

(1)

DHARMSINH DESAI UNIVERSITY, NADIAD
FACULTY OF TECHNOLOGY
ONLINE SESSIONAL EXAMINATION

B. Tech (CCE) Sem: 7

Subject : Image Processing

Roll. No. : CE 129

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Signature : @dodha

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11:15 AM

Total Pages : 18

Question -2 (b)
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↳ Objectives of canny's edge detector.

1. Low error rate. All edges should be found and there should be no spurious responses. detected edges must be as close as possible to the true edges.
2. Edge points should be well localized. The edges must be as close as possible to the true edges.

(2)

3. Single point edge response: The detector should return only one point for each true edge point. The number of local maxima around the true edge should be minimum.

⇒ Algorithm.

1. Smooth the input image with a Gaussian filter.
2. Compute the Gradient magnitude and angle images.
3. Apply non maximum suppression to the gradient magnitude image
4. Use double thresholding and connectivity analysis to detect and link edges.

Step : 1

First derivative of the Gaussian approximates the operator that optimizes the product of signal-to-noise ratio localization.

$$\frac{d}{dx} e^{-\frac{x^2}{2\sigma^2}} = \frac{-x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}$$

(3)

Image - $f(x, y)$
Gaussian fn. $G(x, y)$.

↳ smoothed image by convolving

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

Step: 2

Gradient Magnitude.

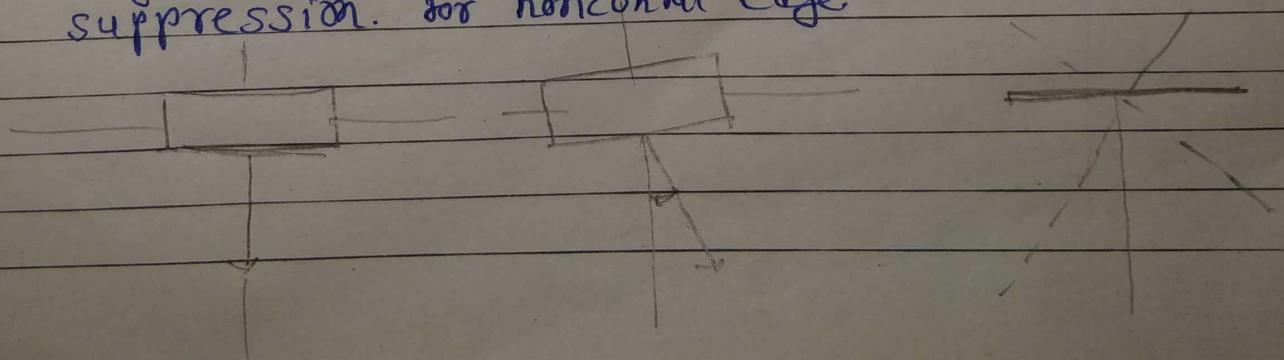
$$M(x, y) = \sqrt{g_x^2 + g_y^2}$$

and

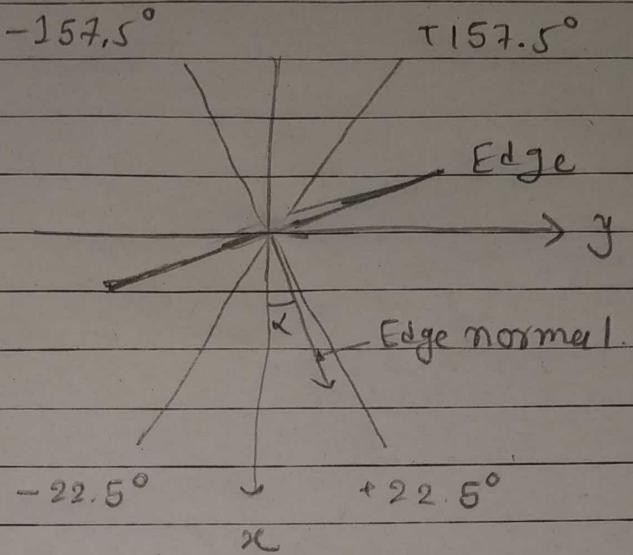
$$\alpha(x, y) = \tan^{-1} \left[\frac{g_y}{g_x} \right]$$

Step: 3 Non maxima suppression:-

Gradient typically contains wide ridges around local maxima - non maxima suppression. for horizontal edge



(4)



lets define horizontal edge. same way for other edges.

