

Engineering Statistics
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Week 3
Lecture no. 15
Conditional PMF and PDF

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Conditional PMF

X_1 and X_2 are discrete with joint PMF $P(X_1 = x_1, X_2 = x_2)$. We may want to know about $X_2 = x_2$ given that $X_1 = x_1$

▶ Conditional PMF is defined as

$$P_{X_2|X_1}(x_2|x_1) = P(X_2 = x_2 | X_1 = x_1) = \frac{P(X_2 = x_2, X_1 = x_1)}{P(X_1 = x_1)}$$

▶ Conditional CDF is $F_{X_2|X_1}(x_2|x_1) = \sum_{x \leq x_2} P_{X_2|X_1}(x|x_1)$

▶ Conditional Expectation:

$$\mathbb{E}(X_2|X_1 = x_1) = \sum_x x P_{X_2|X_1}(x|x_1)$$

Now conditional probability mass functions. This is nothing but the joint probabilities. But earlier we were interested in joint, but now you want to condition on one happening over the other. We already talked about this conditioning on the events. What was that? How did you define if we have E and F are 2 events? And I am trying to find what is the conditional probability of F given that F has occurred. How did you define it? This is nothing but...

Student: (()) (0:51)

Professor: Divided by F. But now, this notion can be extended to random variables. Ultimately, if I am going to say that X equals to some small some number X1. This is corresponding to some event. So, now asking instead of specific events, I can ask those questions in terms of the random variables, we already have defined joint probability mass functions. Now, we want to extend this to the conditional probability mass functions.

Suppose I want to know what is the probability that the second random variable is going to take value X2 given that the first random variable has taken the value X1. This is a notation for this, but exact definition is this. And this is what it represents that is joint probability of

X_1 taking value small x_1 and a second and a variable taking X_2 given divided by probability that X_1 is going to take the value x_1 .

Now, so notice that now, focus on the notations like every time we may simply write like this. Here it is kind of giving already by this notation, we mean that we are asking for the question that my second random variable, this the one in the numerator, this portion corresponds to this, and X_1 corresponds to this random variable. And here, this vertical, this is like this is a notation to say given it is not like a division in this.

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Example:

$P(X_1, X_2)$	$X_2 = 2$	$X_2 = 4$	$X_2 = 5$
$X_1 = 1$.1	.05	.2
$X_1 = 2$.1	.1	.15
$X_1 = 3$.15	.1	0.05

$P_{X_2|X_1}(4|1) =$
 $\mathbb{E}(X_2 | X_1 = 1) =$
 $P_{X_2|X_1}(5|2) =$
 $\mathbb{E}(X_2 | X_1 = 2) =$

Handwritten notes: (x_1, x_2) , $\sum_{x_1} \sum_{x_2} P(x_1, x_2) = 1$, $\sum_{x_2} P(x_2 | x_1) = 1$

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 $\mathbb{E}(X_2 | X_1 = 1) =$
 $P_{X_2|X_1}(5|2) =$
 $\mathbb{E}(X_2 | X_1 = 2) =$

Handwritten calculations: $\frac{.05}{.35} = \frac{1}{7}$, $\frac{.1}{.35} = \frac{2}{7}$, $\frac{.1}{.35} = \frac{2}{7}$, $\frac{.15}{.35} = \frac{3}{7}$

Now, earlier, we argued that if let us say P is a, some discrete random variable, let us say X_1 and X_2 has some probability mass function. And we said X_1 , and X_2 takes value X_2 and

being if you add it or all possible values of X_1 , let us say this X_1 here, and X_2 here, what this value is going to be?

Student: 1.

Professor: It will be 1. But now suppose I am interested in this conditional probability, like fix some X_1 and now I am looking into this probability that X_2 is going to take X_2 given X_1 has already happened X_1 . Now this is fixed. And now I am looking into the possible values the X_2 can take. And now I want to take the summation of this X_2 , what this value can be? This one, no, I just interpret, what is this? This is I am asking the question. I have already telling that there are 2 random variables. Let us take for simplicity, X_1 is one coin, X_2 is another coin. They will be dependent, right now I am not saying anything dependent or independent.

The first coin I said it has already taken head or let us say it is 1, X_1 is 1 I have taught. Now X_2 I am asking, now I am asking X_2 after my first coin has taken value 1, X_2 can take still head or tail. And now I am asking the sum over those values. It has to add up to 1 like after the first one has shown had, the second one either has to take head or tail, one of these possibilities have to there is no other way. And there is some has to be 1. So, this is again going to be 1 actually. And that is why these conditional probabilities are themselves probability mass functions.

