

Multiple Random Variables

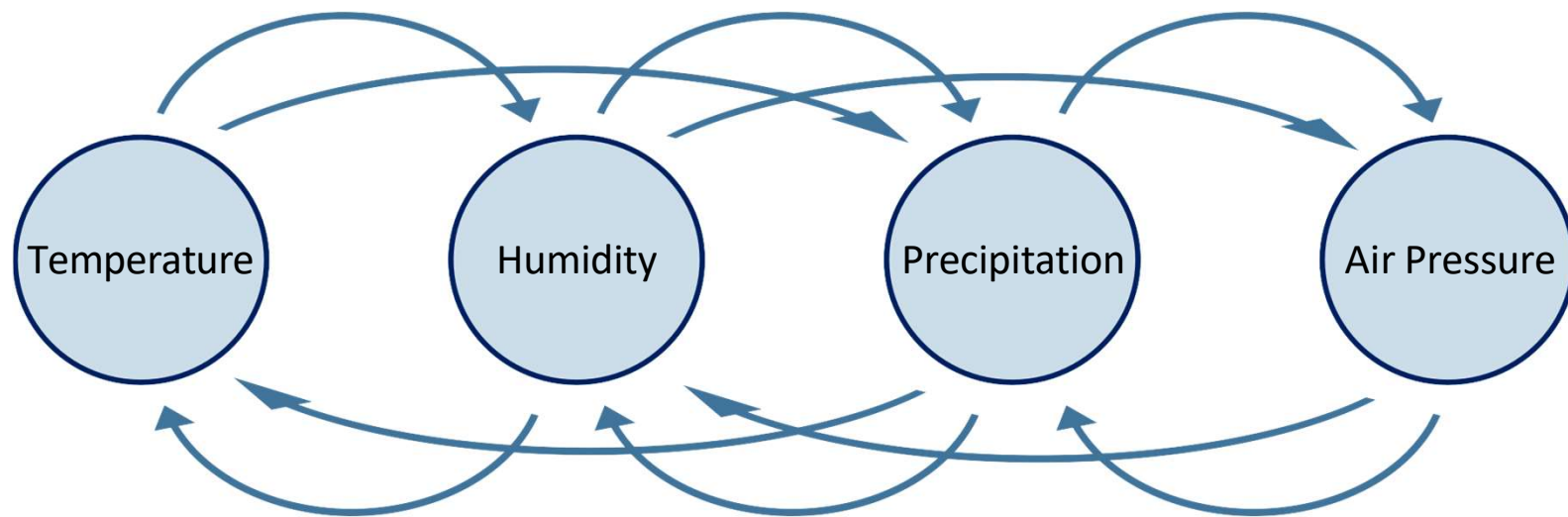
What Are We Missing?



The world is full of interesting probability problems...
...and many of them involve *multiple* random variables, being random *together*

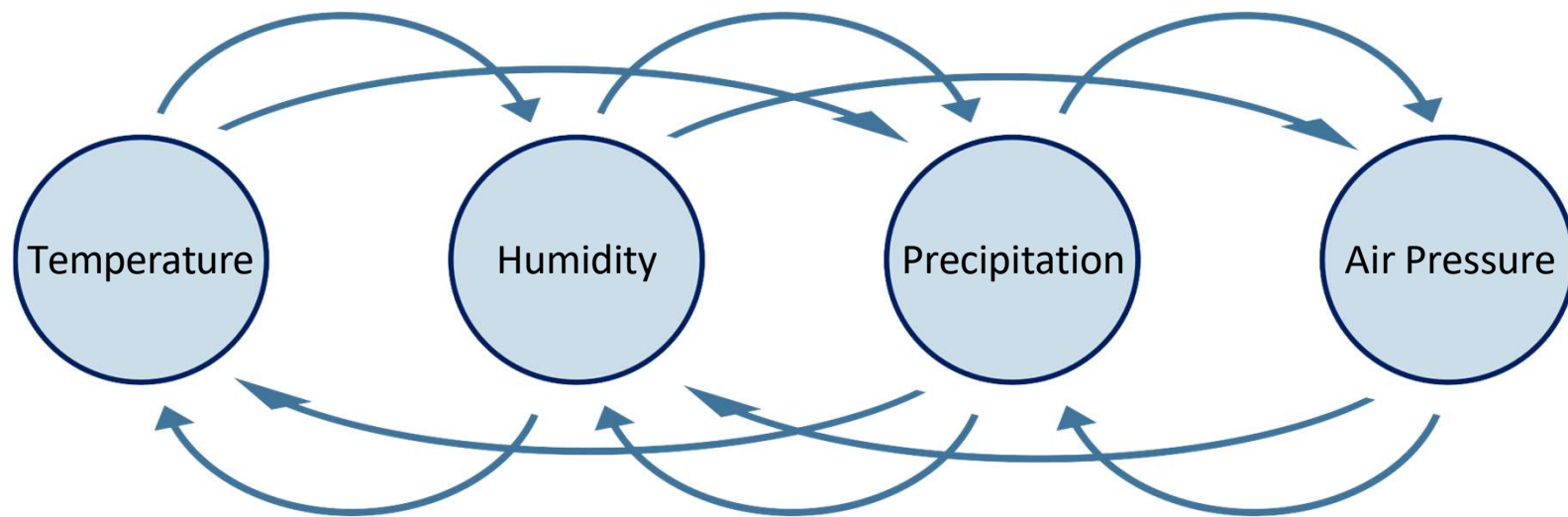
How Do We Model Multiple Random Variables Together?