Non maximal suppression

let d_1, d_2, d_3 and d_4 denote four basic edge horizontal, vertical, $\pm 45^\circ$ respectively.

non maximal suppression for 3×3 region centered at every point (x_i, y_j) in $d(x_i, y_j)$:

1. Find the direction d_k that is closest to $d(x_i, y_j)$
2. If the value of $M(x_i, y_j)$ is less than at least one of its two neighbors along d_k , let $S_N(x_i, y_j) = 0$ (suppression)
otherwise, let $S_N(x_i, y_j) = M(x_i, y_j)$

④ ⑤

Step 4

Thresholding for canny edge detection.

Two additional images are created.

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

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

$g_{NH}(x,y)$ will have fewer nonzero pixels than $g_{NL}(x,y)$ in general, but all pixels in $g_{NH}(x,y)$ will be continuous.

so that

$$g_{NL}(x,y) = g_{NL}(x,y) - g_{NH}(x,y)$$

$g_{NH}(x,y)$ strong edge pixel

$g_{NL}(x,y)$ weak edge pixel.

In above there is some gaps longer edges are formed using the following procedure

- Locate the next unvisited edge pixel p in $g_{NH}(x,y)$

(c)

- b) Mark as valid edge pixels all the weak pixels in $g_{NL}(x, y)$ that are connected to p using say 8-connectivity
- c) If all nonzero pixels in $g_{NL}(x, y)$ have been visited go to step d.
Else step a
- d) Set to zero all pixels in $g_{NL}(x, y)$ that were not marked as valid edge pixels.

Question 2 = C

10	10	10	10	60	10	10	10
10	10	10	60	70	10	10	10
59	10	60	64	59	56	60	
10	59	10	60	70	10	10	62
10	60	59	65	67	10	10	65
10	10	50	10	10	10	10	10
10	10	10	10	10	10	10	10

Here |seed - $f(x, y)$ | ≤ 5

7

segmentation for horizontal & vertical
means & connected

Applying fork connected in above

Now applying region for horizontal, vertical & diagonal means & connected

⑧

59	60	60	64	59	56	60
59		60	70		62	
60	59	65	67		65	

(3)

Question :- 1

(a) Translation property of 2-D Discrete Fourier transform.

$$f(x, y) e^{j2\pi \left(u_0 x/M + v_0 y/N \right)} \Leftrightarrow F(u - u_0, v - v_0)$$

$$f(x - x_0, y - y_0) \Leftrightarrow f(u, v) e^{-j2\pi \left(x_0 u/M + y_0 v/N \right)}$$

Proof :-

$$f(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi \left(\frac{ux}{M} + \frac{vy}{N} \right)}$$

$$F = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{j2\pi \left(\frac{xu_0}{M} + \frac{yu_0}{N} \right)} e^{-j2\pi \left(\frac{ux}{M} - \frac{uy}{N} \right)}$$

$$F = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi \left(\frac{x(u-u_0)}{M} + \frac{y(v-v_0)}{N} \right)}$$

$$= F(u - u_0, v - v_0)$$

(b) False, Butterworth achieves more blurring than Gaussian.

For Butterworth

$$H(u, v) = \frac{1}{1 + \left(\frac{|g(u, v)|}{\Omega_0} \right)^{2n}};$$

where Ω_0 = cut-off frequency

For Gaussian

$$H(u, v) = \frac{e^{-\Omega^2 (u, v)}}{2 \Omega_0^2}$$

(c) Moire patterns are generated from sampling scenes with periodic or nearly periodic components whose spacing is comparable to spacing b/w samples.

Anti aliasing filter is used to reduce moire patterns.

(d) Basic conditions that must be fulfilled during image segmentation are:

- segmentation must be complete.
- pixels in a region must be connected.
- Regions must be disjoint.
- pixels in a region must be share same property

(e) Gradient angle image is computed by

$$\nabla f = \text{grad}(f) = \begin{bmatrix} g_x \\ g_y \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

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

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

⇒ constant intensity indicate that there is slow variation of intensities.

