

STA 103 Lecture 4: Joint Distributions

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Announcement

Midterm 1 Format Announcement

Hi all,

The midterm will be held next Wed from 12:10 pm - 1:00 pm (50 minutes). Below is the general information about the Midterm 1 format.

There will be **four Big problems** with each **3 sub problems**. We have a total of 12 subproblems, i.e., 1(a), 1(b), 1(c), 2(a), 2(b), 2(c), ... 4(a), 4(b), 4(c). The subproblems within the same main problem **may or may not** build on each other (e.g., (a) or (b) might or might not be used to solve (c)), but each main problem (1, 2, 3, 4) is **independent** from the others.

It means that the format (or style) will be similar to **Assignment 1**. However, the difficulty of Midterm 1 is adjusted so the exam can be completed in 50 minutes. (Assignment 1 was designed to take at least 3 hours, so you can expect each midterm problem to be more manageable.)

Details:

- Midterm 1 will cover Lecture 1 - 4.
- Each subproblems will be similar to one of (lecture note examples, assignment 1, and problem set 1).
- You will not encounter completely unfamiliar topics on the exam.
- Each question will have a clear, well-defined answer. No open-ended questions like "give your thoughts," or "discuss somethings."
- However, **showing your work is essential**: grader will check your derivations and may award partial credit even if your final answer is not correct.
- Likewise, you may not receive full credit if you just drop the answer without any derivation process.
- A printed formula sheet (with everything you need) will be provided at the start of the exam—**no need to print** your own.
- Please bring your **own calculator**.
- Please bring your **student ID**.

Tips for preparation:

I recommend mastering first both the material and the worked examples on the **lecture notes**. This will give you the solid foundation you need for the exams. But, to complete 12 subproblems in 50 minutes, you'll need not only a clear understanding of each concept but also enough practice to solve each problem in **about 4 minutes**. To build that speed, working through Assignment 1 and Problem Set 1 will be especially helpful.

Announcement

- To help you prepare for Midterm 1 next Wednesday, I will move my office hour next week only from Thursday (2:00–3:00 pm) to Tuesday (2:00–3:00 pm) in MSB 1143.

Dependence between two RVs X and Y

- **Motivation:** Suppose you roll two 6-sided dice, one red one blue. Let X = the sum of the numbers showing for the red and blue die, Y = the value of just the red die.
- **Question 1:** Suppose your betting that the red die is 1, i.e., $Y = 1$. What is the probability of winning?
- **Question 2:** Suppose you know the information that $X = 2$. What is the probability of $Y = 1$ given $X = 2$?
- **Question 3:** Suppose you know the information that $X = 7$. What is the probability of $Y = 1$ given $X = 7$?

Dependence between two RVs X and Y

- **Answer for Q2:** It turns out that if $X = 2$, then the adjusted PMF for Y is

$$p(y \mid X = 2) = \begin{cases} 1 & \text{if } y = 1, \\ 0 & \text{if } y = 0. \end{cases}$$

(Why?)

- **Answer for Q3:** It turns out that if $X = 7$, then there is no adjustment of PMF for Y is

$$p(y \mid X = 7) = p(y) = 1/6, \quad \text{for } y = 1, 2, 3, 4, 5, 6.$$

(Why?)

- **Message:** The fact that the conditional PMF of Y changes with different information on X indicates that X and Y are **dependent** random variables.
- More explicitly two random variables X, Y are dependent if knowing the value of one changes the PMF of the other.

Joint PMF for X and Y

- **Motivation:** We have seen that a PMF provides the blueprint for a single random variable X . **However**, when we have a pair of random variables (X, Y) , it is **no longer sufficient** to use the PMF of each $p_X(x)$ (PMF of X), $p_Y(y)$ (PMF of Y) to characterize the randomness in the pair.
- **Definition 1:** The Joint Probability Mass Function (JPMF) $p(x, y)$ for a pair of random variables (X, Y) , given in a **table form** is the following:

$X \backslash Y$	0	1	
0	1/8	0	
1	2/8	1/8	
2	0	3/8	
3	1/8	0	
			1

- **Example:** $p(3, 0) = P(X = 3, Y = 0) = \frac{1}{8}$.

