

S&DS 241 Lecture 17

Joint and marginal distributions

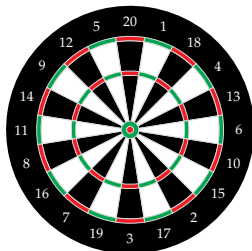
B-H: 7.1, 7.2

Motivating example

- So far we have been focusing on a single continuous random variable

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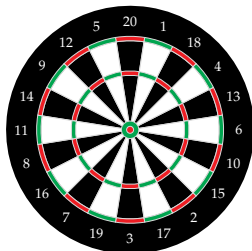
- So far we have been focusing on a single continuous random variable
- Let's consider the following example:



Suppose you are an amateur player and your dart lands uniformly at random on the board. What is the chance of hitting the bullseye? What is your average score?

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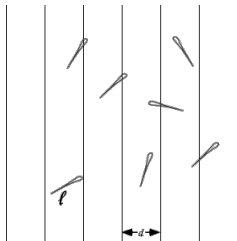
Suppose you are an amateur player and your dart lands uniformly at random on the board. What is the chance of hitting the bullseye?

What is your average score?

- Position of the dart (X, Y) is a pair of **continuous** random variables (RVs)

Buffon's needle

Suppose we have a floor made of parallel strips of wood, each of unit width, and we drop a needle of unit length onto the floor. What is the probability that the needle will touch a line between two strips?



- Source of randomness here: position and orientation of the needle — both are **continuous** random variables
- We will show that the answer is $\frac{2}{\pi}$.

Specific questions

- How to describe multiple continuous RVs?
- How to compute probability of events involving multiple RVs?
- How to make sense of conditioning and independence for continuous RVs?

Joint CDF & PDF

CDF

- The **joint cumulative distribution function (CDF)** of random variables X and Y is

$$F_{XY}(x, y) = P(X \leq x, Y \leq y)$$

(Reminder: comma means intersection i.e. AND)

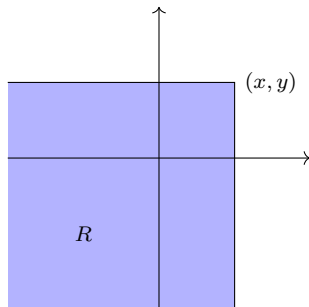
- This defines a function $F_{XY} : \mathbb{R}^2 \rightarrow [0, 1]$, which is non-decreasing in each coordinate when the other is held fixed.

CDF

Equivalently,

$$F_{XY}(x, y) = P((X, Y) \in R)$$

where $R = (-\infty, x] \times (-\infty, y]$ is the following region



Recall: continuous random variable and PDF

A random variable X is **continuous** if its CDF can be expressed as an integral:

$$F_X(x) = \int_{-\infty}^x f_X(s)ds, \quad \text{for all } x$$

We call f_X the **PDF** of X , given by

$$f_X(x) = F'_X(x)$$

Continuous random variables and Joint PDF

A pair of random variables (X, Y) is **continuous** if its joint CDF can be expressed as a double integral, i.e., there exists a nonnegative function $f_{XY} : \mathbb{R}^2 \rightarrow \mathbb{R}$ such that

$$F_{XY}(x, y) = \int_{-\infty}^x \int_{-\infty}^y f_{XY}(s, t) ds dt, \quad \text{for all } x, y$$

We call f_{XY} the **joint PDF** of (X, Y) , given by

$$f_{XY}(x, y) = \frac{\partial^2 F_{XY}(x, y)}{\partial x \partial y}.$$

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$$P(a \leq X \leq b, c \leq Y \leq d) = \int_c^d \int_a^b f_{XY}(x, y) dx dy$$

Proof: LHS = $F_{XY}(b, d) - F_{XY}(a, d) - (F_{XY}(b, c) - F_{XY}(a, c))$

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- ③ For region $A \subset \mathbb{R}^2$, e.g., rectangle or disk

$$P((X, Y) \in A) = \iint_A f_{XY}(x, y) dx dy$$

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Intuition: can approximate A by many rectangles

- ④ For any $x, y \in \mathbb{R}$,

$$P(X = x, Y = y) = 0$$

Moreover, $P((X, Y) \in A) = 0$ for any A with zero area. For example, $P(X = Y) = 0$.

