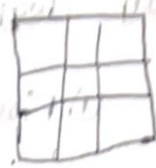


Quiz-2

①

18	22	33	25	32	24
34	128	24	172	28	23
22	19	32	31	28	26



filter

0	0	0	0	0	0
0	18	22	33	25	32
0	34	128	24	172	26
0	22	19	32	31	28
0	0	0	0	0	0

→ calculate median (0, 0, 0, 0, 18, 22, 34, 128)

0	18	22	24	23	0
18	18	19	24	23	23
0	19	19	24	23	0

2) conditions for region splitting and merging

1) Split into four disjoint quadrants any region R_i for which $Q(R_i) = \text{FALSE}$

2) when no further splitting is possible, merge any adjacent regions R_j and R_k for which

$$Q(R_j \cup R_k) = \text{TRUE}$$

3) Stop when no further merging is possible

3) No, if all the pixels in an image are shuffled, there will be no change in the histogram. Since we take how many times a same pixel value appears the values remains same.

4) Abrupt local changes in intensities are detected by derivatives

① First order derivative [produce thicker edge]

$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

② second order derivative [stronger response to fine detail]

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

For 2D Edge detection: derivatives are

i) 1st order

$$\nabla f(x, y) = \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y}$$

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

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

5) Yes, chain code compresses the description information of an object contour by representing the contour as sequence of directional codes, where each code corresponds to the direction of successive contour points.

6) i) Medical imaging

2) Facial Recognition

3) Fingerprint Recognition

4) Object detection

2) Canny - Edge Detection:-

Three objectives

- 1) Low Error rate
- 2) Edge points should be well localized
- 3) Single edge point response

Basic Steps

- 1) smooth the input image with gaussian filter.

$$G(x, y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

$$f_s(x, y) = G(x, y) * f(x, y)$$

- 2) compute the gradient magnitude and angle images

Gradient image

$$M(x, y) = \sqrt{g_x^2 + g_y^2}$$

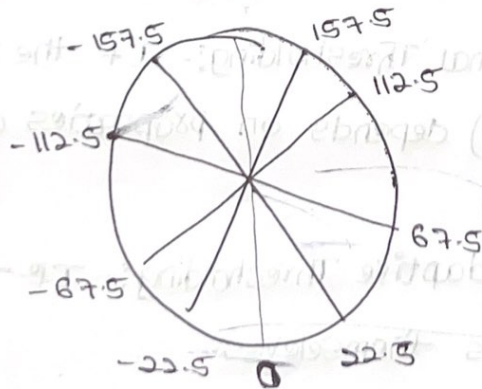
Gradient angle

$$\alpha(x, y) = \tan^{-1}\left(\frac{g_y}{g_x}\right)$$

$$g_x = \frac{\partial f_s}{\partial x}$$

$$g_y = \frac{\partial f_s}{\partial y}$$

- 3) Apply nonmaxima suppression to the gradient magnitude image



→ Let d_1, d_2, d_3, d_4 denote 4 directions: horizontal, -45° , vertical, $+45^\circ$.

→ Find the direction d that is closest to $\alpha(x, y)$

- 4) Use double thresholding and connectivity analysis to detect and link Edges

$$g_{NH}(x, y) = g_N(x, y) \geq T_H$$

$$g_{NL}(x, y) = g_N(x, y) \geq T_L$$

e)

$$f = \{0, 0, 0, 1, 0, 0, 0\}$$

$$w = \{1, 2, 3, 2, 1\}$$

convolution - $\{0, 2, 3, 2, 1\}$ (rotate 180°)

$f =$ By padding 2-zeros

1 \Rightarrow

0	0	0	0	0	0
0	2	3	2	1	0

 $\{1, 0, 0, 0, 0, 0\}$

applying filter

$0 \times 0 + 2 \times 0 + 3 \times 0 + 2 \times 0 + 1 \times 0 = 0$

2 \Rightarrow 00000100000
82321

3 \Rightarrow 00000100000
82321

applying filter = 2

4 \Rightarrow 00000100000
82321 \Rightarrow 3

5 \Rightarrow 00000100000
82321

6 \Rightarrow 00000100000
82321 \Rightarrow 8

7 \Rightarrow 00000100000
82321 \Rightarrow 0

8 \Rightarrow 00000

Final convolution result
0123280

$$\{0.001, 0.003\}$$
$$\begin{bmatrix} 24 & 22 & 33 & 25 & 32 & 54 \\ 34 & 255 & 24 & 0 & 26 & 23 \\ 23 & 21 & 32 & 31 & 28 & 26 \end{bmatrix}$$

