CSE 312

Foundations of Computing II

Lecture 14: Joint Distributions



Aleks Jovcic

Slide Credit: Based on Stefano Tessaro's slides for 312 19au incorporating ideas from Alex Tsun, Rachel Lin, Hunter Schafer & myself ©

- 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.

 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)

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Joint PMFs and Joint Range

Definition. Let *X* and *Y* be discrete random variables. The **Joint PMF** of *X* and *Y* is

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

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

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

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.

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

In this problem, the joint PMF is

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

| X\Y | 1 | 2 | 3 | 4 |
|-----|------|------|------|------|
| 1 | 1/16 | 1/16 | 1/16 | 1/16 |
| 2 | 1/16 | 1/16 | 1/16 | 1/16 |
| 3 | 1/16 | 1/16 | 1/16 | 1/16 |
| 4 | 1/16 | 1/16 | 1/16 | 1/16 |

and the joint range is (since all combinations have non-zero probability)

$$\Omega(X,Y) = \Omega(X) \times \Omega(Y)$$

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Independence

Definition. Let *X* and *Y* be discrete random variables. The **Joint PMF** of *X* and *Y* is

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

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

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

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

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

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 $W = \max(X, Y)$

$$\Omega(U) = \{1,2,3,4\} \text{ and } \Omega(W) = \{1,2,3,4\}$$

$$\Omega(U, W) = \{(u, w) \in \Omega(U) \times \Omega(W) : u \le w \} \ne \Omega(U) \times \Omega(W)$$

Poll:

What is $p_{U,W}(1,3) = Pr(U = 1, W = 3)$?

- *a.* 1/16
- *b.* 2/16
- *c.* 1/2
- d. Not sure

| U\W | 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 $W = \max(X, Y)$

$$\Omega(U) = \{1,2,3,4\} \text{ and } \Omega(W) = \{1,2,3,4\}$$

$$\Omega(U, W) = \{(u, w) \in \Omega(U) \times \Omega(W) : u \le w \} \ne \Omega(U) \times \Omega(W)$$

The joint PMF
$$p_{U,W}(u, w) = Pr(U = u, W = w)$$
 is

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

| U\W | 1 | 2 | 3 | 4 |
|-----|------|------|------|------|
| 1 | 1/16 | 2/16 | 2/16 | 2/16 |
| 2 | 0 | 1/16 | 2/16 | 2/16 |
| 3 | 0 | 0 | 1/16 | 2/16 |
| 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 $\Pr(U = u)$ directly. Can we figure it out if we know $p_{U,W}(u,w)$?

$$p_{U}(u) = \begin{cases} u = 1 \\ u = 2 \\ u = 3 \\ u = 4 \end{cases}$$

| U\W | 1 | 2 | 3 | 4 |
|-----|------|------|------|------|
| 1 | 1/16 | 2/16 | 2/16 | 2/16 |
| 2 | 0 | 1/16 | 2/16 | 2/16 |
| 3 | 0 | 0 | 1/16 | 2/16 |
| 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 $\Pr(U = u)$ directly. Can we figure it out if we know $p_{U,W}(u,w)$?

$$p_U(u) = \begin{cases} 7/16, & u = 1\\ 5/16, & u = 2\\ 3/16, & u = 3\\ 1/16, & u = 4 \end{cases}$$

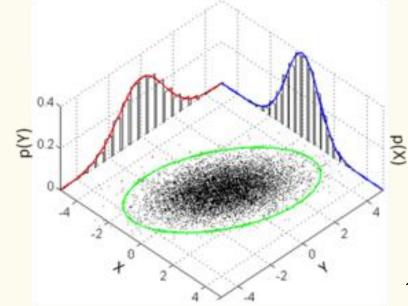
| U\W | 1 | 2 | 3 | 4 |
|-----|------|------|------|------|
| 1 | 1/16 | 2/16 | 2/16 | 2/16 |
| 2 | 0 | 1/16 | 2/16 | 2/16 |
| 3 | 0 | 0 | 1/16 | 2/16 |
| 4 | 0 | 0 | 0 | 1/16 |

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



Visual (for continuous X and Y)

Marginal PMF 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?

