

Sample Path Regularity for One-Dimensional Diffusion Processes

Stanley Sawyer — Washington University — Vs. September 17, 2007

1. Basic Assumptions. Assume that

- (i) $\{T_t\}$ is a strongly-continuous semigroup of bounded linear operators on the Banach space

$$B = C_0(I) = \{f \text{ continuous on } I = [0, 1] : f(0) = f(1) = 0\}$$

- (ii) If $f(x) \geq 0$ and $f \in B$, then $T_t f(x) \geq 0$ and $T_t f(x) \leq \max_{y \in I} f(y)$.

In general, the infinitesimal generator of a strongly-continuous semigroup of linear operators T_t on any Banach space B is defined by

$$Af = \lim_{h \rightarrow 0} (T_h f - f)/h \tag{1.1}$$

on the set

$$\mathcal{D} = \mathcal{D}(A) = \{f : \text{The limit in (1.1) exists in the norm of } B\}$$

Suppose further that

- (iii) The operator A and set $\mathcal{D}(A)$ satisfy the following conditions. Let

$$Lf(x) = \frac{d}{dm(x)} \frac{d}{ds(x)} f(x) \tag{1.2}$$

for $f \in C^2(I_0)$, where $C^2(I_0)$ is the set of all functions on $I_0 = (0, 1)$ that are twice-continuously differentiable on I_0 . In (1.2), $s(x)$ is strictly increasing, continuously differentiable, and bounded on $(0, 1)$, $s'(x) > 0$ on $(0, 1)$, and the measure $m(dx) = m(x)dx$ where $m(x)$ is continuous and $m(x) > 0$ on $(0, 1)$. Thus we can also write

$$Lf(x) = \frac{1}{m(x)} \frac{d}{dx} \frac{1}{s'(x)} \frac{d}{dx} f(x) \tag{1.3}$$

If $f \in C^2(I_0) \cap C_0(I)$ and $Lf \in C_0(I)$, then $f \in \mathcal{D}(A)$ and $Af = Lf$.

- (iv) Since $s(x)$ is strictly increasing and bounded on $(0, 1)$, we can assume without loss of generality that $s(0) = s(0+) = 0$ and $s(1) = s(1-) = 1$. We also assume

$$\int_0^{x_0} s(x)m(dx) + \int_{x_0}^1 (1 - s(x))m(dx) < \infty \tag{1.4}$$

whenever $0 < x_0 < 1$. In terms of diffusion process theory, this is the condition that the endpoints are regular or exit.

2. Some Useful Identities. (1) Define

$$g(x, y) = \frac{(s(1) - s(x \vee y))(s(x \wedge y) - s(0))}{s(1) - s(0)} \tag{2.1}$$

where $0 \leq x, y \leq 1$, $x \wedge y = \min\{x, y\}$, and $x \vee y = \max\{x, y\}$. Let

$$f(x) = \int_0^1 g(x, y)h(y)m(dy) \tag{2.2}$$

$$\begin{aligned} &= \frac{s(1) - s(x)}{s(1) - s(0)} \int_0^x (s(y) - s(0))h(y)m(dy) \\ &\quad + \frac{s(x) - s(0)}{s(1) - s(0)} \int_x^1 (s(1) - s(y))h(y)m(dy) \end{aligned} \tag{2.3}$$

where $h(y)$ is bounded on I_0 . By (2.1)

$$\begin{aligned} g(x, y) &\leq \min\{g(x, x), g(y, y)\} \\ g(y, y) &\leq \min\{s(1) - s(y), s(y) - s(0)\} \end{aligned} \tag{2.4}$$

If $h(y) \geq 0$, then by (2.2) and (2.4)

$$f(x) \leq \int_0^1 g(y, y)h(y)m(dy) \tag{2.5}$$

$$\begin{aligned} &\leq \int_0^1 \min\{s(1) - s(y), s(y) - s(0)\}h(y)m(dy) \\ &= \int_0^{x_0} (s(y) - s(0))h(y)m(dy) + \int_{x_0}^1 (s(1) - s(y))h(y)m(dy) \end{aligned} \tag{2.6}$$

where x_0 is determined by $s(x_0) - s(0) = (1/2)(s(1) - s(0))$. It then follows from (1.4) that $f(x)$ in (2.2) is uniformly bounded with

$$|f(x)| \leq C \max_y |h(y)| \tag{2.7}$$

where C is the constant on the right-hand-side value of (2.6) if $h(y)$ is replaced by 1. It follows similarly from (2.4) and the dominated convergence theorem that $f(x) \in C(I)$ with $f(0) = f(1) = 0$.

(2) If $h(x) \in C_0(I)$, then $f(x)$ in (2.2) is continuous differentiable with

$$f'(x) = s'(x) \left(\int_x^1 \frac{s(1) - s(y)}{s(1) - s(0)} h(y)m(dy) - \int_0^x \frac{s(y) - s(0)}{s(1) - s(0)} h(y)m(dy) \right)$$

Hence $f'(x)/s'(x)$ is continuously differentiable with

$$Lf(x) = \frac{1}{m(x)} \frac{d}{dx} \left(\frac{f'(x)}{s'(x)} \right) = -h(x)$$

Then, by Assumption (iii), $f \in \mathcal{D}(A)$ and $Af = Lf = -h$. In fact, it is not difficult to show that if $f(x) \in C^2(I_0) \cap C_0(I)$ and $Lf = -h \in C_0(I)$, then $f(x)$ is given by (2.2). (**Exercise:** Prove the last statement. Lemma 8.1 in Section 8 below has a second proof using martingales.)

(3) Another useful identity is

$$T_t f(x) = f(x) + \int_0^t T_s(Af)(x) ds, \quad \text{any } f \in \mathcal{D}(A) \tag{2.8}$$

This follows immediately from the definition (1.1) and the strong continuity of the semi-group.

