

Stat 134 Lec 33

Last time:

Sec 5.4 Convolution formula for density of $\frac{Y}{X}$

$$Z = \frac{Y}{X} \quad X > 0, Y > 0$$

$$f(z) = \int_{x=0}^{\infty} f(x, zx) x dx = \int_{x=0}^{\infty} f(x) f_Y(zx) x dx$$

- if x, y indep.

ex $U_1, \dots, U_n \stackrel{iid}{\sim} U(0,1)$

$Y = U_{(k)}, X = U_{(m)}$ with $k < m$



$$\frac{U_{(k)} \text{ in } n}{U_{(m)} \text{ in } n} \stackrel{D}{=} U_{(k)} \text{ in } m-1 \quad \left(\text{both are Beta}(k, m-k) \right)$$

Today

- ① Sec 5.4 Review student explanations from content test.
- ② Sec 6.1 Conditional Distribution: Discrete case.
- ③ Sec 6.2
 - a) Conditional expectation: $E(T|X=x)$
 - b) Rule of average conditional expectations,

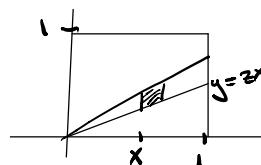
$$E(T) = E(E(T|X))$$

Let $X \sim U(0, 1)$ and $Y \sim U(0, 1)$ be independent. The density of $Z = Y/X$ for $0 < z < 1$ is:

- a $1/(2z)$
- b** $1/2$
- c $1/(2z^2)$
- d none of the above

Discuss with your neighbor for 1 minute how you did this.

b Integrate from 0 to 1 because the slope is between 0 to 1 so it goes to the end of where x can be in the joint



b I used the previous formula and integrated $(fx=1)*(fy=1)*x$ from 0 to 1; intuitively, however, I have no idea what's going on

$$f_z(z) = \int_{x=0}^{x=1} f_x(x) f_y(zx) x dx = \frac{1}{2} \int_0^1 x dx = \frac{1}{2}$$

d y/x is uniform over 0 to 1

Problem is that $P(0 < z < 1) = 1/2$
Since $0 < z < 1$ iff $x > y$

TRUE

$$0 < z < 1 \Leftrightarrow x > y$$

$$P(0 < z < 1) = 1/2$$

$$\text{For } 0 < z < 1, x > y$$



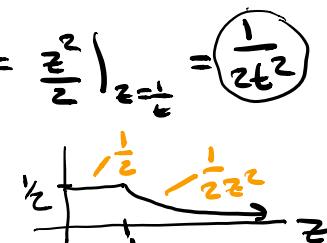
$$\frac{1}{2} = \frac{U(1) - U(0)}{U(1) + U(0)} = \frac{1}{2}$$

$Z \sim U(0,1)$ half the time so $f(z) = 1/2$.

b For $0 < z < 1$, then $X > Y$.
Since they're both uniform and by symmetry
 $P(X > Y) = P(Y > X)$
So the density should be 1/2 since it'll happen half the time uniformly

$$\text{Let } T = \frac{Y}{X} \text{ i.e. } T = \frac{Z}{Z+1}$$

$$f_T(t) = \frac{1}{1-t} \cdot f_Z(z) \Big|_{z=\frac{t}{t-1}} = \frac{1}{1-t} \Big|_{z=\frac{t}{t-1}} = \frac{1}{1-t}$$



② sec 6.1 Conditional Distribution: Discrete case.

let X, N discrete RVs w/ joint distribution $P(X=x, N=n)$.

Bayes rule

$$P(X=x | N=n) = \frac{P(X=x, N=n)}{P(N=n)}$$

or $P(X=x, N=n) = P(X=x | N=n)P(N=n)$

Rule of average conditional probabilities

marginal Prob of X

$$\begin{aligned} P(X=x) &= \sum_n P(X=x, N=n) \\ &= \sum_n P(X=x | N=n)P(N=n) \end{aligned}$$

ex

Let N have Poisson (λ) distribution. Let X be a random variable with the following property: for every n , the conditional distribution of X given $(N = n)$ is binomial (n, p) . Find the unconditional distribution of X and state its parameter(s). Show all your work for full credit.

$$P(N=n) = \frac{\bar{e}^\lambda \lambda^n}{n!}$$

$$P(X=x | N=n) = \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

Find $P(X=x)$

$$\begin{aligned} P(X=x) &= \sum_{n=x}^{\infty} P(X=x | N=n)P(N=n) \\ &= \sum_{n=x}^{\infty} \frac{n!}{x!(n-x)!} p^x q^{n-x} \frac{\bar{e}^\lambda \lambda^n}{n!} \\ &= \frac{\bar{e}^\lambda p^x}{x!} \sum_{n=x}^{\infty} \frac{\lambda^{n-x}}{(n-x)!} \end{aligned}$$

