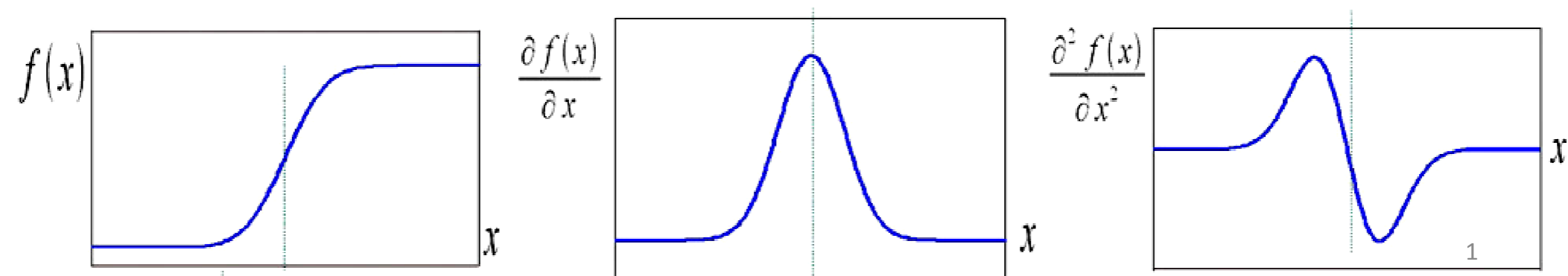


# Edge detection

Edge detection is a technique for reducing information in images, which consists of transforming the image into a set of curves, not necessarily closed, forming the significant boundaries of the image. If the extracted structures are easy to handle (fine, regular, stable curves ...), they can be useful for image matching (robotics, indexing, ...).

Edges generally correspond to sudden changes in the physical or geometric properties of the perceived image and thus form very important attributes for analysis.

several local operators (directional or not) of **first order and second order derivation are used**. They are then followed respectively by a **search for local maxima and zero crossing**



# Gradient operator

$$\frac{\partial f}{\partial x} \approx \Delta_x f(i, j) = f(i + 1, j) - f(i, j) \quad (\text{V.5})$$

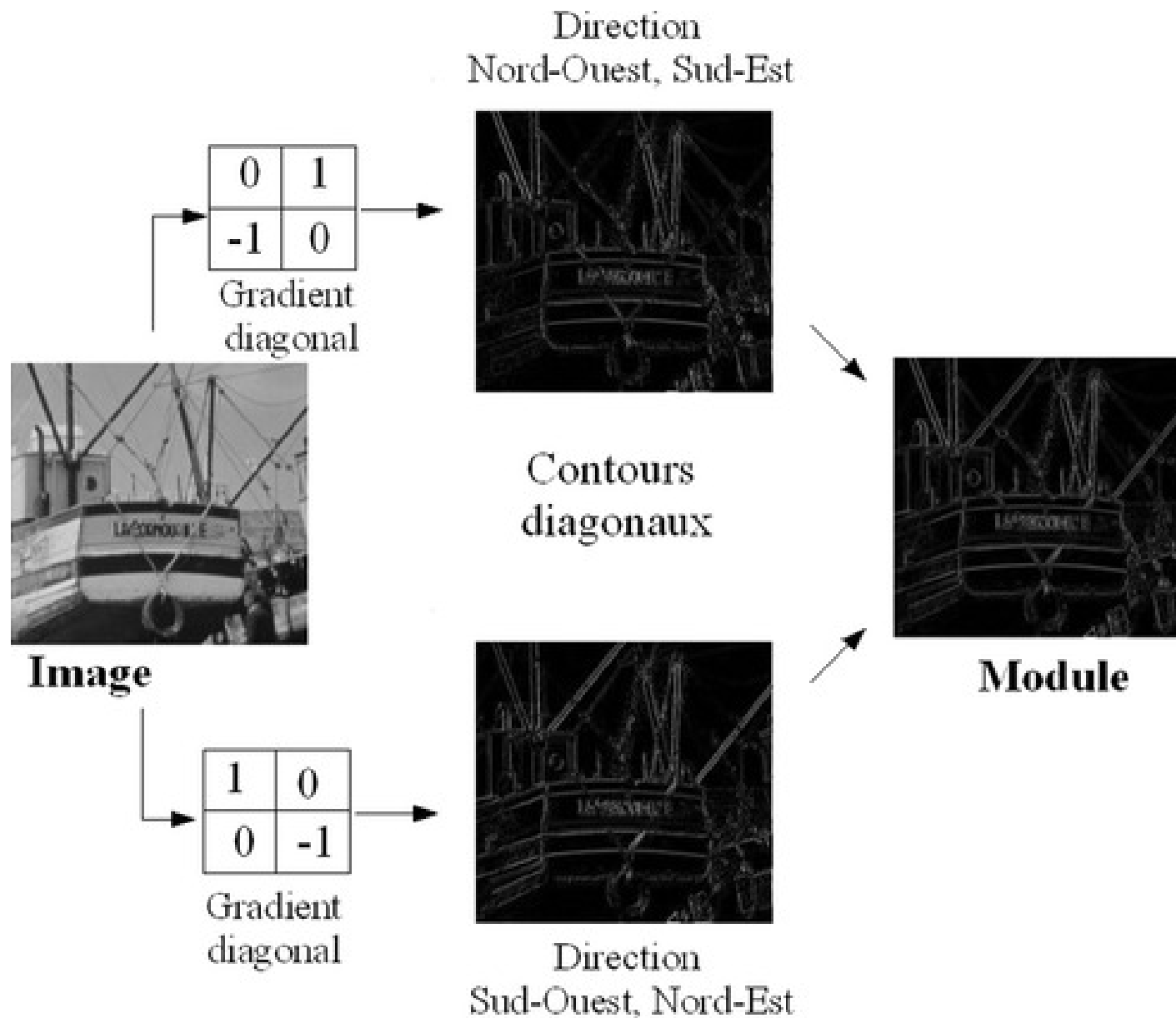
$$\frac{\partial f}{\partial y} \approx \Delta_y f(i, j) = f(i, j + 1) - f(i, j) \quad (\text{V.6})$$

