













Convolution (Sharpening)

- · 1st order derivative produces thick edges.
- 2nd order derivative give better response especially in image regions with details such as small edges and noisy pixels
- Since the 2nd order derivative produce better response at the edges and image regions with details, it is preferred in sharpening mostly

Utilization of 2nd order derivatives - Laplacian filter

$$\partial^2 f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

In the discrete form:

$$\frac{\partial^2 f}{\partial^2 x^2} = f(x+1,y) + f(x-1,y) - 2f(x,y)$$

$$\frac{\partial^2 f}{\partial^2 y^2} = f(x, y+1) + f(x, y-1) - 2f(x, y)$$

Convolution (Sharpening)



• A compact form of 2-D Laplacian filter :

$$\nabla^{2} f = \left[f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y) \right]$$

$$\downarrow \\
\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & -8 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ -1 & 4 & -1 \\ 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

Convolution (Sharpening)

• After 2-D Laplacian filtering, the sharper image can be obtained:

 $g\left(x,y\right) = \begin{cases} f\left(x,y\right) - \nabla^2 f\left(x,y\right) & \text{, if the center value in kernel is negative} \\ f\left(x,y\right) + \nabla^2 f\left(x,y\right) & \text{, if the center value in kernel is positive} \end{cases}$







Convolution - Edge Detection

Utilization of 1st order derivative (The Gradient):

· Gradient as 2-D column vector:

$$\nabla \mathbf{f} = \begin{bmatrix} G_x \\ G_y \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

Amplitude of this vector:
$$\nabla f = \text{mag}(\nabla f) = \left[G_x^2 + G_y^2\right]^{1/2} = \left[\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2\right]^{1/2}$$

In order to eliminate computation cost of square and square-root operations absolute value operation is generally used in practice.

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$$\nabla f \approx |G_x| + |G_y|$$

Convolution - Edge Detection

$$\begin{bmatrix} z_1 & z_2 & z_3 \\ z_4 & z_5 & z_6 \\ z_7 & z_8 & z_9 \end{bmatrix}$$

$$G_x = Z_8 - Z_5$$

$$G_x = Z_5 - Z_5$$



• Robert cross gradient operator: $G_x = z_9 - z_5$

$$G_x = z_9 - z_5$$

 $\nabla f \approx |G_x| + |G_y|$ $\approx |z_9 - z_5| + |z_8 - z_6|$

 $\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \qquad \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$

• In stead of even numbered kernels such as 2x2, let us use a 3x3 kernel:

 $\nabla f \approx \left| \left(z_7 + 2 z_8 + z_9 \right) - \left(z_1 + 2 z_2 + z_3 \right) \right| + \left| \left(z_3 + 2 z_6 + z_9 \right) - \left(z_1 + 2 z_4 + z_7 \right) \right|$



-2 0 2

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