## CSE 312 <br> Foundations of Computing II

## Lecture 14: Joint Distributions

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## Agenda

- Cartesian Products
- Joint PMFs/PDFs/CDFs and Joint Range
- Independence
- Marginal Distributions
- Expectation
- Joint Continuous Random Variables


## Why joint distributions?

- Given all of its User's ratings for different movies, and any preferences you have expressed,', Netflix wants to recommend a new movie for you.

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- Given a bunch of medical data correlating symptoms and personal history with diseases, predict what is ailing a person with a particular medical history and set of symptoms.
- Given current traffic, pedestrian locations, weather, lights, etc. decide whether a self-driving car should slow down or come to a stop.


## Review Cartesian Product

Definition. Let $A$ and $B$ be sets. The Cartesian product of $A$ and $B$ is denoted

$$
A \times B=\left\{(a, b): a \varrho_{\natural}^{\kappa} A, b \in B\right\}
$$

Example. $\stackrel{A}{d} \quad \stackrel{B}{d} \quad|A \times B|=6$

$$
\frac{\{1,2,3\}}{\| A=3} \times \frac{\{4,5\}}{\mid B-2}=\{(1,4),(1,5),(2,4),(2,5),(3,4),(3,5)\}
$$

If $A$ and $B$ are finite sets, then $|A \times B|=|A| \cdot|B|$.
The sets don't need to be finite! You can have $\frac{\mathbb{R} \times \mathbb{R}\left(\text { often denoted } \mathbb{R}^{2}\right)}{L^{\swarrow}}$

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## Joint PMFs and Joint Range

$\Omega(x):\{1,2,3\}$
$\Omega(y)\{4,5\}$
Definition. Let $X$ and $Y$ be discrete random variables. The Joint PMF of $X$ and $Y$ is

$$
\stackrel{\left.\stackrel{\downarrow}{p_{X, Y}(a, b}\right)}{\underline{b}} \mathbf{\operatorname { P r } ( X \stackrel { \checkmark } { = } , Y = b )}
$$

Definition. The joint range of $p_{X, Y}$ is

$$
\left.\underline{\Omega(X, Y})=\left\{\underline{(c, d)} \underline{c}^{4}\right): \underline{p_{X, Y}}(c, d)>0\right\} \subseteq \Omega(X) \times \Omega(Y)
$$

Note that


Example: Weird Dice $P_{x, y}(1,1)=\mathbb{P}(x=1, y=1)$

$$
=\mathbb{T}(x=1) \pi(/=1)=\frac{1}{4}
$$

Suppose I roll two fair 4 -sided die independently. Let $X$ be the value of the first die, and $Y$ be the value of the second die.

$$
\underline{\Omega(X)}=\{\underline{1,2,3,4} \mathbf{\}} \text { and } \underline{\Omega(Y)}=\{\underline{1,2,3,4\}}
$$

In this problem, the joint PMF is

$$
p_{X, Y}(x, y)=\left\{\begin{array}{cc}
1 / 16, & x, y \in \Omega(X, Y) \\
0, & \text { otherwise }
\end{array}\right.
$$


and the joint range is (since all combinations have non-zero probability) $\Omega(X, Y)=\Omega(X) \times \Omega(Y)$

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Independence $\quad \begin{aligned} \quad \mathbb{T}_{X, Y}(X, Y) & =\mathbb{P}(X=x, Y=y) \\ & =\mathbb{P}(X: x) \cdot \mathbb{P}(Y=y)\end{aligned}$
Definition. Let $X$ and $Y$ be discrete random variables. The Joint PMF of $X$ and $Y$ is

$$
p_{X, Y}(a, b)=\operatorname{Pr}(X=a, Y=b)
$$

Definition. The joint range of $p_{X, Y}$ is

$$
\Omega(X, Y)=\left\{(c, d): p_{X, Y}(c, d)>0\right\} \subseteq \Omega(X) \times \Omega(Y)
$$

Definition. $X$ and $Y$ are independent iff for all $a, b$

$$
\operatorname{Pr}(X=a, Y=b)=\operatorname{Pr}(X=a) \cdot \operatorname{Pr}(Y=b) \quad \Omega_{x, y}=\Omega_{x} X \Omega_{y}
$$

