

From semigroup to Markov process

last time: constructed $\{\tilde{X}_t: t \in \mathbb{Q}^+\}$ with "correct" distribution.

Next step: Extract càdlàg-version of this process.

Key idea: $e^{-\lambda t} T(t) (R(\lambda)g)^f$

$$= \int_t^\infty e^{-\lambda s} T(s)g \leq \int_0^\infty e^{-\lambda s} T(s)g = R(\lambda)g. \quad (\text{by pos. pres.})$$

For f we then get

$$e^{-\lambda(t+s)} T(t+s)f \leq e^{-\lambda s} T(s)f$$

Rewrite using $\tilde{X}$: $E\left(e^{-\lambda(t+s)} f(\tilde{X}_{t+s}) \middle| \mathcal{F}_s\right) \leq e^{-\lambda s} f(\tilde{X}_s)$

$\mathcal{F}_s = \sigma(\tilde{X}_u: u \leq s)$

So $\{e^{-\lambda t} f(\tilde{X}_t) : t \geq 0\}$ is ≥ 0 supermartingale.

Now: Let M_t be ≥ 0 supermartingale.

Then

$$P\left(\forall t \geq 0: \lim_{\substack{s \downarrow t \\ (s \in \mathbb{Q})}} M_s \text{ exists} \wedge \forall t \geq 0: \lim_{\substack{s \uparrow t \\ (s \in \mathbb{Q})}} M_s \text{ exists}\right) = 1.$$

and $\forall t \geq 0: P(M_t > 0) = 1$

implies

$$P(\forall t \geq 0: M_t > 0) = 1$$

Using test

V_{XES} :

$$P^x(\forall t \geq 0:$$

V_t

Now define

X_t :

and conclude

Using test functions we now conclude

$\forall x \in S$:

$$P^x \left(\forall t > 0: \lim_{\substack{s \uparrow t \\ s \in \mathbb{Q}}} \tilde{X}_s \text{ exists in } S \right)$$

$$= 1$$

Now define

$$X_t := \lim_{\substack{s \uparrow t \\ s \in \mathbb{Q}}} \tilde{X}_s$$

and conclude $t \mapsto X_t$ càdlàg a.s.

Ex

Def

Then

asso

Thm 1
(1)

(2)

(3)

Ex Standard Brownian motion:

Def Let $S = \mathbb{R}$, $T(t)f(x) = \mathbb{E}(f(x+N(0,t)))$
Then $\{T(t)\}$ is prob. semigroup and the
associated M.P. is called standard Brownian
SBM motion

Thm Let $X = \text{SBM}$. Then

- (1) $\forall n \geq 1 \forall 0 < t_0 < t_1 \dots < t_n$:
 $\{X_{t_i} - X_{t_{i-1}}\}_{i=1}^n$ are independent.
- (2) $\forall 0 \leq s < t$: $X_t - X_s = N(0, t-s)$
- (3) $t \mapsto X_t$ is a.s. continuous.

Pf: MP definition: $\forall s, t \geq 0$:

$$E^x(f(X_{t+s}) | \mathcal{F}_s) \stackrel{\text{MP}}{=} E^{\bar{X}_s}(f(X_t))$$

$$= T(t)f(\bar{X}_s) = E(f(\cdot + N(0, t)) | \bar{X}_s)$$

Now apply (7) to $f(x) = e^{i\lambda x}$.

$$E^x(e^{i\lambda X_{t+s}} | \mathcal{F}_s) = e^{i\lambda \bar{X}_s} \underbrace{E(e^{i\lambda N(0, t)})}_{e^{-\frac{\lambda^2}{2}t}}$$

So $\forall s \leq t$: $E^x(e^{i\lambda(\bar{X}_t - \bar{X}_s)} | \mathcal{F}_s) = e^{-\frac{\lambda^2}{2}(t-s)}$

This gives $E^x\left(e^{i\sum_{j=1}^n \lambda_j (X_{t_j} - X_{t_{j-1}})}\right) = e^{-\frac{1}{2}\sum_{j=1}^n \lambda_j^2 (t_j - t_{j-1})}$

proving (1-2). For (3), need Kolmogorov-Censtov Thm (2750)

Analogue of Q-matrix $\rightarrow$ generator

Def Given any family $\{T(t)\}_{t \geq 0}$ of bounded linear operators on Banach space $\mathcal{B}$, define

$$\text{Dom}(\mathcal{L}) := \left\{ f \in \mathcal{B} : \lim_{t \downarrow 0} \frac{T(t)f - f}{t} \text{ exists} \right\}$$

and for $f \in \text{Dom}(\mathcal{L})$ set

$$\mathcal{L}f := \lim_{t \downarrow 0} \frac{T(t)f - f}{t}$$

$\nwarrow$ linear space

Note For conti. semigroups this will be densely defined, linear, (unbounded).

