

Lecture 18. Proof of Theorem 17.6. Other forms of the strong Markov property.

Proof of Theorem 17.6. It's enough to prove that for every $f \in \mathbf{C}$ and for every $(\mathcal{F}_{\leq t+})$ -stopping time τ

$$E(f(\xi_{\tau+s})|\mathcal{F}_{\leq s+}) = P^s f(\xi_\tau), \tag{18.1}$$

where we replace $f(\xi_\infty)$, $P^t f(\xi_\infty)$ with 0.

This means two things: the simpler one, that the right-hand side is $\mathcal{F}_{\leq \tau+}$ -measurable; the more complicated one, that for every $A \in \mathcal{F}_{\leq \tau+}$

$$E(I_A \cdot f(\xi_{\tau+s})) = E(I_A \cdot P^s f(\xi_\tau)). \tag{18.2}$$

Let us introduce new stopping times, discrete ones: for every $h > 0$ take

$$\tau^h = ih \quad \text{for } (i-1)h \leq \tau < ih, \quad \tau^h = \infty \quad \text{if } \tau = \infty. \tag{18.3}$$

We have:

$$\{\tau^h \leq t\} = \bigcup_{i: ih \leq t} \{(i-1)h \leq \tau < ih\} = \bigcup_{i: ih \leq t} [\{\tau < ih\} \setminus \{\tau < (i-1)h\}]. \tag{18.4}$$

Since τ is a $(\mathcal{F}_{\leq t+})$ -stopping time, the event $\{\tau < ih\}$ in the right-hand side belongs to $\mathcal{F}_{\leq ih}$, and $\{\tau < (i-1)h\}$ belongs to $\mathcal{F}_{\leq (i-1)h}$. All these events belong to $\mathcal{F}_{\leq t}$, and $\{\tau^h \leq t\} \in \mathcal{F}_{\leq t}$. So the random variables τ^h are stopping times *with respect to the non-extended σ -algebras $\mathcal{F}_{\leq t}$* .

Our next move: the random variable ξ_{τ^h} is $\mathcal{F}_{\leq \tau^h}$ -measurable. This means that for every $C \in \mathcal{X} = \mathcal{B}_X$ and every $t \in [0, \infty)$

$$\{\xi_{\tau^h} \in C\} \cap \{\tau^h \leq t\} \in \mathcal{F}_{\leq t}. \tag{18.5}$$

We have:

$$\begin{aligned} \{\xi_{\tau^h} \in C\} \cap \{\tau^h \leq t\} &= \bigcup_{i: ih \leq t} \{\xi_{\tau^h} \in C\} \cap \{\tau^h = ih\} \\ &= \bigcup_{i: ih \leq t} \{\xi_{ih} \in C\} \cap [\{\tau < ih\} \setminus \{\tau < (i-1)h\}]. \end{aligned} \tag{18.6}$$

All events here belong to $\mathcal{F}_{\leq t}$, which proves (18.5).

And the next move: $\tau^h \rightarrow \tau$ as $h \rightarrow 0^+$, $\tau^h > \tau$ if $\tau < \infty$, so by the right continuity $\xi_\tau = \lim_{h \rightarrow 0^+} \xi_{\tau^h}$ (for $\tau < \infty$). For every $\delta > 0$ the random variables τ^h , $0 < h \leq \delta$, are measurable with respect to $\mathcal{F}_{\leq \tau+\delta}$, so their limit is also measurable with respect to this

σ -algebra. Since δ is an arbitrary positive number, we get that the random variable ξ_τ is $\mathcal{F}_{\leq \tau+}$ -measurable.

Now to (18.2). For $f \in \mathbf{C}$, $P^s f \in \mathbf{C}$ we have: $f(\xi_{\tau+s}) = \lim_{h \rightarrow 0^+} f(\xi_{\tau^h+s})$, $P^s f(\xi_{\tau+s}) = \lim_{h \rightarrow 0^+} P^s f(\xi_{\tau^h+s})$; so it's enough to prove

$$E(I_A \cdot f(\xi_{\tau^h+s})) = E(I_A \cdot P^s f(\xi_{\tau^h})). \quad (18.7)$$

But this follows from Theorem 17.5.

In fact, we repeated, with some additions, the proof of Theorem 14.4.

A particular case of Theorem 17.6: all our continuous-time Markov chains are strong Markov (the space \mathbf{C} consists of *all* bounded functions, so $P^t \mathbf{C} \subseteq \mathbf{C}$).

