

Aim : If $x \in L^1(\mathcal{A})$, \mathcal{F} σ -algebra
existence + uniqueness of $E[x | \mathcal{F}]$?
↳ proved

Tool: Consider $\nu \ll \mu$. Then
 $\exists f \geq 0$, measurable such
that for all $g \in \mathcal{B}_b$
 $\nu(g) = \mu(fg)$

(In particular $\nu(1_A) = \mu(f 1_A)$)

Conditional expectation: existence

Proposition 12.

On the probability space $(\Omega, \mathcal{F}_0, \mathbf{P})$ consider

- A σ -algebra $\mathcal{F} \subset \mathcal{F}_0$.
- $X \in \mathcal{F}_0$ such that $\mathbf{E}[|X|] < \infty$.

Then the random variable

satisfies \checkmark

$\mathbf{E}[X|\mathcal{F}]$ satisfies

(i) $\mathbf{E}[X|\mathcal{F}] \in \mathcal{L}^1$

(ii) $\mathbf{E}[\mathbf{E}[X|\mathcal{F}] 1_A] = \mathbf{E}[X 1_A]$

$\forall A \in \mathcal{F}$

exists and is uniquely defined.

Proof Hyp: $x \geq 0$

(1) Define a measure ν on (Ω, \mathcal{F})
by setting

$$\nu(A) = \mathbb{E}[X \cdot 1_A] \quad , \quad \forall A \in \mathcal{F}$$

Claim: ν is a measure, i.e.

(a) $\nu(\emptyset) = 0$  disjoint union

(b) $\nu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \nu(A_i)$

(c) $\nu(A) \geq 0 \quad \forall A \in \mathcal{F}$

Here (a) and (c) are clear.
Moreover, if A_i 's are disjoint

$$\begin{aligned}\nu\left(\bigcup_{i=1}^{\infty} A_i\right) &= E\left[X \mathbb{1}_{\bigcup_{i=1}^{\infty} A_i}\right] \\ &\stackrel{\text{disjoint}}{=} E\left[X \sum_{i=1}^{\infty} \mathbb{1}_{A_i}\right] \\ &= E\left[\sum_{i=1}^{\infty} X \mathbb{1}_{A_i}\right] \\ &\stackrel{\text{Fubini}}{=} \sum_{i=1}^{\infty} E\left[X \mathbb{1}_{A_i}\right] \\ &= \sum_{i=1}^{\infty} \nu(A_i) \Rightarrow \text{(b) verified}\end{aligned}$$

Thus ν is a measure

Recall : $\nu(A) = \mathbb{E}[X \mathbb{1}_A]$

Take $\mu = \mathbb{P}$. μ is a probability
on (Ω, \mathcal{F}_0) , but also on (Ω, \mathcal{F})

We can write $\mu(A) = \mathbb{E}[\mathbb{1}_A]$

Question : Do we have $\nu \ll \mu$?

If $\mu(A) = 0$, then $\mathbb{1}_A = 0$ a.s.

$\Rightarrow X \mathbb{1}_A = 0$ a.s.

yes!

$\Rightarrow \nu(A) = \mathbb{E}[X \mathbb{1}_A] = 0$

Radon-Nykodym: Since $\nu \ll \mu$,
 $\exists Y \in \mathcal{F}$ s.t. $\forall A \in \mathcal{F}$

$$\nu(\mathbb{1}_A) = \mathbb{E}[Y \mathbb{1}_A]$$

Thus $Y \in \mathcal{F}$ and for all $A \in \mathcal{F}$

$$\mathbb{E}[Y \mathbb{1}_A] = \mathbb{E}[Y \mathbb{1}_A]$$

Conclusion: $Y = \mathbb{E}[Y | \mathcal{F}]$

Proof of existence

Hypothesis: We have

- A σ -algebra $\mathcal{F} \subset \mathcal{F}_0$.
- $X \in \mathcal{F}_0$ such that $\mathbf{E}[|X|] < \infty$.
- $X \geq 0$.

Defining two measures: we set

- 1 $\mu = P$, measure on (Ω, \mathcal{F}) .
- 2 $\nu(A) \equiv \mathbf{E}[X \mathbf{1}_A] = \int_A X d\mathbf{P}$.

Then ν is a measure (owing to Beppo-Levi).

Proof of existence (2)

Absolute continuity: we have

$$\begin{aligned}\mathbf{P}(A) = 0 &\Rightarrow \mathbf{1}_A = 0 \quad P\text{-a.s.} \\ &\Rightarrow X \mathbf{1}_A = 0 \quad P\text{-a.s.} \\ &\Rightarrow \nu(A) = 0\end{aligned}$$

Thus $\nu \ll P$

Conclusion: invoking Radon-Nykodym, there exists $f \in \mathcal{F}$ such that, for all $A \in \mathcal{F}$, we have $\nu(A) = \int_A f \, d\mathbf{P}$.

