## Chapter 5: JOINT PROBABILITY DISTRIBUTIONS

## Part 1: Joint Discrete Probability Distributions... <br> Marginal Distributions <br> Conditional Distributions Independence

Sections 5-1.1 to 5-1.4

Recall a discrete probability distribution (or probability mass function)

$$
\begin{array}{l|l|l|l}
x & 0 & 1 & 2 \\
\hline f(x) & 0.50 & 0.20 & 0.30
\end{array}
$$

Sometimes we're simultaneously interested in two or more discrete variables in a random experiment.

## Examples

- Year in college vs. Number of credits taken
- Count of plants grown in a tray vs. Count of healthy plants
- Number of cigarettes smoked per day vs. Age of cancer onset

In general, if $X$ and $Y$ are two random variables, the probability distribution that defines their simultaneous behavior is called a joint probability distribution.

If $X$ and $Y$ are discrete, this distribution can be described with a joint probability mass function (this section).

If $X$ and $Y$ are continous, this distribution can be described with a joint probability density function (next section).

- Example: Plastic covers for CDs


Measurements for the length and width of a rectangular plastic covers for CDs are rounded to the nearest $m m$ (so they are discrete).

Let $X$ denote the length and
$Y$ denote the width.

The possible values of $X$ are 129, 130, and 131 mm . The possible values of $Y$ are 15 and 16 mm .

Both $X$ and $Y$ are discrete.

There are 6 possible pairs $(X, Y)$.
We show the probability for each pair in the following table:

The sum of all the probabilities is 1.0 .
The combination with the highest probability is $(130,15)$.

The combination with the lowest probability is $(131,16)$.

The joint probability mass function is the function $f_{X Y}(x, y)=P(X=x, Y=y)$. For example, we have $f_{X Y}(129,15)=0.12$.

If we are given a joint probability distribution for $X$ and $Y$, we can obtain the individual probability distribution for $X$ or for $Y \ldots$

- Example: Continuing plastic covers for CDs

Find the probability that a CD cover has length of 129 mm (i.e. $\mathrm{X}=129$ ).

$$
\begin{gathered}
\mathrm{x}=\text { length } \\
\begin{array}{|l|lll|} 
& 129 & 130 & 131 \\
\cline { 1 - 4 } & \mathrm{y}=\text { width } \\
15 & \mathbf{0 . 1 2} & 0.42 & 0.06 \\
16 & \mathbf{0 . 0 8} & 0.28 & 0.04 \\
\hline
\end{array} \\
\begin{aligned}
P(X=129) & =\quad P(X=129 \text { and } Y=15) \\
& +P(X=129 \text { and } Y=16) \\
& =0.12+0.08=0.20
\end{aligned}
\end{gathered}
$$

What is the probability distribution of $X$ ?

|  | $\mathrm{x}=$ length |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | 129 | 130 | 131 |
| $\mathrm{y}=$ width | 15 | 0.12 | 0.42 | 0.06 |
|  | 16 | 0.08 | 0.28 | 0.04 |
| column totals |  | 0.20 |  | 0.10 |

The probability distribution for $X$ appears in the column totals...

$$
\begin{array}{l|l|l|l}
x & 129 & 130 & 131 \\
\hline f_{X}(x) & 0.20 & 0.70 & 0.10
\end{array}
$$

* NOTE: We've used a subscript $X$ in the probability mass function of X , or $f_{X}(x)$, for clarification since we're considered more than one variable at a time now.

We can do the same for the $Y$ random variable.
row

|  | $\mathrm{x}=$ length |  |  | totals |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 129 | 130 | 131 |  |  |
|  | $\mathrm{y}=$ width | 15 | 0.12 | 0.42 | 0.06 | $\mathbf{0 . 6 0}$ |
|  | 16 | 0.08 | 0.28 | 0.04 | $\mathbf{0 . 4 0}$ |  |
|  | $\mathbf{0 . 2 0}$ | $\mathbf{0 . 7 0}$ | $\mathbf{0 . 1 0}$ | $\mathbf{1}$ |  |  |
|  |  |  |  |  |  |  |

$$
\begin{array}{l|l|l}
y & 15 & 16 \\
\hline f_{Y}(y) & 0.60 & 0.40
\end{array}
$$

Because the the probability mass functions for $X$ and $Y$ appear in the margins of the table (i.e. column and row totals), they are often referred to as the marginal distributions for $X$ and $Y$.