Why is the process $\tilde{\xi}_t$ of Example 17.2 not strong Markov? Its trajectories are continuous, so the only thing is that the transition probabilities (17.15) don't satisfy the Feller property: indeed, $\tilde{P}(t, x, \bullet)$ does not depend on x in a weakly continuous way: the normal distributions $\tilde{P}(t, x, \bullet)$ do not converge to $\tilde{P}(t, 0, \bullet) = \delta_0(\bullet)$ as $x \rightarrow 0$, but rather to the normal distribution with parameters $(0, t)$.

Other forms of the strong Markov property.

Just as from the ordinary Markov property we deduced

$$P(\theta_t^{-1} A | \mathcal{F}_t) = P_{\xi_t}(A) \quad (18.8)$$

first for $A = \{(\xi_{s_1}, \dots, \xi_{s_n}) \in C_n\}$, and then for general $A \in \mathcal{F}_{\geq 0}$, we can deduce other forms of the strong Markov property. But first we need to consider the shift operators θ_τ and θ_τ^{-1} .

The operator θ_τ is defined as taking an $\omega \in \Omega$ to $\theta_{\tau(\omega)}\omega$. But there is no such thing as a shift by an infinite space $\theta_\infty\omega$; so $\theta_{\tau(\omega)}\omega$ is *not defined* for ω 's with $\tau(\omega) = \infty$. So the operator θ_τ is defined on the set $\{\tau < \infty\}$ only: $\theta_\tau: \{\tau < \infty\} \mapsto \Omega$ (not necessarily onto Ω). The inverse image $\theta_\tau^{-1}A$ of every subset $A \subset \Omega$ is a subset of the event $\{\tau < \infty\}$.

So we formulate the new form of the strong Markov property:

$$P(\theta_\tau^{-1} A | \mathcal{F}_\tau) = P_{\xi_\tau}(A), \quad (18.9)$$

where $P_{\xi_\tau}(A)$ is replaced with 0 if $\tau = \infty$ (and ξ_τ makes no sense). It is proved in exactly the same way as the similar form of the ordinary Markov property, for $A = \{(\xi_{s_1}, \dots, \xi_{s_n}) \in C_n\}$ first, and for arbitrary $A \in \mathcal{F}_{\geq 0}$ after that.

Example 18.1. For continuous-time Markov chains, taking $\tau = \tau_1$ and $A = \{\tau_1 \in C\}$, $C \in \mathcal{B}_{(0, \infty)}$, we have: $\theta_{\tau_1}^{-1}A = \{\tau_2 - \tau_1 \in C\}$; and the strong Markov property yields

$$P\{\tau_2 - \tau_1 \in C | \mathcal{F}_{\tau_1}\} = P_{\xi_{\tau_1}}\{\tau_1 \in C\} = \int_C v_{\xi_{\tau_1}} e^{-v_{\xi_{\tau_1}} s} ds: \quad (18.10)$$

the conditional distribution is exponential with parameter $v_{\xi_{\tau_1}}$. In particular,

$$\begin{aligned} P\{\tau_2 - \tau_1 \in C | \tau_1, \xi_{\tau_1}\} &= E(P\{\tau_2 - \tau_1 \in C | \mathcal{F}_{\leq \tau_1}\} | \tau_1, \xi_{\tau_1}) \\ &= E(P_{\xi_{\tau_1}}\{\tau_1 \in C\} | \tau_1, \xi_{\tau_1}) = P_{\xi_{\tau_1}}\{\tau_1 \in C\}. \end{aligned} \quad (18.11)$$

We have already got it in Theorem 9.2; in fact its proof and that of Theorem 9.1 was using the strong Markov property for the discrete Markov chain $\xi_0, \xi_{1/2^n}, \xi_{2/2^n}, \dots$, and then the limit passage to go from the discrete stopping times τ_1^n to the continuous τ_1 – just the same as in the proof of Theorem 17.6.

Example 18.2. Not only the Wiener process starting from 0 at time 0 changes its sign infinitely many times in any small interval of time from 0 to δ , but this also occurs almost surely after the first time $\tau_0 = \min\{t: \xi_t = 0\}$ of reaching 0.

Another form of the strong Markov property:

Let $u \in T$; let τ be a stopping time. Then for $C \in \mathcal{X}$ almost surely on the set $\{\tau \leq u\}$

$$P\{\xi_u \in C \mid \mathcal{F}_\tau\} = P(u - \tau, \xi_\tau, C) \quad (18.12)$$

(on the complement $\{\tau > u\}$ of this set the left-hand-side conditional probability is obviously equal to the random variable $I_{\{\xi_u \in C\}} = I_C(\xi_u)$).

