

Q5) Let  $G(K)$  denote the gaussian kernel with mean 0 and standard deviation  $\sigma$ . Then for a given image  $I(x)$ , if we apply the filter  $G(K)$ , the output  $J(x)$  would be the convolution  $(G * I)(x)$ .

$$J(x) = (G * I)(x)$$

$$J(x) = \sum_{K=-\infty}^{+\infty} I(x-K) G(K)$$

$$= \sum_{K=-\infty}^{+\infty} (cx - cK + d) G(K)$$

$$= \sum_{K=-\infty}^{+\infty} cx G(K) - \sum_{K=-\infty}^{+\infty} cK G(K) + d \sum_{K=-\infty}^{+\infty} G(K)$$

$$= cx \left( \sum_{K=-\infty}^{+\infty} G(K) \right) - c \left( \sum_{K=-\infty}^{+\infty} K G(K) \right) + d \left( \sum_{K=-\infty}^{+\infty} G(K) \right)$$

Note that for any PMF,  $\sum_{K=-\infty}^{+\infty} G(K) = 1$ .

Also note that  $\sum_{K=-\infty}^{+\infty} K G(K) = \cancel{0} 0$  because  $G(K) = G(-K)$ .

So, we can say that  $J(x) = cx(1) - 0 + d(1)$

$$\boxed{J(x) = cx + d}$$

Now, for a Bilateral Filter, we know that

$$BF[I]_p = \frac{1}{W_p} \sum_{q \in S} G_{\sigma_s}(\|p - q\|) G_{\sigma_r}(I_p - I_q) I_q$$

$$\text{where } W_p = \sum_{q \in S} G_{\sigma_s}(\|p - q\|) G_{\sigma_r}(I_p - I_q)$$

Note that the image is of infinite extent. So  $q$  goes from  $-\infty$  to  $+\infty$ .

Let  $q = K$  and  $p = x$  for convenience.

So,  $K$  goes from  $-\infty$  to  $+\infty$ .

$$\begin{aligned}
 BF[I]_x &= \frac{\sum_{k=-\infty}^{+\infty} I(k) e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{(I(x)-I(k))^2}{2\sigma_r^2}}}{\sum_{k=-\infty}^{+\infty} e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{(I(x)-I(k))^2}{2\sigma_r^2}}} \\
 &= \frac{\sum_{k=-\infty}^{+\infty} (ck+d) e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{c^2(x-k)^2}{2\sigma_r^2}}}{\sum_{k=-\infty}^{+\infty} e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{c^2(x-k)^2}{2\sigma_r^2}}}
 \end{aligned}$$

$$= d + c \left( \frac{\sum_{k=-\infty}^{+\infty} k e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{c^2(x-k)^2}{2\sigma_r^2}}}{\sum_{k=-\infty}^{+\infty} e^{-\frac{(x-k)^2}{2\sigma_s^2}} e^{-\frac{c^2(x-k)^2}{2\sigma_r^2}}} \right)$$

$$= d + c \left( \frac{\sum_{k=-\infty}^{+\infty} k e^{-\left[\frac{(x-k)^2}{2}\right] \left[\frac{1}{\sigma_s^2} + \frac{c^2}{\sigma_r^2}\right]}}{\sum_{k=-\infty}^{+\infty} e^{-\left[\frac{(x-k)^2}{2}\right] \left[\frac{1}{\sigma_s^2} + \frac{c^2}{\sigma_r^2}\right]}} \right)$$

$$\text{Let } \boxed{\frac{1}{d^2} = \frac{1}{\sigma_s^2} + \frac{c^2}{\sigma_r^2}}$$

$$BF[I]_x = d + c \left( \frac{\sum_{k=-\infty}^{+\infty} k e^{-\frac{(x-k)^2}{2d^2}}}{\sum_{k=-\infty}^{+\infty} e^{-\frac{(x-k)^2}{2d^2}}} \right)$$

$$= d + c \left( \frac{\sum_{k=-\infty}^{+\infty} k e^{-\frac{(k-x)^2}{2d^2}}}{\sum_{k=-\infty}^{+\infty} e^{-\frac{(k-x)^2}{2d^2}}} \right)$$

The denominator is the PDF of a gaussian with mean =  $x$  and standard deviation  $d$ .

Since sum of PDF must be 1, the denominator is 1.



The numerator is convolution between  $f(x) = x$  and  $g(x) = e^{-x^2/2\sigma^2}$ .

So, we can also write it as  $\sum_{k=-\infty}^{+\infty} (x-k) e^{-k^2/2\sigma^2}$  because convolution is commutative.

$$\text{Numerator} = x \sum_{k=-\infty}^{+\infty} e^{-k^2/2\sigma^2} - \sum_{k=-\infty}^{+\infty} k e^{-k^2/2\sigma^2}$$

$$= x(1) - 0$$

$$= x$$

PDF  
sums  
to 1

odd function. So,  
sum from  $-\infty$  to  $+\infty$   
will be zero.

$$\therefore BF[I]_x = d + \frac{c(x)}{(1)}$$

$$\boxed{BF[I]_x = (x + d)}$$