Properties of Joint PDF

- ① Non-negativity:

$$f_{XY}(x, y) \geq 0$$

- ② Normalization:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = 1$$

since

$$P(-\infty < X < \infty, -\infty < Y < \infty) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = 1$$

Properties of Joint PDF

- Again, joint PDF is not a probability (can exceed 1); need to integrate to get probability.

For discrete RVs

Recall the notation $p_X(x) = P(X = x)$ denotes PMF of discrete X .

We can also define

- joint PMF

$$p_{XY}(x, y) = P(X = x, Y = y)$$

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- joint CDF

$$F_{XY}(x, y) = \sum_{a \leq x} \sum_{b \leq y} p_{XY}(a, b)$$

(analogous to the connection between CDF and PMF in Lec 13)

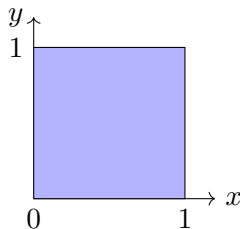
Running example

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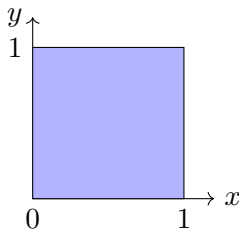
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$$f_{XY}(x, y) = \begin{cases} 1 & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{else} \end{cases}$$



- This is in fact the uniform distribution over the unit square.

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What is the chance that Alice arrives earlier than Bob?

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- In further hindsight, we could have used symmetry to conclude

$$P(X < Y) = \frac{1}{2}$$

Lunch example (2)

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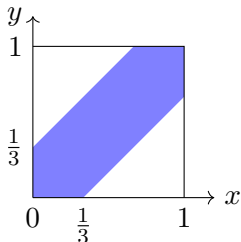
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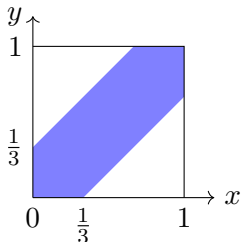
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- Thus

$$P(\text{lunch takes place}) = \text{area} \left(\begin{array}{|c|} \hline \text{blue shaded region} \\ \hline \end{array} \right) = 5/9$$

LOTUS in 2-D

Let $Z = g(X, Y)$. As usual, two ways to find $E(Z)$:

- 1 Find CDF F_Z , then PDF f_Z , then $E(Z) = \int z f_Z(z) dz$.

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- 1 Find CDF F_Z , then PDF f_Z , then $E(Z) = \int z f_Z(z) dz$.
- 2 Invoke LOTUS rule:

$$E(g(X, Y)) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{XY}(x, y) dx dy$$

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Let Z denote the arrival time of the **later** one of the two (Alice and Bob). Find the pdf of Z and its mean.

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- Then $E(Z) = \int_0^1 2z^2 dz = \frac{2}{3} > E(X) = E(Y) = \frac{1}{2}$

Alternatively, using LOTUS

Since $Z = \max(X, Y)$,

$$\begin{aligned} E(Z) &= \iint \max(x, y) f_{XY}(x, y) dx dy = \int_0^1 \int_0^1 \max(x, y) dx dy \\ &= \int_0^1 \left(\int_0^y y dx \right) dy + \int_0^1 \left(\int_y^1 x dx \right) dy \\ &= \int_0^1 y^2 dy + \int_0^1 \frac{1 - y^2}{2} dy = \frac{2}{3} \end{aligned}$$

Marginalization

Marginal PDF

- The joint distribution of X, Y contains more information than the (marginal) distribution of X or Y individually.
- Joint PDF determines the marginal PDFs, but **not** vice versa

How to recover the distribution of X from the joint distribution of X, Y ?