(4) If $\|T_t\| \leq Ce^{Mt}$ and

$$R_\lambda f(x) = \int_0^\infty e^{-\lambda t} T_t f(x) dt, \quad \lambda > M \tag{2.9}$$

then applying R_λ to the identity (2.8) and rearranging terms implies

$$R_\lambda(\lambda I - A)f = f, \quad f \in \mathcal{D}(A)$$

Similarly, if $g = R_\lambda f$ for $f \in B$, it follows from the strong continuity of $\{T_t\}$ that $g \in \mathcal{D}(A)$ and $Ag = \lambda g - f$. Thus $(\lambda I - A)R_\lambda f = f$ for all $f \in B$. It follows that $\lambda I - A$ on $\mathcal{D} = \mathcal{D}(A)$ is a two-sided inverse of R_λ on B .

3. A Probability Space for T_t . By Assumptions (i) and (ii), we can write

$$T_t f(x) = \int_0^1 f(y) P(t, x, dy) \tag{3.1}$$

where $P(t, x, dy) \geq 0$ are Borel measures on the open unit interval $I_0 = (0, 1)$. The semigroup property $T_{t+s}f = T_t T_s f$ for $\{T_t\}$ is equivalent to the *Chapman-Kolmogorov equations*

$$P(t + s, x, B) = \int_0^1 P(t, y, B) P(s, x, dy) \tag{3.2}$$

for $P(t, x, dy)$. The assumptions in Section 1 imply $0 \leq P(t, x, I_0) \leq 1$, but do not exclude the possibility $P(t, x, I_0) < 1$.

The first step in finding a stochastic process corresponding to the semi-group (3.1) is to extend $P(t, x, A)$ on $I_0 = (0, 1)$ to a larger space I_Δ such that (i) $P(t, x, I_\Delta) = 1$ for all $x \in I_\Delta$ and (ii) the Chapman-Kolmogorov equations (3.2) hold on I_Δ . To do this, we first define an abstract “death point” $\Delta \notin I_0$ and set

$$I_\Delta = I_0 \cup \{\Delta\}, \quad I_0 = (0, 1) \tag{3.3}$$

We extend $P(t, x, A)$ for $x \in I_0, A \subseteq I_0$ to $x \in I_\Delta, A \subseteq I_\Delta$ by

$$\begin{aligned} P(t, x, \{\Delta\}) &= 1 - P(t, x, I_0) \\ P(t, \Delta, \{\Delta\}) &= 1, \quad P(t, \Delta, I_0) = 0 \end{aligned} \tag{3.4}$$

Then the extended function $P(t, x, A)$ for $x \in I_\Delta$ and $A \subseteq I_\Delta$ satisfies (i) $P(t, x, A) \geq 0$, (ii) $P(t, x, I_\Delta) = 1$ for all $x \in I_\Delta$, and (iii) the Chapman-Kolmogorov relations

$$P(t + s, x, B) = \int_{I_\Delta} P(t, y, B)P(s, x, dy) \tag{3.5}$$

for $x \in I_\Delta$ and $B \subseteq I_\Delta$. (**Exercise:** Prove (i,ii,iii) using (3.4) and (3.2) on I_0 .)

Let $Q = \{k/2^m : k \geq 0, m \geq 0\}$ be the set of nonnegative dyadic rationals. We will use the Kolmogorov Consistency Theorem (KCT) to construct a probability space $(\Omega, \mathcal{F}, P_x)$ with random variables $\{X_q(\omega) \in I_\Delta : q \in Q\}$ that form a Markov process consistent with (3.3), (3.4), and (3.5). The subscript x in P_x is a parameter $x \in I_\Delta$ for which $P_x(X_0 = x) = 1$. In the Kolmogorov representation, the basic sample space is the infinite product

$$\Omega = (I_C)^Q = \{\omega : \omega = \omega(r) \in I_C, r \in Q\} \tag{3.6}$$

where $I_C = [0, 1] \cup \{\Delta\}$ is a compact version of I_Δ . The random variables $X_r(\omega) = \omega(r)$ are the coordinate functions in the infinite product Ω . The sigma-algebra $\mathcal{F} = \mathcal{B}\{X_r : r \in Q\}$ is the smallest sigma-algebra of subsets of Ω with respect to which all X_r are measurable. This is also the sigma algebra generated by the *cylinder sets*

$$\Gamma(r_1, \dots, r_n, A_1, \dots, A_n) = \{\omega : X_{r_i}(\omega) \in A_i \text{ for } 1 \leq i \leq n\} \tag{3.7}$$

for $0 \leq r_1 < r_2 < \dots < r_n$ for $r_i \in Q$, sets $A_i \subseteq I_C$, and all n . The first step in applying the KCT is to define P_x for cylinder sets. The appropriate

definition in this case turns out to be the somewhat complex

$$\begin{aligned}
 &P_x\left(\Gamma(r_1, \dots, r_n, A_1, \dots, A_n)\right) \\
 &= \int_{A_1} \dots \int_{A_n} P(r_1, x, dy_1)P(r_2 - r_1, y_1, dy_2) \dots \\
 &\quad \dots \quad P(r_n - r_{n-1}, y_{n-1}, dy_n) \tag{3.8}
 \end{aligned}$$

We assume in (3.8) that $P_x(X_0 = x) = 1$ if $r_1 = 0$, $P(t, 0, \{0\}) = P(t, 1, \{1\}) = 1$ for $t \geq 0$, and $P(t, x, \{0, 1\}) = 0$ for $x \in I_\Delta = (0, 1) \cup \{\Delta\}$.

The next step is to notice that if (for example) $A_{i_0} = I_C$ in (3.7), then $P_x(\Gamma(r_1, \dots, A_1, \dots))$ is the same as if we dropped the i_0 -th coordinate in (3.7) and (3.8) and computed the probability of the resulting cylinder set with $n - 1$ conditions. (This is the *consistency condition* in the KCT.) Here the consistency condition follows from the Chapman-Kolmogorov relation

$$\int_{I_C} P(r_{i_0+1} - r_{i_0}, y, A)P(r_{i_0} - r_{i_0-1}, x, dy) = P(r_{i_0+1} - r_{i_0-1}, x, A)$$

which follows from (3.5).