The equality (18.12) can be deduced from (17.12), (18.9); but the proof could be rather long and complicated, so better I formulate the following couple of theorems:

Theorem 18.1. Let $\xi_t, t \in T = \{0, h, 2h, \dots, nh, \dots\}$, be a Markov process; τ a stopping time for it, and $u \in T$. Then (18.12) is satisfied almost surely on the event $\{\tau \leq u\}$.

The **proof** is the standard one: we subdivide the probabilities and expectations involved into their parts on the events $\{\tau = ih\}$, and apply to each summand the ordinary Markov property with respect to the non-random time ih .

Before I formulate the next theorem, let me tell you this. If the process $\xi_t, t \geq 0$, has right-continuous trajectories, the function $P^t f(x), f \in \mathbf{C}$, is right-continuous in t for every x : $\lim_{t' \rightarrow t+} P^{t'} f(x) = \lim_{t' \rightarrow t+} E_x f(\xi_{t'}) = E_x \lim_{t' \rightarrow t+} f(\xi_{t'}) = E_x f(\xi_t) = P^t f(x)$. If $P^t \mathbf{C} \subseteq \mathbf{C}$, the function $P^t f(x)$ is continuous in x for every t . But continuity in x for every t , plus (right) continuity in t for every x is not the same as continuity in the pair (t, x) (from the right for the argument t). I did not take this into account when I formulated Theorem 18.2 in the lecture. I am correcting this in the lecture note.

Theorem 18.2. Let $\xi_t, t \geq 0$, be a Markov process on a metric space X having right-continuous trajectories and such that $\lim_{s \rightarrow t+, y \rightarrow x} P^s f(y) = P^t f(x)$; let τ be a stopping time with respect to the σ -algebras $\mathcal{F}_{\leq t+}$, and $u \in [0, \infty)$. Then (18.12) is satisfied almost surely on the event $\{\tau \leq u\}$ with $\mathcal{F}_{\leq \tau+}$ as \mathcal{F}_τ .

Proof. We reformulate the statement in terms of expectations:

$$E(f(\xi_u) \mid \mathcal{F}_{\leq \tau+}) = P^{u-\tau} f(\xi_\tau) \quad (18.13)$$

on $\{\tau \leq u\}$ for every $f \in \mathbf{C}$. This means $\mathcal{F}_{\leq \tau+}$ -measurability of the right-hand side, plus

$$E(I_B \cdot f(\xi_u)) = E(I_B \cdot P^{u-\tau} f(\xi_\tau)) \quad (18.14)$$

for every event $B \in \mathcal{F}_{\leq \tau+}, B \subseteq \{\tau \leq u\}$.

Then we define τ^h by (18.3), and $u^h = (i+1)h$. For every $h > 0$ the random variable τ^h is a stopping time with respect to the non-extended σ -algebras $\mathcal{F}_{\leq t}$; and the event $B \in \mathcal{F}_{\leq \tau^h}$. By Theorem 18.1,

$$E(f(\xi_{u^h}) \mid \mathcal{F}_{\leq \tau^h}) = P^{u^h - \tau^h} f(\xi_{\tau^h}) \quad (18.15)$$

on the event $\{\tau^h \leq u^h\}$. The event $B \subseteq \{\tau \leq u\} \subseteq \{\tau^h \leq u^h\}$, so we have:

$$E(I_B \cdot f(\xi_{u^h})) = E(I_B \cdot P^{u^h - \tau^h} f(\xi_{\tau^h})). \quad (18.16)$$

Then we take $h \rightarrow 0^+$. We have $\tau^h \rightarrow \tau$, $\tau^h > \tau$, $\xi_{\tau^h} \rightarrow \xi_\tau$, $u^h \rightarrow u$, $u^h > u$, $u^h - \tau^h \rightarrow u - \tau$, $u^h - \tau^h > u - \tau$ (this is why we added the additional h defining u^h). We have $\lim_{h \rightarrow 0^+} P^{u^h - \tau^h} f(\xi_{\tau^h}) = P^{u - \tau} f(\xi_\tau)$; and by the dominated-convergence theorem we get (18.14).

The semigroup P^t associated with the Wiener process satisfies the condition of (t, x) -continuity of $P^t f(x)$ for every $f \in \mathbf{C}$, so this form of the strong Markov property can be applied to it. The example will be considered in the next lecture; in particular, the distribution of the first time τ_0 of reaching 0 will be found.