

Q2)

$$\text{Ans (a)} \quad g(x, y) = (h * f)(x, y) \quad \text{--- (1)}$$

Taking fourier transform of both sides

$$G(u, v) = H(u, v) \cdot F(u, v) \quad \text{--- (2)}$$

$$\text{Now, } \hat{F}(u, v) = \frac{G(u, v)}{H(u)} = F(u, v) \quad \text{--- (3)}$$

$$\text{So, } f(x, y) = F^{-1}(\hat{F}(u, v))$$

The fundamental difficulties we will face will be that as $h(x)$ is a convolution kernel to represent gradient, hence $H(u)$ is a high pass filter.

So at low frequency $H(u)$ tends to 0 which causes $\hat{F}(u, v)$ to blow up.

If noise is considered, $H(u)$ blows up amplifying noise in $f(x, y)$

(b) For 2-d gradient,

$$g_x(x, y) = (h_x * f)(x, y) \quad \text{--- (1)}$$

$$g_y(x, y) = (h_y * f)(x, y) \quad \text{--- (2)}$$

Taking fourier transform of eq (1) & (2),

$$G_x(u, v) = H_x(u, v) F(u, v) \quad \text{--- (3)}$$

$$G_y(u, v) = H_y(u, v) F(u, v) \quad \text{--- (4)}$$

Using eq (3) & (4)

$$\hat{F}(u, v) = \frac{G_x(u, v)}{H_x(u, v)} = F(u, v) \quad \text{--- (5)}$$

$$F(f(x, y)) = \hat{F}(u, v) = \frac{G_y(u, v)}{H_y(u, v)} = F(u, v) \quad \text{--- (6)}$$

$$f(x, y) = F^{-1}(\hat{F}(u, v))$$

Similar to the previous part, at low frequency $H_x(u, v) = H_y(u, v) \rightarrow 0$ and $\hat{F}(u, v)$ blows up.

To estimate good values of $\hat{F}(u, v)$ we can use eq (5) & (6) for appropriate boundary conditions.

$h_x(x, y)$ is gradient in x direction while $h_y(x, y)$ is gradient in y direction. $H_x(u, v)$ is a high pass filter in u while $H_y(u, v)$ is high pass filter in v .

(i) when $u \ll 1$ is small

Use equation (6) to estimate $f(x, y)$.

(ii) when $v \ll 1$ is small

Use eq. (5)

(iii) when u & $v \ll 1$ are small.

We cannot estimate $f(x, y)$ accurately as $\hat{F}(x, y)$ blows up in both eq.