

Conditional expectation (end)

Example. Let $(X_{n,m})_{n,m \geq 0}$ a double sequence of i.i.d. r.v. with values in $\mathbb{N}_{\geq 0} = \{0, 1, 2, \dots\}$ (in particular $\mathbb{P}(X_{1,1} < \infty) = 1$). Let $Z_0 = 1$ and then I define recursively

$$Z_n = X_{n,1} + \dots + X_{n,Z_{n-1}},$$

for $n \geq 1$. Note that Z_n is a integer valued random variable.

The r.v.s $(Z_n)_{n \geq 0}$ model the evolution of the size of population of individuals where at every step n the m -th individual give rise to $X_{n,m} \sim X_{1,1}$ individuals which go the next generation.

We want to compute the generating function $f_n(\theta) := \mathbb{E}[\theta^{Z_n}]$ of Z_n for every $\theta \in (0, 1)$.

Note that we can more rigorously define

$$Z_n = \sum_{m \geq 1} \mathbb{1}_{Z_{n-1} \geq m} X_{n,m},$$

and observe that Z_0 it is a.s. finite and if Z_{n-1} is a.s. finite then the above sum is also made a.s. of finitely many summands and therefore Z_n is a.s. finite if $\mathbb{P}(X_{1,1} < \infty) = 1$, i.e. this proves that $\mathbb{P}(Z_n < \infty) = 1$ for all $n \geq 0$ by induction, assuming $\mathbb{P}(X_{1,1} < \infty) = 1$.

We know that $f_0(\theta) = \mathbb{E}[\theta^{Z_0}] = \theta$. Let us call $f(\theta) := \mathbb{E}[\theta^{X_{1,1}}]$. How we compute

$$f_n(\theta) = \mathbb{E}[\theta^{Z_n}] = \mathbb{E}[\theta^{X_{n,1} + \dots + X_{n,Z_{n-1}}}]?$$

We note that $(X_{n,m})_{m \geq 1}$ is independent of $Z_{n-1} \hat{=} \sigma(X_{k,m} : 0 \leq k \leq n-1, m \geq 1)$ since the family $(X_{n,m})_{n,m \geq 0}$ is iid. In this case we can condition the average on the value of Z_{n-1} : by the theorem we proved in the last lecture (Prop 14 in Note 2)

$$\mathbb{E}[\theta^{X_{n,1} + \dots + X_{n,Z_{n-1}}}] = \mathbb{E}[\mathbb{E}[\theta^{X_{n,1} + \dots + X_{n,Z_{n-1}}} | Z_{n-1}]] = \mathbb{E}[\varphi(Z_{n-1})]$$

with

$$\varphi(z) = \mathbb{E}[\theta^{X_{n,1} + \dots + X_{n,z}}], \quad z = 0, 1, 2, \dots$$

But now we can compute this easily since cond. exp. has disappeared thanks to independence, for $z \geq 0$

$$\varphi(z) = \mathbb{E}[\theta^{X_{n,1}}] \dots \mathbb{E}[\theta^{X_{n,z}}] = (\mathbb{E}[\theta^{X_{1,1}}])^z = (f(\theta))^z$$

which means that

$$f_n(\theta) = \mathbb{E}[\theta^{X_{n,1} + \dots + X_{n,Z_{n-1}}}] = \mathbb{E}[\varphi(Z_{n-1})] = \mathbb{E}[(f(\theta))^{Z_{n-1}}] = f_{n-1}(f(\theta)).$$

Therefore we have shown that f_n solve the recursive equation

$$f_0(\theta) = \theta, \quad f_n(\theta) = f_{n-1}(f(\theta))$$

which has unique solution $f_n(\theta) = f^{\circ n}(\theta)$. This is very useful if one would like to understand what happens to Z_n as $n \rightarrow \infty$, i.e. how the population behave on long time. Interesting question: does it become extinct with probability 1. Note that if $Z_n = 0$ then $Z_k = 0$ for all $k \geq n$.

Regular conditional probabilities

Given a σ -algebra $\mathcal{G} \subseteq \mathcal{F}$ we can consider the family of r.v.

$$A \in \mathcal{F} \mapsto \mathbb{P}(A|\mathcal{G}) := \mathbb{E}[\mathbb{1}_A|\mathcal{G}] \in L^\infty(\Omega, \mathcal{G}, \mathbb{P}) \subseteq L^\infty(\Omega, \mathcal{F}, \mathbb{P})$$

These r.v. satisfy **almost surely** the following equalities:

- a) $\mathbb{P}(\emptyset|\mathcal{G}) = 0, \mathbb{P}(A^c|\mathcal{G}) = 1 - \mathbb{P}(A|\mathcal{G})$
- b) $\mathbb{P}(\cup_n A_n|\mathcal{G}) = \sum_n \mathbb{P}(A_n|\mathcal{G})$ for a family $(A_n)_{n \geq 1} \subseteq \mathcal{F}$ of pairwise disjoint events.

These relations shows that the map $A \in \mathcal{F} \mapsto \mathbb{P}(A|\mathcal{G})$ behaves like a probability measure. So we would like to think to $\mathbb{P}(\cdot|\mathcal{G})$ as a random probability measure $\mathbb{P}_{\mathcal{G}}$, i.e. something like

$$\mathbb{P}_{\mathcal{G}}: \omega \in \Omega \mapsto \mathbb{P}_{\mathcal{G}}(\omega) \in \Pi(\Omega, \mathcal{F})$$

where $\Pi(\Omega, \mathcal{F})$ is the set of all probability measures on (Ω, \mathcal{F}) . This is in this generality not possible, because the properties a),b) are true only a.s., that is for any choice of A or of $(A_n)_n$ one has different exceptional set $\mathcal{N}_A, \mathcal{N}_{(A_n)_n}$ in which the property is not satisfied and unfortunately one cannot construct an exceptional universal measurable set valid for all the possible choices of $A, (A_n)$, because we have uncountably many choices here.