Often, all the random variables involved are not independent of each other.



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So we can't just have a single distribution for each random variable — we need a way to talk about all the random variables at the same time.

The “Joint” Distribution of Multiple Random Variables

For *discrete* random variables X and Y , we have a **joint probability mass function**:

$$P(X = x, Y = y)$$

The joint is the “and” between an assignment to X , and an assignment to Y

The same as $P(A \text{ and } B)$ for events A and B !

The “Joint” Distribution of Multiple Random Variables

For *discrete* random variables X and Y , we have a **joint probability mass function**:

$$P(X = x, Y = y) \rightarrow 0.5134 \dots$$

$X = 2, Y = 4$ \rightarrow

$P(\underbrace{X = \text{male}}_{\text{gender}}, \underbrace{Y = 5.9 \text{ feet}}_{\text{height}})$

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The “Joint” Distribution of Multiple Random Variables

For *discrete* random variables X and Y , we have a **joint probability mass function**:

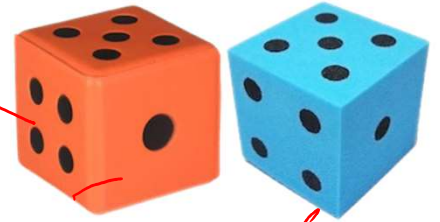
$$P(X = x, Y = y)$$

For *continuous* random variables, we have a **joint probability density function**:

$$f(X = x, Y = y) \quad P\{(X, Y) \in C\} = \iint_{(x,y) \in C} f(x, y) dx dy$$

Example Joint PMF: Two Dice

Roll two 6-sided dice, yielding values X and Y .

 X

random variable

$$P(X = 1)$$

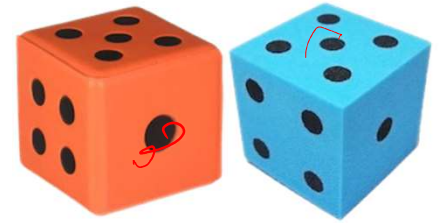
probability of
an event

$$P(X = k)$$

probability mass function

Example Joint PMF: Two Dice

Roll two 6-sided dice, yielding values X and Y .

 X

random variable

$$P(X = 1)$$

probability of
an event

$$P(X = k)$$

probability mass function

 X, Y

random variables

$$P(X = 1, Y = 6)$$

probability of the intersection
of two events

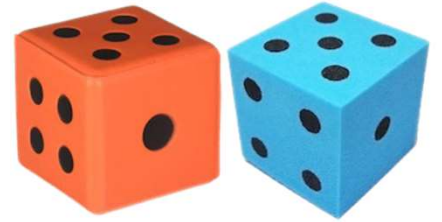
$$P(X = x, Y = y)$$

joint probability mass function

Example Joint PMF: Two Dice

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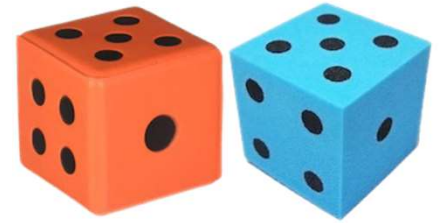
What is $P(X = x, Y = y)$?



Example Joint PMF: Two Dice

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What is $P(X = x, Y = y)$?



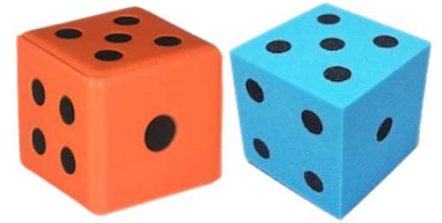
$$P(\underline{X = x}, \underline{Y = y}) = \frac{1}{\underline{\underline{36}}}$$

$$(x, y) \in \{(1,1), \dots, (6,6)\}$$

Example Joint PMF: Two Dice

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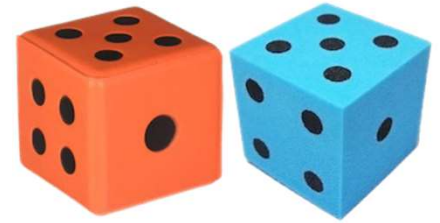
$$(x, y) \in \{(1,1), \dots, (6,6)\}$$

	X					
	1	2	3	4	5	6
1	1/36	1/36
2
3
4
5
6	1/36	1/36

Example Joint PMF: Two Dice

Roll two 6-sided dice, yielding values X and Y .