\hookrightarrow We set $f = \mathbf{E}[X|\mathcal{F}]$.

Outline

1 Definition

- Baby conditional distributions: discrete case
- Baby conditional distributions: continuous case
- Definition with measure theory

2 Examples

3 Existence and uniqueness

4 Conditional expectation: properties

5 Conditional expectation as a projection

6 Conditional regular laws

Linearity, expectation

Proposition 13.

Let $X \in L^1(\Omega)$. Then

still a r.v.

$$\mathbf{E}\{\mathbf{E}[X|\mathcal{F}]\} = \mathbf{E}[X].$$

Proposition 14.

Let $\alpha \in \mathbb{R}$, and $X, Y \in L^1(\Omega)$. Then

$$\mathbf{E}[\alpha X + Y|\mathcal{F}] = \alpha \mathbf{E}[X|\mathcal{F}] + \mathbf{E}[Y|\mathcal{F}] \quad \text{a.s.}$$

Proof of linearity. Set

$$z = \alpha E[X|F] + E[Y|F]$$

We want to prove (i) and (ii) for z

(i) $z \in F$, since

$$E[X|F] \in F, \quad E[Y|F] \in F$$

and z is a linear combination of the two.

$$Z = \alpha E[X|F] + E[Y|F]$$

(ii) Take $A \in F$. Then

$$E[Z \mathbb{1}_A] \quad \text{r.v.} \quad \text{r.v.}$$

$$= E\left\{ (\alpha E[X|F] + E[Y|F]) \mathbb{1}_A \right\}$$

E linear

$$= \alpha E\{E[X|F] \mathbb{1}_A\} + E\{E[Y|F] \mathbb{1}_A\}$$

(ii)

$$= \alpha E[X \mathbb{1}_A] + E[Y \mathbb{1}_A]$$

linearity

$$= E[(\alpha X + Y) \mathbb{1}_A]$$

\Rightarrow (ii) verified

Proof That $\mathbb{E}\{\underbrace{\mathbb{E}[X|F]}_Y\} = \mathbb{E}[X]$

We have that $\forall A \in F$,

$$\mathbb{E}[Y \mathbb{1}_A] = \mathbb{E}[X \mathbb{1}_A]$$

In particular, $\Omega \in F$. We get

$$\mathbb{E}[Y \mathbb{1}_\Omega] = \mathbb{E}[X \mathbb{1}_\Omega]$$

$$\Rightarrow \boxed{\mathbb{E}[Y] = \mathbb{E}[X]}$$

Proof

Strategy: Check (i) and (ii) in the definition for the r.v

$$Z \equiv \alpha \mathbf{E}[X|\mathcal{F}] + \mathbf{E}[Y|\mathcal{F}].$$

Verification: we have

(i) Z is a linear combination of $\mathbf{E}[X|\mathcal{F}]$ and $\mathbf{E}[Y|\mathcal{F}]$

$$\hookrightarrow Z \in \mathcal{F}.$$

(ii) For all $A \in \mathcal{F}$, we have

$$\begin{aligned} \mathbf{E}[Z \mathbf{1}_A] &= E\{(\alpha \mathbf{E}[X|\mathcal{F}] + \mathbf{E}[Y|\mathcal{F}]) \mathbf{1}_A\} \\ &= \alpha E\{\mathbf{E}[X|\mathcal{F}] \mathbf{1}_A\} + E\{\mathbf{E}[Y|\mathcal{F}] \mathbf{1}_A\} \\ &= \alpha \mathbf{E}[X \mathbf{1}_A] + \mathbf{E}[Y \mathbf{1}_A] \\ &= \mathbf{E}[(\alpha X + Y) \mathbf{1}_A]. \end{aligned}$$

Monotonicity

Proposition 15.

Let $X, Y \in L^1(\Omega)$ such that $X \leq Y$ almost surely. We have

$$\mathbf{E}[X|\mathcal{F}] \leq \mathbf{E}[Y|\mathcal{F}],$$

almost surely.

Proof: Along the same lines as proof of uniqueness for the conditional expectation. For instance if we set

$$A_\varepsilon = \{\mathbf{E}[X|\mathcal{F}] - \mathbf{E}[Y|\mathcal{F}] \geq \varepsilon > 0\},$$

then it is readily checked that

$$\mathbf{P}(A_\varepsilon) = 0.$$

Proof. Take $X \leq Y$. We want
to prove that

$$E[X|F] \leq E[Y|F]$$

It is enough to prove that $\forall \varepsilon > 0$,
if

$$A_\varepsilon = (E[X|F] - E[Y|F] \geq \varepsilon),$$

Then

$$P(A_\varepsilon) = 0$$

$$A_\epsilon = (\mathbb{E}[X|F] - \mathbb{E}[Y|F] \geq \epsilon) \in F$$

We have

$\geq \epsilon$ on A_ϵ

$$\epsilon \mathbb{P}(A_\epsilon) \leq \mathbb{E} \{ (\mathbb{E}[X|F] - \mathbb{E}[Y|F]) \mathbb{1}_{A_\epsilon} \}$$