The KCT then implies that there exists a unique probability measure P_x on (Ω, \mathcal{F}) such that $P_x(\Gamma(\dots))$ is given by (3.8) for cylinder sets (3.7). One can then check from (3.8) that

$$\begin{aligned}
 P_x(X_r \in I_\Delta) &= 1, & x \in I_\Delta \subset I_C, r \in Q \\
 P_x(X_r \in A) &= P(r, x, A), & x \in I_\Delta, A \subseteq I_\Delta \quad \text{and} \\
 P_x(X_q \in A \mid \mathcal{B}_r)(\omega) &= P(q - r, X_r(\omega), A) \tag{3.9}
 \end{aligned}$$

for $r, q \in Q$, $r < q$, and $x \in I_\Delta$. Here $\mathcal{B}_r = \mathcal{B}\{X_a : a \leq r, a \in Q\}$ is the smallest σ -algebra $\mathcal{B}_r \subseteq \mathcal{F}$ such that $X_a(\omega)$ is \mathcal{B}_r -measurable for $a \leq r, a \in Q$. Here $P_x(X_q \in A \mid \mathcal{B}_r)(\omega)$ is the conditional-probability random variable. That is, $g(\omega) = P_x(X_q \in A \mid \mathcal{B}_r)(\omega)$ is the unique bounded \mathcal{B}_r -measurable function such that

$$E(I_A(X_q)h) = E((P(q - r, X_r, A)h)$$

for all bounded \mathcal{B}_r -measurable random variables $h(\omega)$. (**Exercise:** Verify (3.9).)

The relations (3.9) imply that, for a set of $\omega \in \Omega$ with P_x -probability one for every $x \in I_\Delta = (0, 1) \cup \{\Delta\}$, the sample paths $\{X_r(\omega) : r \in Q\}$ start

in and remain in I_Δ . Thus $\{X_r : r \in Q\}$ is a Markov stochastic process with values in I_Δ and Markov transition function $P(t, x, A)$.

Similarly, (3.4) implies that, again for a set of ω with P_x -probability one for every $x \in I_\Delta$, $X_r(\omega) = \Delta$ for any ω implies $X_q(\omega) = \Delta$ for every $q > r$. Set $\zeta(\omega) = \min\{r \in Q : X_r(\omega) = \Delta\}$. We call $\zeta(\omega)$ the *death time* for a sample path $X_r(\omega)$ in I_0 . For these $\omega \in \Omega$, $X(r, \omega) = \Delta$ whenever $r > \zeta(\omega)$. Thus, within sets of probability zero,

$$\{\omega : \zeta(\omega) > r\} = \{\omega : X_r(\omega) \in I_0\}$$

This implies that by (3.1) and (3.3)–(3.4)

$$\begin{aligned} T_r f(x) &= \int_0^1 f(y)P(r, x, dy) = E_x(f(X_r)I_{[X_r \in I_0]}) \\ &= E_x(f(X_r)I_{[\zeta > r]}) \end{aligned} \tag{3.10}$$

The purpose for restricting r to a countable set is so that events involving the sample paths $X_r(\omega)$ depend on at most a countably infinite number of random variables. Otherwise, simple events involving sample-path continuity or even Lebesgue measurability may not be measurable.

The next step will be to extend $\{X_r : r \in Q\}$ to random variables $\{X_t : t \geq 0\}$ defined on the same probability space in such a way that events involving the process $\{X_t : t \geq 0\}$ will depend only on $\{X_r : r \in Q\}$.

4. Sample Path Regularity. As in (2.2),

$$\begin{aligned} f(x) &= \int_0^1 g(x, y)h(y)m(dy) \\ &= (1 - s(x)) \int_0^x s(y)h(y)m(dy) + s(x) \int_x^1 (1 - s(y))h(y)m(dy) \end{aligned} \tag{4.1}$$

for $h \in C_0(I)$. Define functions $\phi_N(x) \in C_0(I)$ for $N \geq 3$ such that $\phi_N(x) = 0$ for $x < 1 - 1/N$, $\phi_N(x) \geq 0$,

$$\text{Supp}(\phi_N) \subseteq \left(1 - \frac{1}{N}, 1 - \frac{1}{2N}\right)$$

and $\int_0^1 (1 - s(y))\phi_N(y)m(dy) = 1$. Define

$$f_N(x) = \int_0^1 g(x, y)\phi_N(y)m(dy) \tag{4.2}$$

Then $f_N \in \mathcal{D}(A)$, $Af_N(x) = -\phi_N(x) \leq 0$, and by (4.1)

$$f_N(x) = s(x) \quad \text{on} \quad \left(0, 1 - \frac{1}{N}\right) \tag{4.3}$$

It follows from the identity (2.8) in Section 2 that

$$\begin{aligned} T_t f_N(x) &= f_N(x) + \int_0^t T_s(Af_N)ds \\ &= f_N(x) - \int_0^t T_s\phi_N(x)ds \end{aligned} \tag{4.4}$$

Since $\phi_N(x) \geq 0$ and $f_N(x) \geq 0$ by (4.2), $0 \leq T_t f_N(x) \leq f_N(x)$ for all $t \geq 0$ and $x \in I_0$. Then by (3.10), (4.4), and the Markov property

$$\begin{aligned} E_x \left(f_N(X_q) I_{[\zeta > q]} \mid \mathcal{B}_r \right) (\omega) &= T_{q-r} f_N(X_r(\omega)) \\ &\leq f_N(X_r(\omega)) I_{[\zeta(\omega) > r]} \end{aligned} \tag{4.5}$$

where $\mathcal{B}_r = \mathcal{B}\{X_a : a \leq r, a \in Q\}$ as before. Since $0 \leq f_N(x) \leq C_N$ by (4.2) and (2.7),

$$Y_r^N(\omega) = f_N(X_r(\omega)) I_{[\zeta(\omega) > r]}, \quad 0 \leq r < \infty, r \in Q$$

is a uniformly bounded supermartingale for each N . Moreover, it follows from (4.3) and the conditional Fatou's inequality

$$E \left(\liminf_{N \rightarrow \infty} H_N \mid \mathcal{B}_r \right) \leq \liminf_{N \rightarrow \infty} E \left(H_N \mid \mathcal{B}_r \right) \text{ a.s.,}$$

for arbitrary random variables $H_N(\omega) \geq 0$ that (4.5) also holds with $f_N(x)$ replaced by $s(x)$. Thus

$$Y_r(\omega) = s(X_r(\omega)) I_{[\zeta(\omega) > r]} \quad 0 \leq r < \infty, r \in Q \tag{4.6}$$

is also a uniformly bounded supermartingale.