What is $P(X = x, Y = y)$?



$$P(X = x, Y = y) = \frac{1}{36}$$

$$(x, y) \in \{(1,1), \dots, (6,6)\}$$

		X					
		1	2	3	4	5	6
Y	1	1/36	1/36
	2
	3
	4
	5
	6	1/36	1/36

$$P(X = 4, Y = 3)$$

This is a **joint probability table**:
it contains the probabilities of all
possible outcomes for a set of
discrete random variables

Another Example

Example 4.3.a. Suppose that 3 batteries are randomly chosen from a group of 3 new, 4 used but still working, and 5 defective batteries. If we let X and Y denote, respectively, the number of new and used but still working batteries that are chosen, then the joint probability mass function of X and Y , $p(i, j) = P\{X = i, Y = j\}$, is given by

$$p(i, j) = \frac{\binom{3}{i} \binom{4}{j} \binom{5}{3-i-j}}{\binom{12}{3}}$$

$$p(0, 0) = \frac{\binom{5}{3}}{\binom{12}{3}} = 10/220$$

$$p(0, 1) = \frac{\binom{4}{1} \binom{5}{2}}{\binom{12}{3}} = 40/220$$

$$p(0, 2) = \frac{\binom{4}{2} \binom{5}{1}}{\binom{12}{3}} = 30/220$$

Table 4.1 $P\{X = i, Y = j\}$

$i \backslash j$	0	1	2	3	Row Sum $= P\{X = i\}$
0	$\frac{10}{220}$	$\frac{40}{220}$	$\frac{30}{220}$	$\frac{4}{220}$	$\frac{84}{220}$
1	$\frac{30}{220}$	$\frac{60}{220}$	$\frac{18}{220}$	0	$\frac{108}{220}$
2	$\frac{15}{220}$	$\frac{12}{220}$	0	0	$\frac{27}{220}$
3	$\frac{1}{220}$	0	0	0	$\frac{1}{220}$
Column Sums $= P\{Y = j\}$	$\frac{56}{220}$	$\frac{112}{220}$	$\frac{48}{220}$	$\frac{4}{220}$	

Handwritten notes and diagrams:

- Red arrows point from the formula to the table, indicating the correspondence between i and j in the binomial coefficients and the table rows and columns.
- Red circles highlight the values 3 in the table, specifically in the row for $i=2$ and column for $j=2$, and in the row for $i=3$ and column for $j=0$.
- Red text labels the columns as "# of new batteries" and "# of used batteries".
- A red box on the right contains a diagram of a 3x3 grid with circles, labeled "new" and "defective".
- Red text at the bottom left says "4 2 3" with arrows pointing to the first three rows of the table.

Example with continuous density

Example 4.3.c. The joint density function of X and Y is given by

$$f(x, y) = \begin{cases} 2e^{-x}e^{-2y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$C \equiv \{X > 1, Y < 1\}$$

$$P\{X > 1, Y < 1\} = \int_0^1 \int_1^\infty 2e^{-x}e^{-2y} dx dy$$

$$= \int_0^1 2e^{-2y} (-e^{-x} \big|_1^\infty) dy$$

$$P(X < Y) = \int \int f(x, y) dy$$

y x

$$C \equiv \{X < Y\}$$

$$P\{X < Y\} = \iint_{(x,y): x < y} 2e^{-x}e^{-2y} dx dy$$

$$= \int_0^\infty \int_0^y 2e^{-x}e^{-2y} dx dy$$

$$= \int_0^\infty 2e^{-2y} (1 - e^{-y}) dy$$

$$= \int_0^\infty 2e^{-2y} dy - \int_0^\infty 2e^{-3y} dy$$

$$= 1 - \frac{2}{3}$$

$$= \frac{1}{3}$$

Marginals

$$\begin{aligned}\underline{P\{X < a\}} &= \int_0^a \int_0^\infty 2e^{-2y} e^{-x} dy dx \\ &= \int_0^a e^{-x} dx \\ &= \underline{1 - e^{-a}} \quad \blacksquare\end{aligned}$$

$$P(X < a) \sim \exp(\tau = 1)$$

Law of total probability

Joint table expresses the complete information about the random variables

$$\underline{P(X = x)} = \sum_{\underline{y}} P(X = x, Y = y)$$

$P(X = x)$ is called the marginal of the joint distribution $P(X, Y)$

$$f(x) = \int_y f(x, y) dy$$

Independent Random Variables

The random variables X and Y are said to be independent if for any two sets of real numbers A and B