Finish

$$= \frac{e^{-\lambda} \lambda^x}{x!} \left(1 + \lambda q + (\lambda q)^2 + \dots \right)$$

$$= \frac{e^{-\lambda(1-p)}}{x!} (\lambda p)^x$$

$$\boxed{X \sim \text{Pois}(\lambda p)}$$

Explanation (Poisson thinning)

$$N = \# \text{ cars in 1 minute} \quad N \sim \text{Pois}(\lambda \cdot 1)$$

p = Prob of car being red

$X|N=n = \# \text{ of red cars given } N=n$

$$X|N=n \sim \text{Bin}(n, p)$$

$X = \# \text{ red cars in 1 minute}$

$X \sim \text{Pois}(\lambda p)$ by Poisson thinning.

2

Sec 6.2 Conditional Expectation (discrete case)

Bayes rule : $P(T=t | S=s) = \frac{P(T=t, S=s)}{P(S=s)}$

recall $\Leftrightarrow (T, S)$ is joint distribution below,

Find $P(T=3 | S=7)$

$$= \frac{P(T=3, S=7)}{P(S=7)} = \frac{0.3}{0.4} = \boxed{0.75}$$

	$T=3$	$T=4$	Sum	\leftarrow marginal ∂S
$S=7$	0.3	0.1	0.4	
$S=6$	0.2	0.2	0.4	
$S=5$	0.1	0.1	0.2	
Sum	0.6	0.4	1.0	
\nearrow marginal of T				

Find $P(T=4 | S=7)$

$$\frac{P(T=4, S=7)}{P(S=7)} = \frac{0.1}{0.4} = \boxed{0.25}$$

Find $E(T) = \sum_{t \in T} t P(T=t) = 3(0.6) + 4(0.4) = \boxed{3.4}$

Find $E(T | S=7)$

$$= \sum_{t \in T} t \cdot P(T=t | S=7) = 3 \cdot P(T=3 | S=7) + 4 \cdot P(T=4 | S=7)$$

$$= 3(0.75) + 4(0.25) = \boxed{3.25}$$

Find $E(T | S=6)$

$$= 3 \cdot P(T=3 | S=6) + 4 \cdot P(T=4 | S=6)$$

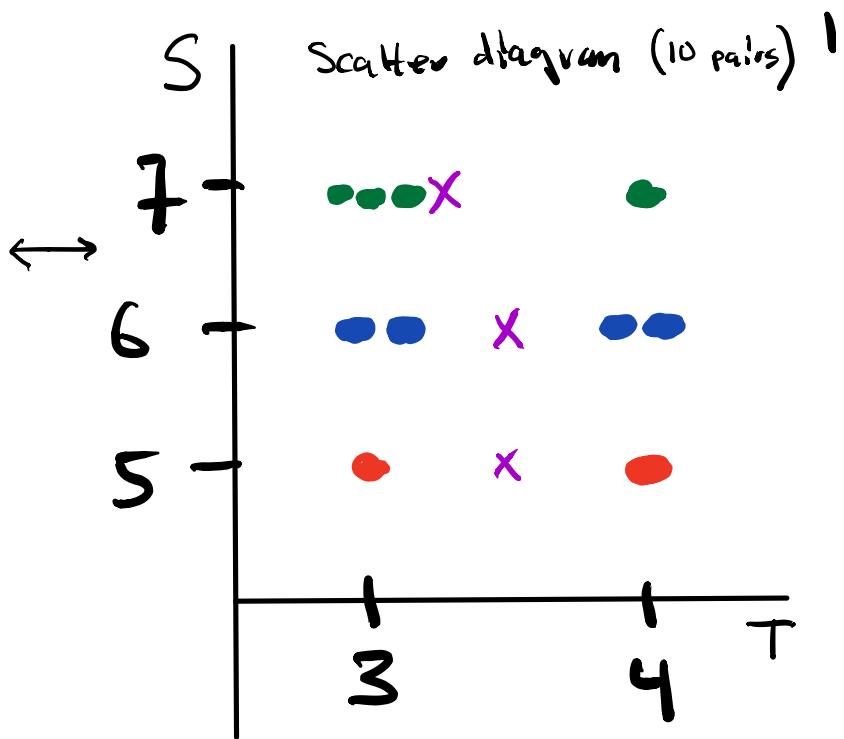
$$= 3 \left(\frac{0.2}{0.4} \right) + 4 \left(\frac{0.2}{0.4} \right) = \underline{\underline{3.5}}$$

