

# Filtrado espacial

Unidad 2

BME423 · Procesamiento de imágenes médicas

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### Convolución

La operación convolución es útil para eliminar ruido.

$$\underbrace{\hat{g}(x,y)}_{\text{imagen filtrada}} = \underbrace{w(x,y)}_{\text{filtro}} * \underbrace{g(x,y)}_{\text{imagen medida}}$$

## Convolución

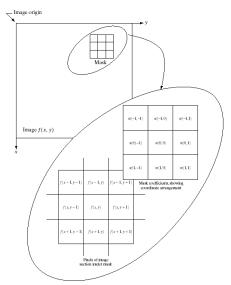
Considere una imagen f(x, y) de  $M \times N$  y un kernel w(s, t) de  $m \times n$ .

La convolución está dada por:

$$\hat{g}(x, y) = w * f = \sum_{s=-a}^{a} \sum_{t=-b}^{b} w(s, t) f(x + s, y + t)$$

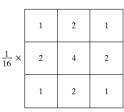
con 
$$a = (m-1)/2$$
 y  $b = (n-1)/2$ 

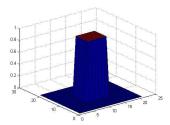
# Convolución



#### Box and weighted average filters.

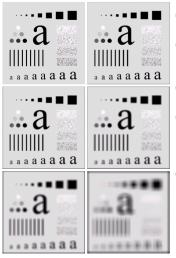
	1	1	1
$\frac{1}{9}$ ×	1	1	1
	1	1	1





#### a b

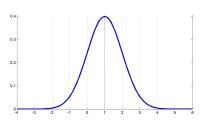
FIGURE 3.34 Two 3 × 3 smoothing (averaging) filter masks. The constant multipli er in front of each mask is equal to the sum of the values of its coefficients, as is required to compute an average.



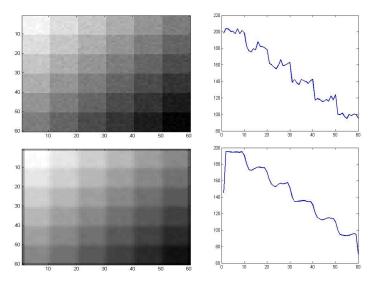
- Máscara de suavizado de n: 3, 5, 9, 15 y 35.
- Tamaños de los cuadrados negros: 3, 5, 9, 15, 25, 35, 45, 55 (separación de 25 píxeles).
- Las letras van desde 10 hasta 24 ptos. en incrementos de 2.
- Las barras verticales tienen un ancho de 5 píxeles, 100 de alto y una separación entre barras de 20 píxeles.
- El diámetro de los círculos es de 25 píxeles, y están separados por 15 píxeles. Sus niveles de gris varían de 0-100% en incrementos de un 20%.

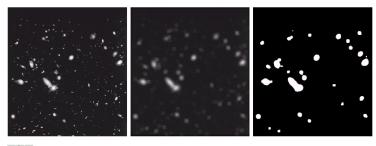
#### Kernel gaussiano

$$h_g(x, y) = e^{\left(-\frac{(x^2+y^2)}{2\sigma^2}\right)} h(x, y) = \frac{h_g(x, y)}{\sum_{x} \sum_{y} h_g(x, y)}$$



0.0751	0.1238	0.0751
0.1238	0.2042	0.1238
0.0751	0.1238	0.0751





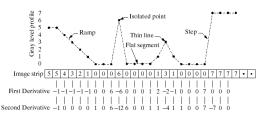
a b c

**FIGURE 3.36** (a) Image from the Hubble Space Telescope. (b) Image processed by a  $15 \times 15$  averaging mask. (c) Result of thresholding (b). (Original image courtesy of NASA.)

# Filtrado - derivadas y detalles



$$\begin{split} \partial f(x,y)/\partial x &= f(x+1,y) - f(x,y) \\ \partial^2 f(x,y)/\partial x^2 &= f(x+1,y) + f(x-1,y) - 2f(x,y) \end{split}$$



# Filtrado - Laplaciano $abla_f^2$

Rosenfeld y Kak, 1982

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

$$\partial f(x,y)/\partial x = f(x+1,y) - f(x,y)$$
  
 $\partial f(x,y)/\partial y = f(x,y+1) - f(x,y)$ 

$$\partial^2 f(x,y) / \partial x^2 = f(x+1,y) + f(x-1,y) - 2f(x,y)$$
  
 $\partial^2 f(x,y) / \partial y^2 = f(x,y+1) + f(x,y-1) - 2f(x,y)$ 

Se puede implementar como una convolución

$$\nabla^2 f = f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y)$$

# Filtrado - Laplaciano $abla_f^2$

$$\nabla^2 f = f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y)$$

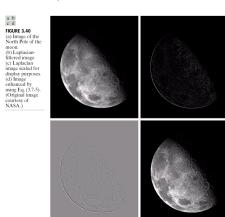
0	1	0	1	1	1
1	-4	1	1	-8	1
0	1	0	1	1	1
0	-1	0	-1	-1	-1
-1	4	-1	-1	8	-1
0	-1	0	-1	-1	-1

a b c d

#### FIGURE 3.39

(a) Filter mask used to implement the digital Laplacian, as defined in Eq. (3.7-4). (b) Mask used to implement an extension of this equation that includes the diagonal neighbors. (c) and (d) Two other implementations of the Laplacian.

