

Joint Probability Distributions Continuous Case

Section 6.2

STAT/MATH 395 Spring 2020

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Lecture 6, April 10th, 2020

Ask questions via [chat on Zoom](#)

Answer quiz via [PollEverywhere](#) (username: vincentroulet)

Announcements

Office hours (After poll)

- ▶ Mondays 14:30 to 15:30 with T.A. Z. Yuan by Zoom
- ▶ Fridays 11:30 to 12:30 with instructor V. Roulet
- ▶ Register in advance to access the zoom session

Lecture material

Updated slides with solutions given at the end of the lecture

Exercise

Roll a normal die 100 times. Find the probability that among the 100 rolls, we observe exactly 22 ones, 17 fives.

Answer Previous Exercise

Exercise

Roll a normal die 100 times. Find the probability that among the 100 rolls, we observe exactly 22 ones, 17 fives.

Solution Denote X_1, X_5 the number of times you get a 1 or a 5 resp. among 100 rolls

We have $\mathbb{P}(\text{"face is 1"}) = \mathbb{P}(\text{"face is 5"}) = 1/6$

We could model $X_1, X_2, X_3, X_4, X_5, X_6$ as a multinomial but that can be simplified

Denote $Y = X_2 + X_3 + X_4 + X_6$ the number of times you get any other face

We have $\mathbb{P}(\text{"face is not 1 or 5"}) = 4/6 = 2/3$

Then $(X_1, X_5, Y) \sim \text{Multinom}(100, 1/6, 1/6, 2/3)$

So $\mathbb{P}(X_1 = 22, X_5 = 17, Y = 100 - (22 + 17)) = \frac{100!}{22!17!61!} \left(\frac{1}{6}\right)^{22} \left(\frac{1}{6}\right)^{17} \left(\frac{2}{3}\right)^{61} \approx 0.0037$

Joint Probability Density Functions

Definition (Joint probability density function)

Random variables X_1, \dots, X_n are **jointly continuous** if there exists a **joint probability density function** $f : \mathbb{R}^n \rightarrow \mathbb{R}$ such that for any¹ $B \subset \mathbb{R}^n$,

$$\mathbb{P}(X_1, \dots, X_n \in B) = \int \dots \int_B f(x_1, \dots, x_n) dx_1 \dots dx_n$$

Note:

- ▶ $f(x_1, \dots, x_n) \geq 0$ and $\int_{-\infty}^{+\infty} \dots \int_{-\infty}^{+\infty} f(x_1, \dots, x_n) dx_1 \dots dx_n = 1$
- ▶ X and Y have a p.d.f. does not imply that (X, Y) is jointly continuous!

¹Think of B as for example $[a, b]^n$. Again a rigorous definition requires B to belong to the Borel algebra of \mathbb{R}^n

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- ▶ X and Y have a p.d.f. does not imply that (X, Y) is jointly continuous!

Example: Take X any continuous r.v., define $Y = X$, s.t. $\mathbb{P}(X = Y) = 1$. If (X, Y) had a joint p.d.f. f , denoting $D = \{(x, y) : x=y\}$, we would have

$$\mathbb{P}(X = Y) = \int_D \int f(x, y) dx dy = \int_{-\infty}^{+\infty} \left(\int_x^x f(x, y) dy \right) dx = 0$$

¹Think of B as for example $[a, b]^n$. Again a rigorous definition requires B to belong to the Borel algebra of \mathbb{R}^n

Joint Probability Density Functions

Lemma

Let X_1, \dots, X_n be n jointly continuous r.v.. Then for any subset $A \subset \mathbb{R}^n$ included in a linear subspace $E \subset \mathbb{R}^n$ of dimension $\dim(E) = m < n$,

$$\mathbb{P}((X_1, \dots, X_n) \in A) = 0$$

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Proof General case requires change of variables, let's consider $A=[a, b]^m \subset \mathbb{R}^n$
Denote f the joint p.d.f. of (X_1, \dots, X_n) ,

$$\mathbb{P}((X_1, \dots, X_n) \in A) = \underbrace{\int_a^b \dots \int_a^b}_{m \text{ times}} \underbrace{\int_0^0 \dots \int_0^0}_{n-m \text{ times}} f(x_1, \dots, x_n) dx_1 \dots dx_n = 0$$

Joint Probability Density Functions

Example (Synthetic)

Assume X, Y have a joint p.d.f.

$$f(x, y) = \begin{cases} \frac{3}{2}(xy^2 + y) & \text{if } 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

1. Check that it is a valid joint p.d.f.

Joint Probability Density Functions

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1. Check that it is a valid joint p.d.f.