When there are two random variables of interest, we also use the term bivariate probability distribution or bivariate distribution to refer to the joint distribution.

## - Joint Probability Mass Function

The joint probability mass function of the discrete random variables $X$ and $Y$, denoted as $f_{X Y}(x, y)$, satisfies
(1) $f_{X Y}(x, y) \geq 0$
(2) $\sum_{x} \sum_{y} f_{X Y}(x, y)=1$
(3) $f_{X Y}(x, y)=P(X=x, Y=y)$

## - Marginal Probability Mass Function

If $X$ and $Y$ are discrete random variables with joint probability mass function $f_{X Y}(x, y)$, then the marginal probability mass functions of $X$ and $Y$ are

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

and

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

where the sum for $f_{X}(x)$ is over all points in the range of $(X, Y)$ for which $X=x$ and the sum for $f_{Y}(y)$ is over all points in the range of $(X, Y)$ for which $Y=y$.

When asked for $E(X)$ or $V(X)$ in a joint probability distribution problem, first calculate the marginal distribution $f_{X}(x)$ and work it as we did in chapter 3 for the univariate case (i.e. for a single random variable).

- Example: Batteries

Suppose that 2 batteries are randomly chosen without replacement from the following group of 12 batteries:

3 new
4 used (working)
5 defective

Let $X$ denote the number of new batteries chosen.

Let $Y$ denote the number of used batteries chosen.
a) Find $f_{X Y}(x, y)$
\{i.e. the joint probability distribution\}.

ANS:
Though $X$ can take on values 0,1 , and 2 , and $Y$ can take on values 0,1 , and 2 , when we consider them jointly, $X+Y \leq 2$. So, not all combinations of $(X, Y)$ are possible.

CASE: no new, no used (so all defective)

$$
f_{X Y}(0,0)=\frac{\binom{5}{2}}{\binom{12}{2}}=10 / 66
$$

CASE: no new, 1 used

$$
f_{X Y}(0,1)=\frac{\binom{4}{1}\binom{5}{1}}{\binom{12}{2}}=20 / 66
$$

CASE: no new, 2 used

$$
f_{X Y}(0,2)=\frac{\binom{4}{2}}{\binom{12}{2}}=6 / 66
$$

CASE: 1 new, no used

$$
f_{X Y}(1,0)=\frac{\binom{3}{1}\binom{5}{1}}{\binom{12}{2}}=15 / 66
$$

CASE: 2 new, no used

$$
f_{X Y}(2,0)=\frac{\binom{3}{2}}{\binom{12}{2}}=3 / 66
$$

CASE: 1 new, 1 used

$$
f_{X Y}(1,1)=\frac{\binom{3}{1}\binom{4}{1}}{\binom{12}{2}}=12 / 66
$$

| $\mathrm{x}=$ number of new chosen |  |  |  |  |
| :---: | :---: | :---: | :--- | :--- |
|  |  | 0 | 1 | 2 |
| $\mathrm{y}=$ number of | 0 | $10 / 66$ | $15 / 66$ | $3 / 66$ |
| used | 1 | $20 / 66$ | $12 / 66$ |  |
| chosen | 2 | $6 / 66$ |  |  |
|  |  |  |  |  |

There are 6 possible ( $\mathrm{X}, \mathrm{Y}$ ) pairs.
And, $\sum_{x} \sum_{y} f_{X Y}(x, y)=1$.

## Conditional Probability Distributions

As we saw before, we can compute the conditional probability of an event given information of another event.

