals using linear transformations, these do not remove the strong dependencies, as discussed in the previous section. The motivation of our new approach is to find a more direct way to compare the structures of the reference and the distorted signals.



**Figure 1:** Diagram of the structural similarity (SSIM) measurement system.

$$SSIM(\mathbf{x}, \mathbf{y}) = \frac{(2\mu_x\mu_y + C_1)(2\sigma_{xy} + C_2)}{(\mu_x^2 + \mu_y^2 + C_1)(\sigma_x^2 + \sigma_y^2 + C_2)}$$

$$S_G(\mathbf{x}) = \frac{2G_1(\mathbf{x}) \cdot G_2(\mathbf{x}) + T_2}{G_1^2(\mathbf{x}) + G_2^2(\mathbf{x}) + T_2}$$

$$S_{PC}(\mathbf{x}) = \frac{2PC_1(\mathbf{x}) \cdot PC_2(\mathbf{x}) + T_1}{PC_1^2(\mathbf{x}) + PC_2^2(\mathbf{x}) + T_1}$$

$$S_L(\mathbf{x}) = [S_{PC}(\mathbf{x})]^\alpha \cdot [S_G(\mathbf{x})]^\beta$$

## 3.FSIM (Feature Similarity Index for Image Quality Assessment)

The computation of FSIM index consists of two stages. In the first stage, the local similarity map is computed, and then in the second stage, pools the similarity map into a single similarity score. The separation of the feature similarity measurement between  $f1(\mathbf{x})$  and  $f2(\mathbf{x})$  into two components, each for PC or GM. First, the similarity measure for  $PC1(\mathbf{x})$  and  $PC2(\mathbf{x})$  is defined as

where  $\Omega$  means the whole image spatial domain.

$$FSIM = \frac{\sum_{\mathbf{x} \in \Omega} S_L(\mathbf{x}) \cdot PC_m(\mathbf{x})}{\sum_{\mathbf{x} \in \Omega} PC_m(\mathbf{x})}$$

Where in the fomula, the value are as follows,

## 6. Conclusion

The key contribution of this paper is to develop a new improved sobel edge detection method which can process images with improved performance. We have applied Diamond masking method on sobel kernel and gradients are processed horizontal and vertical wise where negative values are made absolute. Performance wise it shows better with

respect to Sobel conventional and other industrial application. Our experiment has shown 4 to 5 fold accuracy compared to other techniques. The processing time is within the acceptance level. The various parameter analyses at algorithmic level show an improvement upto 3 times in comparison with conventional.

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