(f) Variable thresholding is a process in which the threshold value varies over the image as a function of local image characteristic.

⇒ Standard deviation and mean of local neighbourhood on every point is used

Eg:-

$$T_{xy} = a \sigma_{xy} + b \bar{M}_{xy}$$

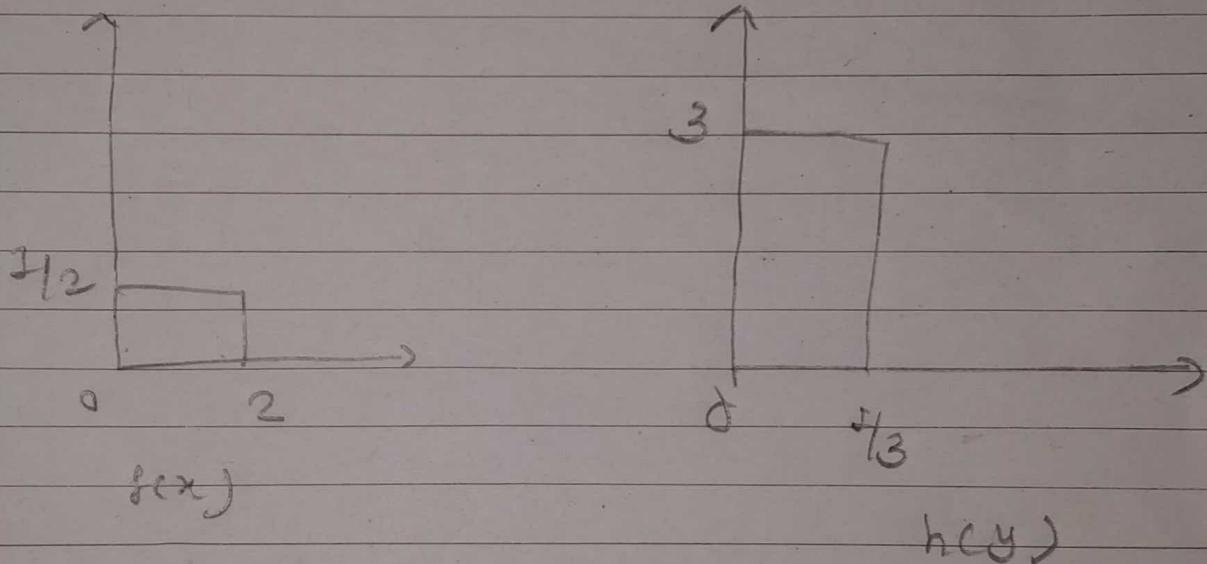
↑ ↓
 standard deviation non negative constant
 ↓ ↓
 mean

Question :- 3

$$(a) \underline{=} f(x) = \begin{cases} 1/2 & , 0 < x < 2 \\ 0 & \text{else} \end{cases}$$