Joint PMF for X and Y

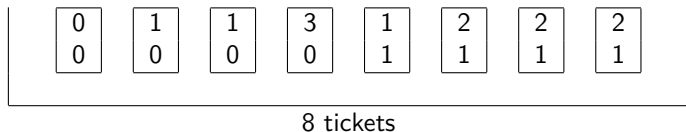
- **Definition 2:** The Joint Probability Mass Function (JPMF) $p(x, y)$ for a pair of random variables (X, Y) , given in a **function form** is the following:

$$p(x, y) = P(X = x \text{ and } Y = y).$$

- The input of a possible (X, Y) pair (x, y) and the output is the chance $P(X = x \text{ and } Y = y)$ that $X = x$ and $Y = y$.

Use the joint PMF to find any probability

- By summing the entries corresponding to the corresponding event.
Box model is best illustrated with an example:



- Equivalently, we have a joint PMF in a table form:

$X \backslash Y$	0	1
0	$1/8$	0
1	$2/8$	$1/8$
2	0	$3/8$
3	$1/8$	0
		1

Use the joint PMF to find any probability

- Suppose we have

$X \backslash Y$	0	1	
0	$1/8$	0	
1	$2/8$	$1/8$	
2	0	$3/8$	
3	$1/8$	0	
			1

- Question 1:** What is $P(X = 2 \text{ and } Y = 1)$?
- Question 2:** What is $P(X = 1 \text{ or } Y = 1)$?
- Question 3:** What is $P(Y = 0)$?
- Question 4:** What is $P(X = 1)$?

Use the joint PMF to find marginal PMF

- Notice that

$$P(X = x) = p_X(x) = \text{row sum } p(x, y) \text{ at row } x,$$

$$P(Y = y) = p_Y(y) = \text{column sum } p(x, y) \text{ at column } y.$$

- These PMFs: $p_X(x)$ and $p_Y(y)$ obtained from a joint PMF $p(x, y)$ are typically called the marginal PMFs for X and Y , respectively.

- **Example:**

$X \backslash Y$	0	1	$p_X(x)$
0	1/8	0	1/8
1	2/8	1/8	3/8
2	0	3/8	3/8
3	1/8	0	1/8
$p_Y(y)$	4/8	4/8	1

Use the joint PMF to find marginal PMF

- **Example:**

$X \backslash Y$	0	1	$p_X(x)$
0	1/8	0	1/8
1	2/8	1/8	3/8
2	0	3/8	3/8
3	1/8	0	1/8
$p_Y(y)$	4/8	4/8	1

- **Question:** What is $E(X)$, and what is $E(Y)$?

Marginal PMFs are **Not** Sufficient, We Need Joint PMF

- Another example with same marginal PMFs:

$X \backslash Y$	0	1	$p_X(x)$
0	1/8	0	1/8
1	2/8	1/8	3/8
2	1/8	2/8	3/8
3	0	1/8	1/8
$p_Y(y)$	4/8	4/8	1

- **Main message:** To provide a mathematical blueprint for characterizing the dependence between two RVs X and Y , we need **joint PMF**. The **marginal PMFs alone is not sufficient**.

Conditional Joint PMF

- Suppose X, Y have joint PMF:

$X \backslash Y$	0	1	$p_X(x)$
0	1/8	0	1/8
1	2/8	1/8	3/8
2	0	3/8	3/8
3	1/8	0	1/8
$p_Y(y)$	4/8	4/8	1

- For a given draw of (X, Y) , suppose you told that $X \geq 1$ and $Y = 1$ but not the exact values of X and Y .
- Question 1:** $P(X = 2 \mid X \geq 1 \text{ and } Y = 1)$?

Conditional Joint PMF

- Set the entries of $p(x, y)$ to zero that do not satisfy the given information, $X \geq 1$ and $Y = 1$.

$X \backslash Y$	0	1	
0	0	0	
1	0	$1/8$	
2	0	$3/8$	
3	0	0	
			$1/8 + 3/8 + 0 = 4/8$

- Re-normalize the entries to add to 1.