Doob's Upcrossing Inequality now applies to the finite subsets $Q_{NM} = \{k/2^N : 0 \leq k \leq M\}$ of Q with a uniform upper bound. This in turn implies that there exists a single null set $E \in \mathcal{F}$ such that for $\omega \notin E$, the limits

$$\lim_{q > t, q \downarrow t} X_q(\omega) = X_{t+}(\omega), \quad \lim_{q < t, q \uparrow t} X_q(\omega) = X_{t-}(\omega) \tag{4.7}$$

exist for all real values $t < \zeta(\omega)$, where the first limit exists for $t = 0$, and the second limit exists at $t = \zeta(\omega)$. It follows from (4.7) that the set

$$\{t : t < \zeta(\omega), X_{t+}(\omega) \neq X_{t-}(\omega)\}$$

is at most countably infinite for each such ω , even though the set of possible t in (4.7) is uncountably infinite. Since $X_r(\omega) = \Delta$ almost surely for $r \geq \zeta(\omega)$, it follows that (4.7) holds for all real $t \geq 0$ and $\omega \notin E_1$ for a larger null set E_1 .

We now define random variables $X_t(\omega)$ for all real $t \geq 0$ in terms of the random variables $\{X_r(\omega) : r \in Q\}$ by

$$\begin{aligned} X_t(\omega) &= \lim_{r>t, r\downarrow t} X_r(\omega) = X_{t+}(\omega), \quad 0 \leq t < \zeta(\omega) \\ &= \Delta, \quad \zeta(\omega) \leq t < \infty \end{aligned} \tag{4.8}$$

Note that this defines an uncountable number of random variables $\{X_t(\omega)\}$ in terms of a countable set $\{X_r(\omega) : r \in Q\}$. The first relation in (4.7) implies that $X_t(\omega)$ is right-continuous in t for all real $t \geq 0$. The relations (4.7) and (4.8) do not exclude $X_t(\omega) \neq X_r(\omega)$ for $t = r$. However, the strong continuity condition on $T_f(x)$ in Section 1 implies

$$\lim_{q>r, q\downarrow r} T_q f(x) = \lim_{q>r, q\downarrow r} E_x(f(X_q)I_{[\zeta>q]}) = T_r f(x) = E_x(f(X_r)I_{[\zeta>r]})$$

uniformly in $x \in I_0$ for all $f \in B = C_0(I)$. This implies $\lim_{q>r, q\downarrow r} X_q = X_r$ weakly P_x for all $x \in (0, 1)$. Since the same limit converges a.s. to X_t by (4.7), it follows that $P_x(X_t = X_r) = 1$ for $t = r$. Thus, with probability one, the sample paths $\{X_t(\omega) : t \geq 0\}$ are right continuous in t . We will use another martingale argument below to show that, in fact, almost every sample path $\{X_t(\omega)\}$ is continuous for $t < \zeta(\omega)$.

A subtlety of the definition (4.8) is that $X_t(\omega)$ is not \mathcal{B}_t -measurable for $\mathcal{B}_t = \mathcal{B}\{X_r : r \leq t\}$ even if $t = r \in Q$, since it involves a right-hand limit. However, X_t is \mathcal{B}_{t+} measurable for $\mathcal{B}_{t+} = \bigcap_{\epsilon>0} \mathcal{B}_{t+\epsilon}$. The right-continuity of sample paths from (4.8) implies that X_t is a Markov process with respect to \mathcal{B}_{t+} : That is,

$$P_x(X_{t+s} \in A \mid \mathcal{B}_{t+}) = P(s, X_t(\omega), A) \text{ a.s.} \tag{4.9}$$

This follows from the relation

$$E_x(\phi(X_{t+s})\psi(X_{t_1}, \dots, X_{t_n})) = E_x(T_{s-\epsilon}\phi(X_{t_n})\psi(X_{t_1}, \dots, X_{t_n})) \tag{4.10}$$

for all $\phi \in C_0(I)$, $\psi \in C_0(I^n)$, and $0 < t_1 < t_2 < \dots < t_n = t + \epsilon < t + s$. The relation (4.10) follows from (3.5) and (3.8). Finally, let $\epsilon \rightarrow 0$ in (4.10) and use the strong continuity of $\{T_t\}$ and the right-continuity of sample paths $X_t(\omega)$. (**Exercise:** Carry out the details.)

5. A Family of Martingales. The key step in the last step was identifying a bounded supermartingale. Most of the following results will also depend on martingale arguments. Before continuing, we state a result about a set of martingales that we will use several times that arises naturally for any Markov process.

To keep things simple, first consider a discrete-time Markov process X_n on a state space J with a transition function

$$P(X_{n+1} \in B \mid X_n = x) = \pi(x, B)$$

Since X_n is a Markov process,

$$P(X_{n+1} \in B \mid \mathcal{B}(X_1, X_2, \dots, X_n)) = \pi(X_n(\omega), B) \tag{5.1}$$

Define $Tf(x) = \int_J f(y)\pi(x, dy) = E_x(f(X_1))$ and $Af(x) = Tf(x) - f(x)$. Then $T^n f(x) = E_x(f(X_n))$ by the Markov property, and

Lemma 5.1. For any bounded function $f(x)$ on J , the process

$$Y_n = f(X_n) - \sum_{k=0}^{n-1} Af(X_k) \tag{5.2}$$

is a martingale with respect to the σ -algebras $\mathcal{B}_n = \mathcal{B}(X_1, X_2, \dots, X_n)$.