In some situations however this is possible. In this case we say that the family $(\mathbb{P}(A|\mathcal{G}))_{A \in \mathcal{F}}$ admits a **regular conditional version**, or that we have a **regular conditional probability** for \mathbb{P} given \mathcal{G} . This means that there exists a map

$$\mathbb{P}_{\mathcal{G}}: \Omega \rightarrow \Pi(\Omega, \mathcal{F}) \subseteq (\Omega \times \mathcal{F} \rightarrow [0, 1])$$

such that for all $A \in \mathcal{F}$ it holds

$$\mathbb{P}_{\mathcal{G}}(\omega, A) = \mathbb{P}(A|\mathcal{G})(\omega), \quad \mathbb{P} - a.e. \omega \in \Omega.$$

In case we have a regular conditional probability, then for any $X \in L^1(\Omega, \mathcal{F}, \mathbb{P})$ we can express the conditional expectation wrt. \mathcal{G} as an integral:

$$\mathbb{E}[X|\mathcal{G}](\omega) = \int X(\omega') \mathbb{P}_{\mathcal{G}}(\omega, d\omega'), \quad \mathbb{P} - a.e. \omega \in \Omega.$$

(this is not difficult to prove from the definition of reg. cond. prob. and cond. exp.)

Regular conditional probabilities are guaranteed to exist when (Ω, \mathcal{F}) is a **Polish space**: i.e. complete, metrisable topological space endowed with the Borel σ -algebra.

Example: $\mathbb{R}, \mathbb{R}^n, \mathbb{R}^{\mathbb{N}}, C([0, 1], \mathbb{R}^n)$.

Filtrations & stopping times

Definition. A (discrete time) stochastic process is just a family $(X_n)_{n \geq 0}$ of random variables indexed by $\mathbb{N}_{\geq 0}$ (or $\mathbb{N}_{\geq 1}$).

We think to the index n at *time* and to the sequence X_1, X_2, \dots as the description of random phenomena which evolves in time. Time take with it a notion of “past”, “present” and “future”.

This is encoded in the notion of *filtration*:

Definition. A filtration $(\mathcal{F}_n)_{n \geq 0}$ is an increasing family of sub- σ -algebras of \mathcal{F} , i.e.

$$\mathcal{F}_n \subseteq \mathcal{F}_{n+1}$$

for all $n \geq 0$. We let $\mathcal{F}_\infty = \sigma(\mathcal{F}_n; n \geq 0)$, that is the smallest σ -algebra which contains all the $(\mathcal{F}_n)_n$.

A filtration represents the flow of time, in the sense that \mathcal{F}_n is the information I dispose at time n .

Example. If $X = (X_n)_{n \geq 0}$ is a stochastic process then we can always consider its *natural filtration* $(\mathcal{F}_n^X)_{n \geq 0}$

$$\mathcal{F}_n^X = \sigma(X_0, X_1, \dots, X_n).$$

It is easy to check that indeed $\mathcal{F}_n^X \subseteq \mathcal{F}_{n+1}^X$. This filtration encode the information given by the observation of the process X as time passes.

Example. Let $\Omega = (0, 1]$ and define

$$\mathcal{F}_n = \{(k/2^n, (k+1)/2^n] : k = 0, \dots, 2^n - 1\} \subseteq \mathcal{F} = \mathcal{B}([0, 1])$$

then $(\mathcal{F}_n)_{n \geq 0}$ is a filtration and $\mathcal{F}_\infty = \mathcal{F}$. Here n represent the precision of our observation of a point in $[0, 1]$.

Definition. We say that a process $X = (X_n)_{n \geq 0}$ is adapted to the filtration $(\mathcal{F}_n)_{n \geq 0}$ iff

$$X_n \hat{\in} \mathcal{F}_n$$

for all $n \geq 0$. A process X is previsible to the filtration $(\mathcal{F}_n)_{n \geq 0}$ iff

$$X_{n+1} \hat{\in} \mathcal{F}_n$$

for all $n \geq 0$.

The natural filtration \mathcal{F}^X of a process X is the smallest filtration for which the process is adapted.

I want to define now stopping times. A stopping time is a rule to determine how to stop given what happened in the past.

Definition. A stopping time $T: \Omega \rightarrow \mathbb{N}^* := \mathbb{N} \cup \{+\infty\}$ for the filtration $(\mathcal{F}_n)_{n \geq 0}$ is a r.v. with values in \mathbb{N}^* such that

$$\{T \leq n\} \in \mathcal{F}_n,$$

for all $n \geq 0$.

This is equivalent to require that $\{T = n\} \in \mathcal{F}_n$ for all $n \geq 0$.

Example. The first time T we observe “head” in a repeated launch of a coin, is a stopping time wrt. the natural filtration of this problem. However, the last time S I observe “head” is not a stopping time.

The notion of stopping time encode a “fair” stopping rule, i.e. a rule which does not use information from future to make a decision (to stop or no).

Definition. The σ -algebra \mathcal{F}_T of the stopping time T wrt. the filtration $(\mathcal{F}_n)_{n \geq 0}$ is defined as

$$\mathcal{F}_T := \{A \in \mathcal{F} : A \cap \{T \leq n\} \in \mathcal{F}_n \text{ for all } n \in \mathbb{N}^*\}$$

Exercise. Show that \mathcal{F}_T is a σ -algebra.