$X \backslash Y$	0	1	
0	0	0	
1	0	$(1/8)/(4/8)$	
2	0	$(3/8)/(4/8)$	
3	0	$0/(4/8)$	
			$(4/8) / (4/8)$

Conditional Joint PMF

The Joint PMF for $P(X = x, Y = y \mid X \geq 1 \text{ and } Y = 1)$

$X \backslash Y$	0	1	$p_X(x \mid X \geq 1 \text{ and } Y = 1)$
0	0	0	$0 + 0 = 0$
1	0	$1/4$	$0 + 1/4 = 1/4$
2	0	$3/4$	$0 + 3/4 = 3/4$
3	0	0	$0 + 0 = 0$
			1

- Getting back to the problem:

- Answer for Question 1:**

$$P(X = 2 \mid X \geq 1 \text{ and } Y = 1) = 0 + 3/4 = 3/4.$$

Conditional Joint PMF

- **Business example (customer behavior):** a small online retailer tracks two variables across 100 recent customers.
 - ▶ X : Number of items purchased (0, 1, or 2)
 - ▶ Y : Whether the customer had a coupon (0 = “No”, 1 = “Yes”).
- The joint distribution is summarized in the following table:

$X \backslash Y$	0 (No Coupon)	1 (Coupon)	
0	0.10	0.05	
1	0.20	0.25	
2	0.10	0.30	
			1

Conditional Joint PMF

- The joint distribution is summarized in the following table:

$X \backslash Y$	0 (No Coupon)	1 (Coupon)	
0	0.10	0.05	
1	0.20	0.25	
2	0.10	0.30	
			1

- Interpretation of marginal PMFs:

- Marginal PMF of Y (Distribution of **Coupon Usage**): sum across rows:

$$p_Y(0) = 0.10 + 0.20 + 0.10 = 0.40$$

$$p_Y(1) = 0.05 + 0.25 + 0.30 = 0.60$$

- Marginal PMF of X (Distribution of **Item Purchased**): sum across columns:

$$p_X(0) = 0.10 + 0.05 = 0.15$$

$$p_X(1) = 0.20 + 0.25 = 0.45$$

$$p_X(2) = 0.10 + 0.30 = 0.40$$

Computing $E(f(X, Y))$

- Recall our shortcut for computing $E(f(X))$:

$$E(f(X)) = \sum_x f(x)p_X(x).$$

- A similar shortcut works with a joint PMF $p(x, y)$ for two RVs X and Y :

$$E(f(X, Y)) = \sum_{(x, y)} f(x, y)p(x, y),$$

$$E(f(X)) = \sum_{(x, y)} f(x)p_X(x),$$

$$E(f(Y)) = \sum_{(x, y)} f(y)p_Y(y).$$

- Let (X, Y) have the joint PMF:

$X \backslash Y$	0	1	
0	0.06	0.04	
1	0.54	0.36	
			1

- Question:** Find $E(X + Y)$, $E(XY)$, and $\text{Var}(XY)$.

Computing $E(f(X, Y))$

Joint PMF

$X \backslash Y$	0	1	
0	0.06	0.04	
1	0.54	0.36	
			1

- Using our shortcut:

$$\begin{aligned} E(X + Y) &= \sum_{(x,y)} (x + y)p(x, y) \\ &= (0 + 0) \times 0.06 + (1 + 0) \times 0.54 \\ &\quad + (0 + 1) \times 0.04 + (1 + 1) \times 0.36. \end{aligned}$$

- Similarly,

$$\begin{aligned} E(XY) &= \sum_{(x,y)} (xy)p(x, y) \\ &= (0 \cdot 0) \times 0.06 + (1 \cdot 0) \times 0.54 \\ &\quad + (0 \cdot 1) \times 0.04 + (1 \cdot 1) \times 0.36. \end{aligned}$$

Computing $E(f(X, Y))$

- We know that $\text{Var}(XY) = E(X^2Y^2) - (E(XY))^2$. Then,

$$\begin{aligned} E(X^2Y^2) &= \sum_{(x,y)} (x^2y^2)p(x,y) \\ &= (0^2 \cdot 0^2) \times 0.06 + (1^2 \cdot 0^2) \times 0.54 \\ &\quad + (0^2 \cdot 1^2) \times 0.04 + (1^2 \cdot 1^2) \times 0.36. \end{aligned}$$

Then,

$$\text{Var}(XY) = E(X^2Y^2) - (E(XY))^2.$$

Conditional Joint PMF

- Revisit the joint PMF for (X, Y) :

$X \backslash Y$	0	1	$p_X(x)$
0	1/8	0	1/8
1	2/8	1/8	3/8
2	0	3/8	3/8
3	1/8	0	1/8
$p_Y(y)$	4/8	4/8	1

- For a given draw of (X, Y) , suppose you told that $X \geq 1$ and $Y = 1$ but not the exact values of X and Y .
- Question 2:** Your best prediction for X given $X \geq 1$ and $Y = 1$, i.e., $E(X \mid X \geq 1 \text{ and } Y = 1)$?
- Question 3:** Your typical prediction error for X given $X \geq 1$ and $Y = 1$, i.e., $\text{sd}(X \mid X \geq 1 \text{ and } Y = 1)$?

