Introduction to Probability Theory 7th week: Multivariate random variables. The variance and covariance of random variables.

DEFINITION. The expectation (or the mean) of a random variable X with atoms $\{x_1, x_2, \ldots\}$ is defined as

$$\mathbb{E}X = \sum_{i} x_i \mathbb{P}\left(X = x_i\right)$$

Remark: If $X : \Omega \to \mathbb{R}$ is a random variable and $g : \mathbb{R} \to \mathbb{R}$ is a (measurable) function, then $g(X) : \Omega \to \mathbb{R}$ is a new random variable.

Theorem (Law of the Unconscious Statistician). If X is a discrete random variable and $g : \mathbb{R} \to \mathbb{R}$ is a (measurable) function, then

$$\mathbb{E}(g(X)) = \sum_{x} g(x)\mathbb{P}(X = x).$$

DEFINITION. The variance of a random variable X is defined as

 $\operatorname{Var} X = \mathbb{E}((X - \mathbb{E}X)^2) \quad (a \text{ useful formula: } \operatorname{Var} X = \mathbb{E}(X^2) - (\mathbb{E}X)^2).$

Remark: For some random variables X the expectation does not exists and then the variance of such X is not defined. It might also happen that $\mathbb{E}X$ exists but the variance Var X does not, since the sum on the right-hand side of the formula does not (absolutely) converge.

DEFINITION. By a k-dimensional random variable we mean a (measurable) function

$$\bar{X} = (X_1, \ldots, X_k) : \Omega \to \mathbb{R}^k,$$

where Ω is the sample space of a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Remark: On this course we will be studying mainly two dimensional random variables.

DEFINITION. A 2-dimensional random variable (X, Y) is discrete, if for some countable set $A = \{(x_1, y_1), (x_2, y_2), \ldots\}$ we have

$$\sum_{x} \sum_{y} \mathbb{P}\left(X = x, Y = y\right) = 1.$$

For a discrete random variable (X, Y) the probabilities $p_{xy} = \mathbb{P}(X = x, Y = y)$ determine the (joint) probability mass function (pmf) which describes the (joint) distribution of (X, Y).

DEFINITION. The (joint) cumulative distribution function (CDF) of a 2-dimensional random variable (X, Y) is defined as the function $F_{X,Y} : \mathbb{R} \to \mathbb{R}$ such that

$$F_{X,Y}(x,y) = \mathbb{P}\left(X \leqslant x, Y \leqslant y\right).$$

Theorem (Law of the Unconscious Statistician for two dimensions). If (X, Y) is a 2-dimensional discrete random variable and $g : \mathbb{R}^2 \to \mathbb{R}$ is a (measurable) function, then

$$\mathbb{E}(g(X,Y)) = \sum_{x} \sum_{y} g(x,y) \mathbb{P} \left(X = x, Y = y \right).$$

DEFINITION. The cumulative distribution function (CDF) of a 2-dimensional random variable (X, Y) is a function $F_{X,Y} : \mathbb{R} \to \mathbb{R}$ defined as

$$F_{X,Y}(x,y) = \mathbb{P}\left(X \leqslant x, Y \leqslant y\right).$$

DEFINITION. The covariance Cov(X, Y) of random variables X and Y is defined as

$$\operatorname{Cov} X = \mathbb{E}[(X - \mathbb{E}X)(Y - \mathbb{E}Y)] = \mathbb{E}(XY) - (\mathbb{E}X)(\mathbb{E}Y),$$

while the correlation coefficient $\rho(X,Y)$ is given by

$$\rho(X,Y) = \frac{Cov(X,Y)}{\sqrt{\operatorname{Var} X \operatorname{Var} Y}}$$

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Remark: For some random variables X and Y the covariance Cov(X, Y) and the correlation coefficient $\rho(X, Y)$ do not exist. However, if $\rho(X, Y)$ exists, then $|\rho(X, Y)| \leq 1$.

DEFINITION. Let (X, Y) be a 2-dimensional random variable. The marginal distribution of X is given by

$$\mathbb{P}\left(X=x_0\right)=\sum_{y}\mathbb{P}\left(X=x_0,Y=y\right)$$

while for marginal distribution of Y we have

$$\mathbb{P}(Y = y_0) = \sum_{x} \mathbb{P}(X = x, Y = y_0)$$

If for every pair (x_0, y_0) we have

$$\mathbb{P}(X = x_0, Y = y_0) = \mathbb{P}(X = x_0) \mathbb{P}(Y = y_0) ,$$

then the random variables X and Y are independent.

Theorem. If X and Y are independent, then

$$\mathbb{E}XY = \mathbb{E}X\mathbb{E}Y.$$

In particular, for independent random variables Y and Y we have Cov(X,Y) = 0 and $\rho(X,Y) = 0$ (provided both Cov(X,Y) and $\rho(X,Y)$ exist).

Remark: It is generally not true that if $\mathbb{E}XY = \mathbb{E}X\mathbb{E}Y$, then X and Y are independent.

