18.175: Lecture 39

Last lecture

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Strong Markov property

Strong Markov property

- Write $\mathcal{F}_s^o = \sigma(B_r : r \leq s)$.
- Write $\mathcal{F}_s^+ = \cap_{t>s} \mathcal{F}_t^o$
- Note right continuity: $\cap_{t>s} \mathcal{F}_t^+ = \mathcal{F}_s^+$.
- ▶ 𝓕⁺_s allows an "infinitesimal peek at future"

► Expectation equivalence theorem If Z is bounded and measurable then for all s ≥ 0 and x ∈ ℝ^d have

$$E_{x}(Z|\mathcal{F}_{s}^{+})=E_{x}(Z|\mathcal{F}_{s}^{o}).$$

- ▶ Proof idea: Consider case that Z = ∑_{i=1}^m f_m(B(t_m)) and the f_m are bounded and measurable. Kind of obvious in this case. Then use same measure theory as in Markov property proof to extend general Z.
- ▶ Observe: If Z ∈ F⁺_s then Z = E_x(Z|F^o_s). Conclude that F⁺_s and F^o_s agree up to null sets.

- If A ∈ F₀⁺, then P(A) ∈ {0,1} (if P is probability law for Brownian motion started at fixed value x at time 0).
- There's nothing you can learn from infinitesimal neighborhood of future.
- ▶ **Proof:** If we have $A \in \mathcal{F}_0^+$, then previous theorem implies

$$1_A=E_x(1_A|\mathcal{F}_0^+)=E_x(1_A|\mathcal{F}_0^o)=P_x(A)\quad P_x\text{a.s.}$$

▶ If $s \ge 0$ and Y is bounded and C-measurable, then for all $x \in \mathbb{R}^d$, we have

$$E_{x}(Y \circ \theta_{s} | \mathcal{F}_{s}^{+}) = E_{B_{s}}Y,$$

where the RHS is function $\phi(x) = E_x Y$ evaluated at $x = B_s$.

Proof idea: First establish this for some simple functions Y (depending on finitely many time values) and then use measure theory (monotone class theorem) to extend to general case.

- If $\tau = \inf\{t \ge 0 : B_t > 0\}$ then $P_0(\tau = 0) = 1$.
- If $T_0 = \inf\{t > 0 : B_t = 0\}$ then $P_0(T_0 = 0) = 1$.
- If B_t is Brownian motion started at 0, then so is process defined by X₀ = 0 and X_t = tB(1/t). (Proved by checking E(X_sX_t) = stE(B(1/s)B(1/t)) = s when s < t. Then check continuity at zero.)

Strong Markov property

Strong Markov property

- A random variable S taking values in [0,∞] is a stopping time if for all t ≥ 0, we have {S > t} ∈ F_t.
- ▶ Distinction between {S < t} and {S ≤ t} doesn't make a difference for a right continuous filtration.</p>
- ► Example: let S = inf{t : B_t ∈ A} for some open (or closed) set A.

Strong Markov property

Let (s, ω) → Y_s(ω) be bounded and R × C-measurable. If S is a stopping time, then for all x ∈ ℝ^d

$$E_x(Y_S \circ \theta_S | \mathcal{F}_S) = E_{B(S)}Y_S \text{ on } \{S < \infty\},$$

where RHS means function $\phi(x, t) = E_x Y_t$ evaluated at x = B(S), and t = S.

- In fact, similar result holds for more general Markov processes (Feller processes).
- Proof idea: First consider the case that S a.s. belongs to an increasing countable sequence (e.g., S is a.s. a multiple of 2⁻ⁿ). Then this essentially reduces to discrete Markov property proof. Then approximate a general stopping time by a discrete time by rounding down to multiple of 2⁻ⁿ. Use some continuity estimates, bounded convergence, monotone class theorem to conclude.
- Extend optional stopping to continuous martingales similarly.
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Continuous martingales

- **Question:** If B_t is a Brownian motion, then is $B_t^2 t$ a martingale?
- **Question:** If B_t and \tilde{B}_t are independent Brownian motions, then is $B_t \tilde{B}_t$ a martingale?
- **Question:** If B_t is a martingale, then is $e^{B_t t/2}$ a martingale?
- ► Question: If B_t is a Brownian motion in C (i.e., real and imaginary parts are independent Brownian motions) and f is an analytic function on C, is f(B_t) a complex martingale?
- ► Question: If B_t is a Brownian motion on ℝ^d and f is a harmonic function on ℝ^d, is f(B_t) a martingale?
- Question: Suppose B_t is a one dimensional Brownian motion, and g_t : C → C is determined by solving the ODE

$$\frac{\partial}{\partial t}g_t(z) = \frac{2}{g_t(z) - 2B_t}, \quad g_0(z) = z.$$

Is
$$\arg(g_t(z) - W_t)$$
 a martingale?

- Course has reached finite stopping time. Process goes on.
- Future probability graduate courses include
 - 18.177: fall 2014 (Jason Miller)
 - 18.177: spring 2015 (Alice Giuonnet)
 - 18.176: fall or spring 2015-16
- Probability seminar: Mondays at 4:15.
- I am happy to help with quals and reading.
- Talk to friendly postdocs: Vadim Gorin, Jason Miller, Jonathon Novak, Charlie Smart, Nike Sun, Hao Wu.
- Thanks for taking the class!

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