

Ex- 2.4

For a uniform distribution:

$$p(x) = \begin{cases} \frac{1}{b-a} & ; \quad a < x \leq b \\ 0 & ; \quad \text{otherwise} \end{cases}$$

$$\mathbb{E}[x] = \int_a^b x \cdot \frac{1}{b-a} dx = \frac{a+b}{2}$$

$$\mathbb{E}[x^2] = \frac{(b-a)^2}{12}$$

\Rightarrow These values depend only on the interval, not the actual value of x .

Gaussian Distribution

$$p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\mathbb{E}[x] = \int_{-\infty}^{\infty} x p(x) dx$$

$$\mathbb{E}[x] = \int_{-\infty}^{\infty} x \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$\text{Substitute } x-\mu = y \Rightarrow x = y+\mu \quad \lambda dy = dx$$

$$= \int_{-\infty}^{\infty} (y+\mu) \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{y^2}{2\sigma^2}} dy$$

$$= \int_{-\infty}^{\infty} y \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{y^2}{2\sigma^2}} dy + \mu \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{y^2}{2\sigma^2}} dy$$

(total probability)

$$= 0 + \mu \times 1 = \mu$$



Ex - 2.8

$$\text{var}[b] = \mathbb{E}[(b(x) - \mathbb{E}[b(x)])^2] \quad (2.44)$$

$$\begin{aligned}\mathbb{E}[(b(x) - \mathbb{E}[b(x)])^2] &= \mathbb{E}[b(x)^2 - 2b(x)\mathbb{E}[b(x)] + \mathbb{E}[b(x)]^2] \\ &= \mathbb{E}[b(x)^2] - 2\mathbb{E}[b(x)]\mathbb{E}[b(x)] + \mathbb{E}[b(x)]^2 \\ &= \mathbb{E}[b(x)^2] - \mathbb{E}[b(x)]^2 \quad (2.45)\end{aligned}$$

as required.

Ex - 2.9

$$\text{cov}[x, y] = \mathbb{E}[xy] - \mathbb{E}[x]\mathbb{E}[y] \quad (2.47)$$

using (2.38)

$$\mathbb{E}[xy] = \sum_x \sum_y p(x,y) xy$$

since x & y are independent $p(x,y) = p(x)p(y)$

$$\begin{aligned}&= \sum_x p(x)x \sum_y p(y)y \\ &= \mathbb{E}[x] \cdot \mathbb{E}[y]\end{aligned}$$

$$\Rightarrow \text{cov}(x, y) = 0$$

