

Week 7
Statistics 251

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Where are we?

Sums of independent r.v.

Expectation of Sums

Covariance

Correlation

Sum of two independent r.v.

Given X, Y independent, continuous r.v. with density function f_X, f_Y , the density function of $X + Y$ can be written as,

$$f_{X+Y}(a) = \int_{-\infty}^{\infty} f_X(a-y)f_Y(y)dy = \int_{-\infty}^{\infty} f_X(x)f_Y(a-x)dx$$

Sums of uniform distributions

If X_1, X_2 are independent identical uniform distributed on $(0, 1)$,
what is the distribution of $X_1 + X_2$?

Continued..

If X_1, X_2, \dots , are independent identical uniform distributed on $(0, 1)$. What is the expectation of N where

$$N = \min\{n : X_1 + X_2 + \dots + X_n > 1\}$$

Let F_n denote cumulative distribution function of $X_1 + \dots + X_n$.
By Mathematical Induction, we first try to prove

$$F_n(x) = x^n/n!, \quad 0 \leq x \leq 1.$$

$$\text{So } \mathbb{P}\{N > n\} = F_n(1)$$

Sums of Normal distribution

Recall density function of a normal distribution with parameters (μ, σ^2) is given by

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right)$$

Proposition: If X_1, X_2, \dots, X_n are independent random variables with respective parameters $(\mu_1, \sigma_1^2), (\mu_2, \sigma_2^2), \dots, (\mu_n, \sigma_n^2)$, then $X_1 + X_2 + \dots + X_n$ is a normal random variable with parameters $(\mu_1 + \mu_2 + \dots + \mu_n, \sigma_1^2 + \sigma_2^2 + \dots + \sigma_n^2)$.

Sums of Gamma distribution

Recall density function of a gamma distribution with parameters (α, λ) is given by

$$f(x) = \begin{cases} \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

Proposition: If X and Y are independent gamma random variables with respective parameters (s, λ) and (t, λ) , then $X + Y$ is a gamma random variable with parameters $(s + t, \lambda)$.

- So α is referred as shape parameter, λ is referred as scale parameter.

You are also advised to read the relevant part (Poisson, binomial, geometric) on Ross book.

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Revision: Expected Value of a function of a r.v.

If X and Y have a joint probability mass function $p(x, y)$, then

$$E[g(X, Y)] = \sum_y \sum_x g(x, y)p(x, y)$$

If X and Y have a joint probability density function $f(x, y)$, then

$$E[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y)f(x, y)dx dy$$

Derivation

Suppose that $E[X]$ and $E[Y]$ are both finite and let $g(X, Y) = X + Y$. Then, in the continuous case,

$$\begin{aligned} E[X + Y] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x + y)f(x, y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xf(x, y) dy dx + \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} yf(x, y) dx dy \\ &= \int_{-\infty}^{\infty} xf_X(x) dx + \int_{-\infty}^{\infty} yf_Y(y) dy \\ &= E[X] + E[Y] \end{aligned}$$

What about $E[X_1 + \dots + X_n]$?

Sample Mean

Let X_1, \dots, X_n be independent and identically distributed random variables having distribution function F and expected value μ .

Then X_1, \dots, X_n is said to constitute a *sample from the distribution F* .

The quantity

$$\bar{X} = \sum_{i=1}^n \frac{X_i}{n}$$

is called the sample mean.

Now what is $E[\bar{X}]$?

Note: when the distribution mean μ is unknown, the sample mean is often used in statistics to estimate it.

Example: A Summation Formula

Consider any nonnegative, integer-valued random variable X . If, for each $i \geq 1$, we define

$$X_i = \begin{cases} 1 & \text{if } X \geq i \\ 0 & \text{if } X < i \end{cases}$$

then $E[X] = \sum_{i=1}^{\infty} E(X_i)$

Note $E(X_i) = P\{X \geq i\}$, so

$$E[X] = \sum_{i=1}^{\infty} P\{X \geq i\}$$

a useful identity.

Example: Sorting elements

Suppose that n elements, $1, 2, \dots, n$ must be stored in a computer in the form of an ordered list. Each unit of time, a request will be made for one of these elements i being requested, independently of the past, with known probability $P(i), i \geq 1, \sum_i P(i) = 1$.

What ordering minimizes the average position in the line of the element requested?

Suppose that the elements are numbered so that $P(1) \geq P(2) \geq \dots \geq P(n)$. Let X denote the position of the requested element. Now, under any ordering, $O = i_1, i_2, \dots, i_n$,

$$P_O\{X \geq k\} = \sum_{j=k}^n P(i_j)$$

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Expected value of multiplication of functions of independent r.v.

If X and Y are independent, then, for any functions h and g ,

$$E[g(X)h(Y)] = E[g(X)]E[h(Y)]$$

Continuous case proof:

Definition of Covariance

Recall definition of Var X :

The covariance between X and Y , denoted by $\text{Cov}(X, Y)$, is defined by

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

Alternative form

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

\Updownarrow

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

Properties

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

$$(ii) \text{Cov}(X, X) = \text{Var}(X)$$

$$(iii) \text{Cov}(aX, Y) = a \text{Cov}(X, Y)$$

$$(iv) \text{Cov}\left(\sum_{i=1}^n X_i, \sum_{j=1}^m Y_j\right) = \sum_{i=1}^n \sum_{j=1}^m \text{Cov}(X_i, Y_j)$$

Variance of sum

$$\text{Var} \left(\sum_{i=1}^n X_i \right) = \sum_{i=1}^n \text{Var} (X_i) + 2 \sum_{i < j} \text{Cov} (X_i, X_j)$$

Example: Sample Variance

Let X_1, \dots, X_n be independent and identically distributed random variables having expected value μ and variance σ^2 . Let

$\bar{X} = \sum_{i=1}^n X_i/n$ be the *sample mean*. Then what is $\text{Var}(\bar{X})$?

The quantities $X_i - \bar{X}$, $i = 1, \dots, n$, are called *deviations*, as they equal the differences between the individual data and the sample mean. The random variable

$$S^2 = \sum_{i=1}^n \frac{(X_i - \bar{X})^2}{n-1}$$

is called the *sample variance*. Find $\mathbb{E}[S^2]$.

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Definition

The correlation of two random variables X and Y , denoted by $\rho(X, Y)$, is defined, as long as $\text{Var}(X) \text{Var}(Y)$ is positive, by

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

Then $-1 \leq \rho(X, Y) \leq 1$.

Example: Deviation and sample mean are uncorrelated

Let X_1, \dots, X_n be independent and identically distributed random variables having variance σ^2 . Then

$$\text{Cov}(X_i - \bar{X}, \bar{X}) = 0$$

Example: if $Y = a + bX$

Given mean and variance of X to be μ and σ^2 , calculate $\rho(X, Y)$, where $Y = a + bX$.