Proof. Assume $n < m$. Clearly $\mathcal{B}_n \subseteq \mathcal{B}_m$ and Y_n is \mathcal{B}_n -measurable. Thus it only remains to prove $E(Y_m \mid \mathcal{B}_n) = Y_n$ in order to show that $\{Y_n, \mathcal{B}_n\}$ is a martingale. By (5.2)

$$\begin{aligned} Y_m &= f(X_m) - \sum_{k=n}^{m-1} Af(X_k) - \sum_{k=0}^{n-1} Af(X_k) \\ &= Y_m^{(1)} - Y_m^{(2)} - Y_m^{(3)} \end{aligned}$$

Then

$$\begin{aligned} E(Y_m^{(1)} \mid \mathcal{B}_n) &= E(f(X_m) \mid \mathcal{B}_n) = T_{m-n}f(X_n) \\ E(Y_m^{(2)} \mid \mathcal{B}_n) &= \sum_{k=n}^{m-1} E(Af(X_k) \mid \mathcal{B}_n) = \sum_{k=n}^{m-1} T_{k-n}Af(X_n) \end{aligned}$$

by (5.1) and the Markov property for X_n . Since $Af = Tf - f$,

$$E(Y_m^{(1)} \mid \mathcal{B}_n) - E(Y_m^{(2)} \mid \mathcal{B}_n) = T^{m-n}f(X_n) - \sum_{k=n}^{m-1} T^{k-n}Af(X_n)$$

$$\begin{aligned}
 &= T^{m-n} f(X_n) - \sum_{k=n}^{m-1} (T^{k-n+1} f(X_n) - T^{k-n} f(X_n)) \\
 &= f(X_n)
 \end{aligned}$$

and $E(Y_m^{(1)} - Y_m^{(2)} \mid \mathcal{B}_n) = f(X_n)$. Since $Y_m^{(3)} = \sum_{k=0}^{n-1} Af(X_k)$ is \mathcal{B}_n -measurable,

$$\begin{aligned}
 E(Y_m \mid \mathcal{B}_n) &= E(Y_m^{(1)} - Y_m^{(2)} \mid \mathcal{B}_n) - E(Y_m^{(3)} \mid \mathcal{B}_n) \\
 &= f(X_n) - \sum_{k=0}^{n-1} g(X_k) = Y_n
 \end{aligned}$$

This completes the proof that $\{Y_n, \mathcal{B}_n\}$ is a martingale.

Lemma 5.2. For any function $f \in \mathcal{D}(A)$, the process

$$Y_t(\omega) = f(X_t(\omega))I_{[\zeta(\omega) > t]} - \int_0^{t \wedge \zeta(\omega)} Af(X_s(\omega))ds \tag{5.3}$$

is a martingale with respect to the sigma-algebras $\{\mathcal{B}_{t+}\}$.

Proof. By (2.8)

$$T_t f(x) = f(x) + \int_0^t T_s Af(x)ds \tag{5.4}$$

for any $f \in \mathcal{D}(A)$. Since Δ is a trap, any $f \in \mathcal{D}(A)$ for the semigroup $T_t f(x)$ for $f(x)$ on $I_0 = (0, 1)$ can be extended to $f \in \mathcal{D}(A)$ for $f(x)$ on $I_\Delta = (0, 1) \cup \{\Delta\}$ by setting $f(\Delta) = 0$. By essentially the same argument as in the proof of Lemma 5.1 with (5.4) replacing a telescoping sum, this implies that the process

$$Y_t = f(X_t) - \int_0^t Af(X_s)ds \tag{5.5}$$

is a continuous-parameter martingale with respect to \mathcal{B}_{t+} . Since Δ is a trap, $Af(\Delta) = 0$ for any $f \in \mathcal{D}(A)$. Thus the martingale (5.5) is the same as the process in (5.3).

Exercise. Complete the proof that (5.5) is a continuous-time martingale with respect to \mathcal{B}_{t+} . Give justifications for interchanging the order of integrals and conditional expectations.

6. An Upper Bound for Exit and Death Times. Let

$$T_N(\omega) = \min \left\{ t : t < \zeta(\omega), X_t(\omega) \in \left(0, \frac{1}{N}\right) \cup \left(1 - \frac{1}{N}, 1\right) \right\} \quad (6.1)$$

be the first time that $X_t(\omega)$ exits from the closed interval $[1/N, 1 - 1/N]$, assuming that it remains within $I_0 = (0, 1)$.

We make the the convention that $T_N(\omega) = \infty$ if the right-hand side of (6.1) is empty; that is, if $X_t(\omega) \notin (0, 1/N) \cup (1 - 1/N, 1)$ for all $t < \zeta(\omega)$. If $T_N(\omega) < \infty$, then $T_N(\omega) < \zeta(\omega)$.

Note that $\{\omega : T_N(\omega) > t\} \in \mathcal{B}_{t+}$. We may not have $\{\omega : T_N(\omega) > t\} \in \mathcal{B}_t$ in general, either because $t \notin Q$ (recall $\mathcal{B}_t = \mathcal{B}\{X_q : q \leq t, q \in Q\}$) or because $t = q \in Q$ but $X_t = X_{q+} \neq X_q$. Thus T_N may not be a \mathcal{B}_t -stopping time, but is a \mathcal{B}_{t+} -stopping time. However, Lemma 5.2 implies that $Y_t(\omega)$ are martingales with respect to \mathcal{B}_{t+} , so that this is enough.

We then have

Theorem 6.1. If C is the constant in (2.7),

$$\max_{0 \leq x \leq 1} E_x(T_\infty \wedge \zeta) \leq C, \quad T_\infty(\omega) = \sup_N T_N(\omega) \quad (6.2)$$

where $T_\infty \wedge \zeta = \min\{T_\infty, \zeta\}$.

Thus, with P_x -probability one for any $x \in I_0$, either $T_\infty(\omega) < \infty$ or $\zeta(\omega) < \infty$. In particular, the only way that $T_\infty(\omega) = \infty$ can happen is if $\zeta(\omega) < \infty$.