$$\left\| \vec{\nabla} f \right\| = \sqrt{\Delta_x f^2 + \Delta_y f^2} \quad (\text{V.6})$$

$$\left\| \vec{\nabla} f \right\| = \max (|\Delta_x f|, |\Delta_y f|) \quad (\text{V.7})$$

Some gradient operators :

# Roberts Operator



# Prewitt Operator

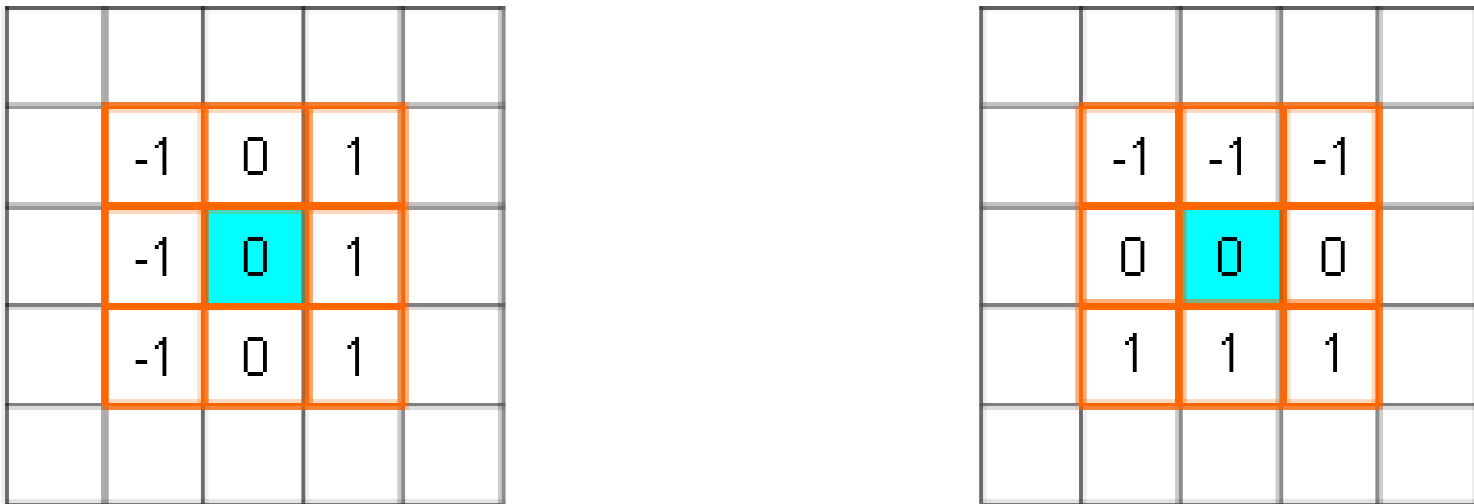
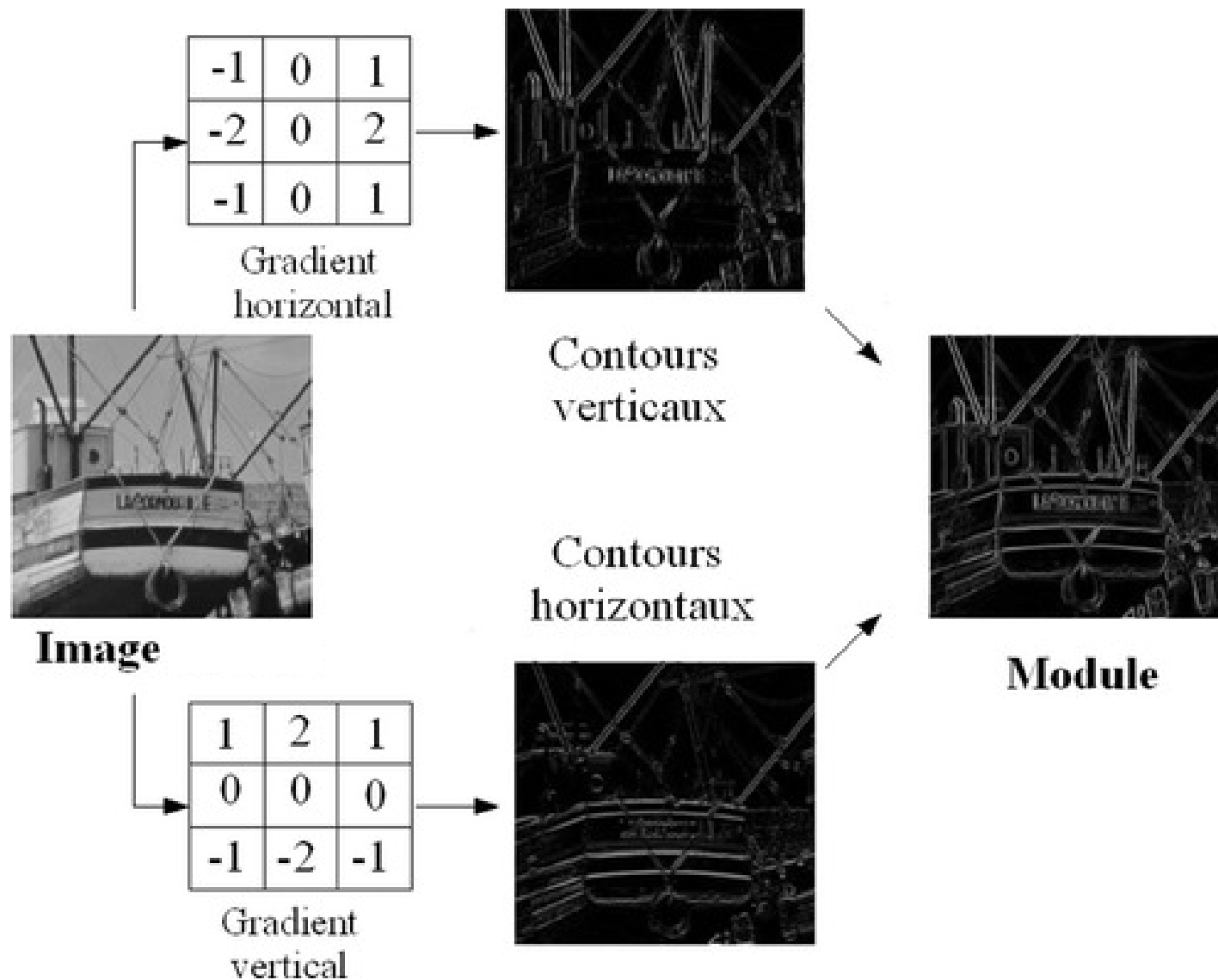