Joint CDF \rightarrow Marginal CDF

Recall:

$$F_{XY}(x, y) = P(X \leq x, Y \leq y)$$

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Here $F_{XY}(x, \infty)$ is understood as $\lim_{y \rightarrow \infty} F_{XY}(x, y)$, $F_{XY}(\infty, y)$ as $\lim_{x \rightarrow \infty} F_{XY}(x, y)$.

Joint PDF \rightarrow Marginal PDF

To find marginal PDF, let's express the marginal CDF as an integral:

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Summary:

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{XY}(x, y) dy \\ f_Y(y) &= \int_{-\infty}^{\infty} f_{XY}(x, y) dx \end{aligned}$$

Joint PMF \rightarrow Marginal PMF

Entirely analogously,

$$\begin{aligned} p_X(x) &= \sum_y p_{XY}(x, y) \\ p_Y(y) &= \sum_x p_{XY}(x, y) \end{aligned}$$

Example: Contingency table

Consider a randomly sampled individual.

- Let X be the indicator of being a smoker.
- Let Y be the indicator of developing lung cancer.

Suppose p_{XY} is given by:

	$Y = 1$	$Y = 0$
$X = 1$	0.05	0.20
$X = 0$	0.01	0.74

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Sum	0.06	0.94	

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	$Y = 1$	$Y = 0$	Sum
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$X = 0$	0.01	0.74	0.75
Sum	0.06	0.94	1

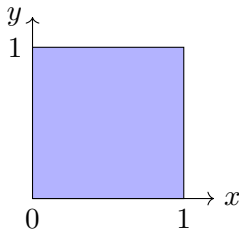
Summary

To marginalize the joint PDF/PMF to one variable, integrate/sum out the other variable.

Lunch example (4)

- Joint PDF of (X, Y) , the arrival time of Alice and Bob:

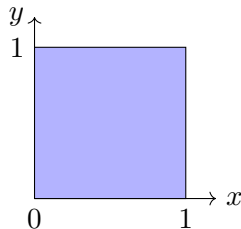
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- Marginal PDF of X :

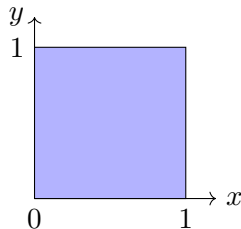
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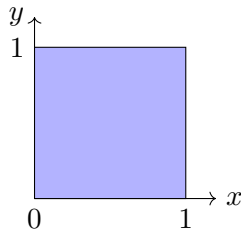
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if $x \in [0, 1]$; otherwise $f_X(x) = 0$.

- Marginal PDF of Y : same.
- Both X and Y are distributed as $\text{Unif}(0, 1)$.

Another (abstract) example

$$\stackrel{\text{LOTUS}}{=} E(X + Y) \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x + y) f_{XY}(x, y) dx dy$$

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This justifies the **linearity of expectation** for continuous random variables.

Uniform distribution in 2D

2-dimensional uniform distribution

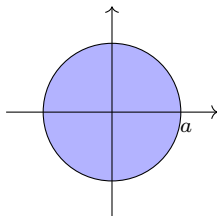
Let $A \subset \mathbb{R}^2$ be a region on the plane with finite area.

- We say (X, Y) is **uniformly distributed over A** if the joint PDF is

$$f_{XY}(x, y) = \begin{cases} \frac{1}{\text{area}(A)} & (x, y) \in A \\ 0 & \text{else} \end{cases}$$

- In the previous lunch example, the arrival times of Alice and Bob are uniform over the unit square $[0, 1]^2$

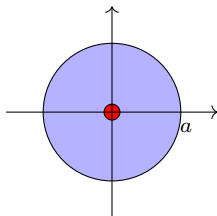
Example: Dart



Let (X, Y) denote the coordinate of the dart, which is uniformly distributed over the board (a disk of radius a). Then

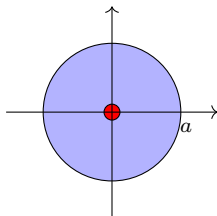
$$f_{XY}(x, y) = \begin{cases} \frac{1}{\pi a^2} & x^2 + y^2 \leq a^2 \\ 0 & \text{else} \end{cases}$$

Example: Dart



What's the probability of hitting the bullseye?