Proof of Theorem 6.1. Define continuous functions $\psi_N \in C_0(I)$ such that $0 \leq \psi_N(y) \leq 1$ and $\psi_N(y) = 1$ on $(1/N, 1 - 1/N)$. If

$$h_N(x) = \int_0^1 g(x, y)\psi_N(y)m(dy) \quad (6.3)$$

then $h_N \in \mathcal{D}(A)$ and $Ah_N = -\psi_N \leq 0$. In particular, $Ah_N(x) = -1$ on $(1/N, 1 - 1/N)$. Thus by Lemma 5.2

$$Y_t^N = h_N(X_t)I_{[\zeta > t]} + \int_0^{t \wedge \zeta} \psi_N(X_s)ds \quad (6.4)$$

is a nonnegative martingale for any $N \geq 3$. Then by the Optional Stopping Theorem,

$$\begin{aligned} Y_{t \wedge T_N}^N &= h_N(X_{t \wedge T_N})I_{[\zeta > t \wedge T_N]} + \int_0^{t \wedge T_N \wedge \zeta} \psi_N(X_s)ds \\ &= h_N(X_{t \wedge T_N})I_{[\zeta > t \wedge T_N]} + t \wedge T_N \wedge \zeta \end{aligned}$$

are also martingales, since $\psi(X_s) = 1$ for $0 \leq s < T_N \wedge \zeta$. This in turn implies

$$E_x(t \wedge T_N \wedge \zeta) = h_N(x) - E_x(h_N(X_{t \wedge T_N \wedge \zeta})) \leq C \tag{6.5}$$

since $0 \leq \psi_N(x) \leq 1$ implies $0 \leq h_N(x) \leq C$ by (6.3) where C is the same constant as in (2.7). Fatou's theorem applied to t and N (in any order) then implies (6.2).

7. Exit Times are Less Than Death Times. Death times $\zeta(\omega)$ are necessary for general Markov processes, but can be shown to be less mysterious here. As in (6.1), let

$$T_N(\omega) = \min \left\{ t : t < \zeta(\omega), X_t(\omega) \in \left(0, \frac{1}{N}\right) \cup \left(1 - \frac{1}{N}, 1\right) \right\} \tag{7.1}$$

We show below that $P_x(T_N < \zeta) = 1$, so that $T_N(\omega) < \zeta(\omega)$ almost surely. This implies that $T_\infty(\omega) = \sup_N T_N(\omega) \leq \zeta(\omega)$ with probability one. We will show in a later section that, in fact, $T_\infty(\omega) = \zeta(\omega)$ almost surely.

The same argument as in Theorem 7.1 can also be used to show that the sample paths $\{X_t(\omega)\}$ are continuous for $0 \leq t < \zeta(\omega)$ (see Theorem 7.2).

Theorem 7.1. For the process $\{X_t\}$ defined above,

$$P_x(T_N < \zeta) = 1 \quad \text{for all } x \in (0, 1) \tag{7.2}$$

Corollary 7.1.

$$\max_{0 \leq x \leq 1} E_x(T_\infty) \leq C, \quad T_\infty(\omega) = \sup_N T_N(\omega) \tag{7.3}$$

Moreover, $\lim_{N \rightarrow \infty} X_{T_N}(\omega) = A$ exists almost surely where A (depending on ω) is either 0 or 1.

Proof. Equations (6.2) and (7.2) imply (7.3), $\lim_{N \rightarrow \infty} X_{T_N}$ exists a.s. by (4.7), and $X_{T_N} \in (0, 1/N) \cup (1 - 1/N, 1)$ by right-continuity of the sample paths.

Proof of Theorem 7.1. It is sufficient to assume $1/N < x < 1 - 1/N$ since $T_N(\omega) = 0$ if $x \in (0, 1/N) \cup (1 - 1/N, 1)$. It follows from (3.10) and the strong continuity of the semi-group $\{T_t\}$ that $P_x(\zeta > 0) = 1$ for arbitrary $x \in I_0$.

Let $\psi_N(x) \in C_0(I) \cap C^2(I)$ such that $0 \leq \psi_N(x) \leq 1$, $\psi_N(x) = 1$ on $(1/N, 1 - 1/N)$, and $0 \leq \psi(x) < 1$ on $(0, 1/N) \cup (1 - 1/N, 1)$. (See the Exercise below.) Then $\psi_N \in \mathcal{D}(A)$ with $A\psi_N(x) = L\psi_N(x)$. By Lemma 5.2

$$Y_t^N(\omega) = \psi_N(X_t(\omega))I_{[\zeta(\omega) > t]} - \int_0^{t \wedge \zeta(\omega)} L\psi_N(X_s(\omega))ds \tag{7.4}$$

is a martingale. By the Optional Stopping Theorem

$$E_x\left(Y_{t \wedge T_N}^N\right) = E_x\left(\psi_N\left(X_{t \wedge T_N}(\omega)\right)I_{[\zeta(\omega) > t \wedge T_N]}\right) = \psi_N(x) \tag{7.5}$$

since $A\psi_N(X_s) = L\psi_N(X_s) = 0$ for $0 \leq s < t \wedge T_N$.

Let $q_N(t, \omega)$ be the integrand in (7.5). Note that $0 \leq t \wedge T_N \leq T_N$. If $T_N < \zeta$, then $\lim_{t \rightarrow \infty} q_N(t, \omega) = \tilde{\psi}_N = \psi_N(X_{T_N})$. If $T_N \geq \zeta$, then $\zeta(\omega) < \infty$ by (6.2) and $\lim_{t \rightarrow \infty} q_N(t, \omega) = 0$. In either case

$$\psi_N(x) = E_x\left(\tilde{\psi}_N I_{[\zeta(\omega) > T_N]}\right) \tag{7.6}$$

where $\tilde{\psi}_N = 0$ (for example) if $T_N \geq \zeta$. If $1/N < x < 1 - 1/N$, then the expected value $\psi_N(x) = 1$. Since the integrand of (7.6) is bounded by one, it must be equal to one a.s. This implies $P_x(\zeta(\omega) > T_N) = 1$, which completes the proof of Theorem 7.1.

