
STOR 435.001 Lecture 14

Jointly distributed Random Variables - II

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Discrete case: Independence is equivalent to

$$p(x, y) = p_X(x)p_Y(y), \quad \text{all } x, y,$$

where $p(x, y)$ is the joint p.m.f. of X and Y , p_X is the p.m.f. of X , p_Y is the p.m.f. of Y .

Jointly continuous case: Independence is equivalent to

$$f(x, y) = f_X(x)f_Y(y), \quad \text{all } x, y,$$

where $f(x, y)$ is the joint p.d.f. of X and Y , f_X is the p.d.f. of X , f_Y is the p.d.f. of Y .

Problem from Final Exam 2011

Suppose that 2 balls are chosen without replacement from an urn containing 5 white and 8 red balls. Let X_1 equal 1 if the 1st ball selected is red, and let it equal 0 otherwise. Let X_2 equal 1 if the 2nd ball selected is red, and let it equal 0 otherwise.

1. Give the joint probability mass function of X_1 and X_2 .
2. Are X_1 and X_2 independent? (Provide a mathematical argument.)

Jointly distributed random variables

Example: Suppose that the number of people who enter Starbucks on a given day is a Poisson random variable with parameter λ . Each person who enters the post office orders a single Latte with probability p and orders something else with probability $1 - p$. Let L and NL be the total number of single order Latte and Non single order Lattes in a day. Show that these are independent Poisson random variables with respective parameters λp and $\lambda(1 - p)$.



Jointly distributed random variables

Problem: Two points are selected randomly on a line of length L so as to be on opposite sides of the midpoint of the line. (In other words, the two points X and Y are independent random variables such that X is uniformly distributed over $(0, L/2)$ and Y is uniformly distributed over $(L/2, L)$.) Find the probability that the distance between the two points is greater than $L/3$.



Problem continued

Jointly distributed random variables

Example: Let $X = U(0, 1)$ and $Y = Exp(1)$ be independent. Find cdf and pdf of $Z = X + Y$.



Problem continued

Jointly distributed random variables

Independence: Random variables X_1, X_2, \dots, X_n are *independent* if, for any sets A_1, A_2, \dots, A_n ,

$$\begin{aligned} &P(X_1 \in A_1, X_2 \in A_2, \dots, X_n \in A_n) \\ &= P(X_1 \in A_1)P(X_2 \in A_2) \dots P(X_n \in A_n) \end{aligned}$$

(that is, the events $\{X_1 \in A_1\}, \{X_2 \in A_2\}, \dots, \{X_n \in A_n\}$ are independent). For example, in the jointly continuous case, this is equivalent to

$$f(x_1, x_2, \dots, x_n) = f_{X_1}(x_1)f_{X_2}(x_2) \dots f_{X_n}(x_n), \quad \text{all } x_1, x_2, \dots, x_n,$$

where $f(x_1, x_2, \dots, x_n)$ is the joint p.d.f. and $f_{X_i}(x_i)$ is a marginal p.d.f. of X_i .

An important operation with combining independent random variables

- ▶ Now suppose we had a bunch of independent random variables say X_1, X_2, \dots, X_n .
- ▶ One important way to combine these is *adding* them i.e. making a new random variable $S_n = X_1 + X_2 + \dots + X_n$.
- ▶ Why important?
- ▶ **Setting one:** We take a random sample of 100 UNC voters. Let X_i be a 1 if i -th individual in sample is voting for Hillary and $X_i = 0$ otherwise. So now we have X_1, X_2, \dots, X_{100} . The total number of people in your sample who are voting for Hillary is?
- ▶ **Setting two:** Researchers are comparing study habits in two universities (call these U and D) and in particular want to study the average amount of time juniors spend studying per day. They take a sample of 40 students from U and 30 from D and see how much time they spend studying in a week. Let X_1, X_2, \dots, X_{40} denote the respective amount of time for the students at U and Y_1, Y_2, \dots, Y_{30} denote the same but for students at D. What would they base the conclusions of their study on?
- ▶ **Punchline:** sums of random variables super important! I will show you two methods for calculating the distribution of such objects.
- ▶ **Method 1: Direct method**
- ▶ **Method 2: Moment generating function**



Sums of independent random variables: E.g. jointly continuous case: if X and Y are independent, X has density f_X and Y has density f_Y , then the density of $X + Y$ is:

$$f_{X+Y}(a) =$$



Name: *convolution* of f_X and f_Y .

Example 3a: If X and Y are independent random variables, both uniformly distributed on $(0, 1)$, calculate the probability density of $X + Y$.