Example: Weirder Dice $\quad(1,3)$

Suppose I roll two fair 4 -sided die independently. Let $X$ be the value of the first die, and $Y$ be the value of the second die. Let $U=\min (X, Y)$ and $W=\max (X, Y)$

$$
\begin{aligned}
& \Omega(U)=\{1,2,3,4\} \text { and } \Omega(W)=\{1,2,3,4\} \quad\left(\begin{array}{c}
u, \\
v \\
v \\
w
\end{array}\right. \\
& \frac{\Omega(U, W)}{\zeta}=\left\{\left(\frac{u, w)}{\prec}\right) \in \Omega(U) \times \Omega(W): u \leq w\right\} \neq \Omega(U) \times \Omega(W)
\end{aligned}
$$



## Example: Weirder Dice



Suppose I roll two fair 4 -sided die independently. Let $X$ be the value of the first die, and $Y$ be the value of the second die. Let $U=\min (X, Y)$ and $W=\max (X, Y)$
$\Omega(U)=\{1,2,3,4\}$ and $\Omega(W)=\{1,2,3,4\}$
$\Omega(U, W)=\{(u, w) \in \Omega(U) \times \Omega(W): u \leq w\} \neq \Omega(U) \times \Omega(W)$

The joint PMF $p_{U, W}(u, w)=\operatorname{Pr}(U=u, W=w)$ is
$p_{U, W}(u, w)=\left\{\begin{array}{rc}2 / 16, & (u, w) \in \Omega(U) \times \Omega(W) \\ 1 / 16, & (u, w) \in \Omega(U) \times \Omega(W) \\ 0, & \text { where } w>u \\ 0, & \text { otherwise }=u\end{array}\right.$

| Ulw | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 0 | $1 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{3}$ | 0 | 0 | $1 / 16$ | $2 / 16$ |
| $\mathbf{4}$ | 0 | 0 | 0 | $1 / 16$ |

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Example: Weirder Dice

Suppose I roll two fair 4-sided die independently. Let $X$ be the value of the first die, and $Y$ be the value of the second die. Let $U=\min (X, Y)$ and $W=\max (X, Y)$

Suppose we didn't know how to compute $\operatorname{Pr}_{\varepsilon_{1}}(U=u)$ directly. Can we figure it out if we know $p_{U, W}(u, w)$ ? $\quad \mathbb{P}(u=2)=\sum_{\omega=1} \mathbb{P}(u=2 n w=\omega)$

| Ulw | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| :---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 0 | $1 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{3}$ | 0 | 0 | $1 / 16$ | $2 / 16$ |
| $\mathbf{4}$ | 0 | 0 | 0 | $1 / 16$ |

## Example: Weirder Dice

Suppose I roll two fair 4 -sided die independently. Let $X$ be the value of the first die, and $Y$ be the value of the second die. Let $U=\min (X, Y)$ and $W=\max (X, Y)$

Suppose we didn't know how to compute $\operatorname{Pr}(U=u)$ directly. Can we figure it out if we know $p_{U, W}(u, w)$ ?

$$
\begin{aligned}
& p_{U}(u)= \begin{cases}7 / 16, & u=1 \\
5 / 16, & u=2 \\
3 / 16, & u=3 \\
1 / 16, & u=4\end{cases} \\
& \mathbb{P}_{u}(3)=\sum_{w \in \Omega(w)=P_{v, \omega}(3, w)}(3,1)+\ldots
\end{aligned}
$$

| Uaw | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 0 | $1 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{3}$ | 0 | 0 | $1 / 16$ | $2 / 16$ |
| $\mathbf{4}$ | 0 | 0 | 0 | $1 / 16$ |

## Marginal PMF

Definition. Let $X$ and $Y$ be discrete random variables and $p_{X, Y}(a, b)$ their joint PMF. The marginal PMF of $X$

$$
p_{X}(a)=\sum_{\underline{b \in \Omega(Y)}} p_{X, Y}(a, b)
$$

Similarly, $p_{Y}(b)=\sum_{a \in \Omega(X)} p_{X, Y}(a, b)$



Marginal PMF Example

Suppose the table below gives us the joint mf of X and Y .