| X\Y | 1 | 2 |
|-----|-----|-----|
| 1 | 0.4 | 0.1 |
| 2 | 0.1 | 0.4 |

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Joint Continuous Random Variables

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

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

Expectation Example





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

What is E(XY)?

| X\Y | 1 | 2 |
|-----|-----|-----|
| 1 | 0.4 | 0.1 |
| 2 | 0.1 | 0.4 |

- Suppose the number of requests Z to a particular web server per hour is Poisson(λ). 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 | | |
| $\Omega_{X,Y}$ | $\{(x,y)\in\Omega_X\times\Omega_Y:p_{X,Y}(x,y)>0\}$ | |
| Joint CDF | $F_{X,Y}(x,y) = \sum_{t \le x,s \le y} p_{X,Y}(t,s)$ | $F_{X,Y}(x,y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f_{X,Y}(t,s) ds dt$ |
| Normalization | $\sum_{x,y} p_{X,Y}(x,y) = 1$ | $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy = 1$ |
| Marginal PMF/PDF | $p_X(x) = \sum_y p_{X,Y}(x,y)$ | $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y)dy$ |
| Expectation | $\mathbb{E}[g(X,Y)] = \sum_{x,y} g(x,y) p_{X,Y}(x,y)$ | $\mathbb{E}[g(X,Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{X,Y}(x,y) dx dy$ |

Independence (continuous random variables)

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

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

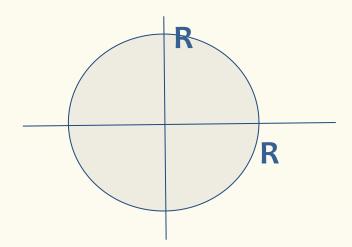
Definition. The **joint range** of $p_{X,Y}$ is

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

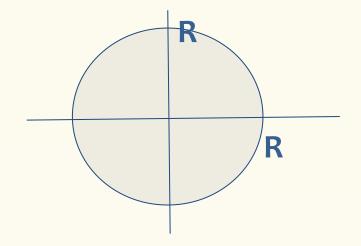
Definition. X and Y are **independent** iff for all a, b

$$f_{X,Y}(a,b) = f_X(a) \cdot f_Y(b)$$

- 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 Ω_X ?

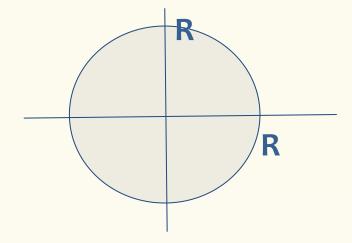
a.
$$[-\sqrt{R^2 - x^2}, \sqrt{R^2 - x^2}]$$

b.
$$[-R, R]$$

c.
$$[-\sqrt{R^2-y^2}, \sqrt{R^2-y^2}]$$

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?



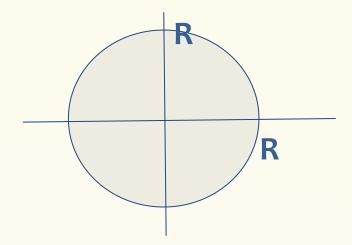
Poll:

Are X and Y independent?

- a. yes
- b. nc

• 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 | | |
| $\Omega_{X,Y}$ | $\{(x,y)\in\Omega_X\times\Omega_Y:p_{X,Y}(x,y)>0\}$ | $\{(x,y)\in\Omega_X\times\Omega_Y:f_{X,Y}(x,y)>0\}$ |
| Joint CDF | $F_{X,Y}(x,y) = \sum_{t \le x,s \le y} p_{X,Y}(t,s)$ | $F_{X,Y}(x,y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f_{X,Y}(t,s) ds dt$ |
| Normalization | $\sum_{x,y} p_{X,Y}(x,y) = 1$ | $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy = 1$ |
| Marginal PMF/PDF | $p_X(x) = \sum_{y} p_{X,Y}(x,y)$ | $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y)dy$ |
| Expectation | $\mathbb{E}[g(X,Y)] = \sum_{x,y} g(x,y) p_{X,Y}(x,y)$ | $\mathbb{E}[g(X,Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{X,Y}(x,y) dx dy$ |