$$h(y) = \begin{cases} 2/3 & , 0 < y < 3 \\ 0 & \text{else} \end{cases}$$

$$T=4$$



$$g(a) = \begin{cases} 3 & , 0 < a < 2/3 \\ 0 & \text{elsewhere} \end{cases}$$

Representation of
 $h(y)$ in another
form

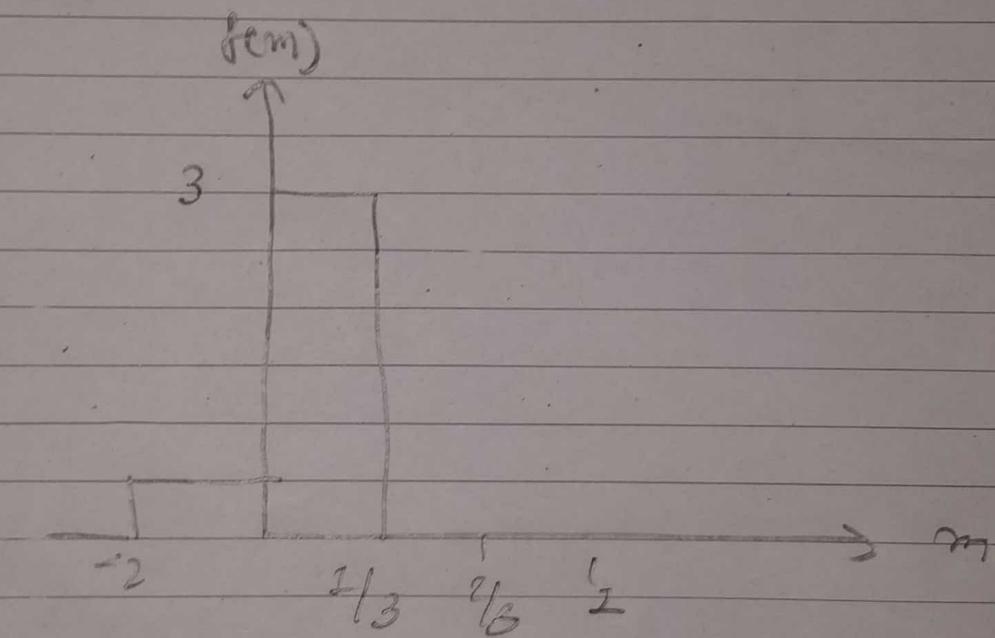
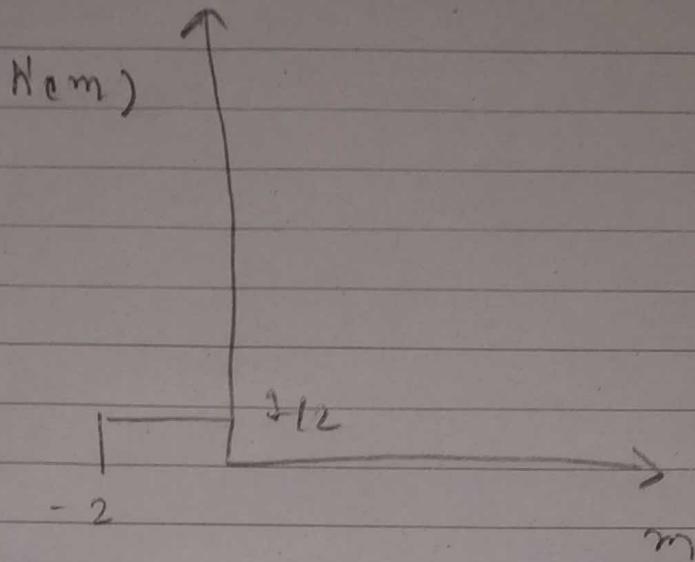
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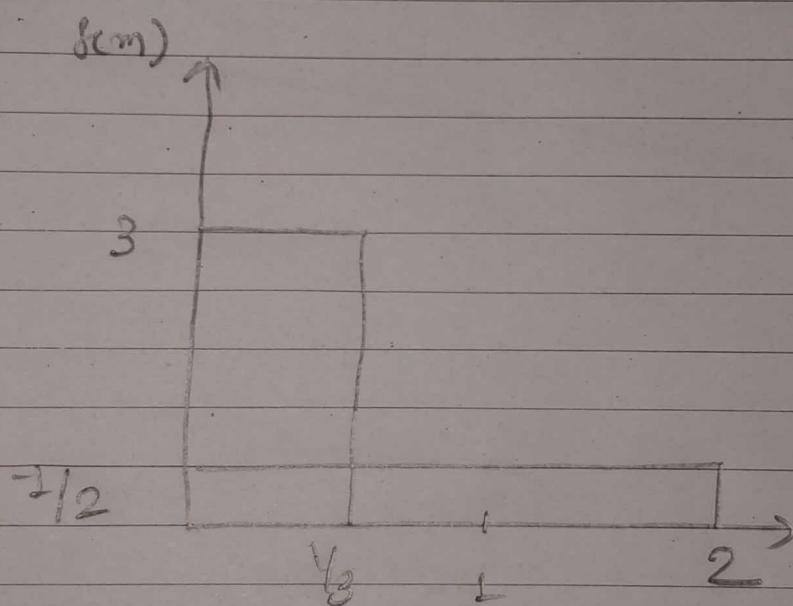
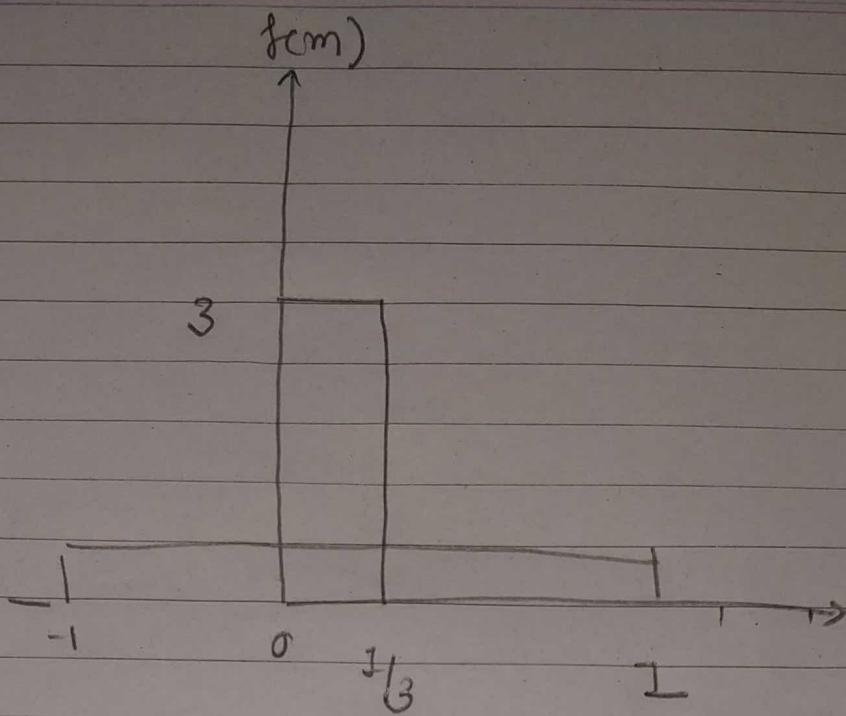
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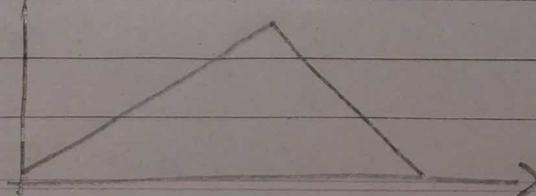
$$f(x) * h(x) = \sum_{m=0}^3 f(m) h'(x-m)$$



