

* Announcement

① Quiz 3: Thu (9/24) 9:00 AM PT
~ Fri (9/25) 9:00 AM PT

Ch4

Waiting time until r th success
 $P(T_r = k), P(T_r \leq k), P(T_r > k)$
Exp. Approximation
No success in n trials, At least one success
Poisson dist'n.
" "
Binomial (n, p) for n large, p small
Poisson formula, independence

STAT 88: Lecture 12

After today's lecture

→ HW 5 Q1(b)(c)(d), Q4, Q3
try

Contents

Section 5.2: Functions of Random Variables

Section 5.3: Method of Indicators

Last time

Sec 5.1 The expectation of a random variable X , denoted $E(X)$, is the average of the possible values of X weighted by their probabilities:

$$E(X) = \sum_{\text{all } x} xP(X = x).$$

(Bernoulli RV)

$$X = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p \end{cases}$$

- Success
- failure.

Then

$$E(X) = 1 \cdot p + 0 \cdot (1 - p) = p.$$

(Uniform distribution) X has a uniform distribution on $\{1, 2, \dots, n\}$ if

$$P(X = k) = \frac{1}{n} \text{ for } k = 1, \dots, n.$$

Then

$$E(X) = \sum_{k=1}^n k \cdot \frac{1}{n} = \frac{n+1}{2}.$$

(Poisson distribution) X has a Poisson distribution if

$$P(X = k) = \frac{e^{-\mu} \mu^k}{k!} \text{ for } k = 0, 1, \dots$$

Then

$$E(X) = \sum_{k=0}^{\infty} k \cdot \frac{e^{-\mu} \mu^k}{k!} = \mu.$$

↖ average of counts

Sec 5.2 Let $Y = g(X)$ be a function of the random variable X . Then

$$E(Y) = E(g(X)) = \sum_{\text{all } x} g(x)P(X = x).$$

Warm up: (Exercise 5.7.1) Let X have the distribution displayed in the table below.

x	-2	-1	0	1
$P(X = x)$	0.2	0.25	0.35	0.2
$g(x) = x-1 $	3	2	1	0
$g(x) = (x-1)^2$	9	4	1	0

(c) Find $E|X - 1|$.

(d) Find $E((X - 1)^2)$.

$$\begin{aligned} \text{(c)} \quad E|X-1| &= 3 * (0.2) + 2 * (0.25) + 1 * (0.35) + 0 * (0.2) \\ &= 1.45 \end{aligned}$$

$$\begin{aligned} \text{(d)} \quad E((X-1)^2) &= 9 * (0.2) + 4 * (0.25) + 1 * (0.35) + 0 * (0.2) \\ &= 3.15 \end{aligned}$$

5.2. Functions of Random Variables (Continued)

Recall that a random variable is a function on the outcome space Ω , $X : \Omega \rightarrow \mathbb{R}$.

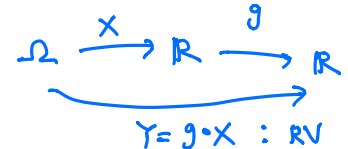
Example: Flip a fair coin twice. $\Omega = \{HH, HT, TH, TT\}$. Let $X = \#$ heads in 2 coin tosses. Then $X(HH) = 2$.

$$\begin{aligned} X(HT) &= 1 = X(TH) \\ Y(HT) &= 1^2 = Y(TH) \end{aligned}$$

Now if $Y = g(X) = X^2$, this is also a function of the outcome space

$$Y(HH) = (X(HH))^2 = 4.$$

So a function of a random variable is itself a random variable.



Joint Distribution Suppose two draws are made at random without replacement from a population that has five elements labeled 1, 2, 2, 3, 3. Define the following random variables:

- X_1 is the number on the first draw 1, 2, 3
- X_2 is the number on the second draw 1, 2, 3

(vector)

The pair (X_1, X_2) is a random variable and we describe its distribution in a table

$$P(X_1 = 1, X_2 = 1) = P(X_1 = 1)P(X_2 = 1|X_1 = 1) = \frac{1}{5} \cdot 0 = 0.$$

$$P(X_1 = 1, X_2 = 2)?$$

$$\begin{aligned} &\text{"and" } \uparrow \text{ Multiplication rule} \\ &P(X_1=1)P(X_2=2|X_1=1) = \frac{1}{5} * \frac{2}{4} = \frac{2}{20} = \frac{1}{10} \end{aligned}$$

This fills out the table:

	$X_2 = 1$	$X_2 = 2$	$X_2 = 3$
$X_1 = 1$	0	$\frac{2}{20}$	$\frac{2}{20}$
$X_1 = 2$	$\frac{2}{20}$	$\frac{2}{20}$	$\frac{4}{20}$
$X_1 = 3$	$\frac{2}{20}$	$\frac{4}{20}$	$\frac{2}{20}$

Sum to 1

$$\text{Find } P(X_1 + X_2 = 4)? \quad P(X_1=1, X_2=3) + P(X_1=2, X_2=2) + P(X_1=3, X_2=1)$$

$$= \frac{2}{20} + \frac{2}{20} + \frac{2}{20} = \frac{6}{20}$$

$$\begin{aligned} P(X_1 = X_2) &= P(X_1=1, X_2=1) + P(X_1=2, X_2=2) + P(X_1=3, X_2=3) \\ &= \frac{4}{20} \end{aligned}$$

Marginal Distribution Note that

$$P(X_1 = 1) = P(X_1 = 1, X_2 = 1) + P(X_1 = 1, X_2 = 2) + P(X_1 = 1, X_2 = 3).$$

↑ Addition rule

So we can recover the distribution of X_1 from the joint distribution.