Figure : Horizontal and vertical Prewitt masks

# Sobel Operator



# Kirsch Operator

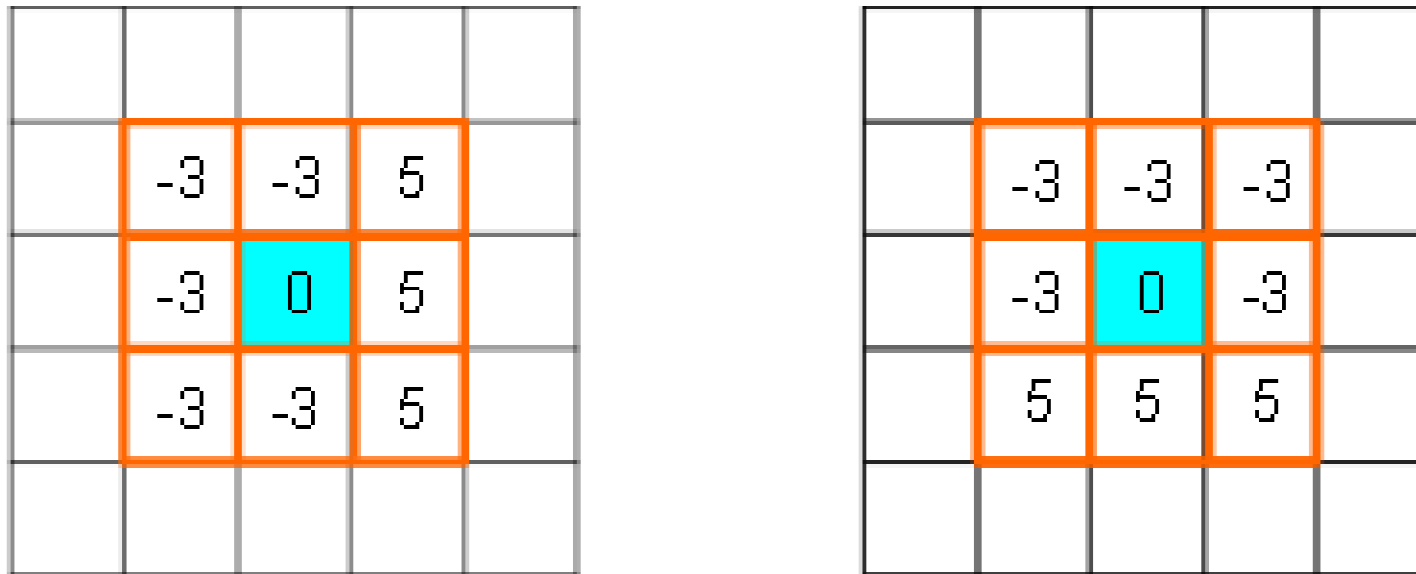


Figure : Horizontal (G) and vertical (D) Kirsch masks

# Thresholding methods

Global Threshold :

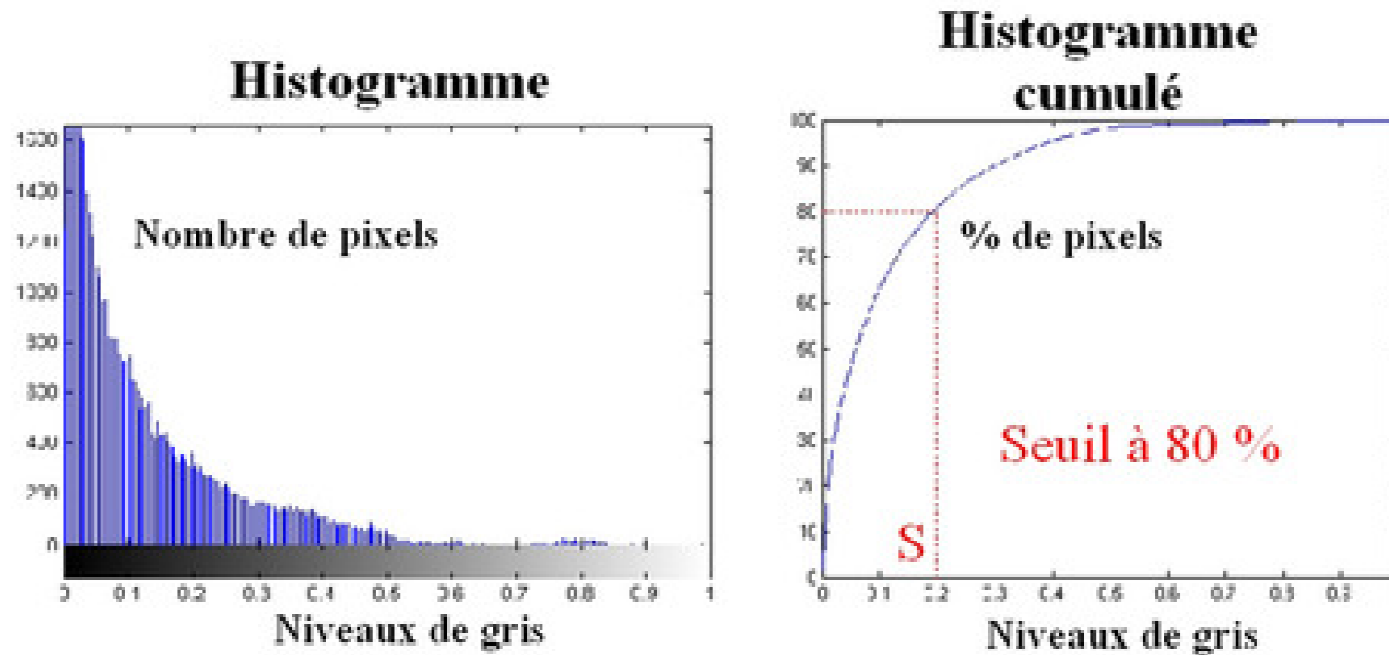
$$\forall i, j \in N \times M \quad I(i, j) = \begin{cases} 1 & \text{si } f(i, j) > S \\ 0 & \text{sinon} \end{cases}$$

