## CSE 312 <br> Foundations of Computing II

## Lecture 20: Joint Distributions

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Hash functions - few more comments

## Agenda

- Joint Distributions
- Cartesian Products
- Joint PMFs/PDFs/CDFs and Joint Range
- Marginal Distributions, etc.


## 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.
- 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

$$
A \times B=\{(a, b): a \in A, b \in B\}
$$

## Example.

$$
\{1,2,3\} \times\{4,5\}=\{(1,4),(1,5),(2,4),(2,5),(3,4),(3,5)\}
$$

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 $\mathbb{R} \times \mathbb{R}$ (often denoted $\mathbb{R}^{2}$ )

## Joint PMFs and Joint Range

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

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

Note that

$$
\sum_{(s, t) \in \Omega(X, Y)} p_{X, Y}(s, t)=1
$$

## Example: Weird 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.
$\Omega(X)=\{1,2,3,4\}$ and $\Omega(Y)=\{1,2,3,4\}$

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

| $x \mid y$ | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $1 / 16$ | $1 / 16$ | $1 / 16$ |
| $\mathbf{2}$ | $1 / 16$ | $1 / 16$ | $1 / 16$ | $1 / 16$ |
| $\mathbf{3}$ | $1 / 16$ | $1 / 16$ | $1 / 16$ | $1 / 16$ |
| $\mathbf{4}$ | $1 / 16$ | $1 / 16$ | $1 / 16$ | $1 / 16$ |

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

## Independence

## 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 $a, b$

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

## 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 $\Omega(W)=\{1,2,3,4\}$
$\Omega(U, W)=\{(u, w) \in \Omega(U) \times \Omega(W): u \leq w\} \neq \Omega(U) \times \Omega(W)$

Poll:
What is $p_{U, W}(1,3)=\operatorname{Pr}(U=1, W=3)$ ?
a. $1 / 16$
b. $2 / 16$
c. $1 / 2$
d. Not sure

| UIW | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: |
| 1 |  |  |  |  |
| 2 |  |  |  |  |
| 3 |  |  |  |  |
| 4 |  |  |  |  |

## 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
$p_{U, W}(u, w)=\left\{\begin{aligned} 2 / 16, & (u, w) \in \Omega(U) \times \Omega(W) \\ 1 / 16, & (u, w) \in \Omega(U) \times \Omega(W) \\ 0, & \text { where } w>u \\ 0, & \text { otherwise }=u\end{aligned}\right.$

| UIW | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 0 | $1 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{3}$ | 0 | 0 | $1 / 16$ | $2 / 16$ |
|  | 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 $p_{U, W}(u, w)$ ?
$p_{U}(u)=\left\{\begin{array}{l}u=1 \\ u=2 \\ u=3 \\ u=4\end{array}\right.$

| U\|w | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 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 $p_{U, W}(u, w)$ ?
$p_{U}(u)= \begin{cases}7 / 16, & u=1 \\ 5 / 16, & u=2 \\ 3 / 16, & u=3 \\ 1 / 16, & u=4\end{cases}$

| UIW | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ |
| ---: | :---: | :---: | :---: | :---: |
| $\mathbf{1}$ | $1 / 16$ | $2 / 16$ | $2 / 16$ | $2 / 16$ |
| $\mathbf{2}$ | 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 $X$

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

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

## 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 $Y$

$$
E[g(X, Y)]=\sum_{a \in \Omega(X)} \sum_{b \in \Omega(Y)} g(a, b) p_{X, Y}(a, b)
$$

## Another example.

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

What is the marginal pmf of $X$ ? What is the marginal pmf of $Y$ ? Are X and Y independent?
What is $\mathrm{E}(\mathrm{XY})$ ?

| $\mathrm{X} \mid \mathrm{Y}$ | $\mathbf{1}$ | $\mathbf{2}$ |
| :---: | :---: | :---: |
| $\mathbf{1}$ | 0.4 | 0.1 |
| $\mathbf{2}$ | 0.1 | 0.4 |

- 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?

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

## 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 $\Omega_{X}$ ?
a. $\left[-\sqrt{R^{2}-x^{2}}, \sqrt{R^{2}-x^{2}}\right]$
b. $[-R, R]$
c. $\left[-\sqrt{R^{2}-y^{2}}, \sqrt{R^{2}-y^{2}}\right]$
d. Not sure
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- 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?


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Poll:
Are X and Y independent?
a. yes
b. no
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- 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 |  |  |

## Brain Break



