

Q. 2

$$g = h \star f$$

Apply DFT,

$$G(\mu) = H(\mu) F(\mu)$$

$$\Rightarrow F(\mu) = \frac{G(\mu)}{H(\mu)}$$

Again When  $H(\mu) = 0$ ,  $F(\mu)$  will shoot up

$h(\cdot)$  is a convolutional kernel

$$\text{Such that } g(x) = f(x+1) - f(x) \quad \text{for } 1 \leq x \leq N$$

Applying DFT, we know that

$$G(\mu) = (e^{j\frac{2\pi}{N}\mu} - 1) F(\mu)$$

$$\Rightarrow F(\mu) = \frac{G(\mu)}{(e^{j\frac{2\pi}{N}\mu} - 1)}$$

Hence for  $\mu = 0$ ,  $F(\mu)$  shoots up

To solve this, we need to assume some boundary condition

$$\text{Let } f(N+1) = 0$$

Then starting from the rightmost pixel  $N$ , obtain  $f(\cdot)$  by reverse integration

For the 2D images, we have

$$G_x(\mu, \sigma) = H_x(\mu, \sigma) F(\mu, \sigma) = (e^{j\frac{2\pi}{N}\mu} - 1) F(\mu, \sigma)$$

$$G_y(\mu, \sigma) = H_y(\mu, \sigma) F(\mu, \sigma) = (e^{j\frac{2\pi}{N}\sigma} - 1) F(\mu, \sigma)$$

Knowing  $G_x$ ,  $\mu = 0 \rightarrow$  shoot up

Knowing  $G_y$ ,  $\sigma = 0 \rightarrow$  shoot up

Hence, we can rewrite as:

$$F(\mu, \sigma) = \frac{G_x - G_y}{H_x - H_y}$$

Hence, even when  $\mu=0$ ,  $\sigma \neq 0$  won't shoot up

Similarly  $\sigma=0$ ,  $\mu \neq 0$  won't shoot up

But  $\mu=0$  and  $\sigma=0$  will still be problematic

Hence, the use of integration based on boundary conditions will be ineffective because integration across rows ( $f_x$ ) will be inconsistent with integration across columns ( $f_y$ )