Conditional Joint PMF

The Joint PMF for $P(X = x, Y = y \mid X \geq 1 \text{ and } Y = 1)$

$X \backslash Y$	0	1	$p_X(x \mid X \geq 1 \text{ and } Y = 1)$
0	0	0	$0 + 0 = 0$
1	0	$1/4$	$0 + 1/4 = 1/4$
2	0	$3/4$	$0 + 3/4 = 3/4$
3	0	0	$0 + 0 = 0$
			1

- **Answer for Question 2:** Using the shortcut

$$\begin{aligned} E(X \mid X \geq 1 \text{ and } Y = 1) &= \sum_x x p_X(x \mid X \geq 1 \text{ and } Y = 1). \\ &= (0 \cdot 0) + (1 \cdot 1/4) + (2 \cdot 3/4) + (3 \cdot 0) \\ &= 7/4. \end{aligned}$$

Conditional Joint PMF

The Joint PMF for $P(X = x, Y = y \mid X \geq 1 \text{ and } Y = 1)$

$X \backslash Y$	0	1	$p_X(x \mid X \geq 1 \text{ and } Y = 1)$
0	0	0	$0 + 0 = 0$
1	0	$1/4$	$0 + 1/4 = 1/4$
2	0	$3/4$	$0 + 3/4 = 3/4$
3	0	0	$0 + 0 = 0$
			1

- **Answer for Question 3:** Using the shortcut

$$\begin{aligned} & \text{Var}(X \mid X \geq 1 \text{ and } Y = 1) \\ &= E(X^2 \mid X \geq 1 \text{ and } Y = 1) - (E(X \mid X \geq 1 \text{ and } Y = 1))^2 \\ &= \sum_x x^2 p_X(x \mid X \geq 1 \text{ and } Y = 1) - \left(\sum_x x p_X(x \mid X \geq 1 \text{ and } Y = 1) \right)^2 \\ &= (0^2 \cdot 0) + (1^2 \cdot 1/4) + (2^2 \cdot 3/4) + (3^2 \cdot 0) - (7/4)^2 \\ &= 13/4 - (7/4)^2 = 3/16. \end{aligned}$$

Conditional Joint PMF

The Joint PMF for $P(X = x, Y = y \mid X \geq 1 \text{ and } Y = 1)$

$X \backslash Y$	0	1	$p_X(x \mid X \geq 1 \text{ and } Y = 1)$
0	0	0	$0 + 0 = 0$
1	0	$1/4$	$0 + 1/4 = 1/4$
2	0	$3/4$	$0 + 3/4 = 3/4$
3	0	0	$0 + 0 = 0$
			1

- **Answer for Question 3 (Continue):** Then we have

$$\begin{aligned}\text{sd}(X \mid X \geq 1 \text{ and } Y = 1) &= \sqrt{\text{Var}(X \mid X \geq 1 \text{ and } Y = 1)} \\ &= \sqrt{\frac{3}{16}} = \frac{\sqrt{3}}{4}.\end{aligned}$$

Independent Random Variables

The **formal** definition of independence for two random variables is that

- $p_X(x \mid Y = y) = p_X(x)$ for all possible x, y , **or**
- $p_Y(y \mid X = x) = p_Y(y)$ for all possible x, y .
- Here is an example of a pair X, Y such that knowing the value of Y does not change the PMF for X :

$X \backslash Y$	0	1	$p_X(x)$
0	0.06	0.04	0.1
1	0.54	0.36	0.9
			1

Independent Random Variables

Marginal PMF for X :

$x \backslash Y$	0	1	$p_X(x)$
0	0.06	0.04	0.1
1	0.54	0.36	0.9
			1

- Given $Y = 0$:

$x \backslash Y$	0	1	$p_X(x Y = 0)$
0	0.06 / 0.6	0	0.1
1	0.54 / 0.6	0	0.9
			0.6 / 0.6

- Given $Y = 1$:

$x \backslash Y$	0	1	$p_X(x Y = 1)$
0	0	0.04 / 0.4	0.1
1	0	0.36 / 0.4	0.9
			0.4 / 0.4

- Since $p_X(x | Y = y) = p_X(x)$ for all possible x, y , the randomness in X is unaffected by Y , then X and Y are independent.