Local threshold

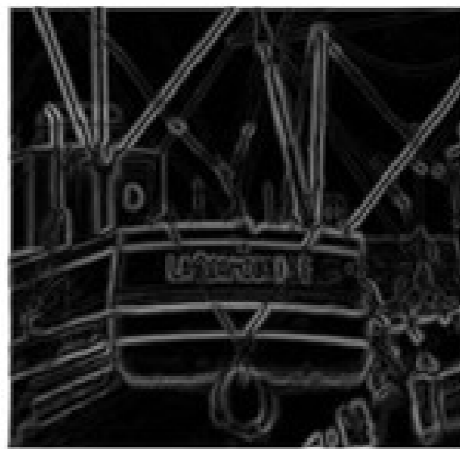
$$S(i, j) = (\max(i, j) + \min(i, j))/2,$$

- $S(i, j)$ : threshold to be applied for point  $i, j$ ;
- $\max(i, j)$ : value of the maximum gray level in a centered window - in  $(i, j)$  of size  $N \times M$ ;
- $\min(i, j)$ : value of the minimum gray level in a window centered in  $(i, j)$  of size  $N \times M$ ;
- $N$  and  $M$  belonging to  $\mathbb{N}$

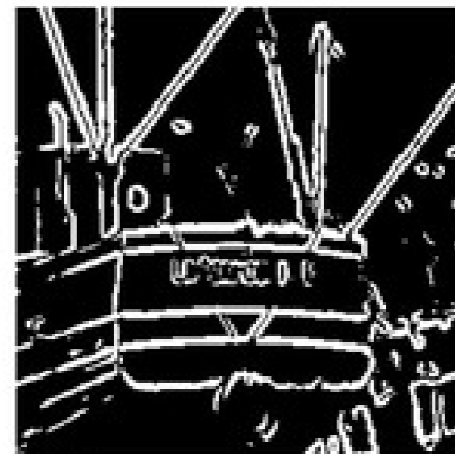
# Global thresholding by histogram



**Module  
du  
gradient**



**Contours**





# The Laplacian operator

$$\nabla^2 f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} = \frac{\partial}{\partial x} \left( \frac{df}{dx} \right) + \frac{\partial}{\partial y} \left( \frac{df}{dy} \right) \quad (\text{V.10})$$

$$\nabla^2 f = \nabla_x (f(x+1, y) - f(x, y)) + \nabla_y (f(x, y+1) - f(x, y)) \quad (\text{V.11})$$

$$\nabla_x (f(x+1, y) - f(x, y)) = f(x+1, y) - f(x, y) - (f(x, y) - f(x-1, y)) \quad (\text{V.12})$$

$$\nabla_x (f(x+1, y) - f(x, y)) = f(x+1, y) + f(x-1, y) - 2f(x, y) \quad (\text{V.13})$$

$$\nabla_y (f(x, y+1) - f(x, y)) = f(x, y+1) - f(x, y) - (f(x, y) - f(x, y-1)) \quad (\text{V.14})$$

$$\nabla_y (f(x, y+1) - f(x, y)) = f(x, y+1) + f(x, y-1) - 2f(x, y) \quad (\text{V.15})$$

$$\nabla^2 f = f(x+1, y) + f(x-1, y) + f(x, y+1) + f(x, y-1) - 4f(x, y) \quad (\text{V.16})$$

# Laplacian Mask

0	-1	0
-1	4	-1
0	-1	0

-1	-1	-1
-1	-8	-1
-1	-1	-1

1	-2	1
-2	4	-2
1	-2	1

$$G = \sqrt{G_h^2 + G_v^2}$$

$$G_s = (G > \text{threshold})$$