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AFRE 802 Statistical Methods for Agricultural, Food, & Resource Economists



Multivariate probability distributions – Part 2 of 3 (WMS Ch. 5.3-5.4)

October 10, 2017 Nicole Mason Michigan State University Fall 2017

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GAME PLAN

- · Return HWs and in-class exercises
- · No graded in-class exercise but expect one on Thursday
- Reminder about mid-term (Thurs., Oct. 19)
- Review

Multivariate probability distributions (Part 2 of 3)

- 1. Finish bivariate probability distributions for continuous RVs
- 2. Marginal probability distributions
- 3. Conditional probability distributions
- 4. Independent random variables

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Reminder: Mid-term is next Thurs., Oct. 19

- Start studying NOW (if you haven't already)
- Will cover through end of Ch. 5 (multivariate probability distributions)
- One (double-sided) 8.5 x 11 inch cheat sheet
- I will provide tables
- Past mid-terms and answers are on D2L
- Discuss good exam prep and test-taking strategies

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Review

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Bivariate probability distributions for discrete RVs

Probability distribution : $p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2)$

1.
$$0 \le p(y_1, y_2) \le 1$$

2. $\sum_{y_1, y_2} p(y_1, y_2) = 1$

EX) p(y₁, y₂) for discrete PDF (# of customers going to supermarket counter 1 vs. 2)

		y_1	
y_2	0	1	2
0	1/9	2/9	1/9
1	1/9 2/9 1/9	2/9 2/9	0
2	1/9	0	0

Review: questions on this?

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Bivariate CDFs for discrete RVs

$$F(y_1, y_2) = P(Y_1 \le y_1, Y_2 \le y_2) = \sum_{t_2 \le y_2} \sum_{t_1 \le y_1} p(t_1, t_2)$$

EX) In the supermarket checkout counter example, find:

- a. F(-1, 2)=0 because Y_1 can't be less than zero
- b. F(1.5, 2)=8/9 because everything but (2,0)
- c. F(5, 7) = 1 because all valid values of Y_1 and Y_2 are less than 5 and 7, respectively

		y 1	
y ₂	0	1	2
0	1/9	2/9	1/9
1	1/9 2/9	2/9 2/9	0
2	1/9	0	0

Review

Both discrete & continuous bivariate CDFs must satisfy similar properties to what we saw in the univariate case:

1.
$$F(-\infty, -\infty) = 0$$
, $F(-\infty, y_2) = 0$, $F(y_1, -\infty) = 0$

$$2. F(\infty, \infty) = 1$$

Bivariate CDF (discrete RVs)

$$F(y_1, y_2) = P(Y_1 \le y_1, Y_2 \le y_2) = \sum_{t_2 \le y_2} \sum_{t_1 \le y_1} p(t_1, t_2)$$

Bivariate CDF (continuous RVs)

$$F(y_1, y_2) = P(Y_1 \le y_1, Y_2 \le y_2) = \int_{-\infty}^{y_2} \int_{-\infty}^{y_1} f(t_1, t_2) dt_1 dt_2$$
where $-\infty < y_1 < \infty$, $-\infty < y_2 < \infty$

Review

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Bivariate PDFs (continuous RVs)

$$f(y_1, y_2)$$

Like PDFs for the univariate case (and bivariate probability distributions for discrete RVs), bivariate PDFs must satisfy similar properties:

$$1. f(y_1, y_2) \ge \theta$$
 for all y_1, y_2

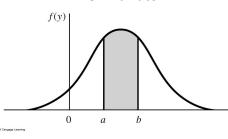
2.
$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(y_1, y_2) dy_1 dy_2 = 1$$

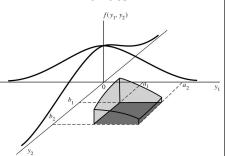
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Graphical representation of univariate vs. bivariate PDFs



Bivariate





Recall that in the univariate case, area under the PDF between a and $b = P(a \le Y \le b)$

$$P(a \le Y \le b)$$

$$= \int_{a}^{b} f(y) dy$$

Bivariate PDFs are 2-dimensional surfaces, so the volume under the surface corresponds to a probability – e.g., above:

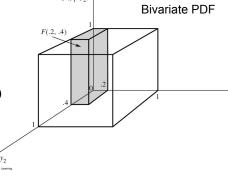
$$P(a_1 \le Y_1 \le a_2, b_1 \le Y_2 \le b_2)$$

$$= \int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$$

EX1) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 1, \ 0 \le y_1 \le 1, \ 0 \le y_2 \le 1 \\ 0, \ \text{elsewhere} \end{cases}$$

- a. Find $F(0.2, 0.4) = P(Y_1 \le 0.2, Y_2 \le 0.4)$
- b. Find $P(0.1 \le y_1 \le 0.3, 0 \le y_2 \le 0.5)$



$$P(a_1 \le Y_1 \le a_2, b_1 \le Y_2 \le b_2)$$

= $\int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$

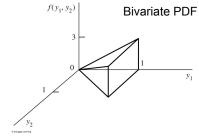
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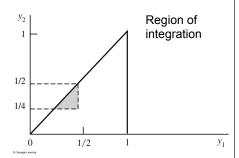
EX1) Finding probabilities from a bivariate PDF (continuous RVs)

EX2) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 3y_1, & 0 \le y_2 \le y_1 \le 1 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(0 \le Y_1 \le 0.5, Y_2 > 0.25)$





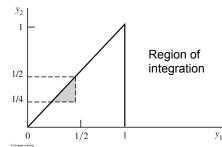
$$P(a_1 \le Y_1 \le a_2, b_1 \le Y_2 \le b_2)$$

= $\int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$

EX2) Finding probabilities from a bivariate PDF (continuous RVs)

 $f(y_1, y_2) = \begin{cases} 3y_1, & 0 \le y_2 \le y_1 \le 1 \\ 0, & \text{elsewhere} \end{cases}$

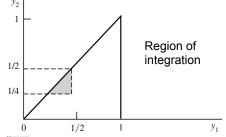
Find $P(0 \le Y_1 \le 0.5, Y_2 > 0.25)$



EX2) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 3y_1, & 0 \le y_2 \le y_1 \le 1 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(0 \le Y_1 \le 0.5, Y_2 > 0.25)$



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Moving beyond the bivariate case

 $\label{lem:continuous} \textbf{Joint probability distributions for discrete RVs:}$

$$p(y_1, y_2, ..., y_n) = P(Y_1 = y_1, Y_2 = y_2, ..., Y_n = y_n)$$

Joint probability density function (PDF) for continuous RVs:

$$f(y_1, y_2, ..., y_n)$$

Joint cumulative distribution function (CDF) for discrete & continuous RVs:

$$F(y_1, y_2, ..., y_n) = P(Y_1 \le y_1, Y_2 \le y_2, ..., Y_n \le y_n)$$

Marginal probability distributions

• Recall the supermarket checkout counter example. We derived the **bivariate** probability distribution, $p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2)$:

		y 1		
y_2	0	1	2	$p_2(y)$
0	1/9	2/9	1/9	— 4/9
1	2/9	2/9	0	4/9
2	1/9	0	0	1/9
$p_1(y_1)$:	4/9	4/9	1/9	

- Can we get the <u>univariate</u> probability distribution for Y_1 (i.e., $p_1(y_1) = P(Y_1 = y_1)$ from this bivariate distribution? In other words, what is $P(Y_1 = 0)$? $P(Y_1 = 1)$? $P(Y_1 = 2)$?
- How about $p_2(y_2) = P(Y_2 = y_2)$?

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Marginal probability distributions

What we just derived are called "marginal probability distributions"

Marginal probability distributions for discrete RVs:

$$p_1(y_1) = \sum_{\text{all } y_2} p(y_1, y_2) \text{ and } p_2(y_2) = \sum_{\text{all } y_1} p(y_1, y_2)$$

• For continuous RVs:

Marginal probability density functions (PDFs) for continuous RVs:

$$f_1(y_1) = \int_{-\infty}^{\infty} f(y_1, y_2) dy_2$$
 and $f_2(y_2) = \int_{-\infty}^{\infty} f(y_1, y_2) dy_1$

EX) Marginal probability distributions from discrete bivariate probability distribution

From a group of three Republicans, two Democrats, and one independent, a committee of two people is to be randomly selected.