And similarly, you can define conditional ones, conditional probabilities are themselves are probability mass functions, and you just treat it as a simple probability mass function on a given condition and for that you can define your CDF in this fashion. Now here, where X_1 is fixed, and now you are looking for the probability that your X_2 is going to take value less than or equals to X_2 . So, I am just adding all the possible values of X_2 till the value x_2 . And once I know these conditional probabilities, I can talk about conditional expectation now. And this is a conditional expectation of X_2 , given X_1 has taken value x_1 , all I need to do is now X_1 has already happened x_1 . So, I do not need to worry about that X_1 is fixed. I now need to worry about all possible values of X_2 .

So, I am taking all possible values of X_2 here, maybe I should written this X_2 here. I multiply it, it is a probability conditional probability. And when you submit over all possible X_2 values, you will get the expected value. It is clear the expectation, those who could not

follow like, go back and refer to the slides. Now, let us quickly do an example to see that how much you could understand.

This table is the same table which we have earlier used for the probability mass function. So, there, you try to compute the marginals but now, we are going to compute...

Student: Conditional probabilities.

Professor: Conditional probabilities. So, now let us compute the probability that X_1 has taken the value 1, what is the probability that X_2 will take value 4? So, this one is simple go back and use this formula, let us compute what is the probability that X_2 takes value 4 and X_1 takes value 1, 4 and 1. So, if I had to use this formula, the numerator I got is 0.05. Now, what is the probability that X_1 equals to 1?

Student: 0.35

Professor: This is the sum of this row. And what is that value?

Student: 0.35.

Professor: This is going to be 1 by 7. Now, I want to compute the expectation of X_2 given that X_1 equals to 1. Now, for that what you need to do? You need to compute the probability mass function of X_2 given that X_1 equals to 1. One value already computed here. You have computed what is the value that X_2 takes value 4 given X_1 equals to 1. Now you need to find out what is that it takes value 2 and 5? Can somebody tell me, what is 2 given 1? So, the denominator remains same. What is going to change is this is going to be 0.1. So, this is like 1 by 35, 10 by 35.

And similarly, what is that, $p_{X_2 \text{ given } X_1 = 5}$ given 1, 20 by 35. So, now you got all the 3 all the possible values of X_2 .

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 $E(X_2 | X_1 = 1) = \frac{10}{35} + 4 + \frac{5}{35} = \frac{24}{35} + 4 + \frac{5}{35} = \frac{180}{35} = 4 + \frac{10}{35} = 4 + \frac{2}{7} = \frac{30}{7}$
 $P_{X_2|X_1}(5|2) = \frac{.15}{.35} = \frac{3}{7}$
 $E(X_2 | X_1 = 2) = \frac{10}{35} + \frac{5}{35} + \frac{20}{35} = \frac{35}{35} = 1$

$E[X_2] = \sum x_2 P(x_2)$
 $= 4$

$P_{X_2|X_1}(x_2|x_1) = \left\{ \begin{matrix} \frac{10}{35} & \frac{5}{35} & \frac{20}{35} \end{matrix} \right\}$
 $E[X_2 | X_1 = 1] = \frac{10}{35} + 4 + \frac{5}{35} = \frac{180}{35} = 4 + \frac{10}{35} = \frac{30}{7}$
 $P_{X_2|X_1}(2|1) = \frac{.1}{.35} = \frac{2}{7}$
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$E[X_2] = \sum x_2 P(x_2)$
 $= 3.7 = \frac{37}{10}$

$P_{X_2|X_1}(x_2|x_1) = \left\{ \begin{matrix} \frac{10}{35} & \frac{5}{35} & \frac{20}{35} \end{matrix} \right\}$
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So, first thing you need to now verify is, let me what we get is now probability that I am putting a dot here, X_1 equals to x_1 , what are the possible value X_2 is taking? It is taking 3 values 2, 4, 5. And now we just computed its probability mass function? When X_2 maybe write me $2 \times X_2$ when X_2 equals to 2, it was how much?

Student: 10/35.

Professor: It was 10 by 35. And when it was 4, how much it was? 1 by 7 or let me write is a 5 by 35 and when X_2 equals to 5, it was 20 by 35, is this a probability mass function the conditional because it is adding up to 1. So, always if it is a conditional probability is better add up to 1. Now, find the expectation now, it is easy. You have all the things. Now, this is

nothing but X_1 equals to x , now this is going to be 1. Now, all you need to do is 2 into 10 by 35 plus 5 into sorry 4 into 5 by 35 plus 5 into 20 by 35 and how much is this? 20, 40, 100, 140 by 35, am I correct? That is it that is expected value.