	$X_2 = 1$	$X_2 = 2$	$X_2 = 3$	Dist of X_1
$X_1 = 1$	0	$\frac{2}{20}$	$\frac{2}{20}$	$\frac{4}{20}$ $= \frac{1}{5}$
$X_1 = 2$	$\frac{2}{20}$	$\frac{2}{20}$	$\frac{4}{20}$	$\frac{8}{20}$ $= \frac{2}{5}$
$X_1 = 3$	$\frac{2}{20}$	$\frac{4}{20}$	$\frac{2}{20}$	$\frac{8}{20}$ $= \frac{2}{5}$
Dist of X_2	$\frac{4}{20}$	$\frac{8}{20}$	$\frac{8}{20}$	

You can get the distribution of X_2 by summing along columns.

Are X_1 and X_2 independent? $P(X_1=2, X_2=2) \neq P(X_1=2)P(X_2=2) \rightarrow \underline{\underline{No}}$
 $\frac{2}{20} \neq \frac{2}{5} \cdot \frac{2}{5} = \frac{4}{25}$

X_1 and X_2 are independent if

$$P(X_1 = a, X_2 = b) = P(X_1 = a)P(X_2 = b)$$

for all cells in the joint table.

Expectations of Functions We can find the expectation of any function g of two random variables X and Y by extending our method for finding the expectation of a function of X .

- Take a cell of the joint distribution table of X and Y . This corresponds to one possible value (x, y) of the pair (X, Y) .
- Apply the function g to get $g(x, y)$.
- Weight this by the probability in the cell, to get the product $g(x, y)P(X = x, Y = y)$.
- Add these products over all the cells of the table.

Example: Find $E|X_1 - X_2|$. $g(x_1, x_2) = |x_1 - x_2|$

	$X_2 = 1$	$X_2 = 2$	$X_2 = 3$
$X_1 = 1$	$0 \overset{g(1,1)=0}{\text{green}} \overset{\text{underline}}{\text{green}}$	$\frac{2}{20} \overset{g(1,2)=1}{\text{green}} \overset{\text{underline}}{\text{green}}$	$\frac{2}{20} \overset{g(1,3)=2}{\text{green}}$
$X_1 = 2$	$\frac{2}{20} \overset{g(2,1)=1}{\text{green}}$	$\frac{2}{20} \overset{g(2,2)=0}{\text{green}}$	$\frac{4}{20} \overset{g(2,3)=1}{\text{green}}$
$X_1 = 3$	$\frac{2}{20} \overset{g(3,1)=2}{\text{green}}$	$\frac{4}{20} \overset{g(3,2)=1}{\text{green}}$	$\frac{2}{20} \overset{g(3,3)=0}{\text{green}}$

$$\begin{aligned}
 \sum_{\text{all cells}} g(x_1, x_2) P(X_1 = x_1, X_2 = x_2) &= 0 * 0 + 1 * \frac{2}{20} + 2 * \frac{2}{20} \\
 &\quad + 1 * \frac{2}{20} + \dots \\
 &= 1
 \end{aligned}$$

Example: A joint distribution for two random variables M and S is given below.

	$M = 2$	$M = 3$
$S = 2$	0	$\frac{1}{3}$
$S = 1$	$\frac{1}{3}$	$\frac{1}{3}$
Dist of M	$\frac{1}{3}$	$\frac{2}{3}$

Find $E(M)$.

Exercise

$$\begin{aligned}
 E(M) &= 2 * \left(\frac{1}{3}\right) + 3 * \left(\frac{2}{3}\right) \\
 &= \frac{8}{3}
 \end{aligned}$$

5.3. Method of Indicators

Preliminary: Additivity of Expectation No matter what the joint distribution of X and Y is, we have

$$E(X + Y) = E(X) + E(Y).$$

This additivity is one of the most important properties of expectation, because it is true whether the random variables are dependent or independent.

Furthermore, for any constants a and b , the linearity also holds:

$$E(aX + bY) = aE(X) + bE(Y). \quad a=1, b=1$$

Method of Indicators to find $E(X)$:

Key idea Counting the number of successful trials is the same as adding zeros and ones.

Example: A success is blue, and failure non blue

B R R G B R B B
1 0 0 0 1 0 1 1

$$\# \text{blue} = 1 + 0 + 0 + 0 + 1 + 0 + 1 + 1 = 4.$$

Recall (Bernoulli (indicator) random variable)

$$X = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p \end{cases}$$

Binomial X is the number of successes in n independent trials all with the same probability p of success. To find $E(X)$, write X as a sum of n indicators.





$$X = I_1 + I_2 + \cdots + I_n,$$

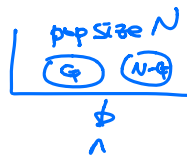
where

$$I_j = \begin{cases} 1 & \text{if } j\text{th trial is success} \\ 0 & \text{otherwise} \end{cases} \quad \text{— Prob } p$$

Then

$$E(X) = E(I_1 + I_2 + \cdots + I_n) = E(I_1) + E(I_2) + \cdots + E(I_n) = np.$$

   
Additivity



Hypergeometric X is the number of good elements in a simple random sample of size n drawn from a population N elements of which G are good. To find $E(X)$, write X as a sum of n indicators.

$$X = I_1 + I_2 + \cdots + I_n,$$

where

$$I_j = \begin{cases} 1 & \text{if } j\text{th draw yields a good element} \\ 0 & \text{otherwise} \end{cases}$$

Then

$$E(X) = E(I_1 + I_2 + \cdots + I_n) = E(I_1) + E(I_2) + \cdots + E(I_n) = n \frac{G}{N}.$$

↑
Additivity

↑
 $E(I_j) = \frac{G}{N}$
~~~~~  
by Symmetry