

1) A)

Clean Image intensity: $I(x, y)$

Let Noisy Image intensity be $I'(x, y)$

Let the Pdf of Intensity I be $P_I(x)$

We can say $I'(x, y) = I(x, y) + N(x, y)$

$$P_N(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{x^2}{2\sigma^2}\right) \quad (\text{Let it be the Pdf of } N(x, y))$$

W.K.T when two independent random variable A and B summation gives

$$\text{i.e. } S = A + B$$

$$P_S(x) = \int_{-\infty}^{\infty} P_A(t) P_B(x-t) dt$$

Proof:-

$$S = A + B$$

$$\text{consider } P_S(x) = P(S \leq x)$$

$$= P(A + B \leq x)$$

$$\Rightarrow P_S(x) = \int_{-\infty}^{\infty} P_B(x-t) P_A(t) dt$$

$$P_S(x) = \frac{d}{dx} P_S(x)$$

$$P_S(x) = \frac{d}{dx} \int_{-\infty}^{\infty} P_B(x-t) P_A(t) dt$$

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Let us use this relation to calculate $P_{I'}(x)$

$$\text{as } I' = I + N$$

$$P_{I'}(x) = \int_{-\infty}^{\infty} P_I(t) P_N(x-t) dt$$

$$= \int_{-\infty}^{\infty} P_{\pm}(t) \cdot e^{\frac{(-x-t)^2}{2\sigma^2}} \times \frac{1}{\sqrt{2\pi}\sigma} dt.$$

→ The resultant Pdf resembles "convolution operation" in Image Processing.

If the Noise is uniform ~~for~~ from $-r$ to r , $r > 0$

$$\text{w.x.t } P_N(x) = \begin{cases} \frac{1}{2r} & -r \leq x \leq r \\ 0 & x > r \text{ or } x < -r. \end{cases}$$

$$I'(x, y) = I(x, y) + N(x, y)$$

$$P_{I'}(x) = \int_{-\infty}^{\infty} P_I(t) \cdot P_N(x-t) dt.$$

$$= \int_{-\infty}^{t-r} P_I(t) \cdot 0 dt + \int_{t-r}^{t+r} P_I(t) \times \frac{1}{2r} dt + \int_{t+r}^{\infty} P_I(t) \cdot 0 dt$$

$$P_{I'}(x) = \frac{1}{2r} \int_{t-r}^{t+r} P_I(t) dt$$