Now as a quick computation this is expectation of X_2 given that X_1 equals 1. Now, can you compute what is the expectation of X_2 just what is the expectation of X_2 . This is unconditional expectation. What is the value of this? This is nothing but x_2 into P of X_2 is taking value X_2 or x_2 and you can compute this marginal from this conditional that joint probabilities you can come compute this marginal. And also how you can compute this. Can somebody quickly compute this and tell me whether you get the same value as this? Are you getting same value, sure?

Student: 4.

Professor: It is just 4. Someone can someone verify this for me?

Student: 4

Professor: This is 4.

Student: (()) (12:38)

Professor: Leave it, why you want to calculate. Just enough know a fraction. I want this value.

Student: 3.7.

Professor: You are getting this as 3.7, you say like 37 by 100, did anybody get 3.7?

Student: (()) (12:59)

Professor: And this is 2. No this is not 2. This is 4. So, conditional expectation is 4 and unconditional expectation is 3.7, so in that way conditional probabilities and unconditional probability need not be the same. Similarly, you can compute it for others, I am just skipping it.

(Refer Slide Time: 13:47)

Conditional PDF

X_1 and X_2 are jointly continuous with PDF $f(X_1 = x_1, X_2 = x_2)$.
 We may want to know PDF of $X_2 = x_2$ given that $X_1 = x_1$

- Conditional PDF is defined as $c = \frac{4}{10}$

$$f_{X_2|X_1}(x_2|x_1) = \frac{f_{X_1, X_2}(x_1, x_2)}{f_{X_1}(x_1)} = f_{X_2|X_1}(x_2|x_1)$$

- Conditional Expectation: $= \frac{f(x_2=2.5)}{f_{X_2}(2.5)}$

$$\mathbb{E}(X_2|X_1 = x_1) = \int_x x f_{X_2|X_1}(x|x_1) dx$$

Example: $X = (X_1, X_2)$ are jointly continuous with PDF given by

$$f_X(x_1, x_2) = \begin{cases} c(1 + x_1 x_2) & \text{if } 2 \leq x_1 \leq 3, 1 \leq x_2 \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

Find $f_{X_2|X_1}(x_2|2.5)$ and $\mathbb{E}(X_2|X_1 = 2.5)$

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- Conditional Expectation: $\mathbb{E}(X_2|X_1 = x_1) = \sum x P_{X_2|X_1}(x|x_1)$

Example:

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$P_{X_2|X_1}(4|1) = \frac{P(X_2=4, X_1=1)}{P(X_1=1)} = \frac{0.05}{0.35} = \frac{1}{7}$
 $\mathbb{E}(X_2|X_1 = 1) = \frac{2 \cdot \frac{10}{35} + 4 \cdot \frac{5}{35} + 5 \cdot \frac{20}{35}}{\frac{30}{35}} = \frac{100}{35} = \frac{20}{7}$
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Once we do it for PMF same thing can be done for probability density function also. Let us say you have this joint continuous probability density function then we want to know the PDF of X_2 given that X_1 has taken some value. And by definition, the conditional probability is defined like this.

That is again the joint PDF divided by the marginal PDF on which you want to condition. And similarly, if you want to compute the expectation, you do the same thing. You take the expectation with respect to the conditional PDF. And here again, as an example, I put the same example that I used earlier, where we try to compute marginal PDF from their joint PDF. Now, can somebody quickly compute me? By the way, do somebody remember what was the value of C, earlier?

Student: 4 by 9.

Professor: 4 by 9.

Student: 4 by 19.

Professor: 4 by 19, 4 by 29.

Student: 4 by 19.

Professor: 4 by 19. Let us take for time being c equals to 4 by 19. Now, what is this value of f of $x_2 \times x_1$ given that your x_2 and x_1 value is 2.5. How to compute this? For this you need to find f of.

Professor: You want to take X_2 equals to X_2 and X_1 is 0.25 divided by f of x at 0.25. This is the value. And notice that both in this case and this case and this case the value on which you want to condition that this probability need to be positive. Otherwise you will end up with divided by 0 condition. If you are trying to condition on some value x_1 it is better be some value in which there is some positive mass. If there is no positive mass that means value 0 means why you want to condition on something which is not going to occur.

So, whenever you want to do condition make sure that you condition on something which has a nonzero mass and the same thing when you want to condition the continuous case on some point make sure that that point has a positive value the PDF is positive at that point. Like similarly again you can compute the expectation here by finding the like first you need to find the conditional PDF and based on that you can find out this conditional expectation.