Theorem. For any real numbers $a, b, c \in \mathbb{R}$ and any functions $f, g : \mathbb{R}^2 \to \mathbb{R}$ we have

$$\mathbb{E}(af(X,Y) + bg(X,Y) + c) = a\mathbb{E}f(X,Y) + b\mathbb{E}g(X,Y) + c.$$

Moreover,

$$\operatorname{Var}(aX+b) = a^2 \operatorname{Var} X$$

If X and Y are independent, then we have also

$$\operatorname{Var}(X+Y) = \operatorname{Var} X + \operatorname{Var} Y.$$

PROBLEMS WE SHALL DO IN CLASS ON APRIL 18TH

Problem 1. The pmf of X is given by

k	0	1	2	3
$\mathbb{P}\left(X=k\right)$	0.4	0.3	0.2	0.1

- a) Find $\mathbb{E}X$ and $\operatorname{Var}X$.
- b) Find the pmf of Y = |2 X| and use it to compute $\mathbb{E}Y$.
- c) Find $\mathbb{E}|2 X|$ using the Law of the Unconscious Statistician.

Problem 2. The pmf of X is the following

Find the pmf of $Y = \operatorname{sgn}(X)$ and calculate $\mathbb{E}Y$ i Var Y.

Problem 3. The joint pmf of (X, Y) is given by

$$\mathbb{P}(X = 0, Y = -1) = 1/2, \ \mathbb{P}(X = 1, Y = 1) = \mathbb{P}(X = -1, Y = 1) = 1/4.$$

Find both marginal distributions and compute $\rho(X, Y)$. Are X and Y independent?

Problem 4. We toss a die n times. Let X_n denote the number of ones, and let Y_n stand for the number of twos. Are these two random variables independent?

Problem 5. The first urn contains 5 white balls and 3 black balls, in the second one we have 2 white balls and 2 black balls. We choose one ball from the first urn and one ball from the second. Let X denote the number of white balls we selected and Y the number of black balls we chose.

- (i) Find the joint pmf of (X, Y).
- (ii) Are X i Y independent?
- (iii) Find $\mathbb{E}X$ i $\mathbb{E}Y$ using the marginal distributions of X and Y.
- (iv) Find $\mathbb{E}X$ representing X as $X = X_1 + X_2$, where X_i is the number of white balls selected from the *i*th urn.
- (v) What is the pmf of Z = X + Y?
- (vi) Find $\rho(X, Y)$. Can one do it in a shorter way, not using the joint pmf of (X, Y) (but using (v))?

Homework assignment for Quiz 5, April 18th

Problem 1. The pmf of X is given by

$$\mathbb{P}(X=8) = \frac{1}{8}$$
 and $\mathbb{P}(X=i) = \frac{1}{2^i}$, for $i = 1, 2, 3$.

Find the pmfs of $Y = (-2)^X$ and Z = 4X + 1. Compute $\mathbb{E}Y$ and $\mathbb{E}Z$ in two ways: directly from their pmfs and using the Law of the Unconscious Statistician.

Problem 2. The pmf of X is given by

$$\frac{k}{\mathbb{P}(X=k)} \frac{-1}{\frac{2}{10}} \frac{0}{\frac{5}{10}} \frac{3}{\frac{2}{20}} \frac{3}{\frac{2}{20}}$$

- (a) Find the pmf of Y = |2 X| and compute $\mathbb{E}Y$ and $\operatorname{Var}Y$.
- (b) Find the pmf of $Z = \sin(\pi X/2)$, and compute $\mathbb{E}Z$ and Var Z.

Problem 3. There are 37 numbers on the French roulette wheel: 18 of them are red, 18 are black, and one is green. Find the expectation and the variance of a win if the player:

- (a) bets on token on red (then, if there will be red, he will win another token, if not, he will lose the token he bet);
- (b) bets one token on 28 (then, in the case of 28, he will get another 35 tokens, otherwise he loses the token he bet).

Problem 4. We toss a die once. The first player wins 2, if the result is even, and loses 1 (i.e. wins -1) if it is odd. The second player wins 2 if the result is at least 4, and loses 1 in all other cases. Let X_i denote the win of the *i*th player.

- (a) Find the joint pmf of (X_1, X_2) .
- (b) Find the marginal ditributions of (X_1, X_2) and check if X_1 i X_2 are independent.
- (c) Compute $\rho(X_1, X_2)$.
- (d) Find $\mathbb{E}X_1^2 X_2^3$.

Problem 5. We select three balls from an urn which contains 2 white and 3 black balls. The first player wins 6 if all three balls are black and loses 1 in all other cases. The second player wins 1 for every black ball (white balls does not affect his win). By X we denote the win of the first player, and Y the gain of the second player.

- (a) Find the joint pmf of (X, Y).
- (b) Find the marginal ditributions of (X, Y) and check if X and Y are independent.
- (c) Find $\rho(X, Y)$.