Solution We have $f(x, y) \geq 0$ and

$$\begin{aligned} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) dx dy &= \frac{3}{2} \int_0^1 \left(\int_0^1 xy^2 + y dx \right) dy \\ &= \frac{3}{2} \int_0^1 \left(\frac{1}{2}y^2 + y \right) dy = \frac{3}{2} \left(\frac{1}{6} + \frac{1}{2} \right) = 1 \end{aligned}$$

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2. Compute $\mathbb{P}(X < Y)$

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2. Compute $\mathbb{P}(X < Y)$

Solution

$$\begin{aligned} \mathbb{P}(X < Y) &= \frac{3}{2} \int_0^1 \left(\int_0^y (xy^2 + y) dx \right) dy \\ &= \frac{3}{2} \int_0^1 \left(\frac{1}{2}y^4 + y^2 \right) dy \\ &= \frac{3}{2} \left(\frac{1}{10} + \frac{1}{3} \right) = 0.65 \end{aligned}$$

Uniform Continuous Random Variable in higher dimensions

Definition (Uniform continuous random variable in dimension 2 or 3)

Let D be a bounded subset of \mathbb{R}^2 s.t. $\text{Area}(D) < +\infty$. The random point (X, Y) is **uniformly distributed on D** if its joint p.d.f. reads

$$f(x, y) = \frac{1}{\text{Area}(D)} \mathbf{1}_D(x, y) = \begin{cases} \frac{1}{\text{Area}(D)} & \text{if } (x, y) \in D \\ 0 & \text{otherwise} \end{cases}$$

Let D be a bounded subset of \mathbb{R}^3 s.t. $\text{Vol}(D) < +\infty$. The random point (X, Y, Z) is **uniformly distributed on D** if its joint p.d.f. reads

$$f(x, y, z) = \frac{1}{\text{Vol}(D)} \mathbf{1}_D(x, y, z) = \begin{cases} \frac{1}{\text{Vol}(D)} & \text{if } (x, y, z) \in D \\ 0 & \text{otherwise} \end{cases}$$

We denote $(X, Y) \sim \text{Unif}(D)$ or $(X, Y, Z) \sim \text{Unif}(D)$.

Uniform Continuous Random Variable in higher dimensions

Lemma

Let $(X, Y) \sim \text{Unif}(D)$ for $D \subset \mathbb{R}^2$, then for any $G \subset D$, (similar for \mathbb{R}^3)

$$\mathbb{P}((X, Y) \in G) = \frac{\text{Area}(G)}{\text{Area}(D)}$$

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Proof

$$\Pr((X, Y) \in G) = \frac{1}{\text{Area}(D)} \int \int \mathbf{1}_G(x, y) \mathbf{1}_D(x, y) dx dy = \int \int \mathbf{1}_G(x, y) dx dy = \frac{\text{Area}(G)}{\text{Area}(D)}$$

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Example

Denote $D_r = \{(x, y) : x^2 + y^2 < r^2\}$ a disk of radius r

Throw a dart uniformly at random on a disk of radius 2

What is the probability that the dart is in the central disk of radius one?

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Denote $D_r = \{(x, y) : x^2 + y^2 < r^2\}$ a disk of radius r

Throw a dart uniformly at random on a disk of radius 2

What is the probability that the dart is in the central disk of radius one?