$$g(x,y) = \begin{cases} f(x,y) - \nabla^2 f(x,y) & \text{Si el coef del centro de } \nabla_f^2 \text{ es negativo} \\ f(x,y) + \nabla^2 f(x,y) & \text{Si el coef del centro de } \nabla_f^2 \text{ es positivo} \end{cases}$$



$$\begin{split} g(x,y) &= f(x,y) - \nabla^2 f \\ g(x,y) &= f(x,y) - [f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y)] \\ g(x,y) &= 5f(x,y) - [f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1)] \end{split}$$

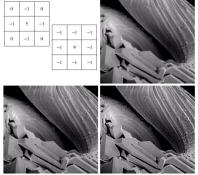


FIGURE 3.41 (a) Composite Laplacian mask. (b) A second composite mask. (c) Scanning electron microscope image. (d) and (e) Results of filtering with the masks in (a) and (b), respectively. Note how much sharper (e) is than (d). (Original image courtesy of Mr. Michael Shaffer, Dearnment of Geological Sciences. University of Orecon, Eusene.)

**Unsharp Masking:** 

$$f_s(x,y) = f(x,y) - \bar{f}(x,y)$$

donde  $f_s(x,y)$  es la imagen agudizada (sharpened) dada por la sustracción de la imagen y su versión suavizada

Una generalización de Unsharp Masking es High-Boost Filtering:

$$f_{
m hb}(x,y) = Af(x,y) - ar{f}\left(x,y
ight)$$

donde A ≥ 1

La ecuación del filtro High-Boost puede ser expresada mediante:

$$f_{
m hb}(x,y)=(A-1)f(x,y)+f(x,y)-ar{f}\left(x,y
ight)$$

donde  $A \ge 1$ . Teniendo en cuenta que:

$$f_s(x,y) = f(x,y) - ar{f}\left(x,y
ight)$$

se obtiene:

$$f_{
m hb}(x,y)=(A-1)f(x,y)+f_s(x,y)$$

Si se emplea el Laplaciano para obtener  $f_s(x,y)$  se tiene:

$$f_{\mathrm{hb}}(x,y) = \begin{cases} Af(x,y) - \nabla^2 f(x,y) & \text{Si el coef del centro de } \nabla_f^2 \text{ es negativo} \\ Af(x,y) + \nabla^2 f(x,y) & \text{Si el coef del centro de } \nabla_f^2 \text{ es positivo} \end{cases}$$

Note que si A = 1,  $f_{hb}(x,y)$  corresponde al agudizado Laplaciano estándar

Si A >> 1, el efecto sharpening se vuelve menos significativo

0	-1	0	-1	-1	-1
-1	A + 4	-1	-1	A + 8	-1
0	-1	0	-1	-1	-1



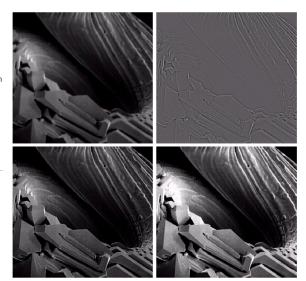
**FIGURE 3.42** The high-boost filtering technique can be implemented with either one of these masks, with  $A \ge 1$ .

a b c d

#### FIGURE 3.43

(a) Same as Fig. 3.41 (c), but darker.
(a) Laplacian of (a) computed with the mask in Fig. 3.42 (b) using A = 0.
(c) Laplacian enhanced image using the mask in Fig. 3.42 (b) with

A = 1. (d) Same as (c), but using A = 1.7.



El vector gradiente de una imagen en las coordenadas (x,y) está dado por:

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

La magnitud es: 
$$\nabla f = \text{mag}(\nabla \mathbf{f})$$

$$= \left[G_x^2 + G_y^2\right]^{1/2}$$

$$= \left[\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2\right]^{1/2}$$

$$\nabla \mathbf{f} \approx |\mathbf{G}_{\mathbf{x}}| + |\mathbf{G}_{\mathbf{v}}|$$

Varios autores han propuesto máscaras para obtener el gradiente de una imagen vía convolución:

La aproximación más simple plantea:

$$G_x = (z_8 - z_5); G_y = (z_6 - z_5)$$

Roberts (1965): 
$$G_x = (z_9 - z_5)$$
;  $G_v = (z_8 - z_6)$ 

Entonces:

$$\nabla f = [(z_9 - z_5)^2 + (z_8 - z_6)^2]^{1/2}$$
 ó  $\nabla f \approx |z_9 - z_5| + |z_8 - z_6|$ 

-1	0	0	-1
0	1	1	0

Roberts cross-gradient operators.

$z_1$	$z_2$	$z_3$
Z <sub>4</sub>	Z <sub>5</sub>	Z <sub>6</sub>
z <sub>7</sub>	$z_8$	Z9

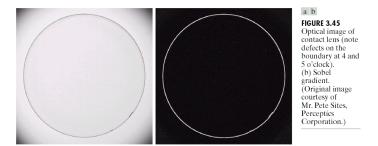
Sobel:

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

-1	-2	-1
0	0	0
1	2	1

_			
	-1	0	1
	-2	0	2
	-1	0	1

#### Sobel operators



Filtro agudizador para la detección de bordes



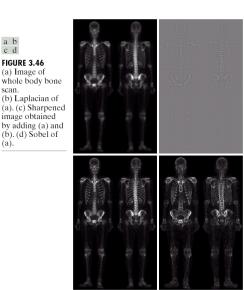


## Filtrado - efectos combinados

a b c d

scan.

(a).



### Filtrado - efectos combinados

