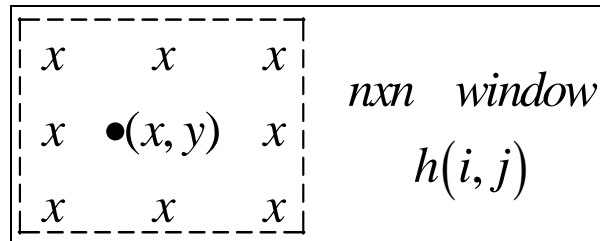


## Enhancement, Restoration, Conversion

### Local vs. Global

**Local:** Only use neighborhood information



**Spatial Smoothing:** Noise removed and reduction of effects due to under sampling.

**Simplest:** 
$$g(i, j) = \frac{1}{M} \sum_s f(i, j)$$

S = M – pixel neighborhood of points surrounding (and perhaps including (x, y) )

S is usually rectangular, e.g. *nxn square*

*nxn window*

$$g(x, y) = \sum_{i=-n/2}^{n/2} \sum_{j=-n/2}^{n/2} h_{sm}(i, j) f(x+i, y+j)$$

$h_{sm}(i, j)$  = smoothing function

e.g.  $= \frac{1}{n^2}$

This is convolution

Take  $n = 3$

$$\begin{bmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{bmatrix}$$

exactly averaging!!

**Remark:** This has continuous version.

Nonlinear

$I_t = \Delta I$  /linear let equation

$$I(x, y, t) = G_t * I(x, y, 0)$$

$$G_t = \frac{1}{\sqrt{2\pi t}} e^{-\left(\frac{x^2+y^2}{t}\right)}$$

**Can see blurring very easily:**

I will choose an odd  $n$  in order for window center to be on sampling grid. Shift origin:

$$k = i + n/2$$

$$l = i + n/2$$

Fourier transform is:

$$H_{sm}(u, x) = \frac{1}{N} \sum_{k=0}^{n-1} \sum_{l=0}^{n-1} h_{sm}(k, l) \exp[-j2\pi(uk + vl)/N]$$

Algebra:

$$H_{sm}(u, x) = \frac{1}{N} \sum_{k=0}^{n-1} \sum_{l=0}^{n-1} \frac{1}{n^2} \exp[-j2\pi uk/N] \cdot \exp[-j2\pi vl/n]$$

Separable transform

Define:

$$H_1(u) = \frac{1}{n} \sum_{k=0}^{n-1} \exp[-j2\pi uk/N]$$

$$\text{Set } Z = \exp[j2\pi u/N]$$

$$H_1(u) = \sum_{k=0}^{n-1} eZ^{-k} \text{ (ignoring scales)}$$

$$= \frac{1}{1-Z^{-1}} - \frac{z^{-n}}{(1-Z^{-1})}$$

$Z$  – transfer of difference two steps ; one occurs at  $k = 0$   
other at  $k = n$

$$= \frac{1-Z^{-n}}{1-Z^{-1}} = Z^{-(n-1)/2} \frac{Z^{n/2} - Z^{-n/2}}{Z^{1/2} - Z^{-1/2}}$$

or

$$H_1(u) = \exp\left[-j\pi u \frac{(n-1)}{N}\right] \frac{\sin\left(\pi u n / N\right)}{\sin(\pi u N)}$$

Similarly  $\mathcal{F}\{H_2(v)\}$

Plot Magnitudes

## Edge Information Removal

*Blurs*

Exercise:

Take smoothing filter from Matlab or  $xv$  (*blur*).

Apply to image at various window sizes.

Hand in ext Wednesday.

## Temporal Smoothing

$\hat{f}_i(x, y) = f_i(x, y) + m_i(x, y)$  zero mean time-uncorrelated noise poor.

$i = 1, 2, \dots, M$  time changes due only + noise process

Sequence of images corrupted by noise.

$$g\left(\begin{matrix} x \\ \theta, y \end{matrix}\right) = \frac{1}{M} \sum_{\text{ensemb}} \hat{f}_i(x, y)$$

Reduce noise effects by reducing variance of (averaged) noise process.

Spatial Sharpening

Simplest approach

Smoothing = low-pass

Sharpening = high-pass

$$g(x, y) = f(x, y) - \underbrace{f_{sm}(x, y)}_{\leftarrow \text{Smoothed version}}$$

Typically:  $f_{sm}$  is taken as local average of eight neighboring pixels surrounding  $(x, y)$  including  $(x, y)$ :

$$f_{sm}(x, y) = \frac{1}{8} \sum_{i=-1}^{i=1} \sum_{j=-1}^{j=1} f(x+i, y+j)$$

$i \neq 0, j \neq 0$

$$\therefore 3 \times 3 \text{ window function for } (a) \text{ is } \begin{bmatrix} -\frac{1}{8} & -\frac{1}{8} & -\frac{1}{8} \\ -\frac{1}{8} & 1 & -\frac{1}{8} \\ -\frac{1}{8} & -\frac{1}{8} & -\frac{1}{8} \end{bmatrix}$$

**Fourier Analysis:**

$$G(u, x) = F(u, x) - F_{sm}(u, \cup)$$
$$= F(u, x) \left[ 1 - \frac{F_{sm}(u, x)}{\underbrace{F(u, x)}_{H_2(u, x)}} \right]$$
$$\underbrace{\hspace{10em}}_{H_{eq}(u, v)}$$

## Nonlinear Operations

Not every operator is linear. Class of enhoncent called “rank” or “median”.

