

**Problem 8.8** Show that the mean and variance of a Gaussian random variable  $X$  with the density function given by Eq. (8.48) are  $\mu_X$  and  $\sigma_X^2$ .

**Solution**

Consider the difference  $\mathbf{E}[X] - \mu_X$ :

$$\mathbf{E}[X] - \mu_X = \int_{-\infty}^{\infty} \frac{(x - \mu_X)}{\sqrt{2\pi}\sigma_X} \exp\left\{-\frac{(x - \mu_X)^2}{2\sigma_X^2}\right\} dx$$

Let  $y = x - \mu_X$  and substitute

$$\begin{aligned} \mathbf{E}[X] - \mu_X &= \int_{-\infty}^{\infty} \frac{y}{\sqrt{2\pi}\sigma_X} \exp\left(-\frac{y^2}{2\sigma_X^2}\right) dy \\ &= 0 \end{aligned}$$

since integrand has odd symmetry. This implies  $\mathbf{E}[X] = \mu_X$ . With this result

$$\begin{aligned} \text{Var}(X) &= \mathbf{E}(x - \mu_X)^2 \\ &= \int_{-\infty}^{\infty} \frac{(x - \mu_X)^2}{\sqrt{2\pi}\sigma_X} \exp\left\{-\frac{(x - \mu_X)^2}{2\sigma_X^2}\right\} dx \end{aligned}$$

In this case let

$$y = \frac{x - \mu_X}{\sigma_X}$$

and making the substitution, we obtain

$$\text{Var}(X) = \sigma_X^2 \int_{-\infty}^{\infty} \frac{y^2}{\sqrt{2\pi}} \exp\left\{-\frac{y^2}{2}\right\} dy$$

Recalling the integration-by-parts, i.e.,  $\int u dv = uv - \int v du$ , let  $u = y$  and

$$dv = y \exp\left(-\frac{y^2}{2}\right) dy. \text{ Then}$$

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Problem 8.8 continued

$$\begin{aligned}\text{Var}(X) &= \sigma_x^2 \left. \frac{(-)y}{\sqrt{2\pi}} \exp\left(-\frac{y^2}{2}\right) \right|_{-\infty}^{\infty} + \sigma_x^2 \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y^2}{2}\right) dy \\ &= 0 + \sigma_x^2 \cdot 1 \\ &= \sigma_x^2\end{aligned}$$

where the second integral is one since it is integral of the normalized Gaussian probability density.