Exercise. Let $k_N(x) = Nk(Nx)$ for

$$k(x) = \begin{cases} C \exp(-1/(1 - x^2)) & \text{for } |x| < 1 \\ 0 & \text{for } |x| \geq 1 \end{cases}$$

where C is chosen so that $\int_{-\infty}^{\infty} k(y)dy = 1$. Prove that (i) $k(x)$ has continuous derivatives of all orders on the real line R (that is, $k \in C^\infty(R)$) and (ii) $\text{Supp}(k) = [-1, 1]$. Now set

$$\psi_N(x) = \int_{1/2N}^{1-1/2N} k_{2N}(x - y)dy$$

Prove that (iii) $\psi_N(x)$ has continuous derivatives of all orders on R , (iv) $0 \leq \psi_N(x) \leq 1$ for all x , (v) $\psi_N(x) = 1$ for $1/N \leq x \leq 1 - 1/N$, and (vi) $\psi_N(x) = 0$ if $x \leq 0$ or $x \geq 1$.

Remark. It follows from (7.3) that

$$\max_{0 \leq x \leq 1} P_x(T_\infty \geq t_0) \leq 1/2, \quad t_0 = 2C \tag{7.7}$$

This implies by the Markov property and standard arguments that

Corollary 7.2. Under the assumptions of Theorem 7.1

$$\max_{0 \leq x \leq 1} P_x(T_\infty \geq t) \leq 2e^{-\alpha t}, \quad t \geq 0 \tag{7.8}$$

for $\alpha = 2C \log 2 > 0$, and also

$$\max_{0 \leq x \leq 1} E_x(e^{aT_\infty}) \leq C_1(a), \quad 0 \leq a < \alpha \tag{7.9}$$

where $C_1(a) < \infty$ for $0 \leq a < \alpha$.

We can use the same argument as in Theorem 7.1 to prove

Theorem 7.2. With probability one, the sample paths $X_t(\omega)$ are continuous functions of t for $0 \leq t < \zeta(\omega)$.

Proof. Since $P_x(T_N < \zeta) = 1$ and $E_x(\zeta) \leq C$, it follows from (7.6) that $\tilde{\psi}_N = \psi_N(X_{T_N}) = 1$ a.s. in (7.6) for $1/N < x < 1 - 1/N$. This implies that $X_{T_N}(\omega) \in \{1/N, 1 - 1/N\}$ almost surely. This means that $X_t(\omega)$ cannot leave $(1/N, 1 - 1/N)$ by a jump beyond the boundary of $(1/N, 1 - 1/N)$. This suggests that, at least, $X_t(\omega)$ cannot be discontinuous at $X_t(\omega) = 1/N$ or $X_t(\omega) = 1 - 1/N$.

If general, define

$$T_{ab}(\omega) = \min\{ t : t < \zeta(\omega), X_t(\omega) \in (0, a) \cup (b, 1) \} \tag{7.10}$$

for rational values a, b with $0 < a < b < 1$ with, as before, the convention $T_{ab}(\omega) = \infty$ if the set of t on the right-hand side of (7.10) is empty. By the same argument as in Theorem 7.1, $P_x(T_{ab} < \infty) = 1$ and $P_x(X_{T_{ab}} \in \{a, b\}) = 1$ for $0 < a < x < b < 1$.

Expand the null set $E \in \mathcal{F}$ in (4.7) to include the null set of all $\omega \in \Omega$ that provide counterexamples to $X_{T_{ab}(\omega)}(\omega) \in \{a, b\}$ for rational $0 < a < b < 1$. It follows that, if $\omega \notin E$, there cannot exist any rational values r such that $X_{t-}(\omega) < r < X_{t+}(\omega)$, since we could then find a counterexample to $X_{T_{ar}(\omega)}(\omega) \in \{a, r\}$ for some rational interval (a, r) .

Similarly there cannot exist any rational values r such that $X_{t+}(\omega) < r < X_{t-}(\omega)$ since we could find a counterexample to $X_{T_{rb}(\omega)}(\omega) \in \{r, b\}$ for some rational interval (r, b) . Since $X_t(\omega)$ has at worst jump discontinuities for $\omega \notin E$ by the argument that led up to (4.7), this implies that $X_t(\omega)$ must be continuous in t for $0 \leq t < \zeta(\omega)$. This completes the proof of Theorem 7.2.

8. Exit Times are Death Times. The purpose of this section is to prove that, in fact, $T_\infty(\omega) = \zeta(\omega)$ almost surely for $T_\infty(\omega) = \sup_N T_N(\omega)$. We begin with two lemmas.

Lemma 8.1. If $f, h \in C_0(I)$, then $h \in \mathcal{D}(A)$ and $Ah = -f$ if and only if

$$h(x) = \int_0^1 g(x, y)f(y)m(dy) \tag{8.1}$$

Proof. If $h(x)$ is given by (8.1), then $h \in \mathcal{D}(A)$ and $Ah = Lh = -f$ as in Section 1. Conversely, assume $h \in \mathcal{D}(A)$ and $Ah = -f$. Define

$$h_1(x) = \int_0^1 g(x, y)f(y)m(dy)$$

Then $h_1 \in \mathcal{D}(A)$ and $Ah_1 = Lh_1 = -f$. Let $g = h - h_1$. Then $g \in \mathcal{D}(A)$ and $Ag = Ah - Ah_1 = 0$.

If $g \in \mathcal{D}(A)$ and $Ag = 0$, it follows from Lemma 5.2 that $g(X_t)I_{[\zeta > t]}$ is a uniformly-bounded martingale. Thus

$$g(x) = E_x(g(X_t)I_{[\zeta > t]})$$

By the relation $P_x(T_N < \zeta) = 1$ and the Optional Stopping Theorem,

$$g(x) = E_x(g(X_{t \wedge T_N}))$$

Since $P_x(T_N < \infty) = 1$ by Theorems 6.1 and 7.1 and $g(x)$ is bounded, we can let $t \rightarrow \infty$ and obtain

$$g(x) = E_x(g(X_{T_N}))$$

However, $\lim_{N \rightarrow \infty} g(X_{T_N}) = 0$ since $g \in C_0(I)$ by Corollary 7.1. This proves that $g(x) = 0$ for $0 < x < 1$. Since $g = h - h_1$, it follows that $h(x) = h_1(x) = \int g(x, y)f(y)m(dy)$. This completes the proof of the lemma.