Idea: Output image intensity of spatial location  $(x, y)$  is chosen on the best of the relative rank or intensity of pixels in neighborhood of  $(x, y)$ .

Given  $N$  pixel intensities obtail over a local region,  $S$ , denoted or  $f_1, \dots, f_N$  order in increasing value:

$$R(x, y) = \{f_1, \dots, f_N\} \\ f_i \leq f_{i+1}$$

Output intensity:

$$g(x, y) \triangleq \text{Rank } j \text{ } R(x, y)$$

Where  $\text{Rank } j$  is the intensity of the  $j$ -th intensity it position or  $\text{rank } j$  in  $R(x, y)$ .

Ex.

i.  $j=1$  yields min filter

$$g(x, y) = \min R(x, y) \\ = \{f(x, y) | (x, y) \in \delta\}$$

ii. max filters is  $j = N$ :

$$g(x, y) = \max R(x, y) \\ = \max \{f(x, y) | x, y \in \delta\}$$

iii.  $N$  is odd:

$$m = \frac{n+1}{2}$$

$$\text{med } R(x, y) \triangleq \text{Rank}_{n+1} (R(x, y))$$

$V_e, \text{ nice}$ : Try in Matlab or

Suppose have image corrupted by spike-like noise.

Low-pass distributes equally.

Median removes noise without degraditive.

Not b two image functions yielding sample sequences  $R_1$  al  $R_2$ .

$$\text{med } (R_1(x, y) + R_2(x, y)) \neq \text{med } R_1(x, y)$$

$$+ \text{med } R_2(x, y)$$

However:

$$\begin{aligned} \text{med}(KR(x, y)) &= K \text{med} R(x, y) \\ \text{med}(K + R(x, y)) &= K \in \text{med} R(x, y) \end{aligned}$$

Properties

1. Median filter reduces variability of intensities in the image.
2. Median filter preserve certain edge shapes.  
Preserves triple pulse
3. No new grey values generated

Edge detection

$$\nabla f(x, y) = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

$$\frac{\partial f}{\partial x} \approx \frac{f(x + \Delta x, y) - f(x, y)}{\Delta x} \leftrightarrow \text{convolution with } \begin{bmatrix} -1 & 1 \end{bmatrix}$$

$$\frac{\partial f}{\partial y} \approx \frac{f(x, y + \Delta y) - f(x, y)}{\Delta y} \leftrightarrow \text{convolution with } \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

How do those work:

$$f(x + \Delta x) = f(x) + \Delta x f^1(x) + \frac{(\Delta x)^2}{2!} f^{11}(x) + \dots$$

$$\therefore f(x + \Delta x) - f(x) = \Delta x f^1(x) + \frac{(\Delta x)^2}{2!} f^{11}(x) + \dots$$

mean value theorem

$$x < \varepsilon < x + \Delta x$$

Define

$$D_1(x) = \frac{f(x + \Delta x) - f(x)}{\Delta x}$$

$$\text{Error } (D_1) \approx \mathcal{O}(\Delta x)$$

$D_1$  is a better approximation to the slope of  $f(x)$  at the midpoint of the interval  $[x, x + \Delta x]$ .

$\therefore$  Using  $D_1$  operator for  $\frac{\partial f}{\partial x}$  and  $\frac{\partial f}{\partial y}$  approximations results in approximations to the gradient not at  $(x, y)$  but at different points in  $(x, y)$  plane.



∴ Use centered difference:

$$D_2(x) = \frac{f(x + \Delta x) - f(x - \Delta x)}{2\Delta x}$$

This has several nice proportions:

1.  $D_2(x)$  and  $D_2(y)$  due to centered routine both comprise good estimates for the respective directions of  $f(x, y)$  of midpoints of interval, i.e.  $(x, y)$ .
2. Error ( $D_2$ )  $\cong \theta((\Delta x)^2)$
3.  $D_2(x)$  and  $D_2(y)$  maybe implanted by convolving image with nests.

$$[-1 \ 0 \ 1]$$

Mery Vanitor

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

Varits: 3x3

$$\underbrace{\begin{bmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{bmatrix}}_{D_2\Delta(x)} \quad \text{and} \quad \underbrace{\begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ -1 & -1 & -1 \end{bmatrix}}_{D_2\Delta(y)}$$

“Smoothed” or “averaged” centered difference operations.

Other smoothed operates with a weighting that emphasizes the central pixel eveth

*Sobel Weighting modes*, denotes by  $D_s(x)$  and  $D_s(y)$  and givenly

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}, \quad \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

Third operator for gradient approximation is **Roberts**:

$$D_x \{ (x, y) \} = f(x + \Delta x, y) - f(x, y)$$

$$D_y \{ (x, y) \} = f(x, y + \Delta y) - f(x, y)$$

Note that this operation:

1. Is a variate of the  $D_1$  operation with derivatives approximate along orthoyond orientation  $45^\circ$  at  $135^\circ$  in image plane;
2. Midpoint of both intervals used for approximation is the same point, i.e. it is the point  $(x + \Delta x/2, y + \Delta y/2)$  located at its center of the rectangle found by its four points used;
3. Correspond to combination with rists

$$D_x = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad D_- = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

Roberts Cross Operation

$$\begin{aligned}
 & Dx = Dy \\
 f_{edge}(x, y) &= \max\{|D_t|, |D_{-1}|\} \\
 &= \max\{|f(x, y + Dy) - f(x + Dx, y)|, |f(x + Dx, y + Dy) - f(x, y)|\}
 \end{aligned}$$

Directionally oriented edge information

K