What is the marginal mf of $X$ ? What is the marginal mf of Y ?
Are $X$ and $Y$ independent?

$$
\begin{aligned}
& P_{x}(x)=\sum_{y \in \Omega_{y}} P_{x, y}(x, y) \\
& P_{x}(1)=P_{x, y}(1,1)+P_{x, y}(1,2) \\
& 0,5
\end{aligned}
$$

$$
\begin{array}{rl}
P_{x}(2) & =P_{x, y}(2,1)+P_{x, y}(2,2) \\
0.1 & 0.0-4 \\
& =0.5
\end{array}
$$



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## Joint Expectation

Definition. Let $X$ and $Y$ be discrete random variables and $p_{X, Y}(a, b)$ their joint PMF. The expectation of some function $g(x, y)$ with inputs $X$ and $Y$

$$
\begin{aligned}
& E[g(X, Y)]=\sum_{\substack{a \in \Omega(X)}} g(a, b) p_{X, Y}(a, b) \\
& \overline{\bar{c}}\left[g(x)=\sum_{r \in \Omega_{x}} g(x) \cdot \mathbb{P}(x=r)\right.
\end{aligned}
$$

Expectation Example

Suppose the table below gives us the joint mf of $X$ and $Y$.

$$
\begin{aligned}
& \text { What is } E(X Y) ? \\
& \begin{aligned}
& E[X Y]: \sum_{x \in \Omega_{x}} \sum_{y \in \Omega y} g(x, 1): x y \\
&=1 \cdot 1 \cdot 0.4=0.4+0.2+0.2+1.6 \\
&+1 \cdot 2 \cdot 0.1 \\
&+2 \cdot 1 \cdot 0.1 \\
&+2 \cdot 2 \cdot 0.7
\end{aligned}
\end{aligned}
$$



- Suppose the number of requests $Z$ to a particular web server per hour is Poisson( $\lambda$ ). And that the request comes from within the US with probability $p$.
- Let $X$ be the number of requests per hour from the US and let $Y$ be the number of requests per hour from outside the US. What is the joint pmf of $X$ and $Y$ ? Are they independent?
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|  | Discrete | Continuous |
| :--- | :--- | :--- |
| Joint PMF/PDF | $p_{X, Y}(x, y)=\mathbb{P}(X=x, Y=y)$ | $f_{X, Y}(x, y) \neq \mathbb{P}(X=x, Y=y)$ |
| Joint range/support | $\left\{(x, y) \in \Omega_{X} \times \Omega_{Y}: p_{X, Y}(x, y)>0\right\}$ | $\left\{(x, y) \in \Omega_{X} \times \Omega_{Y}: f_{X, Y}(x, y)>0\right\}$ |
| $\Omega_{X, Y}$ | $\left\{(x, y)(x, y)=\sum_{t \leq x, s \leq y} p_{X, Y}(t, s)\right.$ | $F_{X, Y}(x, y)=\int_{-\infty}^{x} \int_{-\infty}^{y} f_{X, Y}(t, s) d s d t$ |
| Joint CDF | $F_{X, Y}$ | $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X, Y}(x, y) d x d y=1$ |
| Normalization | $\sum_{x, y} p_{X, Y}(x, y)=1$ | $f_{X}(x)=\int_{-\infty}^{\infty} f_{X, Y}(x, y) d y$ |
| Marginal PMF/PDF | $p_{X}(x)=\sum_{y} p_{X, Y}(x, y)$ | $\mathbb{E}[g(X, Y)]=\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X, Y}(x, y) d x d y$ |
| Expectation | $\mathbb{E}[g(X, Y)]=\sum_{x, y} g(x, y) p_{X, Y}(x, y)$ |  |