$$P\{X \in A, Y \in B\} = P\{X \in A\}P\{Y \in B\} \quad (4.3.7)$$

This also implies that

$$P(X \leq a, Y \leq b) = P(X \leq a)P(Y \leq b)$$

$$\text{Or } F_{X,Y}(a, b) = F_X(a)F_Y(b)$$

In the jointly continuous case, the condition of independence is equivalent to

$$\underline{f(x, y)} = \underline{f_X(x)f_Y(y)} \quad \text{for all } x, y$$

Example 4.3.d. Suppose that X and Y are independent random variables having the common density function

$$f(x) = \begin{cases} e^{-x} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$f(y) = \begin{cases} e^{-y} & y > 0 \\ 0 & \text{otherwise} \end{cases}$$

Find the density function of the random variable X/Y .

$$f(z = \frac{x}{y})$$

$$P(z \leq a) = P\left(\frac{x}{y} \leq a\right) = P(x \leq ay) = \int_0^\infty \int_0^{ay} f(x,y) dx dy$$

$$= \int_0^\infty e^{-y} \int_0^{ay} e^{-x} dx dy = F_2(a) = \frac{a}{a+1}$$

$$f(z) = \frac{\partial}{\partial a} F_2(a)$$

$$f_z(z) = f_x(T^{-1}(z)) \frac{\partial T^{-1}}{\partial z}$$

$z = T(x) \quad f_x(x)$

Conditional Probability

Given two discrete random variables X, Y . The conditional probability of X given a specific value of Y is given as:

$$P(X = x | Y = y) = P(X = x, Y = y) / P(Y = y)$$

For continuous variables with joint density of X, Y as $f(x, y)$:

$$f_{X|Y}(x|y) = f(x, y) / f(y)$$

$$\begin{aligned} f_{X|Y}(x|y) dx &= \frac{f(x, y) dx dy}{f_Y(y) dy} \\ &\approx \frac{P\{x \leq X \leq x + dx, y \leq Y \leq y + dy\}}{P\{y \leq Y \leq y + dy\}} \\ &= P\{x \leq X \leq x + dx | y \leq Y \leq y + dy\} \end{aligned}$$

Example 4.3.h. The joint density of X and Y is given by

$$f(x, y) = \begin{cases} \frac{12}{5}x(2-x-y) & 0 < x < 1, 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

Compute the conditional density of X , given that $Y = y$, where $0 < y < 1$.

Solution. For $0 < x < 1, 0 < y < 1$, we have

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f(x, y)}{f_Y(y)} \\ &= \frac{f(x, y)}{\int_{-\infty}^{\infty} f(x, y) dx} \\ &= \frac{x(2-x-y)}{\int_0^1 x(2-x-y) dx} \\ &= \frac{x(2-x-y)}{\frac{2}{3} - y/2} \\ &= \frac{6x(2-x-y)}{4-3y} \end{aligned}$$

$$f_Y(y) = \int_{x=0}^1 f(x, y) dx$$

Joint distribution of n random variables

If X_1, X_2, \dots, X_n are n random variables. Their joint distribution is defined for the discrete case as

$$p(x_1, x_2, \dots, x_n) = P\{X_1 = x_1, X_2 = x_2, \dots, X_n = x_n\}$$

Further, the n random variables are said to be jointly continuous if there exists a function $f(x_1, x_2, \dots, x_n)$, called the joint probability density function, such that for any set C in n -space

$$P\{(X_1, X_2, \dots, X_n) \in C\} = \int \int_{(x_1, \dots, x_n) \in C} \dots \int f(x_1, \dots, x_n) dx_1 dx_2 \dots dx_n$$

In particular, for any n sets of real numbers A_1, A_2, \dots, A_n

$$\begin{aligned} &P\{X_1 \in A_1, X_2 \in A_2, \dots, X_n \in A_n\} \\ &= \int_{A_n} \int_{A_{n-1}} \dots \int_{A_1} f(x_1, \dots, x_n) dx_1 dx_2 \dots dx_n \end{aligned}$$

Example 4.3.e. Suppose that the successive daily changes of the price of a given stock are assumed to be independent and identically distributed random variables with probability mass function given by

$$P\{\text{daily change is } i\} = \begin{cases} -3 & \text{with probability .05} \\ -2 & \text{with probability .10} \\ -1 & \text{with probability .20} \\ 0 & \text{with probability .30} \\ 1 & \text{with probability .20} \\ 2 & \text{with probability .10} \\ 3 & \text{with probability .05} \end{cases}$$

$P(X_i)$

X_i

Then the probability that the stock's price will increase successively by 1, 2, and 0 points in the next three days is

$$P\{X_1 = 1, X_2 = 2, X_3 = 0\} = (.20)(.10)(.30) = .006$$

where we have let X_i denote the change on the i th day. ■