

① Properties of Expected Value: { $E[\cdot]$: is a linear operator }

Ⓐ $E[c] = c$

Ⓑ $E[cX] = cE[X]$

Ⓒ $E[aX + bY] = aE[X] + bE[Y]$

② Properties of Variance:

Ⓙ $Var[X] = E[(X - E[X])^2]$

Ⓚ $Var[X] = Cov[X, X]$

Ⓛ For two jointly distributed real-valued random variables X and Y with finite second moments { $E[X^2] < \infty$, $E[Y^2] < \infty$ }, we have that:

$Cov[X, Y] = E[(X - E[X])(Y - E[Y])]$

(*) Derivation:

$Cov[X, Y] = E[(X - E[X])(Y - E[Y])] \Rightarrow$

$Cov[X, Y] = E[X \cdot Y - X \cdot E[Y] - Y \cdot E[X] + E[X]E[Y]] \Rightarrow$

$Cov[X, Y] = E[X \cdot Y] - E[X \cdot E[Y]] - E[Y \cdot E[X]] + E[E[X]E[Y]]$

$Cov[X, Y] = E[X \cdot Y] - E[X]E[Y] - E[X]E[Y] + E[X]E[Y] \Rightarrow$

$Cov[X, Y] = E[X \cdot Y] - E[X]E[Y]$

Ⓜ Apparently, by setting $X = Y$, we can get that:

$Cov[X, X] = Var[X] = E[X^2] - E[X]^2$

(v) $\text{Var}[X] \geq 0$ (Variance is non-negative)

(vi) $\text{Var}[c] = 0$ (Variance of a constant is zero)

(vii) $\text{Var}[X] = 0 \iff \exists a \in \mathbb{R} : P(X=a) = 1$

(viii) $\text{Var}[a+X] = \text{Var}[X]$

(ix) $\text{Var}[aX] = a^2 \text{Var}[X]$

(*) Proof of Property ix

$$\text{Var}[aX] = E[a^2 X^2] - E[aX]^2 \Rightarrow$$

$$\text{Var}[aX] = a^2 E[X^2] - (a E[X])^2 \Rightarrow$$

$$\text{Var}[aX] = a^2 E[X^2] - a^2 E[X]^2 \Rightarrow$$

$$\text{Var}[aX] = a^2 \{E[X^2] - E[X]^2\} \Rightarrow$$

$$\text{Var}[aX] = a^2 \text{Var}[X]$$

(*) Variance of sum of random variables:

$$\text{Var}[aX + \beta Y] = a^2 \text{Var}[X] + \beta^2 \text{Var}[Y] + 2a\beta \text{Cov}[X, Y]$$

$$\text{Var}[aX - \beta Y] = a^2 \text{Var}[X] + \beta^2 \text{Var}[Y] - 2a\beta \text{Cov}[X, Y]$$

$$\text{Var}[aX + \beta Y] = E[\underbrace{((aX + \beta Y) - E[aX + \beta Y])^2}_{\text{(By Definition)}}] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = E[\underbrace{((aX + \beta Y) - (E[aX] + E[\beta Y]))^2}_{\text{(By Definition)}}] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = E[\underbrace{((aX + \beta Y) - (aE[X] + \beta E[Y]))^2}_{\text{(By Definition)}}] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = E[(a(X - E[X]) + \beta(Y - E[Y]))^2] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = E[a^2(X - E[X])^2 + \beta^2(Y - E[Y])^2 + 2a\beta(X - E[X])(Y - E[Y])] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = a^2 E[(X - E[X])^2] + \beta^2 E[(Y - E[Y])^2] + 2a\beta E[(X - E[X])(Y - E[Y])] \Rightarrow$$

$$\text{Var}[aX + \beta Y] = a^2 \text{Var}[X] + \beta^2 \text{Var}[Y] + 2a\beta \text{Cov}[X, Y]$$