Thm Let $\{T(t)\}_{t \geq 0}$ be strongly continuous
semigroup of contractions on Banach space B .

Then:

(1) $\text{Dom}(L)$ is dense in B

(2) $\forall \lambda > 0: \text{Ran}(1 - \lambda^{-1}L) = B$

(3) $\forall \lambda > 0: \|(1 - \lambda^{-1}L)^{-1}\| \leq 1$

(4) $\forall t > 0 \forall f \in B: T(t)f = \lim_{n \rightarrow \infty} (1 - \frac{t}{n}L)^{-n} f$

(5) $\forall t > 0 \forall f \in \text{Dom}(L):$

$$\frac{d}{dt} T(t)f = T(t)Lf = L T(t)f$$

and $T(t)f \in \text{Dom}(L)$ $\nearrow$

Pf Key tool: ^{resolvent}

$$R(\lambda)f := \int_0^\infty e^{-\lambda t} T(t)f \, dt$$

(defined as improper Riemann integral)

Note Set $f := \lambda R(\lambda)g$.

Then

$$\begin{aligned} T(t)f - f &= \lambda e^{\lambda t} \int_t^\infty e^{-\lambda s} T(s)g \, ds \\ &\quad - \lambda \int_0^\infty e^{-\lambda s} T(s)g \, ds \\ &= (e^{\lambda t} - 1)f - \lambda e^{\lambda t} \int_0^t e^{-\lambda s} T(s)g \, ds \end{aligned}$$

$$\text{So } \frac{T(t)f - f}{t} \xrightarrow{t \downarrow 0} \lambda f - \lambda g$$

So f
proving

Fact to

Then we

$$\| (1 - \lambda R(\lambda))f \|$$

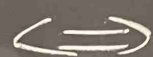
$$\text{So } \| (1 - \lambda R(\lambda))f \|$$

For (5):

$$(1 - \lambda R(\lambda))f$$

So $f = \lambda R(\lambda)g \Rightarrow f \in \text{Dom}(L) \wedge g = f - \lambda^{-1}Lf$
 proving (1) and (2).

Fact to prove now:



holds here.
 (by showing (4) first)

Then we get

$$\begin{aligned} \|(1 - \lambda^{-1}L)^{-1}g\| &= \|f\| = \|\lambda R(\lambda)g\| \\ &\leq \lambda \int_0^\infty e^{-\lambda t} \underbrace{\|T(t)g\|}_{\leq \|g\|} dt \leq \|g\| \end{aligned}$$

So $\|(1 - \lambda^{-1}L)^{-1}\| \leq 1$

For (5):

$$\begin{aligned} (1 - \lambda^{-1}L)^{-n}f &= \lambda^n R(\lambda)^n f \\ &= \int_{\mathbb{R}_+^n} dt_1 \cdots dt_n \lambda^n e^{-\lambda(t_1 + \cdots + t_n)} T(t_1 + \cdots + t_n)f \end{aligned}$$

$$\text{So } (1 - \lambda^{-1} L)^{-n} f = \mathbb{E} \left(T \left(\frac{1}{\lambda} (Y_1 + \dots + Y_n) \right) f \right)$$

where $Y_1, \dots, Y_n = \text{iid Exp}(1)$.

$$\text{So } (1 - \frac{t}{n} L)^{-n} f = \mathbb{E} \left(T \left(t \underbrace{\frac{Y_1 + \dots + Y_n}{n}}_{\xrightarrow[n \rightarrow \infty]{\text{SLW}} 1} \right) f \right)$$

$$\xrightarrow[n \rightarrow \infty]{} T(t)f. \quad \boxed{\times}$$

Note:

$$e^{tL}$$

OR

$$\left(1 + \frac{t}{n} L \right)^n$$

doesn't make sense in general

$$\left(1 - \frac{t}{n} L \right)^{-n} \text{ does.}$$