As stated before,

$$
P(A \mid B)=\frac{P(A \cap B)}{P(B)}
$$

- Example: Continuing the plastic covers...

| $\mathrm{y}=$ width | $\mathrm{x}=$ length |  |  |  |  | totals |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 129 | 9 | 130 | 131 |  |
|  | 15 | 0.1 | 2 | 0.42 | 0.06 | 0.60 |
|  | 16 | 0.0 | . 08 | 0.28 | 0.04 | 0.40 |
| column totals |  |  | 20 | 0.70 | 0.10 | 1 |

a) Find the probability that a CD cover has a length of 130 mm GIVEN the width is 15 mm .

| $\mathrm{y}=$ width | $\mathrm{x}=$ length |  |  |  | $\begin{array}{r} \text { row } \\ \text { totals } \\ \hline \end{array}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 129 | 130 | 131 |  |
|  | 15 | 0.12 | 0.42 | 0.06 | 0.60 |
|  | 16 | 0.08 | 0.28 | 0.04 | 0.40 |
| column totals |  | 0.20 | 0.70 | 0.10 | 1 |

$$
\text { ANS: } \begin{aligned}
P(X=130 \mid Y & =15)=\frac{P(X=130, Y=15)}{P(Y=15)} \\
& =0.42 / 0.60=0.70
\end{aligned}
$$

b) Find the conditional distribution of $X$ given $Y=15$.

$$
\begin{aligned}
& P(X=129 \mid Y=15)=0.12 / 0.60=0.20 \\
& P(X=130 \mid Y=15)=0.42 / 0.60=0.70 \\
& P(X=131 \mid Y=15)=0.06 / 0.60=0.10
\end{aligned}
$$

Once you're GIVEN that $Y=15$, you're in a 'different space'.

We are now considering only the CD covers with a width of 15 mm . For this subset of the covers, how are the lengths $(X)$ distributed.

The conditional distribution of $X$, or $f_{X \mid Y}(x)$, given $Y=15$ :

$$
\begin{array}{l|ccc}
x & 129 & 130 & 131 \\
\hline f_{X \mid 15}(x) & 0.20 & 0.70 & 0.10
\end{array}
$$

Notice that the sum of these probabilities is 1 , and this is a legitimate probability distribution .

* NOTE: Again, we use the subscript $X \mid Y$ for clarity to denote that this is a conditional distribution.


## - Conditional Probability Mass Function

Given discrete random variables X and Y with joint probability mass function $f_{X Y}(x, y)$ the conditional probability mass function of Y given $\mathrm{X}=\mathrm{x}$ is

$$
f_{Y \mid X}(y)=\frac{f_{X Y}(x, y)}{f_{X}(x)} \quad \text { for } \quad f_{X}(x)>0
$$

The conditional probability is the joint probability over the marginal probability.

Notice that we can define $f_{X \mid Y}(x)$ in a similar manner if we are interested in that conditional distribution.

Because a conditional probability mass function $f_{Y \mid X}(y)$ is a probability mass function, the following properties are satisfied:
(1) $f_{Y \mid X}(y) \geq 0$
(2) $\sum_{y} f_{Y \mid X}(y)=1$
(3) $f_{Y \mid X}(y)=P(Y=y \mid X=x)$

- Conditional Mean and Variance

The conditional mean of Y given $\mathrm{X}=\mathrm{x}$, denoted as $E(Y \mid x)$ or $\mu_{Y \mid x}$ is

$$
\begin{aligned}
E(Y \mid x) & =\sum_{y} y f_{Y \mid X}(y) \\
& =\mu_{Y \mid x}
\end{aligned}
$$

and the conditional variance of Y given $\mathrm{X}=\mathrm{x}$, denoted as $V(Y \mid x)$ or $\sigma_{Y \mid x}^{2}$ is

$$
\begin{aligned}
V(Y \mid x) & =\sum_{y}\left(y-\mu_{Y \mid x}\right)^{2} f_{Y \mid X}(y) \\
& =\sum_{y} y^{2} f_{Y \mid X}(y)-\mu_{Y \mid x}^{2} \\
& =E\left(Y^{2} \mid x\right)-[E(Y \mid x)]^{2} \\
& =\sigma_{Y \mid x}^{2}
\end{aligned}
$$