Independent Random Variables

- But there is an easier way to check if X and Y are independent:
- Fact:** Two RVs X and Y are independent whenever

$$p(x, y) = p_X(x)p_Y(y)$$

- Return to the example $p(x, y)$:

$X \backslash Y$	0	1	
0	0.06	0.04	
1	0.54	0.36	
			1

- The product of two PMFs $p_X(x)p_Y(y)$:

$X \backslash Y$	0	1	$p_X(x)$
0	0.6×0.1	0.4×0.1	0.1
1	0.6×0.9	0.4×0.9	0.9
$p_Y(y)$	0.6	0.4	1

- Note that two tables are **identical**.

Independent Random Variables

- **Summary:** Since the marginal PMFs are given by the row and column sums of the joint, this means X and Y are independent whenever each entry of the joint PMF equals the product of the corresponding row and column sum.
- If X and Y are **not** independent then they are said to be **dependent**.

Covariance

- **Motivation:** If two RVs X and Y are **not** independent (i.e., they are **dependent**), there are many ways to summarize how dependent they are. The most common is with covariance.
- **Definition:** If X and Y are two RVs then the covariance between X and Y is defined as

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) \quad (\text{easy to compute}),$$

$$\text{Cov}(X, Y) = E((X - E(X))(Y - E(Y))) \quad (\text{easy to understand}).$$

- **Interpretation:** It measures how much X and Y “co-vary together.”

Properties of Covariance

- We will see that Cov is very closely related to Var. Indeed notice the similarity of the definitions:

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y)$$

$$\text{Var}(X) = E(X \cdot X) - E(X)E(X) = E(X^2) - (E(X))^2.$$

- Notice therefore

$$\text{Cov}(X, X) = \text{Var}(X).$$

- If X and Y are independent, then

$$\text{Cov}(X, Y) = 0.$$

However, $\text{Cov}(X, Y) = 0$ does not imply the independence between X and Y .

Properties of Covariance

- **Formula:**

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y),$$

$$\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y) - 2\text{Cov}(X, Y).$$

- If X and Y are **independent**, then

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y),$$

$$\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y).$$

- **Example:**

1	2	0
2	1	1

Suppose $\begin{bmatrix} X_1 \\ Y_1 \end{bmatrix}$ and $\begin{bmatrix} X_2 \\ Y_2 \end{bmatrix}$ denote these two random tickets chosen with replacement from a box containing the following 3 tickets.

Properties of Covariance

- Selecting tickets with **replacement** implies that

$$\begin{array}{|c|} \hline X_1 \\ Y_1 \\ \hline \end{array}$$

and

$$\begin{array}{|c|} \hline X_2 \\ Y_2 \\ \hline \end{array}$$

- Question 1 (HW):** Find $\text{Var}(X_1 - Y_1)$.

Hint: Let $W = X_1 - Y_1$, then generate PMF function of W in a table form:

w	$p(w)$
\vdots	\vdots

- Question 2 (HW):** Find $\text{Var}(X_1 - Y_2)$.

Hint: $\text{Var}(X_1 - Y_2) = \text{Var}(X_1) + \text{Var}(Y_2) - 2\text{Cov}(X_1, Y_2)$. Note that X_1 and Y_2 are independent, so that $\text{Cov}(X_1, Y_2) = 0$. Then we have

$$\text{Var}(X_1 - Y_2) = \text{Var}(X_1) + \text{Var}(Y_2).$$

The problem boils down to find marginal PMFs of X_1 and Y_2 .

Properties of Covariance

- Selecting tickets with **replacement** implies that $\begin{bmatrix} X_1 \\ Y_1 \end{bmatrix}$ and $\begin{bmatrix} X_2 \\ Y_2 \end{bmatrix}$ are independent.
- **Question 3 (HW):** Are X_1 and Y_1 independent?

Hint: To determine the independence, we need to create joint PMF of X and Y and check each entry of the joint PMF equals the product of the corresponding row and column sum (Lecture Note page 17 – 18).