0	0	0
0	24	22
0	34	255

Applying 3×3 median filter

median

$$\begin{bmatrix} 0 & 22 & 0 & 0 & 0 & 0 \\ 21 & 21 & 24 & 26 & 25 & 23 \\ 0 & 21 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$\epsilon = \text{rat} / 19 - \text{priv} / 990$

By applying median filter \Rightarrow all the intensity values are nearer to each other reducing the salt and pepper noise

10)

10	12	8	9
10	12	12	14
12	13	10	9
14	12	10	12

000000/000000-12

15868

Highest intensity - 14¹⁵

$$00000100000 \leftarrow +$$

10863

requires 4 bits to

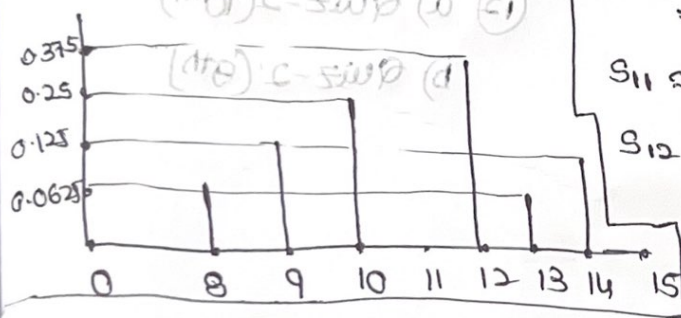
represent so

Intensity. Value lies in

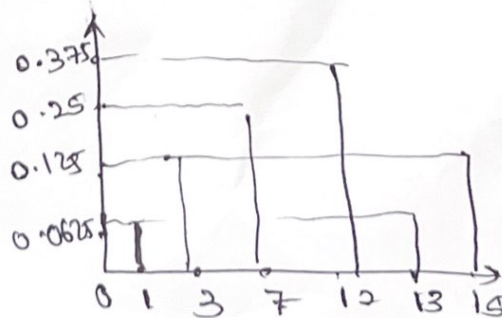
range 0,15

n_k	A_k	$P_x(x_k)$
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
5	0	0
6	0	0
7	0	0
8	1	$1/16 = 0.0625$
9	2	$2/16 = 0.125$
10	4	$4/16 = 0.25$
11	0	0
12	6	$6/16 = 0.375$
13	1	0.0625
14	2	0.125
15	0	0

Histogram for input



s_k	n_k	$P_x(s_k)$
0	0	0
1	1	0.0625
2	0	0
3	0	0
4	2	0.125
5	0	0
6	0	0
7	0	0
8	4	0.25
9	0	0
10	0	0
11	0	0
12	6	0.375
13	1	0.0625
14	0	0
15	2	0.125



For equalization

$$S_k = \sum_{j=0}^k (L-1) P_x(x_j)$$

$$\Rightarrow (L-1) \sum_{k=0}^K P_x(x_k)$$

$$S_0 \Rightarrow 15 * P_0 \Rightarrow 0$$

Similarly $S_1, S_2, S_3, S_4, S_5, S_6,$
 $S_7 \Rightarrow 0$

$$S_8 = 15 * [P_0 + P_1 + P_2 + \dots + P_8]$$

$$= 15 * [0 + 0.0625] = 0.9375$$

$$S_9 = 15 * [P_0 + P_1 + \dots + P_9] \approx 1$$

$$= 15 * [0 + 0.0625 + 0.125]$$

$$= 2.8125 \approx 3$$

$$S_{10} = 15 * [0 + 0.0625 + 0.125 + 0.25]$$

$$= 6.5625 \approx 7$$

$$S_{11} \approx S_{10} = 7$$

$$S_{12} = 15 * [0 + 0.0625 + 0.125 + 0.25 + 0.375]$$

$$= 12.1875 \approx 12$$

$$S_{13} = 13.125 \approx 13$$

$$S_{14} = 15$$

$$S_{15} = 15$$

(12)	9	10
0	0	
2000-0	10	
0	0	
201-0	50	
0	0	
0	0	
0	0	
200-0	11	
0	0	
0	0	
0	0	
0	0	
200-0	2	
2000-0	1	
0	0	
201-0	5	