

(X, Y) - random vector.

Conditional distribution of Y , given X .

Fix $x \in \mathbb{R}$. $B \subset \mathbb{R}$.

Suppose $Q(x, B)$ satisfies:

- (*) {
- 1) for each $x \in \mathbb{R}$, $Q(x, B)$ is a probability in B .
 - 2) \forall borel $B \subset \mathbb{R}$, $Q(x, B)$ is a measurable function.
 - 3) \forall borel $B, C \subset \mathbb{R}$,
- $$P(Y \in B, X \in C) = E(Q(X, B) \cdot \mathbf{1}_C(x))$$

$$Q(x, B) = P(Y \in B \mid X=x)$$

(interpretation)

Two "trivial" cases:

1) X is discrete.

Say, with DCR countable,
satisfying $P(X=x) > 0 \quad \forall x \in \text{DC } \mathbb{R}$.

$$\text{Define } Q(x, B) = P(Y \in B \mid X=x) = \begin{cases} \frac{P(Y \in B, X=x)}{P(X=x)}, & x \in D. \\ \delta_{\{x\}}(B), & x \notin D. \end{cases}$$

Dirac mass.

(doesn't matter!!!)

2. (X, Y) jointly absolutely continuous with joint density f

has joint density.

$$\text{Define } g(x, y) = \int f(x, y) \quad , \quad \text{if } \int f(x, y) dy > 0$$

Define $g(x, y) = \begin{cases} f(x, y) & , \text{ if } \int f(x, y) dy > 0 \\ \int f(x, y) dy & \\ \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}y^2} & , \text{ if } \int f(x, y) dy = 0. \end{cases}$

Marginal density of x .

Can be anything else as well !!

Define $Q(x, B) = \int_B g(x, y) dy$

A Special Case: X, Y - independent r.v.s.

$h: \mathbb{R}^2 \rightarrow \mathbb{R}$ is a measurable function.

$\therefore h(X, Y)$ is a random variable.

Q. What is the conditional distribution of $h(X, Y)$ given X .

- Proposed conditional distribution:

define $Q(x, B) = P(h(X, Y) \in B)$.

to show: this $Q(x, B)$ follows all the 3 results (*)

Consider the following setting:

Let $A \subset \mathbb{R}^2$

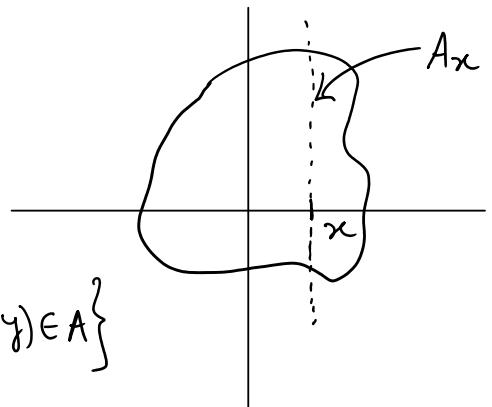
↳ borel subset of \mathbb{R}^2 . "x-section"

For $x \in \mathbb{R}$, denote $A_x = \{y \in \mathbb{R} \mid (x, y) \in A\}$

* $\forall A \subset \mathbb{R}^2$ borel,

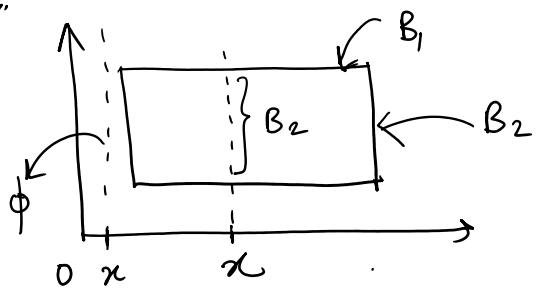
$A_x \subset \mathbb{R}$ is borel $\forall x \in \mathbb{R}$.

[ie, pick a borel in \mathbb{R}^2 . take an x-section.
that $A_x \subset \mathbb{R}$ is also borel.]



Take $A = B_1 \times B_2 \leftarrow$ "rectangle" $\cap \mathbb{R}^2$

Take $A = B_1 \times B_2$ ← "rectangle"
 Clearly, $A_x = \begin{cases} B_2, & \text{if } x \in B_1 \\ \emptyset, & \text{if } x \notin B_1 \end{cases}$



So, we can talk about

$P(Y \in A_x)$, \forall borel $A \subset \mathbb{R}^2$
 $\& x \in \mathbb{R}$.

$$= \varphi(x, A)$$

Now, fix a borel $A \subset \mathbb{R}^2$, & keep on varying x .
 look at $x \mapsto \varphi(x, A)$ ↗ ie, look at x-section.