R code to simulate this

```
x<- runif(10^5)
y<- runif(10^5)
z<- x+y
hist(z)
```

Note: For independent discrete random variables we could carry out similar calculations but **not** integrating but “summing”:

Example: If $X = Pois(\lambda_1)$ and $Y = Pois(\lambda_2)$ are independent, then

$$P(X + Y = n) =$$



Method II for finding distributions of sums of independent random variables

- ▶ The previous slides we directly calculated the distributions of sums. Turns out there is another powerful tool to do the same. The tool uses two ingredients.
- ▶ **Ingredient 1: Moment generating functions** $M_X(t) := E(e^{tX})$.
- ▶ **Amazing math fact:** Moment generating functions uniquely characterize the distribution. So suppose we do not know the distribution of a random variable but have somehow managed to compute the MGF and recognize it as the MGF of a known distribution, this means that the random variable has that distribution.

We have computed earlier:

MGF of some standard distributions

1. $X = \text{Bin}(n, p)$

$$M_X(t) = (1 - p + pe^t)^n$$

2. $X = \text{Pois}(\lambda)$

$$M_X(t) = e^{\lambda(e^t - 1)}.$$

3. $X = N(\mu, \sigma^2)$

$$M_X(t) = e^{t\mu + t^2\sigma^2/2}$$

4. $X = \text{Gamma}$ with parameters (α, λ) then $M_X(t)$ is finite only when $t < \lambda$ and then

$$M_X(t) = \left(\frac{\lambda}{\lambda - t}\right)^\alpha$$

Ingredient 2

Suppose X and Y are **independent** random variables with marginal pdf f_X and f_Y . Let $g(\cdot), h(\cdot)$ be functions. Show that

$$\mathbb{E}(g(X)h(Y)) = \mathbb{E}(g(X))\mathbb{E}(h(Y))$$

The same fact holds for discrete independent random variables and not just 2 but any number n independent random variables.



Jointly distributed random variables

Combining the two ingredients

Suppose X_1, X_2, \dots, X_n are independent random variables with the mgf of X_i given by M_{X_i} . Let S be the sum of these random variables namely $S = X_1 + X_2 + \dots + X_n$. Then MGF of S

$$\begin{aligned}M_S(t) &= E(e^{tS}) = E(e^{t(X_1+X_2+\dots+X_n)}) = E(e^{tX_1+tX_2+\dots+tX_n}) \\ &= E(e^{tX_1}e^{tX_2}\dots e^{tX_n}) \\ &= E(e^{tX_1})E(e^{tX_2})\dots E(e^{tX_n}) \quad \text{by independence} \\ &= M_{X_1}(t)M_{X_2}(t)\dots M_{X_n}(t).\end{aligned}$$

Punchline: MGF of the **sum** of a bunch of **independent** random variables is the **product** of the individual MGFs.

Method II of computing distribution of sums of independent rvs

1. Compute the individual MGfs $M_{X_i}(t)$.
2. Compute the MGF of the sum S using the above formula.
3. See if M_S matches the MGF of one of the known distributions. If it does then S has that distribution.

Jointly distributed random variables



Let $0 < p < 1$. If $X_i, i = 1, 2, \dots, m$ are independent with $X_i = \text{Bin}(n_i, p)$ what is the distribution of $\sum_1^m X_i$? **Note: Same p for all the rvs.**





Let $\lambda_1, \lambda_2, \dots, \lambda_m > 0$. If $X_i, i = 1, 2, \dots, m$ are independent Poisson λ_i what is the distribution of $\sum_{i=1}^n X_i$?





If $X_i, i = 1, 2, \dots, n$ are independent and identically distributed exponential random variables with parameter λ , what is the distribution of $\sum_{j=1}^n X_i$?





If X_i are independent normal random variables with parameters μ_i, σ_i^2 , what is the distribution of $\sum_{i=1}^n X_i$?



Jointly distributed random variables

Problem: Motivated by

https://www.tkiryl.com/Elementarystatistics/Chapter_9.pdf.

The following is taken almost verbatim from the above site. One of the larger species of **tarantulas** is the *Grammostola mollicoma*, whose common name is the Brazilian giant tawny red. A tarantula has two body parts. The anterior part of the body is covered above by a shell, or carapace. From a recent article by F. Costa and F. Perez-Miles titled Reproductive Biology of Uruguayan Theraphosids (The Journal of Arachnology, Vol. 30, No. 3, pp. 571-587):

*The carapace length of the adult male *G. mollicoma* is normally distributed with mean 18.14 mm and standard deviation 1.76 mm.*

Suppose you find three such males and measure them. Assume the lengths are independent random variables. Find the probability that the **average** length of the carpace is greater than 19 mm.



Solution continued.

Conclusion

- ▶ We thought about operations on independent random variables.
- ▶ Focussed on sums of independent random variables.
- ▶ Found that in a number of cases the sum of such random variables can be explicitly evaluated.
- ▶ Two methods. Method 1: Direct calculation of the pmf or pdf of the sum.
- ▶ Method 2: Using moment generating functions.