- Y₁: number of Republicans on the committee
- Y₂: number of Democrats on the committee

The bivariate probability distribution is given below. (This is a hypergeometric distribution problem, which we didn't cover.) Use the information in the table below to find the marginal probability distribution of Y_1 and Y_2 .

Table 5.2 Joint probability function for Y_1 and Y_2 , Example 5.5

	<i>y</i> ₁			$p_2(y_2)$:
y_2	0	1	2	Total
0	0	3/15	3/15	6/15
1	2/15	6/15	0	8/15
2	1/15	0	0	1/15
Total	3/15	9/15	3/15	1

 $p_1(y_1)$:

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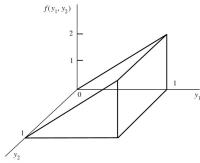
 $f_1(y_1) = \int_{-\infty}^{\infty} f(y_1, y_2) dy_2$ and $f_2(y_2) = \int_{-\infty}^{\infty} f(y_1, y_2) dy_1$

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EX) Marginal PDFs from continuous bivariate PDF

Find the marginal PDFs for Y_1 and Y_2 from the continuous bivariate PDF:

$$f(y_1, y_2) = \begin{cases} 2y_1, & 0 \le y_1 \le 1, \\ 0, & \text{elsewhere} \end{cases}$$



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Conditional probability distributions for discrete RVs:

What were some formulas we saw for $P(A \cap B)$ in terms of conditional and unconditional probabilities?

$$P(A \cap B) = P(A)P(B \mid A) = P(B)P(A \mid B)$$

So how can write $p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2)$, an intersection of 2 events? $p(y_1, y_2) = p_1(y_1)p(y_2 | y_1) = p_2(y_2)p(y_1 | y_2)$

Solve for the conditional probabilities:

$$p(y_1 | y_2) = \frac{p(y_1, y_2)}{p_2(y_2)}$$

and

$$p(y_2 | y_1) = \frac{p(y_1, y_2)}{p_1(y_1)}$$

Keep in mind what these expressions mean:

$$P(Y_1 = y_1 | Y_2 = y_2) = \frac{P(Y_1 = y_1, Y_2 = y_2)}{P(Y_2 = y_2)}$$

and

$$P(Y_2 = y_2 | Y_1 = y_1) = \frac{P(Y_1 = y_1, Y_2 = y_2)}{P(Y_1 = y_1)}$$

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EX) Conditional probability distributions for discrete RVs

Continuing with our previous example about a 2-person committee drawn from Democrats, Republicans, and independents (with bivariate probability distribution below), find the conditional probability distribution of Y_1 given that Y_2 =1.

Table 5.2 Joint probability function for Y_1 and Y_2 , Example 5.5

	Уı			$p_2(y_2)$:	
y_2	0	1	2	Total	
0	0	3/15	3/15	6/15	
1	2/15	6/15	0	8/15	
2	1/15	0	0	1/15	
Total	3/15	9/15	3/15	1	

 $p(y_1 | y_2) = \frac{p(y_1, y_2)}{p_2(y_2)}$

Conditional probability distributions for continuous RVs:

Recall that for **discrete RVs**, these are:

$$p(y_1 | y_2) = \frac{p(y_1, y_2)}{p_2(y_2)}$$
 and $p(y_2 | y_1) = \frac{p(y_1, y_2)}{p_1(y_1)}$

Conditional PDF:

$$f(y_1 | y_2) = \frac{f(y_1, y_2)}{f_2(y_2)}$$
 and $f(y_2 | y_1) = \frac{f(y_1, y_2)}{f_1(y_1)}$

Conditional CDF:

 $F(y_1 | y_2)$ means $P(Y_1 \le y_1 | Y_2 = y_2)$ Obtain these values by integrating the conditional PDF over the relevant range.