- Example: Continuing the plastic covers...

| $\mathrm{y}=$ width | $\mathrm{x}=$ length |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 129 | 130 | 131 |  |
|  | 15 | 0.12 | 0.42 | 0.06 | 0.60 |
|  | 16 | 0.08 | 0.28 | 0.04 | 0.40 |
| column totals |  | 0.20 | 0.70 | 0.10 | 1 |

a) Find the $E(Y \mid X=129)$ and

$$
V(Y \mid X=129)
$$

ANS:
We need the conditional distribution first...

$$
\begin{array}{r|rr}
y & 15 & 16 \\
\hline f_{Y \mid 129}(y) &
\end{array}
$$

## Independence

As we saw earlier, sometimes, knowledge of one event does not give us any information on the probability of another event.

Previously, we stated that if $A$ and $B$ were independent, then

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

In the framework of probability distributions, if $X$ and $Y$ are independent random variables, then $f_{Y \mid X}(y)=f_{Y}(y)$.

## - Independence

For discrete random variables $X$ and $Y$, if any of the following properties is true, the others are also true, and $X$ and $Y$ are independent.
(1) $f_{X Y}(x, y)=f_{X}(x) f_{Y}(y) \quad$ for all x and y
(2) $f_{Y \mid X}(y)=f_{Y}(y)$
for all x and y with $f_{X}(x)>0$
(3) $f_{X \mid Y}(x)=f_{X}(x)$
for all x and y with $f_{Y}(y)>0$
(4) $P(X \in A, Y \in B)=P(X \in A) \cdot P(Y \in B)$ for any sets $A$ and $B$ in the range of X and Y .

Notice how (1) leads to (2):

$$
f_{Y \mid X}(y)=\frac{f_{X Y}(x, y)}{f_{X}(x)}=\frac{f_{X}(x) f_{Y}(y)}{f_{X}(x)}=f_{Y}(y)
$$

- Example: Continuing the battery example

Two batteries were chosen without replacement.

Let $X$ denote the number of new batteries chosen.

Let $Y$ denote the number of used batteries chosen.

|  | $\mathrm{x}=$ number of new chosen |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | 0 | 1 | 2 |
| $\mathrm{y}=$ number |  |  |  |  |
| of used | 0 | $10 / 66$ | $15 / 66$ | $3 / 66$ |
| chosen | 1 | $20 / 66$ | $12 / 66$ |  |
| 2 | $2 / 66$ |  |  |  |
|  |  |  |  |  |

a) Without doing any calculations, can you tell whether $X$ and $Y$ are independent?

- Example: Independent random variables

Consider the random variables $X$ and $Y$, which both can take on values of 0 and 1 .

|  | X |  | $\begin{array}{r} \text { row } \\ \text { totals } \end{array}$ |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | 0 | 1 |  |
| y | 0 | 0.40 | 0.10 | 0.50 |
|  | 1 | 0.40 | 0.10 | 0.50 |
| column totals |  | 0.80 | 0.20 | 1 |

a) Are $X$ and $Y$ independent?

$$
\begin{array}{r|rr}
y & 0 & 1 \\
\hline f_{Y \mid 0}(y) & &
\end{array}
$$

$$
\begin{array}{r|ll}
y & 0 & 1 \\
\hline f_{Y \mid 1}(y) & &
\end{array}
$$

Does $f_{Y \mid X}(y)=f_{Y}(y)$ for all $\mathrm{x} \& \mathrm{y}$ ?

Does $f_{X Y}(x, y)=f_{X}(x) f_{Y}(y)$ for all $\mathrm{x} \& \mathrm{y}$ ?

|  | x |  | totals |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | 0 | 1 |  |
| y | y 0 | 0.40 | 0.10 | 0.50 |
|  | 1 | 0.40 | 0.10 | 0.50 |
| column totals |  | 0.80 | 0.20 | 1 |

i.e. Does $P(X=x, Y=y)$

$$
=P(X=x) \cdot P(Y=y) ?
$$