## Independence (continuous random variables)

Definition. Let $X$ and $Y$ be continuous random variables. The joint pdf of $X$ and $Y$ is

$$
f_{X, Y}(a, b) \neq \operatorname{Pr}(X=a, Y=b)
$$

Definition. The joint range of $p_{X, Y}$ is

$$
\Omega(X, Y)=\left\{(c, d): p_{X, Y}(c, d)>0\right\} \subseteq \Omega(X) \times \Omega(Y)
$$

Definition. $X$ and $Y$ are independent iff for all $a, b$
$f_{X, Y}(a, b)=f_{X}(a) \cdot f_{Y}(b)$

- Suppose that the surface of a disk is a circle with area R centered at the origin and that there is a single point imperfection at a location with is uniformly distributed across the surface of the disk. Let $X$ and $Y$ be the $x$ and $y$ coordinates of the imperfection (random variables) and let $Z$ be the distance of the imperfection from the origin.
- What is their joint density $f(x, y)$ ?

- Suppose that the surface of a disk is a circle with area R centered at the origin and that there is a single point imperfection at a location with is uniformly distributed across the surface of the disk. Let $X$ and $Y$ be the $x$ and $y$ coordinates of the imperfection (random variables) and let $Z$ be the distance of the imperfection from the origin.
- What is the range of $X \& Y$ and the marginal density of $X$ and of $Y$ ?


```
Poll:
What is \mp@subsup{\Omega}{X}{}}\mathrm{ ?
a. }[-\sqrt{}{\mp@subsup{R}{}{2}-\mp@subsup{x}{}{2}},\sqrt{}{\mp@subsup{R}{}{2}-\mp@subsup{x}{}{2}}
b. [-R,R]
c. [-\sqrt{}{\mp@subsup{R}{}{2}-\mp@subsup{y}{}{2}},\sqrt{}{\mp@subsup{R}{}{2}-\mp@subsup{y}{}{2}}]
d. Not sure
```

- Suppose that the surface of a disk is a circle with area R centered at the origin and that there is a single point imperfection at a location with is uniformly distributed across the surface of the disk. Let $X$ and $Y$ be the $x$ and $y$ coordinates of the imperfection (random variables) and let $Z$ be the distance of the imperfection from the origin.
- Are X and Y independent?


```
Poll:
Are }X\mathrm{ and }Y\mathrm{ independent?
a. yes
b. no
```

- Suppose that the surface of a disk is a circle with area R centered at the origin and that there is a single point imperfection at a location with is uniformly distributed across the surface of the disk. Let $X$ and $Y$ be the $x$ and $y$ coordinates of the imperfection (random variables) and let $Z$ be the distance of the imperfection from the origin.
- What is $E(Z)$ ?



## All of this generalizes to more than 2 random variables

|  | Discrete | Continuous |
| :--- | :--- | :--- |
| Joint PMF/PDF | $p_{X, Y}(x, y)=\mathbb{P}(X=x, Y=y)$ | $f_{X, Y}(x, y) \neq \mathbb{P}(X=x, Y=y)$ |
| Joint range/support | $\left\{(x, y) \in \Omega_{X} \times \Omega_{Y}: p_{X, Y}(x, y)>0\right\}$ | $\left\{(x, y) \in \Omega_{X} \times \Omega_{Y}: f_{X, Y}(x, y)>0\right\}$ |
| $\Omega_{X, Y}$ | $F_{X, Y}(x, y)=\sum_{t \leq x, s \leq y} p_{X, Y}(t, s)$ | $F_{X, Y}(x, y)=\int_{-\infty}^{x} \int_{-\infty}^{y} f_{X, Y}(t, s) d s d t$ |
| Joint CDF | $\sum_{x, y} p_{X, Y}(x, y)=1$ | $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X, Y}(x, y) d x d y=1$ |
| Normalization | $p_{X}(x)=\sum_{y} p_{X, Y}(x, y)$ | $f_{X}(x)=\int_{-\infty}^{\infty} f_{X, Y}(x, y) d y$ |
| Marginal PMF/PDF | $\mathbb{E}[g(X, Y)]=\sum_{x, y} g(x, y) p_{X, Y}(x, y)$ | $\mathbb{E}[g(X, Y)]=\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X, Y}(x, y) d x d y$ |
| Expectation |  |  |