15



Convolution Result
 $\uparrow f(x) * g(x)$



Q. 1-3

(b)

1. Let P be the sequence of ordered, distinct points of binary image. Select two stationary points (A, B) .
2. Given a threshold T . Create two empty stacks, namely OPEN, CLOSE.
3. If points in P corresponds to a closed curve put it in OPEN, put B in both.
else
 A in OPEN
 B in CLOSE.
4. calculate complete distance of the line formed in step 3 from all point in P .
Select a point with max. distance.
Let max. dist. = d_{max}
Max. dist. point = V_{max} .

4. Compute the parameters of the line passing from the last vertex in closed to last vertex in open

5. calculate distance of the line formed in step 4 from all point in P

select a point with max
distance.

Let maximum dist. d_{max}
Max. dist point v_{max}

if $d_{max} > t$

put v_{max} at the end of
the open stack as new vertex
GO to step 4.

Step 5, if d_{max} does not exceed "T"

7. Else remove the last vertex from
OPEN and insert it as the last
vertex CLOSED.

8. If OPEN is not empty go
to step 4.

g. Else, exit. The vertices in CLOSED are the vertices of the polygonal fit to the points in t.

CLOSED	OPEN	Curve segment processes	vertex generate
B	B, A	-	A, B
B	B, A	(BA)	A, B C
B	B, A, C	(BC)	B -
B, C	B, A	(CA)	-
B, C, A	B	(AB)	= Ø
B, C, A	B, D	(AD)	B -
B, C, A, D	B	(DB)	-
B, C, A, D, B	Empty	-	-