Linearity of $\mathbb{E}[\cdot|F]$

=

$$\mathbb{E} \{ \mathbb{E}[(X-Y)|F] \mathbb{1}_{A_\epsilon} \}$$

(iii)

$$= \mathbb{E} [(X-Y) \mathbb{1}_{A_\epsilon}]$$

≤ 0 since $X \leq Y$

$$\leq 0$$


$$\text{Thus } \epsilon \mathbb{P}(A_\epsilon) \leq 0$$

$$\Rightarrow \boxed{\mathbb{P}(A_\epsilon) = 0}$$

Monotone convergence

Proposition 16.

Let $\{X_n; n \geq 1\}$ be a sequence of random variables such that

- $X_n \geq 0$  $X_{n+1} \geq X_n$ a.s. and $X_n \rightarrow X$ a.s.
- $X_n \nearrow X$ almost surely
- $\mathbf{E}[X] < \infty$.

Then

$$\mathbf{E}[X_n | \mathcal{F}] \nearrow \mathbf{E}[X | \mathcal{F}] \text{ a.s.}$$

Recall: $X_n \nearrow X$ a.s. We set

$Y_n = X - X_n \stackrel{\geq 0}{\geq} 0$. We have $Y_n \searrow 0$ a.s.

Set $Z_n = E[Y_n | \mathcal{F}]$. We have

(i) Since $E[\cdot | \mathcal{F}]$ is monotone and $Y_n \searrow$, we have

$$Z_n \searrow \quad (Z_{n+1} \leq Z_n \text{ a.s.})$$

(ii) $Y_n \geq 0 \Rightarrow E[Y_n | \mathcal{F}] \stackrel{Z_n}{\geq} 0$

Then $\exists Z_\infty \geq 0$ s.t. $Z_n \searrow Z_\infty$

Summary : $Y_n \equiv X - X_n$
 $Z_n \equiv \mathbb{E}[Y_n | \mathcal{F}]$
 $Z_n \searrow Z_\infty$ a.s.
 $Z_\infty \geq 0$ a.s.

We wish to prove : $Z_\infty = 0$

Hint: If $Z_\infty \geq 0$, in order to prove that $Z_\infty = 0$, it is enough to prove $\mathbb{E}[Z_\infty] = 0$

$$\begin{aligned}\text{However } \mathbb{E}[Z_n] &= \mathbb{E}\{\mathbb{E}[Y_n | \mathcal{F}]\} \\ &= \mathbb{E}[Y_n]\end{aligned}$$

In addition, $Y_n \searrow 0 \xRightarrow{\text{Beppo-Levi}} \mathbb{E}[Y_n] \searrow 0$

We have obtained

$$\mathbb{E}[z_n] \rightarrow 0$$

In addition,

$$z_n \rightarrow z_\infty \text{ a.s.}$$

Beppo-Levi

$$\Rightarrow \mathbb{E}[z_n] \rightarrow \mathbb{E}[z_\infty]$$

Thus

$$\mathbb{E}[z_\infty] = 0 \Rightarrow z_\infty = 0 \text{ a.s.}$$

$$\Rightarrow \boxed{\text{Beppo-Levi for } \mathbb{E}[\cdot | \mathcal{F}]}$$

Proof

Strategy: Set $Y_n \equiv X - X_n$. We are reduced to show $Z_n \equiv \mathbf{E}[Y_n|\mathcal{F}] \searrow 0$.

Existence of a limit: $n \mapsto Y_n$ is decreasing, and $Y_n \geq 0$
 $\hookrightarrow Z_n$ is decreasing and $Z_n \geq 0$.
 $\hookrightarrow Z_n$ admits a limit a.s, denoted by Z_∞ .

Aim: Show that $Z_\infty = 0$.

Proof (2)

Expectation of Z_∞ : we will show that $\mathbf{E}[Z_\infty] = 0$. Indeed

- X_n converges a.s. to X .
- $0 \leq X_n \leq X \in L^1(\Omega)$.

Thus, by dominated convergence, $\mathbf{E}[X_n] \rightarrow \mathbf{E}[X]$.

We deduce:

- $\mathbf{E}[Y_n] \rightarrow 0$
- Since $\mathbf{E}[Y_n] = \mathbf{E}[Z_n]$, we also have $\mathbf{E}[Z_n] \rightarrow 0$.
- By monotone convergence, we have $\mathbf{E}[Z_n] \rightarrow \mathbf{E}[Z_\infty]$

This yields $\mathbf{E}[Z_\infty] = 0$.

Conclusion: $Z_\infty \geq 0$ and $\mathbf{E}[Z_\infty] = 0$

$\hookrightarrow Z_\infty = 0$ almost surely.