	$T=3$	$T=4$	Sum	\leftarrow marginal of S
$S=7$	0.3	0.1	0.4	
$S=6$	0.2	0.2	0.4	
$S=5$	0.1	0.1	0.2	
Sum	0.6	0.4	1.0	
\nearrow marginal of T				

$$\left. \begin{aligned} E(T | S=7) &= 3.25 \\ E(T | S=6) &= 3.5 \\ E(T | S=5) &= 3.5 \end{aligned} \right\} \text{function of } S$$

Picture

		joint distribution			
		T=3	T=4	Sum	
		S=7	0.3	0.1	0.4
		S=6	0.2	0.2	0.4
		S=5	0.1	0.1	0.2
		Sum	0.6	0.4	1



Two main points:

- ① $E(T|S)$ is a function of S .
- ② $E(T|S)$ is a RV so it has an expectation.

Next we explore the expectation of $E(T|S)$,

$$\text{Let } g(S) = E(T|S)$$

$$g(7) = E(T|S=7) = 3.25$$

$$g(6) = 3.5$$

$$g(5) = 3.5$$

$$E(g(S)) = \sum_{S \in S} g(S) P(S=s)$$

$$= 3.25(.4) + 3.5(.4) + 3.5(.2)$$

$$= 3.4 \leftarrow \text{this is } E(T).$$

it is the weighted average of all of the group averages,

In other words,

$$E(E(T|S)) = E(T)$$

This is called the property of iterated expectations.

Intuitively,

If you have a class that is $\frac{2}{3}$ girls and $\frac{1}{3}$ boys and the girls weigh on average 100 lbs and boys weigh 200 lbs then the average weight of the class should be $\frac{2}{3}(100) + \frac{1}{3}(200)$. i.e., we take the weighted average of the averages.

Rule of average conditional expectation

For any random variable Y with finite expectation and any discrete RV X ,

$$E(T) = \sum_{\text{all } S} E(T|S=s) \cdot P(S=s)$$

(see end of this lecture for a formal proof)

Ex 8 transistors (type 1) are distributed $\text{Exp}(\frac{1}{100})$ and 4 transistors (type 2) are $\text{Exp}(\frac{1}{200})$.

Let T be the lifetime of a randomly picked transistor.

Find $E(T)$.

Soln

Let $X = \text{type of transistor}$

$$\begin{aligned} E(T) &= E(E(T|X)) \\ &= E(T|X=1)P(X=1) + E(T|X=2)P(X=2) \\ &= 100 \cdot \frac{8}{12} + 200 \cdot \frac{4}{12} = 133.3 \end{aligned}$$

Ex

(2 pts) Let $X \sim \text{Geom}(p)$ on $\{1, 2, \dots\}$. Suppose Y is a variable such that given $X = x$, Y is uniformly distributed on $\{1, 2, \dots, x\}$. Find $E(Y)$. (Hint: for $U \sim \text{Unif}\{a, a+1, \dots, b\}$, $E(U) = \frac{a+b}{2}$.)

$$E(Y) = E(E(Y|X))$$

$$Y|X \sim \text{Unif}\{1, 2, \dots, x\}$$

$$E(Y|X) = \frac{1+x}{2}$$

$$E(Y) = E(E(Y|X)) = E\left(\frac{1+x}{2}\right) = E\left(\frac{1}{2}\right) + E\left(\frac{x}{2}\right)$$

$$\begin{array}{c} E(x) = \frac{y_p}{2} \\ | \\ \frac{1}{2} \\ " \\ \frac{1}{2} \\ " \\ \frac{1}{2p} \end{array}$$

$$= \boxed{\frac{1}{2} + \frac{1}{2p}}$$

Appendix

Iterated Expectation

We show $E(Y) = E(E(Y|X))$:

$$\begin{aligned} E(Y) &= \sum_{\text{all } y} y P(Y=y) \\ &= \sum_y \sum_{\text{all } x} P(X=x, Y=y) \\ &= \sum_{\text{all } y} y \sum_{\text{all } x} \frac{P(X=x, Y=y)}{P(X=x)} P(X=x) \\ &= \sum_{\text{all } y} y \sum_{\text{all } x} P(Y=y | X=x) P(X=x) \\ &= \sum_{\text{all } x} \left(\sum_{\text{all } y} y P(Y=y | X=x) \right) \cdot P(X=x) \\ &= \sum_{\text{all } x} E(Y | X=x) \cdot P(X=x) \\ &= E(E(Y | X)) \end{aligned}$$

□