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EX) Conditional probability distributions for continuous RVs

Find the conditional probability, $P(Y_1 \le 0.5 \mid Y_2 = 1.5)$, for the continuous bivariate PDF:

$$f(y_1, y_2) = \begin{cases} 0.5, \ 0 \le y_1 \le y_2 \le 2 \\ 0, \text{ elsewhere} \end{cases}$$
Conditional PDF:
$$f(y_1 | y_2) = \frac{f(y_1, y_2)}{f_2(y_2)}$$

Conditional PDF:

$$f(y_1 | y_2) = \frac{f(y_1, y_2)}{f_2(y_2)}$$

Independent random variables

What is $P(A \cap B)$ if events A and B are **independent**?

$$P(A \cap B) = P(A)P(B)$$

Can use similar approach to see if two RVs, Y_1 and Y_2 , are independent:

CDF: $F(y_1, y_2) = F_1(y_1)F_2(y_2)$

Discrete probability distribution : $p(y_1, y_2) = p_1(y_1)p_2(y_2)$

Continuous probability distribution: $f(y_1, y_2) = f_1(y_1) f_2(y_2)$

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EX) Checking for independence – discrete RVs

Continuing with our previous example about a 2-person committee drawn from Democrats, Republicans, and independents (with bivariate probability distribution below), is Y_1 independent of Y_2 ? Can check any point, but let's try (0,0).

 $p_1(y)$

Table 5.2 Joint probability function for Y_1 and Y_2 , Example 5.5

	y_1			$p_2(y_2)$:
y_2	0	1	2	Total
0	0	3/15	3/15	6/15
1	2/15	6/15	0	8/15
2	1/15	0	0	1/15
Total	3/15	9/15	3/15	1

Discrete probability distribution if independent : $p(y_1, y_2) = p_1(y_1)p_2(y_2)$

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EX) Checking for independence – continuous RVs

Suppose Y_1 and Y_2 have the continuous bivariate PDF below. Are these two RVs independent?

$$f(y_1, y_2) = \begin{cases} 6y_1 y_2^2, & 0 \le y_1 \le 1, \\ 0, & \text{elsewhere} \end{cases}$$

Continuous probability distribution if indepedent: $f(y_1, y_2) = f_1(y_1) f_2(y_2)$

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Another way to check for independence

 Y_1 and Y_2 are independent RVs if:

$$f(y_1, y_2) = g(y_1)h(y_2)$$

where $g(y_1)$ is a nonnegative function of y_1 alone and $h(y_2)$ is a nonnegative function of y_2 alone ***AND only if

$$f(y_1, y_2) > 0$$
 for $a \le y_1 \le b$ and $c \le y_2 \le d$ for constants $a, b, c, d ***$.

EX) Suppose Y_1 and Y_2 have the continuous bivariate PDF below. Are these two RVs independent?

$$f(y_1, y_2) = \begin{cases} 2y_1, & 0 \le y_1 \le 1, \\ 0, & \text{elsewhere} \end{cases}$$

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Cannot always use this method – e.g.:

$$f(y_1, y_2) = \begin{cases} 8y_1 y_2, & 0 \le y_1 \le 1, \\ 0, & \text{elsewhere} \end{cases}$$

We <u>cannot</u> use the alternative method here because the range over which y_2 has positive probability is a function of y_1 .

If we found the marginal distributions for y_1 and y_2 , we would see that:

$$f_I(y_I) = 4y_I^3$$

and

$$f_2(y_2) = 4y_2(1 - y_2^2)$$

So $f(y_1, y_2) \neq f_1(y_1) f_2(y_2)$, thus Y_1 and Y_2 not independent

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Homework:

- WMS Ch. 5 (part 2 of 3)
 - Marginal & conditional probability distributions: 5.19, 5.20, 5.26
 - Independent random variables: 5.45, 5.46, 5.52
- Reminder: Ch. 5 HW will <u>not</u> be collected

Next class:

- Multivariate probability distributions, cont'd (Part 3 of 3)
 - · Expected values, variances, and covariances

Reading for next class:

• WMS Ch. 5 (sections 5.5-5.8, 5.11, 5.12)