∴ This is measurable $\& A \subset \mathbb{R}^2$ borel.

clearly, this map $x \mapsto \varphi(x, A)$
 is a fn from borel \mathbb{R} to $[0, 1]$

∴ for $A = B_1 \times B_2$.

$$P(Y \in A_x) = P(Y \in B_2) \cdot \mathbb{1}_{B_1}(x)$$

Now, to show,

$$\alpha = \left\{ A \subset \mathbb{R}^2 : \begin{array}{l} \varphi(A) \text{ is measurable in } x \\ \& B_1 \times B_2 \in \alpha \end{array} \right\}$$

"Monotone class Theorem"

Theorem (Monotone class Theorem):

Ω is a set. \mathcal{F} is a field on Ω
 ↗ NOT σ -field !!
 → closed under

- complementation
- finite intersection.

If M is a class of sets such that,

$\mathcal{F} \subset M$, and M is closed under
 monotone limits, then $M \supset \sigma(\mathcal{F})$

$\gamma \subset M$, and M is closed under monotone limits, then $M \supset \sigma(\mathcal{F})$

\downarrow

σ -field generated by field \mathcal{F} .

In our case, we have to first show -
the set of rectangles in \mathbb{R}^2 form a field.

Where do we stand now? let $A \subset \mathbb{R}^2$ any Borel Set.
 $\forall x \in \mathbb{R}, A_x = \{y \in \mathbb{R} \mid (x, y) \in A\}$

Step-1: We showed:

$A_x \subset \mathbb{R}$ is Borel $\forall x$.

(first for rectangles, then ...)

then, $P(Y \in A_x)$

Borel
Set

\therefore we can talk about
this probability

Step-2: $x \mapsto \varphi(x, A)$ is measurable.

Step-3: $P((X, Y) \in A)$

Borel in \mathbb{R}^2
Hence, (X, Y) is an event
So, we can talk about
this probability.

Claim:

$$P((X, Y) \in A) = E(\varphi(X, A)).$$

\downarrow
a r.v. [\because from Step-2:
 $x \mapsto \varphi(x, A)$ is measurable]
hence, we can
talk about this
expectation.

expectation.

Main thing: We can't do serious Probability without Measure Theory !!

Here, $A \subset \mathbb{R}^2$.

if $A = B_1 \times B_2$.

$$\text{LHS} = P((X, Y) \in A) = P((X, Y) \in B_1 \times B_2)$$

$$= P(X \in B_1, Y \in B_2)$$

$$[\because X, Y \text{ independent}] = P(X \in B_1) \cdot P(Y \in B_2).$$

$$\&, \text{ RHS} = E(\varphi(x, A)) \quad \&, \quad \varphi(x, A) = P(Y \in A_x)$$

$$= E(P(Y \in B_2) \cdot 1_{B_1}(x)) \quad \therefore \varphi(x, A) = P(Y \in B_2).$$

$$= P(Y \in B_2) \cdot E(1_{B_1}(x))$$

$$= P(Y \in B_2) \cdot P(X \in B_1).$$

the
X-section.

$\therefore \text{LHS} = \text{RHS}$ is immediate if $A = B_1 \times B_2 \subset \mathbb{R}^2$
Borel.

Now, back to our last qn from last lecture (Lec-2)

proposed $Q(x, B) := P(h(x, Y) \in B)$

Is $Q(x, B)$ measurable in x ?

$$Q(x, B) = \varphi(x, A), \text{ where } A = h^{-1}(B).$$

$$= \{(x, y) \mid h(x, y) \in B\}$$

$$\therefore \text{the X-section,}$$

$$A_x = \{y \mid h(x, y) \in B\}.$$

$$\therefore \varphi(x, A) = P(Y \in A_x)$$

$$= P(h(x, Y) \in B), \text{ which is measurable.}$$

$\therefore Q(x, B)$ is measurable.

