



Probability Theory – Part 18

Properties of variance and standard deviation:

Let X, Y be independent random variables where $\mathbb{E}(X^2)$ and $\mathbb{E}(Y^2)$ exist.

Then: (a) $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$

(b) $\text{Var}(\lambda \cdot X) = \lambda^2 \cdot \text{Var}(X)$ for every $\lambda \in \mathbb{R}$

(c) $\sigma(\lambda \cdot X) = |\lambda| \cdot \sigma(X)$ for every $\lambda \in \mathbb{R}$

Proof: (a)
$$\begin{aligned} \text{Var}(X+Y) &= \mathbb{E}((X+Y)^2) - \mathbb{E}(X+Y)^2 \\ &= \mathbb{E}(X^2 + 2XY + Y^2) - (\mathbb{E}(X) + \mathbb{E}(Y))^2 \\ &= \mathbb{E}(X^2) + 2\mathbb{E}(XY) + \mathbb{E}(Y^2) - \mathbb{E}(X)^2 - 2\mathbb{E}(X)\mathbb{E}(Y) - \mathbb{E}(Y)^2 \\ &= \text{Var}(X) + \text{Var}(Y) + 2 \cdot (\underbrace{\mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y)}_{\substack{= \mathbb{E}(X)\mathbb{E}(Y) \\ \text{independence}}}) \end{aligned}$$

(b)
$$\begin{aligned} \text{Var}(\lambda \cdot X) &= \mathbb{E}((\lambda \cdot X)^2) - \mathbb{E}(\lambda \cdot X)^2 \\ &= \lambda^2 \mathbb{E}(X^2) - \lambda^2 \mathbb{E}(X)^2 = \lambda^2 \cdot (\mathbb{E}(X^2) - \mathbb{E}(X)^2) \\ &= \lambda^2 \cdot \text{Var}(X) \end{aligned}$$

(c)
$$\sigma(\lambda \cdot X) = \sqrt{\text{Var}(\lambda \cdot X)} \stackrel{(b)}{=} |\lambda| \cdot \sigma(X)$$