(xi) Linear Combinations:

$$\text{Var} \left[\sum_{i=1}^n x_i \right] = \sum_{i=1}^n \sum_{j=1}^n \text{Cov} [x_i, x_j] + \sum_{i \neq j} \text{Cov} [x_i, x_j]$$

or in its more general form:

$$\begin{aligned} \text{Var} \left[\sum_{i=1}^n a_i x_i \right] &= \sum_{i=1}^n \sum_{j=1}^n a_i a_j \text{Cov} [x_i, x_j] = \\ &= \sum_{i=1}^n a_i^2 \text{Var} [x_i] + \sum_{i \neq j} a_i a_j \text{Cov} [x_i, x_j] \end{aligned}$$

(*). Apparently, when the random variables x_i are pairwise independent, such that:

$$\text{Cov} [x_i, x_j] = 0, \quad \forall i \neq j$$

we can write that:

$$\text{Var} \left[\sum_{i=1}^n x_i \right] = \sum_{i=1}^n \text{Var} [x_i]$$

► Άσκηση : Αν (x_1, x_2, \dots, x_n) τυχαίο δείγμα από άπειρο ή πεπερασμένο πληθυσμό με μέσο (μ) και διακύμανση (σ^2) όπου τα x_i είναι ανεξάρτητα τότε:

$$(i): E[\bar{X}] = \frac{1}{n} E[x_1 + x_2 + \dots + x_n] = \mu$$

$$(ii): \text{Var}[\bar{X}] = \frac{1}{n^2} \text{Var}[x_1 + x_2 + \dots + x_n] \stackrel{\text{iid}}{=} \frac{\sigma^2}{n}$$

► Λύση: Για το (i) έχουμε ότι: $\bar{X} = \frac{1}{n} \sum_{i=1}^n x_i$ και

$$E[x_i] = \mu, \forall i \in \{1, \dots, n\}$$

Επομένως, έχουμε ότι:

$$\begin{aligned} E[\bar{X}] &= E\left[\frac{1}{n} \sum_{i=1}^n x_i\right] = \frac{1}{n} E\left[\sum_{i=1}^n x_i\right] = \\ &= \frac{1}{n} \sum_{i=1}^n E[x_i] = \frac{1}{n} \cdot n \cdot \mu = \mu \end{aligned}$$

Για το (ii) έχουμε ότι:

$$\text{Var}[x_i] = \sigma^2, \forall i \in \{1, \dots, n\}$$

$$\text{Var}[\bar{X}] = \text{Var}\left[\frac{1}{n} \sum_{i=1}^n x_i\right] = \frac{1}{n^2} \text{Var}\left[\sum_{i=1}^n x_i\right] = \frac{1}{n^2} \sum_{i=1}^n \text{Var}[x_i] \Rightarrow$$

$$\text{Var}[\bar{X}] = \frac{1}{n^2} \cdot n \cdot \sigma^2 = \frac{\sigma^2}{n}$$

► Problem: A poker hand (5 cards) is dealt off the top of a well-shuffled deck of 52 cards.

Let X be the number of diamonds in the hand.

Let Y be the number of hearts in the hand.

- (a): Do you think $\text{Cov}[X, Y]$ is positive, negative or zero.
- (b): Let D_i ($i = 1, 2, 3, 4, 5$) be a random variable that is 1 if the i -th card is a diamond and 0 otherwise. What is $E[D_i]$?
- (c): Let H_i ($i = 1, 2, 3, 4, 5$) be a random variable that is 1 if the i -th card is a heart and 0 otherwise. Prove that $E[H_i]$ is the same as $E[D_i]$. What is $\text{Cov}[D_i, H_i]$? What is $\text{Cov}[D_i, H_j]$?, when $i \neq j$. Keep in mind that D_i and H_i are indicator random variables that take on values 0 or 1.
- [Hint: make a table for the joint p.m.f. There are only 4 possible outcomes]
- (d): Use the previous answers on questions (b), (c) to calculate $\text{Cov}[X, Y]$?