Solution $(X, Y) \sim \text{Unif}(D_2)$

$$\mathbb{P}((X, Y) \in D_1) = \frac{\pi 1^2}{\pi 2^2} = \frac{1}{4}$$

Joint Probability Density Functions

Lemma

Let $g : \mathbb{R}^n \rightarrow \mathbb{R}$ and let X_1, \dots, X_n be jointly continuous r.v. with joint p.d.f. f ,

$$\mathbb{E}[g(x_1, \dots, x_n)] = \int_{-\infty}^{+\infty} \dots \int_{-\infty}^{+\infty} g(x_1, \dots, x_n) f(x_1, \dots, x_n) dx_1 \dots dx_n$$

Example

Throw a dart uniformly at random on a square of edge size 2 centered on 0

Assume your score is equal to the square distance to the center

What is your average score?

Joint Probability Density Functions

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Example

Throw a dart uniformly at random on a square of edge size 2 centered on 0

Assume your score is equal to the square distance to the center

What is your average score?

Solution $(X, Y) \sim \text{Unif}(S)$ with $S = \{(x, y) : -1 \leq x \leq 1, -1 \leq y \leq 1\}$

Score is $g(x, y) = x^2 + y^2$

Average score

$$\mathbb{E}[g(X, Y)] = \frac{1}{4} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} g(x, y) \mathbf{1}_S(x, y) dx dy = \frac{1}{4} \int_{-1}^1 \int_{-1}^1 (x^2 + y^2) dx dy = 1/3$$

Marginal Probability Density Function

Definition (Marginal probability density function)

Let X, Y be jointly continuous r.v. and denote $f_{X,Y}$ their joint p.d.f. then the p.d.f. of X exists and is given by

$$f_X(X) = \int_{-\infty}^{+\infty} f_{X,Y}(x, y) dy$$

Marginal Probability Density Function

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Let X, Y be jointly continuous r.v. and denote $f_{X,Y}$ their joint p.d.f. then the p.d.f. of X exists and is given by

$$f_X(X) = \int_{-\infty}^{+\infty} f_{X,Y}(x, y) dy$$

Proof We have by definition of the joint p.d.f. an expression of the c.d.f. of X as

$$F_X(t) = \mathbb{P}(X \leq t) = \mathbb{P}(X \leq t, -\infty \leq Y \leq +\infty) = \int_{-\infty}^t \int_{-\infty}^{+\infty} f_{X,Y}(x, y) dy dx$$

Therefore $f_X(x) = F'_X(x) = \int_{-\infty}^{+\infty} f_{X,Y}(x, y) dy$

Marginal Probability Density Function

Example

Consider a disk of radius r , $D_r = \{(x, y) : x^2 + y^2 \leq r\}$ and $(X, Y) \sim \text{Unif}(D_r)$.
What is the marginal p.d.f. of X ?

Marginal Probability Density Function

Example

Consider a disk of radius r , $D_r = \{(x, y) : x^2 + y^2 \leq r^2\}$ and $(X, Y) \sim \text{Unif}(D_r)$. What is the marginal p.d.f. of X ?

Solution Joint p.d.f. is $f_{X,Y}(x, y) = \frac{1}{\pi r^2} \mathbf{1}_{D_r}(x, y)$ where $D_r = \{(x, y) : x^2 + y^2 \leq r^2\}$. Marginal density is then $f_X(x) = 0$ for $|x| > r$, and for $|x| \leq r$,

$$f_X(x) = \int_{-\infty}^{+\infty} f_{X,Y}(x, y) dy = \frac{1}{\pi r^2} \int_{-\sqrt{r^2-x^2}}^{\sqrt{r^2-x^2}} dy = \frac{2}{\pi r^2} \sqrt{r^2 - x^2}$$

Marginal Probability Density Function

Definition (Marginal probability density function)

Let X_1, \dots, X_n be jointly continuous and denote f their joint p.d.f..