Lemma 8.2. For any $f \in C_0(I)$ with $f(x) \geq 0$,

$$\int_0^1 g(x, y)f(y)m(dy) = \int_0^\infty T_s f(x)ds = E_x\left(\int_0^{\zeta(\omega)} f(X_s)ds\right) \quad (8.2)$$

Proof. Define

$$h_\lambda(x) = R_\lambda f(x) = \int_0^\infty e^{-\lambda s} T_s f(x)ds$$

As in Section 2, $h_\lambda(x) \in \mathcal{D}(A)$ and $(\lambda I - A)h_\lambda(x) = f(x)$. Thus $Ah_\lambda = \lambda h_\lambda - f = -(f - \lambda h_\lambda)$. It then follows from Lemma 8.1 that

$$\begin{aligned} h_\lambda(x) &= \int_0^\infty e^{-\lambda s} T_s f(x)ds = \int_0^1 g(x, y)(f(y) - \lambda h_\lambda(y))m(dy) \quad (8.3) \\ &\leq \int_0^1 g(x, y)f(y)m(dy) \leq C \sup_y f(y) \end{aligned}$$

since $f(x) \geq 0$ and $h_\lambda(y) \geq 0$, where C is the constant in (2.7). Thus $h_\lambda(x) \leq C_1$ uniformly in $\lambda > 0$. Since then $\lambda h_\lambda(x) \leq C_1 \lambda$, taking the limit $\lambda \rightarrow 0$ in (8.3) for $f(x) \geq 0$ implies

$$\begin{aligned} h_0(x) &= \int_0^\infty T_s f(x) ds = \int_0^1 g(x, y) f(y) m(dy) \\ &= E_x \left(\int_0^\infty f(X_s) I_{[\zeta > s]} ds \right) = E_x \left(\int_0^{\zeta(\omega)} f(X_s) ds \right) \end{aligned}$$

This implies (8.2).

Theorem 8.1. Define T_N , T_∞ , and ζ as before. Then $P_x(T_\infty = \zeta) = 1$ for all $x \in I_0$.

Proof. Let

$$h(x) = \int_0^1 g(x, y) \psi(y) m(dy) \tag{8.4}$$

for arbitrary $\psi \in C_0(I)$ with $\psi(y) \geq 0$. Then $Ah = -\psi$ and

$$Y_t(\omega) = h(X_t(\omega)) I_{[\zeta(\omega) > t]} + \int_0^{t \wedge \zeta(\omega)} \psi(X_s(\omega)) ds$$

is a nonnegative martingale by Lemma 5.2. Thus

$$E_x \left(h(X_t) I_{[\zeta > t]} + \int_0^{t \wedge \zeta} \psi(X_s) ds \right) = h(x)$$

for all $t \geq 0$ and $x \in I_0$. By the Optional Stopping Theorem

$$\begin{aligned} h(x) &= E_x \left(h(X_{t \wedge T_N}) I_{[\zeta > t \wedge T_N]} + \int_0^{t \wedge T_N \wedge \zeta} \psi(X_s) ds \right) \\ &= E_x \left(h(X_{t \wedge T_N}) + \int_0^{t \wedge T_N} \psi(X_s) ds \right) \end{aligned}$$

since $P_x(T_N < \zeta) = 1$ by (7.2). Since $h(x)$ is bounded, $\psi(y) \geq 0$, and $T_N < \infty$ almost surely by (6.2), we can let $t \rightarrow \infty$ and conclude

$$h(x) = E_x \left(h(X_{T_N}) + \int_0^{T_N} \psi(X_s) ds \right) \tag{8.5}$$

As $N \rightarrow \infty$, $h(X_{T_N}) \rightarrow 0$ since $h \in C_0(I)$. Since $0 \leq T_N \uparrow T_\infty$ and $\psi(X_s) \geq 0$, we conclude from (8.5) that

$$h(x) = E_x \left(\int_0^{T_\infty(\omega)} \psi(X_s) ds \right) \tag{8.6}$$

In contrast, by Lemma 8.2 and (8.4)

$$h(x) = E_x \left(\int_0^{\zeta(\omega)} \psi(X_s) ds \right) \tag{8.7}$$

Recall $T_\infty(\omega) \leq \zeta(\omega)$ almost surely by (7.2). Both integrals in (8.6) and (8.7) are finite since $\psi(X_s) \geq 0$ and $h(x)$ is finite. We can then subtract (8.6) from (8.7) to conclude

$$E_x \left(\int_{T_\infty(\omega)}^{\zeta(\omega)} \psi(X_s) ds \right) = 0$$

This holds for all $\psi \in C_0(I)$ with $\psi(x) \geq 0$. We can approximate $\psi(x) \equiv 1$ by nonnegative functions $\psi_N \in C_0(I)$ by setting (for example) $\psi_0(x) = 2 \min\{x, 1 - x\}$ and $\psi_N(x) = \psi_0(x)^{1/N}$. Then $0 \leq \psi_N(x) \leq 1$ and $0 \leq \psi_N(x) \uparrow 1$ as $N \rightarrow \infty$. This implies

$$E_x \left(\zeta(\omega) - T_\infty(\omega) \right) = 0, \quad T_\infty(\omega) \leq \zeta(\omega) \text{ a.s.} \tag{8.8}$$

Thus $\zeta(\omega) - T_\infty(\omega) = 0$ almost surely with respect to P_x for all $x \in I_0$. This completes the proof of Theorem 8.1. (**Exercise:** Prove or disprove: (8.8) also holds for $x = \Delta$. Use the precise definitions of $X_r(\omega)$, $X_t(\omega)$, and $T_N(\omega)$. Discuss.)

References.

1. Feller, W. (1952) The parabolic differential equations and the associated semi-groups of transformations. *Annals of Mathematics* **55**, 2nd Ser., 468–519.
2. Ito, K., and J. P. McKean, Jr (1965) Diffusion processes and their sample paths. Springer-Verlag.
3. Sawyer, S. A. (1974) A Fatou theorem for the general one-dimensional parabolic equation. *Indiana Univ. Math. J.* **24**, 451–498.