

Expectation & Variance

REVIEW

Expected Value

For discrete RV X , the **expected value** is defined as

$$E(X) = \sum_x x p_x(x)$$

for PMF $p_x(x)$

↳ If you repeat an experiment n times, and the observed samples are (c_1, \dots, c_n) , the average of the n values is

$$\frac{1}{n} \sum_i c_i = \sum_x x \cdot \frac{(\# \text{ times } c_i = x)}{n} \approx E(X) \text{ as } n \rightarrow \infty$$

Example

Let $X \sim \text{Bernoulli}(p)$ ($p_x(x) = p^x (1-p)^{1-x}$, $x \in \{0, 1\}$)

By definition,

$$E(X) = \sum_x x p_x(x) = 0 \cdot p^0 (1-p)^{1-0} + 1 \cdot p^1 (1-p)^{1-1} = p$$

Suppose you roll a fair, 6-sided die nine times, and your outcomes are $(1, 2, 1, 5, 6, 6, 5, 5, 4)$

$$\frac{1}{n} \sum_i c_i = \frac{1}{n} \sum_x x \cdot (\# \text{ times } c_i = x) = \frac{1}{9} (2 \cdot 1 + 1 \cdot 2 + 0 \cdot 3 + 1 \cdot 4 + 3 \cdot 5 + 2 \cdot 6) = \frac{35}{9}$$

Properties of Expected Value

• For $Y = g(X)$, for some function $g(\cdot)$

$$E(Y) = E(g(X)) = \sum_x g(x) p_x(x)$$

$$(E(X^2) = \sum_x x^2 p_x(x))$$

• For multivariate function $Z = g(X, Y)$ of discrete RVs X, Y

$$E(Z) = E(g(X, Y)) = \sum_x \sum_y g(x, y) p_{X,Y}(x, y) \quad (E(X+Y) = \sum_x \sum_y (x+y) p_{X,Y}(x, y))$$

- For functions $g(\cdot), h(\cdot)$, and RVs $X, Y, a, b \in \mathbb{R}$

$$E(ag(X) + bh(Y)) = aE(g(X)) + bE(h(Y)) \quad (E(2X^2 - Y^3) = 2E(X^2) - E(Y^3))$$
- For functions $g(\cdot), h(\cdot)$, and RVs X, Y (where X, Y are independent)

$$E(g(X) \cdot h(Y)) = E(g(X))E(h(Y)) \quad (E(X^2 Y^2) = E(X^2)E(Y^2))$$

Variance

The **variance** (measure of spread) for RV X is

$$V(X) = E((X - \mu_X)^2) = E(X^2) - \mu_X^2$$

Used to measure how much a random variable deviates from its mean.

and is often denoted by σ^2 .

↳ From the variance, we obtain the **standard deviation**

$$SD(X) = \sqrt{\sigma^2}$$

and is often denoted by σ .

Example

Let $X \sim \text{Bernoulli}(p)$ ($p_X(x) = p^x (1-p)^{1-x}, x \in \{0, 1\}$)

Recall $E(g(x)) = \sum_x g(x) p_X(x)$

First calculate $E(X^2)$ (We have $E(X) = p$)

$$E(X^2) = \sum_x x^2 p_X(x) = 0^2 p^0 (1-p)^{1-0} + 1^2 p^1 (1-p)^{1-1} = p$$

$$V(X) = E(X^2) - (E(X))^2 = p - p^2 = p(1-p)$$

Covariance

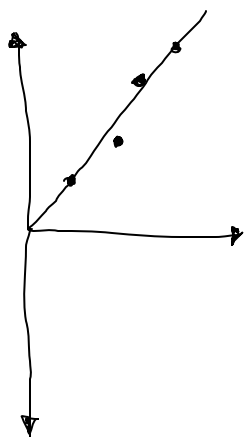
Sometimes, you may want to measure how random variables vary together. Consider the **covariance**.

$$\text{Cov}(X, Y) = E((X - \mu_X)(Y - \mu_Y)) = E(XY) - \mu_X \mu_Y$$

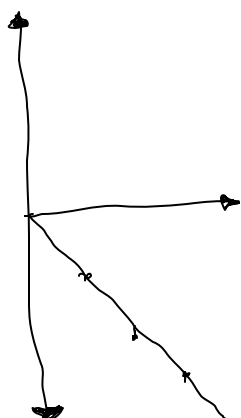
And the **correlation**.