Then for any $j \in \{1, \dots, n\}$, X_j is a continuous random variable with p.d.f.

$$f_{X_j}(x) = \int_{-\infty}^{+\infty} \dots \int_{-\infty}^{+\infty} f(x_1, \dots, x_{j-1}, x, x_{j+1}, \dots, x_n) dx_1 \dots dx_{j-1} dx_{j+1} \dots dx_n$$

($n-1$ integrals)

Joint Cumulative Distribution

Definition (Joint cumulative distribution)

The **joint cumulative distribution** of r.v. X_1, \dots, X_n is defined as

$$\begin{aligned} F(t_1, \dots, t_n) &= \mathbb{P}(\{X_1 \leq t_1\} \cap \dots \cap \{X_n \leq t_n\}) \\ &\triangleq \mathbb{P}(X_1 \leq t_1, \dots, X_n \leq t_n) \end{aligned}$$

Lemma

1. If (X, Y) are jointly continuous with joint p.d.f. f ,

$$F(t, s) = \int_{-\infty}^t \int_{-\infty}^s f(x, y) dx dy$$

2. If (X, Y) are jointly continuous (i.e. there exists a joint p.d.f.) with joint c.d.f. F

$$\left. \frac{\partial^2}{\partial t \partial s} F(t, s) \right|_{s=x, t=y} = f(x, y)$$

Borel algebra in \mathbb{R}^{n*}

Formal details

- ▶ Until now, we defined proba. distributions on any $B \subset \mathbb{R}^n$ for $n=1$ or $n>1$.
- ▶ Formal definitions require to restrict our focus to subsets $B \subset \mathbb{R}^n$ that form a σ -algebra \mathcal{B}

Definition (σ -algebra)

Let Ω be a set, a σ -algebra \mathcal{F} on Ω is a subset of $2^\Omega = \{B \subset \Omega\}$ such that

1. $\Omega \in \mathcal{F}$
2. (Stable by complementarity) For any $A \in \mathcal{F}$, $A^c \triangleq \Omega \setminus A \in \mathcal{F}$
3. (Stable by countable union) For any $A_1, A_2, \dots \in \mathcal{F}$, $\bigcup_{i \in \mathbb{N}} A_i \in \mathcal{F}$

Why introducing σ -algebra?

You want the probability measure to satisfy that

- ▶ the measure is non-negative
- ▶ the measure of the union of disjoint sets is the sum of the measure of union sets

Then you can build a union of sets V_k (see e.g. Vitali set on Wikipedia) s.t.

$$[0, 1] \subset \bigcup_{k=1}^{+\infty} V_k \subset [-1, 2] \quad \mathbb{P}(V_k) = \lambda \geq 0 \quad \text{for all } k$$

which leads to $1 \leq \sum_{k=1}^{+\infty} \mathbb{P}(V_k) \leq 3$ which is impossible

Borel algebra in \mathbb{R}^n *

Formally, we restrict our focus on the Borel algebra of \mathbb{R}^n

Definition (Borel algebra in \mathbb{R}^n)

The Borel algebra in \mathbb{R}^n , denoted \mathcal{B}_n , is the smallest σ -algebra (in terms of inclusion) that contains

- ▶ all product of intervals $[a_1, b_1] \times \dots \times [a_n, b_n]$ for $a_i \leq b_i \in \mathbb{R}$

or equivalently defined as the smallest σ -algebra that contains

- ▶ all product of intervals of the form $(-\infty, a_1] \times \dots \times (-\infty, a_n]$ for $a_i \in \mathbb{R}$.

Consequence

1. If we can measure all intervals of the form $(-\infty, a_1] \times \dots \times (-\infty, a_n]$ for $a_i \in \mathbb{R}$, then we can measure all subsets of interests, i.e. all $B \in \mathcal{B}_n$,
→ we know all the information necessary to describe the proba distribution
2. All the information necessary to describe any r.v. is contained in its c.d.f.

Quiz for next lecture

Exercise

I am shooting an arrow on a target on a wall $W = \{(x, y) : -1 \leq x \leq 1, 0 \leq y \leq 1\}$. A wind affects my shoot from the left and the gravity also affects my shoot such that the position of the arrow has a p.d.f. proportional to $\frac{e^x}{\sqrt{y+1}}$

*What is the probability
that I touch the target $T = \{(x, y) : -0.1 \leq x \leq 0.1, 0.4 \leq y \leq 0.6\}$?*