Example: Dart

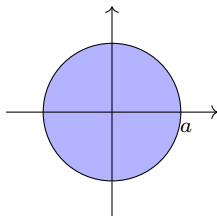


What's the probability of hitting the bullseye?

- For typical dartboard: diameter = 17.75", bullseye diameter = 0.5"

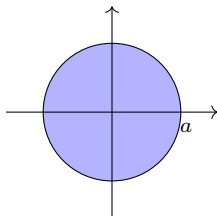
$$\begin{aligned} P(\text{bullseye}) &= \iint_{\text{bullseye}} f_{XY}(x, y) dx dy = \frac{\text{area}(\text{bullseye})}{\text{area}(\text{board})} \\ &= \frac{0.5^2}{17.75^2} \approx \frac{1}{1260} \end{aligned}$$

Example: Dart



What's the distribution of the horizontal coordinate X ?

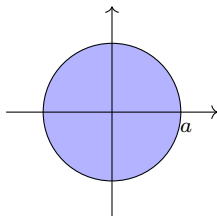
Example: Dart



What's the distribution of the horizontal coordinate X ?

- X takes values in $[-a, a]$

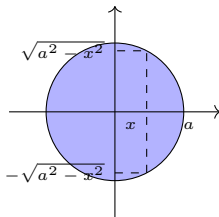
Example: Dart



What's the distribution of the horizontal coordinate X ?

- X takes values in $[-a, a]$
- Uniform?

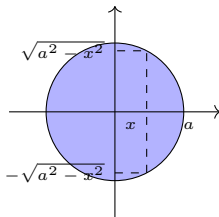
Example: Dart



Find the marginal PDF of X : for $x \in [-a, a]$

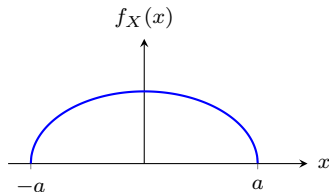
$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy = \int_{-\sqrt{a^2-x^2}}^{\sqrt{a^2-x^2}} \frac{1}{\pi a^2} dy = \frac{2\sqrt{a^2-x^2}}{\pi a^2}$$

Example: Dart



Find the marginal PDF of X : for $x \in [-a, a]$

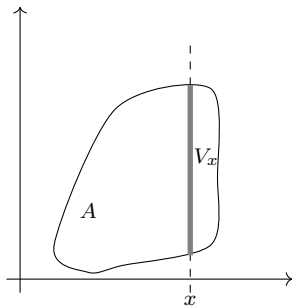
$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy = \int_{-\sqrt{a^2-x^2}}^{\sqrt{a^2-x^2}} \frac{1}{\pi a^2} dy = \frac{2\sqrt{a^2-x^2}}{\pi a^2}$$



Not uniform: more likely to be near the middle than ends. Why?

More generally: marginalize a uniform distribution

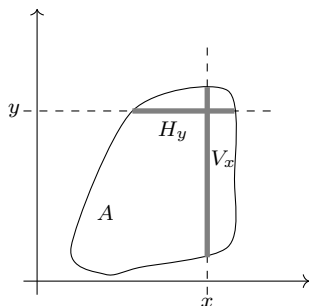
Let (X, Y) be uniformly distributed over a region A :



$$f_X(x) = \int_{V_x} f_{XY}(x, y) dy = \frac{\text{length}(V_x)}{\text{area}(A)}$$

More generally: marginalize a uniform distribution

Let (X, Y) be uniformly distributed over a region A :



$$f_X(x) = \int_{V_x} f_{XY}(x, y) dy = \frac{\text{length}(V_x)}{\text{area}(A)}$$
$$f_Y(y) = \int_{H_y} f_{XY}(x, y) dx = \frac{\text{length}(H_y)}{\text{area}(A)}$$

where H_y and V_x denote the horizontal and vertical segment intersecting A , respectively.