

1. (a) Linear  $y[i] = \frac{x[i-1] + x[i] + x[i+1]}{3}$

4+3+3

Scaling or  
 (i) Homogeneity: In case of  $kx$

output =  $\frac{kx[i-1] + kx[i] + kx[i+1]}{3} = k \cdot y[i]$

Superposition or  
 (ii) Additivity:  $x_1 + x_2$

$$y[i] = \frac{1}{3} \left( \sum_{k=i-1}^{i+1} x_1[k] + x_2[k] \right)$$

$$= \frac{1}{3} \sum_{k=i-1}^{i+1} x_1[k] + \frac{1}{3} \sum_{k=i-1}^{i+1} x_2[k]$$

$$= y_1[i] + y_2[i]$$

Shift invariance

(iii)  $x[i+s]$ ; output =  $\frac{1}{3}(x[i+s-1] + x[i+s] + x[i+s+1]) = y[i+s]$

2. (b)  $\frac{1}{3} (\delta[n-1] + \delta[n] + \delta[n+1])$

(c)  $\frac{1}{5} (\delta[n-2] + \delta[n-1] + \delta[n] + \delta[n+1] + \delta[n+2])$

4x3

2. (a)  $x[n] * h[n]$

$$= \delta[n-a] * \delta[n-1] + \delta[n-a] * \delta[n+1]$$

$$+ \delta[n-b] * \delta[n-1] + \delta[n-b] * \delta[n+1]$$

$$= \delta[n-a-1] + \delta[n-a+1] + \delta[n-b-1] + \delta[n-b+1]$$

(b)  $x[n] * h[n] = e^n * \delta[n+2] = e^{n+2}$

(c)  $x[n] * h[n] = e^{-n} * \delta[n-2] = e^{-n+2}$

(d)  $x[n] * h[n] = e^{-n} * \delta[n] - e^{-n} * \delta[n-1]$

$$= e^{-n} - e^{-n+1} = e^{-n} (1-e)$$

7

2+3

3. (a)

	$x-1$	$x$	$x+1$
$y$	-1	+1	0

(b)

$y$	0	-1	1
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(c)

	$x-1$	$x$	$x+1$
$y-1$	-1	0	1
$y$	-2	0	+2
$y+1$	-1	0	1

4.

impulse response

(a)

$a[n] = \delta[n]$

← unit impulse

(b)

inverted output =  $-b[n]$

(c)

$a[n] + b[n] = \delta[n] + b[n]$

(d)

$a[n] - b[n] = \delta[n] - b[n]$

(e)

(d) is a high pass filter

(f)

for c, in case of low freq



ghosting artifact

for d,

2+3+3

2+3

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5.

blurs  $B = \begin{bmatrix} 1/4 & 1/4 \\ 1/4 & 1/4 \end{bmatrix}$

, Horizontal gradient,  $G = \begin{bmatrix} -1 & +1 \\ -1 & +1 \end{bmatrix}$

4+3+4+4

(a) gradient is (in 1D)

$$g[n] = x[n-1] - x[n]$$

i) Homogeneity:  $k x[n]$

$$g_{\text{gradient}} = k x[n-1] - k x[n] = k g[n]$$

ii) Additivity:  $x[n] = x_1[n] + x_2[n]$

$$\begin{aligned} g_{\text{gradient}} &= x_1[n-1] + x_2[n-1] - x_1[n] - x_2[n] \\ &= g_1[n] + g_2[n] \end{aligned}$$

iii) Shift invariance:  $x[n+s]$

$$g_{\text{gradient}} = x[n+s-1] - x[n+s] = g[n+s]$$

(b) Convolution is commutative, so same output

(c)  $S = B * G$

(d) 
$$\begin{aligned} I' &= I * B + I * G \\ &= I * (B + G) \end{aligned}$$

$\therefore S = B + G$

- (a) curvature based  $\rightarrow$  Laplacian, so it will take single conv  
6.(c) (b) others multiple convolution

The low pass filter is employed to remove high frequency noise from a digital image. i.e. it is employed for smoothing before applying the curvature or gradient filter.

6.(d)

After passing through the low-pass filter for smoothing, the pixel values of the resulting signal will no longer be independent. Rather, smoothing was used as a method to predict a pixel's value from the values of its neighbors. However, if we increase the width of the low pass filter, then the pixels will tend to look <sup>more</sup> like their neighbors and the derivatives will become smaller (because they measure the tendency of pixels to look different from their neighbors).

~~6(a)~~

~~C will require multiple convolution operation. The output value of the edge finding detector in pixel  $(i, j)$  is~~

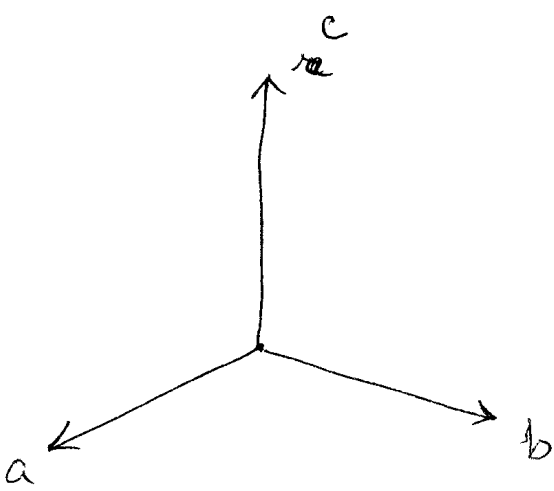
$$~~f(i, j) = \max [0, \max_k (g * h_k)]~~$$

~~For a particular size, we have to consider all possible orientations.~~

6(b) 6(a & b)

The gradient operators which are rotationally invariant (e.g. Laplacian) are computed from single mask only. Others, use multiple masks.

7. (a) 3D



(b) Initialize

For each edge pixel  $(x, y)$  in image

For each radius  $r$

For  $\theta = 0$  to  $360$

$$a = x + r \cos \theta$$

$$b = y + r \sin \theta$$

$$c = r$$

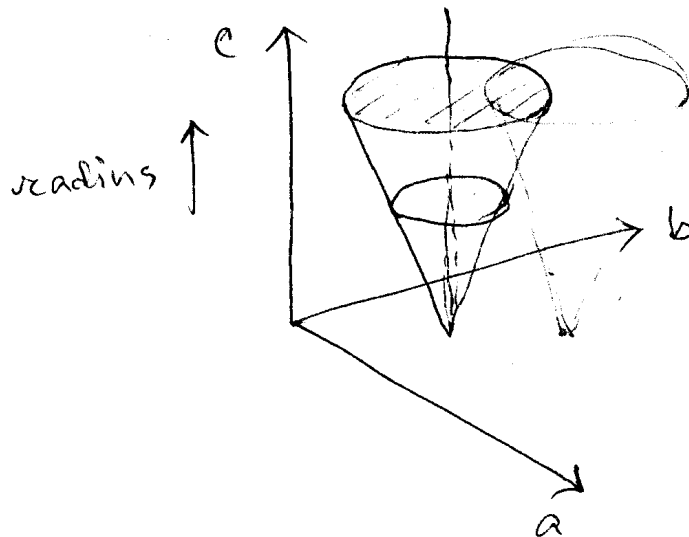
$$H(a, b, c) += 1$$

Or,

$$\begin{aligned} x &= a + r \cos \theta \\ y &= b + r \sin \theta \\ z &= c \end{aligned}$$

end  
end  
end

(c)



center(a, b)