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \quad (\text{note } \text{Corr}(X, Y) \in [-1, 1])$$

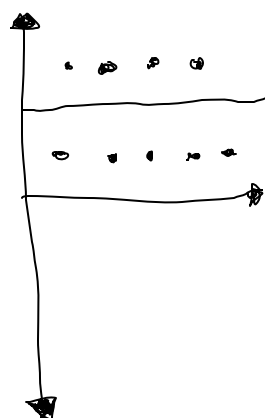
Intuition



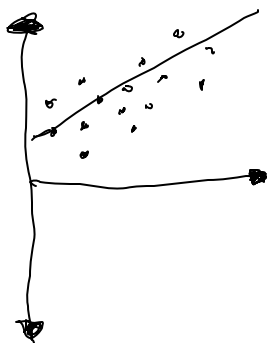
$$\begin{aligned}\text{Corr}(X, Y) &\approx 1 \\ \text{Cov}(X, Y) &> 0\end{aligned}$$



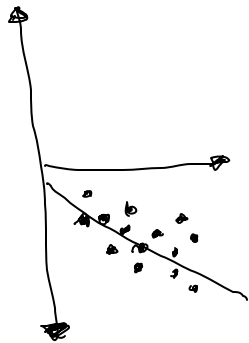
$$\begin{aligned}\text{Corr}(X, Y) &\approx -1 \\ \text{Cov}(X, Y) &< 0\end{aligned}$$



$$\begin{aligned}\text{Corr}(X, Y) &\approx 0 \\ \text{Cov}(X, Y) &\approx 0\end{aligned}$$



$$\begin{aligned}\text{Cov}(X, Y) &> 0 \\ \text{Corr}(X, Y) &> 0\end{aligned}$$



$$\begin{aligned}\text{Corr}(X, Y) &< 0 \\ \text{Cov}(X, Y) &< 0\end{aligned}$$

Covariance Properties

- For RV X

$$\text{Cov}(X, X) = V(X)$$

- For RVs X, Y

$$V(X+Y) = V(X) + V(Y) + 2\text{Cov}(X, Y)$$

- For RVs X, Y , where X and Y are independent,

$$\text{Cov}(X, Y) = 0$$

(However, you cannot assume that if $\text{Cov}(X, Y) = 0$, then X and Y are independent.)

Questions

2. a) X : money won from contest

$$X = 16000$$

$$E(X) = 16000 \quad E(X^2) = E(16000^2) = 16000^2$$

$$\begin{aligned} V(X) &= E(X^2) - (E(X))^2 \\ &= 16000^2 - 16000^2 \\ &= 0 \end{aligned}$$

2.b) Lifeline applied on first question

$$X \in \{1000, 32000, 64000\}$$

$$p_x(x) = \begin{cases} \frac{1}{2} & , x=1000 \\ \frac{1}{2} \cdot \frac{3}{4} & , x=32000 \\ \frac{1}{2} \cdot \frac{1}{4} & , x=64000 \end{cases}$$

$$\begin{aligned} E(X) &= 1000 \cdot \frac{1}{2} + 32000 \cdot \frac{1}{2} \cdot \frac{3}{4} + 64000 \cdot \frac{1}{2} \cdot \frac{1}{4} \\ &= 20500 \end{aligned}$$

$$\begin{aligned} E(X^2) &= 1000^2 \cdot \frac{1}{2} + 32000^2 \cdot \frac{1}{2} \cdot \frac{3}{4} + 64000^2 \cdot \frac{1}{2} \cdot \frac{1}{4} \\ &= 896500000 \end{aligned}$$

$$V(X) = E(X^2) - (E(X))^2 = 896500000 - 20500^2$$

2.c) $X = \{1000, 32000, 64000\}$

$$p_x(x) = \begin{cases} \frac{3}{4} & , x=1000 \\ \frac{1}{4} \cdot \frac{1}{2} & , x=32000 \\ \frac{1}{4} \cdot \frac{1}{2} & , x=64000 \end{cases}$$

$$E(X) = 1000 \cdot \frac{3}{4} + 32000 \cdot \frac{1}{8} + 64000 \cdot \frac{1}{8}$$

$$E(X^2) = 1000^2 \cdot \frac{3}{4} + 32000^2 \cdot \frac{1}{8} + 64000^2 \cdot \frac{1}{8}$$

$$V(X) = E(X^2) - (E(X))^2$$

3. Fair coin - keep flipping until head
 Suppose you win 2^x , x is number of flips before head

a)

$$E(X) = \frac{1}{\frac{1}{2}} = 2$$

$$X \sim \text{Geometric}(\frac{1}{2})$$

b)

$$\text{Expected Payoff} = \sum_{i=1}^{\infty} P(X=i) 2^i$$

$$= \sum_{i=1}^{\infty} \left(\frac{1}{2}\right)^i 2^i$$

$$= \sum_{i=1}^{\infty} \left(\frac{2}{2}\right)^i = \sum_{i=1}^{\infty} 1 = \infty$$

c)

$$E_P = \sum_{i=1}^M P(X=i)2^i + \sum_{i=M+1}^{\infty} P(X=i)2^M$$

$$= \sum_{i=1}^M \left(\frac{1}{2}\right)^i 2^i + 2^M \sum_{i=M+1}^{\infty} \left(\frac{1}{2}\right)^i$$

$$= \sum_{i=1}^M 1 + 2^M \frac{1}{2} \sum_{i=1}^{\infty} \left(\frac{1}{2}\right)^i$$

$$= M + 1$$