Now, Is $Q(x, B)$ as probability in B ?

$$\text{on } (\mathbb{R}, \mathcal{B}) = P(h(x, Y) \in B). \quad \text{if } B = \mathbb{R},$$

- $\hookrightarrow Q(x, B) \rightsquigarrow$ $P(h(x, Y) \in B)$
- if $B = \mathbb{R}$,
 $P(h(x, Y) \in \mathbb{R}) = 1 \checkmark$
 - take $B, B^c \subset \mathbb{R}$.
 $P(h(x, Y) \in B) =$
 $= 1 - P(h(x, Y) \in B^c) \checkmark$
 - B_1, B_2 - disjoint borels in \mathbb{R} .
 $P(h(x, Y) \in B_1 \sqcup B_2) = \checkmark$
 $P(h(x, Y) \in B_1) + P(h(x, Y) \in B_2)$

& now, finally, is

$$P(h(x, Y) \in B, x \in C) = E(Q(x, B) \cdot 1_C(x)) ??$$

$$\text{take } A = \{(x, y) \mid h(x, y) \in B, x \in C\}.$$

$$\therefore A_x = \begin{cases} \{y \mid h(x, y) \in B\}, & x \in C \\ \emptyset, & x \notin C \end{cases}$$

(empty set)

Now, the 3 steps above can be put to use.

Little relief: 

All these proofs are not a part of the syllabus.
Only the final result (ie, the proposition) is !!.

A consequence :

X, Y - independent r.v.s.

$$h(x, y) = x + y.$$

finding dist' of $Z = X + Y$

$$F_Z(a) = P(Z \leq a) = P(X + Y \leq a)$$

$$= F(\psi(x) \mid \text{where } \psi(x) = P(X + Y \leq a))$$

$$f_Z(a) = P(Z \leq a) = E(\varphi(x)) \quad \text{where } \varphi(x) = P(X+Y \leq a)$$

$$= P(Y \leq a-x)$$

$$= F_Y(a-x)$$

$$= E(F_Y(a-x))$$

$$= E(F_X(a-Y))$$

$$\downarrow$$

usually
called

$$F_X * F_Y$$

This is called
the convolution.

So, in general, X has distⁿ f F
 Y " " " G .

Then, $X+Y$ has dist. $F * G$

Fact: take distⁿ of X degenerate at 0.
 If acts as identity !!

Also, in general, F does not have an inverse.
 But, if F is degenerate, then inverse exists.

Prove that: No other distⁿ will have an inverse.

Suppose F has density f $X \sim F$, X, Y -ind.
 G " " " g . $Y \sim G$.

$$f * g(a) = P(X+Y \leq a)$$

$$= E(G(a-x))$$

$$= \int g(a-x) f(x) dx$$

$$= \int_{-\infty}^{\infty} \int_{a-x}^{x} f(u) du dx$$

fixed.

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(y) dy f(x) dx \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^a g(v-x) \cdot f(x) dy dx \\
 &= \int_{-\infty}^a \underbrace{\int_{-\infty}^{\infty} g(v-x) \cdot f(x) dx}_{\text{Acts as density !!}} dy
 \end{aligned}$$

$\therefore f * g$ has density given by—

$$f * g(u) = \int_{-\infty}^{\infty} g(u-x) \cdot f(x) dx.$$

More general cases:

Given general X, Y .

Try to find $P(Y \leq a | X=x)$

Take x s.t.

$$P(X \in (x-\varepsilon, x+\varepsilon)) > 0 \quad \forall \varepsilon > 0$$

Find $P(Y \leq a | X \in (x-\varepsilon, x+\varepsilon)) \quad \forall \varepsilon > 0$
using "classical definition"

In general,
the classical defⁿ
does not work
as the conditioning
event $X=x$
might not have
positive probability.

Now, we'll do $\varepsilon \downarrow 0$ & check if a limit exists.

(If it exists, then it works!!
ie, it gives a candidate
for the conditional dist'.)
 $P(Y \leq a | X=x)$.

[Lebesgue
Differentiation
Theorem
(out
of
syllabus)]

Eg: $U \sim \text{Exp}(1)$ } Independent.

Eg: $U \sim \text{Exp}(1)$ } Independent
 $V \sim \text{Exp}(2)$

$$X = \max\{U, V\}$$

$$Y = V$$

to find: $P(Y \leq a \mid X=x)$

Fix $x > 0$. $P(Y \leq a \mid X=x) = 1$ if $a \geq x$

for $a < x$, take $\varepsilon > 0$ st.

$$\begin{aligned} P(Y \leq a \mid x \in (x-\varepsilon, x+\varepsilon)) \\ = P(Y \leq a, X \in (x-\varepsilon, x+\varepsilon)) \\ \equiv V \leq a, U \in (x-\varepsilon, x+\varepsilon) \end{aligned}$$

$$\therefore f(a|x) = \begin{cases} \text{density} & , a < x \\ 1 & , a \geq x \end{cases}$$

